GNU Scientific Library

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The money raised from sales of the manual helps support the development of GSL.

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Chapter 1: Introduction

1

1 Introduction

The GNU Scientific Library (GSL) is a collection of routines for numerical computing. The routines have been written from scratch in C, and present a modern Applications Programming Interface (API) for C programmers, allowing wrappers to be written for very high level languages. The source code is distributed under the GNU General Public License.

1.1 Routines available in GSL

The library covers a wide range of topics in numerical computing. Routines are available for the following areas,

Complex Numbers Roots of Polynomials Special Functions Vectors and Matrices

Permutations Combinations
Sorting BLAS Support
Linear Algebra CBLAS Library
Fast Fourier Transforms Eigensystems
Random Numbers Quadrature

Random Distributions Quasi-Random Sequences

Histograms Statistics
Monte Carlo Integration N-Tuples

Differential Equations Simulated Annealing

Numerical Differentiation Interpolation

Series Acceleration Chebyshev Approximations Root-Finding Discrete Hankel Transforms

Least-Squares Fitting Minimization
IEEE Floating-Point Physical Constants

Basis Splines Wavelets

The use of these routines is described in this manual. Each chapter provides detailed definitions of the functions, followed by example programs and references to the articles on which the algorithms are based.

Where possible the routines have been based on reliable public-domain packages such as FFTPACK and QUADPACK, which the developers of GSL have reimplemented in C with modern coding conventions.

1.2 GSL is Free Software

The subroutines in the GNU Scientific Library are "free software"; this means that everyone is free to use them, and to redistribute them in other free programs. The library is not in the public domain; it is copyrighted and there are conditions on its distribution. These conditions are designed to permit everything that a good cooperating citizen would want to do. What is not allowed is to try to prevent others from further sharing any version of the software that they might get from you.

Specifically, we want to make sure that you have the right to share copies of programs that you are given which use the GNU Scientific Library, that you receive their source code

or else can get it if you want it, that you can change these programs or use pieces of them in new free programs, and that you know you can do these things.

To make sure that everyone has such rights, we have to forbid you to deprive anyone else of these rights. For example, if you distribute copies of any code which uses the GNU Scientific Library, you must give the recipients all the rights that you have received. You must make sure that they, too, receive or can get the source code, both to the library and the code which uses it. And you must tell them their rights. This means that the library should not be redistributed in proprietary programs.

Also, for our own protection, we must make certain that everyone finds out that there is no warranty for the GNU Scientific Library. If these programs are modified by someone else and passed on, we want their recipients to know that what they have is not what we distributed, so that any problems introduced by others will not reflect on our reputation.

The precise conditions for the distribution of software related to the GNU Scientific Library are found in the GNU General Public License (see [GNU General Public License], page 476). Further information about this license is available from the GNU Project webpage Frequently Asked Questions about the GNU GPL,

```
http://www.gnu.org/copyleft/gpl-faq.html
```

The Free Software Foundation also operates a license consulting service for commercial users (contact details available from http://www.fsf.org/).

1.3 Obtaining GSL

The source code for the library can be obtained in different ways, by copying it from a friend, purchasing it on CDROM or downloading it from the internet. A list of public ftp servers which carry the source code can be found on the GNU website,

```
http://www.gnu.org/software/gsl/
```

The preferred platform for the library is a GNU system, which allows it to take advantage of additional features in the GNU C compiler and GNU C library. However, the library is fully portable and should compile on most systems with a C compiler.

Announcements of new releases, updates and other relevant events are made on the info-gsl@gnu.org mailing list. To subscribe to this low-volume list, send an email of the following form:

```
To: info-gsl-request@gnu.org
Subject: subscribe
```

You will receive a response asking you to reply in order to confirm your subscription.

1.4 No Warranty

The software described in this manual has no warranty, it is provided "as is". It is your responsibility to validate the behavior of the routines and their accuracy using the source code provided, or to purchase support and warranties from commercial redistributors. Consult the GNU General Public license for further details (see [GNU General Public License], page 476).

1.5 Reporting Bugs

A list of known bugs can be found in the BUGS file included in the GSL distribution or online in the GSL bug tracker.¹ Details of compilation problems can be found in the INSTALL file.

If you find a bug which is not listed in these files, please report it to bug-gsl@gnu.org. All bug reports should include:

- The version number of GSL
- The hardware and operating system
- The compiler used, including version number and compilation options
- A description of the bug behavior
- A short program which exercises the bug

It is useful if you can check whether the same problem occurs when the library is compiled without optimization. Thank you.

Any errors or omissions in this manual can also be reported to the same address.

1.6 Further Information

Additional information, including online copies of this manual, links to related projects, and mailing list archives are available from the website mentioned above.

Any questions about the use and installation of the library can be asked on the mailing list help-gsl@gnu.org. To subscribe to this list, send an email of the following form:

```
To: help-gsl-request@gnu.org
Subject: subscribe
```

This mailing list can be used to ask questions not covered by this manual, and to contact the developers of the library.

If you would like to refer to the GNU Scientific Library in a journal article, the recommended way is to cite this reference manual, e.g. M. Galassi et al, GNU Scientific Library Reference Manual (3rd Ed.), ISBN 0954612078.

If you want to give a url, use "http://www.gnu.org/software/gsl/".

1.7 Conventions used in this manual

This manual contains many examples which can be typed at the keyboard. A command entered at the terminal is shown like this,

\$ command

The first character on the line is the terminal prompt, and should not be typed. The dollar sign '\$' is used as the standard prompt in this manual, although some systems may use a different character.

The examples assume the use of the GNU operating system. There may be minor differences in the output on other systems. The commands for setting environment variables use the Bourne shell syntax of the standard GNU shell (bash).

¹ http://savannah.gnu.org/bugs/?group=gsl

2 Using the library

This chapter describes how to compile programs that use GSL, and introduces its conventions.

2.1 An Example Program

The following short program demonstrates the use of the library by computing the value of the Bessel function $J_0(x)$ for x = 5,

```
#include <stdio.h>
#include <gsl/gsl_sf_bessel.h>

int
main (void)
{
   double x = 5.0;
   double y = gsl_sf_bessel_J0 (x);
   printf ("J0(%g) = %.18e\n", x, y);
   return 0;
}
```

The output is shown below, and should be correct to double-precision accuracy,¹

```
J0(5) = -1.775967713143382920e-01
```

The steps needed to compile this program are described in the following sections.

2.2 Compiling and Linking

The library header files are installed in their own gsl directory. You should write any preprocessor include statements with a gsl/ directory prefix thus,

```
#include <gsl/gsl_math.h>
```

If the directory is not installed on the standard search path of your compiler you will also need to provide its location to the preprocessor as a command line flag. The default location of the gsl directory is /usr/local/include/gsl. A typical compilation command for a source file example.c with the GNU C compiler gcc is,

```
$ gcc -Wall -I/usr/local/include -c example.c
```

This results in an object file example.o. The default include path for gcc searches /usr/local/include automatically so the -I option can actually be omitted when GSL is installed in its default location.

2.2.1 Linking programs with the library

The library is installed as a single file, libgsl.a. A shared version of the library libgsl.so is also installed on systems that support shared libraries. The default location of these files is /usr/local/lib. If this directory is not on the standard search path of your linker you will also need to provide its location as a command line flag.

¹ The last few digits may vary slightly depending on the compiler and platform used—this is normal.

To link against the library you need to specify both the main library and a supporting CBLAS library, which provides standard basic linear algebra subroutines. A suitable CBLAS implementation is provided in the library libgslcblas.a if your system does not provide one. The following example shows how to link an application with the library,

```
$ gcc -L/usr/local/lib example.o -lgsl -lgslcblas -lm
```

The default library path for gcc searches /usr/local/lib automatically so the -L option can be omitted when GSL is installed in its default location.

The option -lm links with the system math library. On some systems it is not needed.²

For a tutorial introduction to the GNU C Compiler and related programs, see $An\ Introduction\ to\ GCC\ (ISBN\ 0954161793).^3$

2.2.2 Linking with an alternative BLAS library

The following command line shows how you would link the same application with an alternative CBLAS library libcblas.a,

```
$ gcc example.o -lgsl -lcblas -lm
```

For the best performance an optimized platform-specific CBLAS library should be used for -lcblas. The library must conform to the CBLAS standard. The ATLAS package provides a portable high-performance BLAS library with a CBLAS interface. It is free software and should be installed for any work requiring fast vector and matrix operations. The following command line will link with the ATLAS library and its CBLAS interface,

```
$ gcc example.o -lgsl -lcblas -latlas -lm
```

If the ATLAS library is installed in a non-standard directory use the -L option to add it to the search path, as described above.

For more information about BLAS functions see Chapter 13 [BLAS Support], page 121.

2.3 Shared Libraries

To run a program linked with the shared version of the library the operating system must be able to locate the corresponding <code>.so</code> file at runtime. If the library cannot be found, the following error will occur:

```
$ ./a.out
./a.out: error while loading shared libraries:
libgsl.so.0: cannot open shared object file: No such
file or directory
```

To avoid this error, either modify the system dynamic linker configuration⁴ or define the shell variable LD_LIBRARY_PATH to include the directory where the library is installed.

For example, in the Bourne shell (/bin/sh or /bin/bash), the library search path can be set with the following commands:

```
$ LD_LIBRARY_PATH=/usr/local/lib
```

^{\$} export LD_LIBRARY_PATH

² It is not needed on MacOS X.

³ http://www.network-theory.co.uk/gcc/intro/

⁴ /etc/ld.so.conf on GNU/Linux systems.

\$./example

In the C-shell (/bin/csh or /bin/tcsh) the equivalent command is,

% setenv LD_LIBRARY_PATH /usr/local/lib

The standard prompt for the C-shell in the example above is the percent character '%', and should not be typed as part of the command.

To save retyping these commands each session they can be placed in an individual or system-wide login file.

To compile a statically linked version of the program, use the -static flag in gcc,

\$ gcc -static example.o -lgsl -lgslcblas -lm

2.4 ANSI C Compliance

The library is written in ANSI C and is intended to conform to the ANSI C standard (C89). It should be portable to any system with a working ANSI C compiler.

The library does not rely on any non-ANSI extensions in the interface it exports to the user. Programs you write using GSL can be ANSI compliant. Extensions which can be used in a way compatible with pure ANSI C are supported, however, via conditional compilation. This allows the library to take advantage of compiler extensions on those platforms which support them.

When an ANSI C feature is known to be broken on a particular system the library will exclude any related functions at compile-time. This should make it impossible to link a program that would use these functions and give incorrect results.

To avoid namespace conflicts all exported function names and variables have the prefix gsl_, while exported macros have the prefix GSL_.

2.5 Inline functions

The inline keyword is not part of the original ANSI C standard (C89) so the library does not export any inline function definitions by default. Inline functions were introduced officially in the newer C99 standard but most C89 compilers have also included inline as an extension for a long time.

To allow the use of inline functions, the library provides optional inline versions of performance-critical routines by conditional compilation in the exported header files. The inline versions of these functions can be included by defining the macro <code>HAVE_INLINE</code> when compiling an application,

\$ gcc -Wall -c -DHAVE_INLINE example.c

If you use autoconf this macro can be defined automatically. If you do not define the macro HAVE_INLINE then the slower non-inlined versions of the functions will be used instead.

By default, the actual form of the inline keyword is extern inline, which is a gcc extension that eliminates unnecessary function definitions. If the form extern inline causes problems with other compilers a stricter autoconf test can be used, see Appendix C [Autoconf Macros], page 461.

When compiling with gcc in C99 mode (gcc -std=c99) the header files automatically switch to C99-compatible inline function declarations instead of extern inline. With other C99 compilers, define the macro GSL_C99_INLINE to use these declarations.

2.6 Long double

In general, the algorithms in the library are written for double precision only. The long double type is not supported for actual computation.

One reason for this choice is that the precision of long double is platform dependent. The IEEE standard only specifies the minimum precision of extended precision numbers, while the precision of double is the same on all platforms.

However, it is sometimes necessary to interact with external data in long-double format, so the vector and matrix datatypes include long-double versions.

It should be noted that in some system libraries the stdio.h formatted input/output functions printf and scanf are not implemented correctly for long double. Undefined or incorrect results are avoided by testing these functions during the configure stage of library compilation and eliminating certain GSL functions which depend on them if necessary. The corresponding line in the configure output looks like this,

```
checking whether printf works with long double... no
```

Consequently when long double formatted input/output does not work on a given system it should be impossible to link a program which uses GSL functions dependent on this.

If it is necessary to work on a system which does not support formatted long double input/output then the options are to use binary formats or to convert long double results into double for reading and writing.

2.7 Portability functions

To help in writing portable applications GSL provides some implementations of functions that are found in other libraries, such as the BSD math library. You can write your application to use the native versions of these functions, and substitute the GSL versions via a preprocessor macro if they are unavailable on another platform.

For example, after determining whether the BSD function hypot is available you can include the following macro definitions in a file config.h with your application,

```
/* Substitute gsl_hypot for missing system hypot */
#ifndef HAVE_HYPOT
#define hypot gsl_hypot
#endif
```

The application source files can then use the include command #include <config.h> to replace each occurrence of hypot by gsl_hypot when hypot is not available. This substitution can be made automatically if you use autoconf, see Appendix C [Autoconf Macros], page 461.

In most circumstances the best strategy is to use the native versions of these functions when available, and fall back to GSL versions otherwise, since this allows your application to take advantage of any platform-specific optimizations in the system library. This is the strategy used within GSL itself.

2.8 Alternative optimized functions

The main implementation of some functions in the library will not be optimal on all architectures. For example, there are several ways to compute a Gaussian random variate and their relative speeds are platform-dependent. In cases like this the library provides alternative implementations of these functions with the same interface. If you write your application using calls to the standard implementation you can select an alternative version later via a preprocessor definition. It is also possible to introduce your own optimized functions this way while retaining portability. The following lines demonstrate the use of a platform-dependent choice of methods for sampling from the Gaussian distribution,

```
#ifdef SPARC
#define gsl_ran_gaussian gsl_ran_gaussian_ratio_method
#endif
#ifdef INTEL
#define gsl_ran_gaussian my_gaussian
#endif
```

These lines would be placed in the configuration header file config.h of the application, which should then be included by all the source files. Note that the alternative implementations will not produce bit-for-bit identical results, and in the case of random number distributions will produce an entirely different stream of random variates.

2.9 Support for different numeric types

Many functions in the library are defined for different numeric types. This feature is implemented by varying the name of the function with a type-related modifier—a primitive form of C++ templates. The modifier is inserted into the function name after the initial module prefix. The following table shows the function names defined for all the numeric types of an imaginary module gsl_foo with function fn,

```
gsl_foo_fn
                          double
gsl_foo_long_double_fn
                          long double
gsl_foo_float_fn
                          float
gsl_foo_long_fn
                          long
gsl_foo_ulong_fn
                          unsigned long
gsl_foo_int_fn
                          int
gsl_foo_uint_fn
                          unsigned int
gsl_foo_short_fn
                          short
gsl_foo_ushort_fn
                          unsigned short
gsl_foo_char_fn
                          char
gsl_foo_uchar_fn
                          unsigned char
```

The normal numeric precision double is considered the default and does not require a suffix. For example, the function gsl_stats_mean computes the mean of double precision numbers, while the function gsl_stats_int_mean computes the mean of integers.

A corresponding scheme is used for library defined types, such as gsl_vector and gsl_matrix. In this case the modifier is appended to the type name. For example, if a module defines a new type-dependent struct or typedef gsl_foo it is modified for other types in the following way,

```
gsl_foo
                          double
gsl_foo_long_double
                          long double
gsl_foo_float
                          float
gsl_foo_long
                          long
gsl_foo_ulong
                          unsigned long
gsl_foo_int
                          int
gsl_foo_uint
                          unsigned int
gsl_foo_short
                          short
gsl_foo_ushort
                          unsigned short
gsl_foo_char
                          char
gsl_foo_uchar
                          unsigned char
```

When a module contains type-dependent definitions the library provides individual header files for each type. The filenames are modified as shown in the below. For convenience the default header includes the definitions for all the types. To include only the double precision header file, or any other specific type, use its individual filename.

```
#include <gsl/gsl_foo.h>
                                        All types
#include <gsl/gsl_foo_double.h>
                                        double
#include <gsl/gsl_foo_long_double.h>
                                        long double
#include <gsl/gsl_foo_float.h>
                                        float
#include <gsl/gsl_foo_long.h>
                                        long
#include <gsl/gsl_foo_ulong.h>
                                        unsigned long
#include <gsl/gsl_foo_int.h>
                                        int
#include <gsl/gsl_foo_uint.h>
                                        unsigned int
#include <gsl/gsl_foo_short.h>
                                        short
#include <gsl/gsl_foo_ushort.h>
                                        unsigned short
#include <gsl/gsl_foo_char.h>
                                        char
                                        unsigned char
#include <gsl/gsl_foo_uchar.h>
```

2.10 Compatibility with C++

The library header files automatically define functions to have extern "C" linkage when included in C++ programs. This allows the functions to be called directly from C++.

To use C++ exception handling within user-defined functions passed to the library as parameters, the library must be built with the additional CFLAGS compilation option - fexceptions.

2.11 Aliasing of arrays

The library assumes that arrays, vectors and matrices passed as modifiable arguments are not aliased and do not overlap with each other. This removes the need for the library to handle overlapping memory regions as a special case, and allows additional optimizations to be used. If overlapping memory regions are passed as modifiable arguments then the results of such functions will be undefined. If the arguments will not be modified (for example, if a function prototype declares them as const arguments) then overlapping or aliased memory regions can be safely used.

2.12 Thread-safety

The library can be used in multi-threaded programs. All the functions are thread-safe, in the sense that they do not use static variables. Memory is always associated with objects and not with functions. For functions which use workspace objects as temporary storage the workspaces should be allocated on a per-thread basis. For functions which use table objects as read-only memory the tables can be used by multiple threads simultaneously. Table arguments are always declared const in function prototypes, to indicate that they may be safely accessed by different threads.

There are a small number of static global variables which are used to control the overall behavior of the library (e.g. whether to use range-checking, the function to call on fatal error, etc). These variables are set directly by the user, so they should be initialized once at program startup and not modified by different threads.

2.13 Deprecated Functions

From time to time, it may be necessary for the definitions of some functions to be altered or removed from the library. In these circumstances the functions will first be declared deprecated and then removed from subsequent versions of the library. Functions that are deprecated can be disabled in the current release by setting the preprocessor definition GSL_DISABLE_DEPRECATED. This allows existing code to be tested for forwards compatibility.

2.14 Code Reuse

Where possible the routines in the library have been written to avoid dependencies between modules and files. This should make it possible to extract individual functions for use in your own applications, without needing to have the whole library installed. You may need to define certain macros such as GSL_ERROR and remove some #include statements in order to compile the files as standalone units. Reuse of the library code in this way is encouraged, subject to the terms of the GNU General Public License.

3 Error Handling

This chapter describes the way that GSL functions report and handle errors. By examining the status information returned by every function you can determine whether it succeeded or failed, and if it failed you can find out what the precise cause of failure was. You can also define your own error handling functions to modify the default behavior of the library.

The functions described in this section are declared in the header file gsl_errno.h.

3.1 Error Reporting

The library follows the thread-safe error reporting conventions of the POSIX Threads library. Functions return a non-zero error code to indicate an error and 0 to indicate success.

```
int status = gsl_function (...)
if (status) { /* an error occurred */
    ....
    /* status value specifies the type of error */
}
```

The routines report an error whenever they cannot perform the task requested of them. For example, a root-finding function would return a non-zero error code if could not converge to the requested accuracy, or exceeded a limit on the number of iterations. Situations like this are a normal occurrence when using any mathematical library and you should check the return status of the functions that you call.

Whenever a routine reports an error the return value specifies the type of error. The return value is analogous to the value of the variable error in the C library. The caller can examine the return code and decide what action to take, including ignoring the error if it is not considered serious.

In addition to reporting errors by return codes the library also has an error handler function gsl_error. This function is called by other library functions when they report an error, just before they return to the caller. The default behavior of the error handler is to print a message and abort the program,

```
gsl: file.c:67: ERROR: invalid argument supplied by user Default GSL error handler invoked.

Aborted
```

The purpose of the gsl_error handler is to provide a function where a breakpoint can be set that will catch library errors when running under the debugger. It is not intended for use in production programs, which should handle any errors using the return codes.

3.2 Error Codes

The error code numbers returned by library functions are defined in the file gsl_erro.h. They all have the prefix GSL_ and expand to non-zero constant integer values. Error codes above 1024 are reserved for applications, and are not used by the library. Many of the error codes use the same base name as the corresponding error code in the C library. Here are some of the most common error codes,

int GSL_EDOM [Macro]

Domain error; used by mathematical functions when an argument value does not fall into the domain over which the function is defined (like EDOM in the C library)

int GSL_ERANGE [Macro]

Range error; used by mathematical functions when the result value is not representable because of overflow or underflow (like ERANGE in the C library)

int GSL_ENOMEM [Macro]

No memory available. The system cannot allocate more virtual memory because its capacity is full (like ENOMEM in the C library). This error is reported when a GSL routine encounters problems when trying to allocate memory with malloc.

int GSL_EINVAL [Macro]

Invalid argument. This is used to indicate various kinds of problems with passing the wrong argument to a library function (like EINVAL in the C library).

The error codes can be converted into an error message using the function gsl_strerror.

```
const char * gsl_strerror (const int gsl_errno)
```

[Function]

This function returns a pointer to a string describing the error code gsl_errno. For example,

```
printf ("error: %s\n", gsl_strerror (status));
```

would print an error message like error: output range error for a status value of GSL_ERANGE.

3.3 Error Handlers

The default behavior of the GSL error handler is to print a short message and call abort. When this default is in use programs will stop with a core-dump whenever a library routine reports an error. This is intended as a fail-safe default for programs which do not check the return status of library routines (we don't encourage you to write programs this way).

If you turn off the default error handler it is your responsibility to check the return values of routines and handle them yourself. You can also customize the error behavior by providing a new error handler. For example, an alternative error handler could log all errors to a file, ignore certain error conditions (such as underflows), or start the debugger and attach it to the current process when an error occurs.

All GSL error handlers have the type gsl_error_handler_t, which is defined in gsl_errno.h,

gsl_error_handler_t

[Data Type]

This is the type of GSL error handler functions. An error handler will be passed four arguments which specify the reason for the error (a string), the name of the source file in which it occurred (also a string), the line number in that file (an integer) and the error number (an integer). The source file and line number are set at compile time using the <code>__FILE__</code> and <code>__LINE__</code> directives in the preprocessor. An error handler function returns type <code>void</code>. Error handler functions should be defined like this,

[Macro]

To request the use of your own error handler you need to call the function gsl_set_error_handler which is also declared in gsl_errno.h,

This function sets a new error handler, new_handler, for the GSL library routines. The previous handler is returned (so that you can restore it later). Note that the pointer to a user defined error handler function is stored in a static variable, so there can be only one error handler per program. This function should be not be used in multi-threaded programs except to set up a program-wide error handler from a master thread. The following example shows how to set and restore a new error handler,

```
/* save original handler, install new handler */
old_handler = gsl_set_error_handler (&my_handler);

/* code uses new handler */
.....

/* restore original handler */
gsl_set_error_handler (old_handler);

To use the default behavior (abort on error) set the error handler to NULL,
old_handler = gsl_set_error_handler (NULL);
```

```
gsl_error_handler_t * gsl_set_error_handler_off ()
[Function]
```

This function turns off the error handler by defining an error handler which does nothing. This will cause the program to continue after any error, so the return values from any library routines must be checked. This is the recommended behavior for production programs. The previous handler is returned (so that you can restore it later).

The error behavior can be changed for specific applications by recompiling the library with a customized definition of the GSL_ERROR macro in the file gsl_errno.h.

3.4 Using GSL error reporting in your own functions

If you are writing numerical functions in a program which also uses GSL code you may find it convenient to adopt the same error reporting conventions as in the library.

To report an error you need to call the function gsl_error with a string describing the error and then return an appropriate error code from gsl_errno.h, or a special value, such as NaN. For convenience the file gsl_errno.h defines two macros which carry out these steps:

```
GSL_ERROR (reason, gsl_errno)
```

This macro reports an error using the GSL conventions and returns a status value of gsl_errno. It expands to the following code fragment,

```
gsl_error (reason, __FILE__, __LINE__, gsl_errno);
return gsl_errno;
```

The macro definition in gsl_errno.h actually wraps the code in a do { ... } while (0) block to prevent possible parsing problems.

Here is an example of how the macro could be used to report that a routine did not achieve a requested tolerance. To report the error the routine needs to return the error code GSL_ETOL.

```
if (residual > tolerance)
  {
    GSL_ERROR("residual exceeds tolerance", GSL_ETOL);
}
```

```
GSL_ERROR_VAL (reason, gsl_errno, value)
```

[Macro]

This macro is the same as GSL_ERROR but returns a user-defined value of value instead of an error code. It can be used for mathematical functions that return a floating point value.

The following example shows how to return a NaN at a mathematical singularity using the GSL_ERROR_VAL macro,

3.5 Examples

Here is an example of some code which checks the return value of a function where an error might be reported,

```
#include <stdio.h>
#include <gsl/gsl_errno.h>
#include <gsl/gsl_fft_complex.h>

...
   int status;
   size_t n = 37;

   gsl_set_error_handler_off();

   status = gsl_fft_complex_radix2_forward (data, stride, n);

if (status) {
   if (status) = GSL_EINVAL) {
      fprintf (stderr, "invalid argument, n=%d\n", n);
   } else {
      fprintf (stderr, "failed, gsl_errno=%d\n",
```

```
status);
}
exit (-1);
}
...
```

The function gsl_fft_complex_radix2 only accepts integer lengths which are a power of two. If the variable n is not a power of two then the call to the library function will return GSL_EINVAL, indicating that the length argument is invalid. The function call to gsl_set_error_handler_off stops the default error handler from aborting the program. The else clause catches any other possible errors.

4 Mathematical Functions

This chapter describes basic mathematical functions. Some of these functions are present in system libraries, but the alternative versions given here can be used as a substitute when the system functions are not available.

The functions and macros described in this chapter are defined in the header file gsl_math.h.

4.1 Mathematical Constants

The library ensures that the standard BSD mathematical constants are defined. For reference, here is a list of the constants:

```
M_E
             The base of exponentials, e
M_LOG2E
             The base-2 logarithm of e, \log_2(e)
             The base-10 logarithm of e, \log_{10}(e)
M_LOG10E
             The square root of two, \sqrt{2}
M_SQRT2
M_SQRT1_2
             The square root of one-half, \sqrt{1/2}
M_SQRT3
             The square root of three, \sqrt{3}
M_PI
             The constant pi, \pi
M_PI_2
             Pi divided by two, \pi/2
             Pi divided by four, \pi/4
M_PI_4
             The square root of pi, \sqrt{\pi}
M_SQRTPI
M_2_SQRTPI
             Two divided by the square root of pi, 2/\sqrt{\pi}
M_1_PI
             The reciprocal of pi, 1/\pi
M_2_PI
             Twice the reciprocal of pi, 2/\pi
             The natural logarithm of ten, ln(10)
M_LN10
M_LN2
             The natural logarithm of two, ln(2)
M_LNPI
             The natural logarithm of pi, \ln(\pi)
M_EULER
             Euler's constant, \gamma
```

4.2 Infinities and Not-a-number

GSL_POSINF [Macro]

This macro contains the IEEE representation of positive infinity, $+\infty$. It is computed from the expression +1.0/0.0.

GSL_NEGINF [Macro]

This macro contains the IEEE representation of negative infinity, $-\infty$. It is computed from the expression -1.0/0.0.

GSL_NAN [Macro]

This macro contains the IEEE representation of the Not-a-Number symbol, NaN. It is computed from the ratio 0.0/0.0.

int gsl_isnan (const double x)

[Function]

This function returns 1 if x is not-a-number.

int gsl_isinf (const double x)

[Function]

This function returns +1 if x is positive infinity, -1 if x is negative infinity and 0 otherwise.¹

int gsl_finite (const double x)

[Function]

This function returns 1 if x is a real number, and 0 if it is infinite or not-a-number.

4.3 Elementary Functions

The following routines provide portable implementations of functions found in the BSD math library. When native versions are not available the functions described here can be used instead. The substitution can be made automatically if you use autoconf to compile your application (see Section 2.7 [Portability functions], page 7).

double gsl_log1p (const double x)

[Function]

This function computes the value of log(1+x) in a way that is accurate for small x. It provides an alternative to the BSD math function log1p(x).

double gsl_expm1 (const double x)

[Function]

This function computes the value of $\exp(x) - 1$ in a way that is accurate for small x. It provides an alternative to the BSD math function expm1(x).

double gsl_hypot (const double x, const double y)

[Function]

This function computes the value of $\sqrt{x^2 + y^2}$ in a way that avoids overflow. It provides an alternative to the BSD math function hypot(x,y).

double gsl_hypot3 (const double x, const double y, const double z) [Function] This function computes the value of $\sqrt{x^2 + y^2 + z^2}$ in a way that avoids overflow.

double gsl_acosh (const double x)

[Function]

This function computes the value of $\operatorname{arccosh}(x)$. It provides an alternative to the standard math function $\operatorname{acosh}(x)$.

double gsl_asinh (const double x)

[Function]

This function computes the value of $\operatorname{arcsinh}(x)$. It provides an alternative to the standard math function $\operatorname{asinh}(x)$.

double gsl_atanh (const double x)

[Function]

This function computes the value of $\operatorname{arctanh}(x)$. It provides an alternative to the standard math function $\operatorname{atanh}(x)$.

¹ Note that the C99 standard only requires the system isinf function to return a non-zero value, without the sign of the infinity. The implementation in some earlier versions of GSL used the system isinf function and may have this behavior on some platforms. Therefore, it is advisable to test the sign of x separately, if needed, rather than relying the sign of the return value from gsl_isinf().

```
double gsl_ldexp (double x, int e)
```

[Function]

This function computes the value of $x * 2^e$. It provides an alternative to the standard math function ldexp(x,e).

```
double gsl_frexp (double x, int * e)
```

Function

This function splits the number x into its normalized fraction f and exponent e, such that $x = f * 2^e$ and $0.5 \le f < 1$. The function returns f and stores the exponent in e. If x is zero, both f and e are set to zero. This function provides an alternative to the standard math function frexp(x, e).

4.4 Small integer powers

A common complaint about the standard C library is its lack of a function for calculating (small) integer powers. GSL provides some simple functions to fill this gap. For reasons of efficiency, these functions do not check for overflow or underflow conditions.

```
double gsl_pow_int (double x, int n) [Function]
double gsl_pow_uint (double x, unsigned int n) [Function]
```

These routines computes the power x^n for integer n. The power is computed efficiently—for example, x^8 is computed as $((x^2)^2)^2$, requiring only 3 multiplications. A version of this function which also computes the numerical error in the result is available as $gsl_sf_pow_int_e$.

```
double gsl_pow_2 (const double x)
                                                                          [Function]
double gsl_pow_3 (const double x)
                                                                          [Function]
double gsl_pow_4 (const double x)
                                                                          [Function]
double gsl_pow_5 (const double x)
                                                                          [Function]
double gsl_pow_6 (const double x)
                                                                          [Function]
double gsl_pow_7 (const double x)
                                                                          [Function]
double gsl_pow_8 (const double x)
                                                                          [Function]
double gsl_pow_9 (const double x)
                                                                          [Function]
```

These functions can be used to compute small integer powers x^2 , x^3 , etc. efficiently. The functions will be inlined when HAVE_INLINE is defined, so that use of these functions should be as efficient as explicitly writing the corresponding product expression.

```
#include <gsl/gsl_math.h>
double y = gsl_pow_4 (3.141) /* compute 3.141**4 */
```

4.5 Testing the Sign of Numbers

 $GSL_SIGN(x)$ [Macro]

This macro returns the sign of x. It is defined as $((x) \ge 0 ? 1 : -1)$. Note that with this definition the sign of zero is positive (regardless of its IEEE sign bit).

4.6 Testing for Odd and Even Numbers

 GSL_IS_ODD (n) [Macro]

This macro evaluates to 1 if n is odd and 0 if n is even. The argument n must be of integer type.

GSL_IS_EVEN (n)

[Macro]

This macro is the opposite of $GSL_IS_ODD(n)$. It evaluates to 1 if n is even and 0 if n is odd. The argument n must be of integer type.

4.7 Maximum and Minimum functions

Note that the following macros perform multiple evaluations of their arguments, so they should not be used with arguments that have side effects (such as a call to a random number generator).

 GSL_MAX (a, b) [Macro]

This macro returns the maximum of a and b. It is defined as ((a) > (b) ? (a) : (b)).

 $GSL_MIN(a, b)$ [Macro]

This macro returns the minimum of a and b. It is defined as ((a) < (b) ? (a):(b)).

extern inline double GSL_MAX_DBL (double a, double b) [Function]

This function returns the maximum of the double precision numbers a and b using an inline function. The use of a function allows for type checking of the arguments as an extra safety feature. On platforms where inline functions are not available the macro GSL_MAX will be automatically substituted.

extern inline double GSL_MIN_DBL (double a, double b) [Function]

This function returns the minimum of the double precision numbers a and b using an inline function. The use of a function allows for type checking of the arguments as an extra safety feature. On platforms where inline functions are not available the macro GSL_MIN will be automatically substituted.

extern inline int GSL_MAX_INT (int a, int b) [Function] extern inline int GSL_MIN_INT (int a, int b)

These functions return the maximum or minimum of the integers a and b using an inline function. On platforms where inline functions are not available the macros GSL_MAX or GSL_MIN will be automatically substituted.

extern inline long double GSL_MAX_LDBL (long double a, long double b) [Function]

extern inline long double GSL_MIN_LDBL (long double a, long double b) [Function]

These functions return the maximum or minimum of the long doubles a and b using an inline function. On platforms where inline functions are not available the macros GSL_MAX or GSL_MIN will be automatically substituted.

4.8 Approximate Comparison of Floating Point Numbers

It is sometimes useful to be able to compare two floating point numbers approximately, to allow for rounding and truncation errors. The following function implements the approximate floating-point comparison algorithm proposed by D.E. Knuth in Section 4.2.2 of Seminumerical Algorithms (3rd edition).

int gsl_fcmp (double x, double y, double epsilon)

[Function]

This function determines whether x and y are approximately equal to a relative accuracy epsilon.

The relative accuracy is measured using an interval of size 2δ , where $\delta = 2^k \epsilon$ and k is the maximum base-2 exponent of x and y as computed by the function frexp.

If x and y lie within this interval, they are considered approximately equal and the function returns 0. Otherwise if x < y, the function returns -1, or if x > y, the function returns +1.

Note that x and y are compared to relative accuracy, so this function is not suitable for testing whether a value is approximately zero.

The implementation is based on the package fcmp by T.C. Belding.

5 Complex Numbers

The functions described in this chapter provide support for complex numbers. The algorithms take care to avoid unnecessary intermediate underflows and overflows, allowing the functions to be evaluated over as much of the complex plane as possible.

For multiple-valued functions the branch cuts have been chosen to follow the conventions of Abramowitz and Stegun in the *Handbook of Mathematical Functions*. The functions return principal values which are the same as those in GNU Calc, which in turn are the same as those in *Common Lisp*, *The Language (Second Edition)*¹ and the HP-28/48 series of calculators.

The complex types are defined in the header file gsl_complex.h, while the corresponding complex functions and arithmetic operations are defined in gsl_complex_math.h.

5.1 Representation of complex numbers

Complex numbers are represented using the type gsl_complex. The internal representation of this type may vary across platforms and should not be accessed directly. The functions and macros described below allow complex numbers to be manipulated in a portable way.

For reference, the default form of the gsl_complex type is given by the following struct,

```
typedef struct
{
  double dat[2];
} gsl_complex;
```

The real and imaginary part are stored in contiguous elements of a two element array. This eliminates any padding between the real and imaginary parts, dat[0] and dat[1], allowing the struct to be mapped correctly onto packed complex arrays.

```
gsl_complex gsl_complex_rect (double x, double y) [Function] This function uses the rectangular Cartesian components (x,y) to return the complex
```

This function uses the rectangular Cartesian components (x,y) to return the complex number z = x + iy. An inline version of this function is used when HAVE_INLINE is defined.

```
gsl_complex gsl_complex_polar (double r, double theta) [Function] This function returns the complex number z = r \exp(i\theta) = r(\cos(\theta) + i\sin(\theta)) from the polar representation (r, theta).
```

```
\operatorname{GSL\_REAL}(z) [Macro] \operatorname{GSL\_IMAG}(z)
```

These macros return the real and imaginary parts of the complex number z.

```
GSL\_SET\_COMPLEX (zp, x, y) [Macro]
```

This macro uses the Cartesian components (x,y) to set the real and imaginary parts of the complex number pointed to by zp. For example,

```
GSL_SET_COMPLEX(&z, 3, 4) sets z to be 3+4i.
```

¹ Note that the first edition uses different definitions.

[Macro]

[Macro]

GSL_SET_REAL (zp,x)GSL_SET_IMAG (zp,y)

These macros allow the real and imaginary parts of the complex number pointed to by zp to be set independently.

5.2 Properties of complex numbers

double gsl_complex_arg (gsl_complex z)

[Function]

This function returns the argument of the complex number z, $\arg(z)$, where $-\pi < \arg(z) \le \pi$.

double gsl_complex_abs $(gsl_complex z)$

[Function]

This function returns the magnitude of the complex number z, |z|.

double gsl_complex_abs2 $(gsl_complex z)$

[Function]

This function returns the squared magnitude of the complex number z, $|z|^2$.

double gsl_complex_logabs (gsl_complex z)

[Function]

This function returns the natural logarithm of the magnitude of the complex number z, $\log |z|$. It allows an accurate evaluation of $\log |z|$ when |z| is close to one. The direct evaluation of $\log(gsl_complex_abs(z))$ would lead to a loss of precision in this case.

5.3 Complex arithmetic operators

- gsl_complex gsl_complex_add ($gsl_complex$ a, $gsl_complex$ b) [Function] This function returns the sum of the complex numbers a and b, z = a + b.
- gsl_complex gsl_complex_sub (gsl_complex a, gsl_complex b) [Function] This function returns the difference of the complex numbers a and b, z = a b.
- gsl_complex gsl_complex_mul ($gsl_complex$ a, $gsl_complex$ b) [Function] This function returns the product of the complex numbers a and b, z = ab.
- gsl_complex gsl_complex_div ($gsl_complex$ a, $gsl_complex$ b) [Function] This function returns the quotient of the complex numbers a and b, z = a/b.
- gsl_complex gsl_complex_add_real (gsl_complex a, double x) [Function] This function returns the sum of the complex number a and the real number x, z = a + x.
- gsl_complex gsl_complex_sub_real (gsl_complex a, double x) [Function] This function returns the difference of the complex number a and the real number x, z = a x.
- gsl_complex gsl_complex_mul_real ($gsl_complex$ a, double x) [Function] This function returns the product of the complex number a and the real number x, z = ax.

- gsl_complex gsl_complex_div_real (gsl_complex a, double x) [Function] This function returns the quotient of the complex number a and the real number x, z = a/x.
- gsl_complex gsl_complex_add_imag ($gsl_complex$ a, double y) [Function] This function returns the sum of the complex number a and the imaginary number iy, z = a + iy.
- gsl_complex gsl_complex_sub_imag (gsl_complex a, double y) [Function] This function returns the difference of the complex number a and the imaginary number a in a
- gsl_complex gsl_complex_mul_imag ($gsl_complex$ a, double y) [Function] This function returns the product of the complex number a and the imaginary number iy, z = a * (iy).
- gsl_complex gsl_complex_div_imag (gsl_complex a, double y) [Function] This function returns the quotient of the complex number a and the imaginary number iy, z = a/(iy).
- gsl_complex gsl_complex_conjugate (gsl_complex z) [Function] This function returns the complex conjugate of the complex number $z, z^* = x iy$.
- gsl_complex gsl_complex_inverse (gsl_complex z) [Function] This function returns the inverse, or reciprocal, of the complex number z, $1/z = (x iy)/(x^2 + y^2)$.
- gsl_complex gsl_complex_negative (gsl_complex z) [Function] This function returns the negative of the complex number z, -z = (-x) + i(-y).

5.4 Elementary Complex Functions

- gsl_complex gsl_complex_sqrt (gsl_complex z) [Function] This function returns the square root of the complex number z, \sqrt{z} . The branch cut is the negative real axis. The result always lies in the right half of the complex plane.
- gsl_complex gsl_complex_sqrt_real (double x) [Function]

 This function returns the complex square root of the real number x, where x may be negative.
- gsl_complex gsl_complex_pow (gsl_complex z, gsl_complex a) [Function] The function returns the complex number z raised to the complex power a, z^a . This is computed as $\exp(\log(z) * a)$ using complex logarithms and complex exponentials.
- gsl_complex gsl_complex_pow_real ($gsl_complex z$, double x) [Function] This function returns the complex number z raised to the real power x, z^x .
- gsl_complex gsl_complex_exp ($gsl_complex z$) [Function] This function returns the complex exponential of the complex number z, exp(z).

- gsl_complex gsl_complex_log ($gsl_complex z$) [Function] This function returns the complex natural logarithm (base e) of the complex number z, $\log(z)$. The branch cut is the negative real axis.
- gsl_complex gsl_complex_log10 ($gsl_complex\ z$) [Function] This function returns the complex base-10 logarithm of the complex number z, $\log_{10}(z)$.
- gsl_complex gsl_complex_log_b ($gsl_complex\ z$, $gsl_complex\ b$) [Function] This function returns the complex base-b logarithm of the complex number z, $\log_b(z)$. This quantity is computed as the ratio $\log(z)/\log(b)$.

5.5 Complex Trigonometric Functions

- gsl_complex gsl_complex_sin (gsl_complex z) [Function] This function returns the complex sine of the complex number z, $\sin(z) = (\exp(iz) \exp(-iz))/(2i)$.
- gsl_complex gsl_complex_cos (gsl_complex z) [Function] This function returns the complex cosine of the complex number z, $\cos(z) = (\exp(iz) + \exp(-iz))/2$.
- gsl_complex gsl_complex_tan ($gsl_complex\ z$) [Function] This function returns the complex tangent of the complex number z, $\tan(z) = \sin(z)/\cos(z)$.
- gsl_complex gsl_complex_sec (gsl_complex z) [Function] This function returns the complex secant of the complex number z, $\sec(z) = 1/\cos(z)$.
- gsl_complex gsl_complex_csc ($gsl_complex\ z$) [Function] This function returns the complex cosecant of the complex number z, $\csc(z) = 1/\sin(z)$.
- gsl_complex gsl_complex_cot ($gsl_complex z$) [Function] This function returns the complex cotangent of the complex number z, $\cot(z) = 1/\tan(z)$.

5.6 Inverse Complex Trigonometric Functions

- gsl_complex gsl_complex_arcsin ($gsl_complex\ z$) [Function] This function returns the complex arcsine of the complex number z, arcsin(z). The branch cuts are on the real axis, less than -1 and greater than 1.
- gsl_complex gsl_complex_arcsin_real (double z) [Function] This function returns the complex arcsine of the real number z, $\arcsin(z)$. For z between -1 and 1, the function returns a real value in the range $[-\pi/2, \pi/2]$. For z less than -1 the result has a real part of $-\pi/2$ and a positive imaginary part. For z greater than 1 the result has a real part of $\pi/2$ and a negative imaginary part.

- gsl_complex gsl_complex_arccos ($gsl_complex\ z$) [Function] This function returns the complex arccosine of the complex number z, arccos(z). The branch cuts are on the real axis, less than -1 and greater than 1.
- gsl_complex gsl_complex_arccos_real (double z) [Function] This function returns the complex arccosine of the real number z, arccos(z). For z between -1 and 1, the function returns a real value in the range $[0, \pi]$. For z less than -1 the result has a real part of π and a negative imaginary part. For z greater than 1 the result is purely imaginary and positive.
- gsl_complex gsl_complex_arctan ($gsl_complex\ z$) [Function] This function returns the complex arctangent of the complex number z, arctan(z). The branch cuts are on the imaginary axis, below -i and above i.
- gsl_complex gsl_complex_arcsec (gsl_complex z) [Function] This function returns the complex arcsecant of the complex number z, arcsec(z) = arccos(1/z).
- gsl_complex gsl_complex_arcsec_real (double z) [Function] This function returns the complex arcsecant of the real number z, $\operatorname{arcsec}(z) = \operatorname{arccos}(1/z)$.
- gsl_complex gsl_complex_arccsc ($gsl_complex$ z) [Function] This function returns the complex arccosecant of the complex number z, arccsc(z) = arcsin(1/z).
- gsl_complex gsl_complex_arccsc_real (double z) [Function] This function returns the complex arccosecant of the real number z, arccsc(z) = arcsin(1/z).
- gsl_complex gsl_complex_arccot (gsl_complex z) [Function] This function returns the complex arccotangent of the complex number z, $\operatorname{arccot}(z) = \arctan(1/z)$.

5.7 Complex Hyperbolic Functions

- gsl_complex gsl_complex_sinh ($gsl_complex\ z$) [Function] This function returns the complex hyperbolic sine of the complex number z, $\sinh(z) = (\exp(z) \exp(-z))/2$.
- gsl_complex gsl_complex_cosh ($gsl_complex$ z) [Function] This function returns the complex hyperbolic cosine of the complex number z, $\cosh(z) = (\exp(z) + \exp(-z))/2$.
- gsl_complex gsl_complex_tanh (gsl_complex z) [Function] This function returns the complex hyperbolic tangent of the complex number z, $\tanh(z) = \sinh(z)/\cosh(z)$.
- gsl_complex gsl_complex_sech ($gsl_complex\ z$) [Function] This function returns the complex hyperbolic secant of the complex number z, $sech(z) = 1/\cosh(z)$.

- gsl_complex gsl_complex_csch ($gsl_complex\ z$) [Function] This function returns the complex hyperbolic cosecant of the complex number z, $csch(z) = 1/\sinh(z)$.
- gsl_complex gsl_complex_coth (gsl_complex z) [Function] This function returns the complex hyperbolic cotangent of the complex number z, $\coth(z) = 1/\tanh(z)$.

5.8 Inverse Complex Hyperbolic Functions

- gsl_complex gsl_complex_arcsinh ($gsl_complex z$) [Function] This function returns the complex hyperbolic arcsine of the complex number z, arcsinh(z). The branch cuts are on the imaginary axis, below -i and above i.
- gsl_complex gsl_complex_arccosh (gsl_complex z) [Function] This function returns the complex hyperbolic arccosine of the complex number z, $\operatorname{arccosh}(z)$. The branch cut is on the real axis, less than 1. Note that in this case we use the negative square root in formula 4.6.21 of Abramowitz & Stegun giving $\operatorname{arccosh}(z) = \log(z \sqrt{z^2 1})$.
- gsl_complex gsl_complex_arccosh_real (double z) [Function] This function returns the complex hyperbolic arccosine of the real number z, $\operatorname{arccosh}(z)$.
- gsl_complex gsl_complex_arctanh (gsl_complex z) [Function]
 This function returns the complex hyperbolic arctangent of the complex number z, arctanh(z). The branch cuts are on the real axis, less than -1 and greater than 1.
- gsl_complex gsl_complex_arctanh_real (double z) [Function] This function returns the complex hyperbolic arctangent of the real number z, $\operatorname{arctanh}(z)$.
- gsl_complex gsl_complex_arcsech ($gsl_complex z$) [Function] This function returns the complex hyperbolic arcsecant of the complex number z, $\operatorname{arcsech}(z) = \operatorname{arccosh}(1/z)$.
- gsl_complex gsl_complex_arccsch ($gsl_complex z$) [Function] This function returns the complex hyperbolic arccosecant of the complex number z, $\operatorname{arccsch}(z) = \arcsin(1/z)$.
- gsl_complex gsl_complex_arccoth ($gsl_complex\ z$) [Function] This function returns the complex hyperbolic arccotangent of the complex number z, $\operatorname{arccoth}(z) = \operatorname{arctanh}(1/z)$.

5.9 References and Further Reading

The implementations of the elementary and trigonometric functions are based on the following papers,

T. E. Hull, Thomas F. Fairgrieve, Ping Tak Peter Tang, "Implementing Complex Elementary Functions Using Exception Handling", ACM Transactions on Mathematical Software, Volume 20 (1994), pp 215–244, Corrigenda, p553

T. E. Hull, Thomas F. Fairgrieve, Ping Tak Peter Tang, "Implementing the complex arcsin and arccosine functions using exception handling", *ACM Transactions on Mathematical Software*, Volume 23 (1997) pp 299–335

The general formulas and details of branch cuts can be found in the following books,

Abramowitz and Stegun, *Handbook of Mathematical Functions*, "Circular Functions in Terms of Real and Imaginary Parts", Formulas 4.3.55–58, "Inverse Circular Functions in Terms of Real and Imaginary Parts", Formulas 4.4.37–39, "Hyperbolic Functions in Terms of Real and Imaginary Parts", Formulas 4.5.49–52, "Inverse Hyperbolic Functions—relation to Inverse Circular Functions", Formulas 4.6.14–19.

Dave Gillespie, Calc Manual, Free Software Foundation, ISBN 1-882114-18-3

6 Polynomials

This chapter describes functions for evaluating and solving polynomials. There are routines for finding real and complex roots of quadratic and cubic equations using analytic methods. An iterative polynomial solver is also available for finding the roots of general polynomials with real coefficients (of any order). The functions are declared in the header file gsl_poly.h.

6.1 Polynomial Evaluation

The functions described here evaluate the polynomial $P(x) = c[0] + c[1]x + c[2]x^2 + \ldots + c[len-1]x^{len-1}$ using Horner's method for stability. Inline versions of these functions are used when HAVE_INLINE is defined.

double gsl_poly_eval (const double c[], const int len, const double x) [Function] This function evaluates a polynomial with real coefficients for the real variable x.

 $gsl_complex gsl_poly_complex_eval (const double c[], const int len, const gsl_complex z)$ [Function]

This function evaluates a polynomial with real coefficients for the complex variable z.

gsl_complex gsl_complex_poly_complex_eval ($const\ gsl_complex$ [Function] c[], $const\ int\ len$, $const\ gsl_complex\ z)$

This function evaluates a polynomial with complex coefficients for the complex variable z.

int gsl_poly_eval_derivs (const double c[], const size_t lenc, const double x, double res[], const size_t lenres) [Function]

This function evaluates a polynomial and its derivatives storing the results in the array res of size lenres. The output array contains the values of $d^k P/dx^k$ for the specified value of x starting with k=0.

6.2 Divided Difference Representation of Polynomials

The functions described here manipulate polynomials stored in Newton's divided-difference representation. The use of divided-differences is described in Abramowitz & Stegun sections 25.1.4 and 25.2.26, and Burden and Faires, chapter 3, and discussed briefly below.

Given a function f(x), an nth degree interpolating polynomial $P_n(x)$ can be constructed which agrees with f at n+1 distinct points $x_0, x_1, ..., x_n$. This polynomial can be written in a form known as Newton's divided-difference representation:

$$P_n(x) = f(x_0) + \sum_{k=1}^n [x_0, x_1, ..., x_k](x - x_0)(x - x_1) \cdots (x - x_{k-1})$$

where the divided differences $[x_0, x_1, ..., x_k]$ are defined in section 25.1.4 of Abramowitz and Stegun. Additionally, it is possible to construct an interpolating polynomial of degree 2n+1 which also matches the first derivatives of f at the points $x_0, x_1, ..., x_n$. This is called the Hermite interpolating polynomial and is defined as

$$H_{2n+1}(x) = f(z_0) + \sum_{k=1}^{2n+1} [z_0, z_1, ..., z_k](x - z_0)(x - z_1) \cdots (x - z_{k-1})$$

where the elements of $z = \{x_0, x_0, x_1, x_1, ..., x_n, x_n\}$ are defined by $z_{2k} = z_{2k+1} = x_k$. The divided-differences $[z_0, z_1, ..., z_k]$ are discussed in Burden and Faires, section 3.4.

This function computes a divided-difference representation of the interpolating polynomial for the points (x, y) stored in the arrays xa and ya of length size. On output the divided-differences of (xa,ya) are stored in the array dd, also of length size. Using the notation above, $dd[k] = [x_0, x_1, ..., x_k]$.

This function evaluates the polynomial stored in divided-difference form in the arrays dd and xa of length size at the point x. An inline version of this function is used when HAVE_INLINE is defined.

This function converts the divided-difference representation of a polynomial to a Taylor expansion. The divided-difference representation is supplied in the arrays dd and xa of length size. On output the Taylor coefficients of the polynomial expanded about the point xp are stored in the array c also of length size. A workspace of length size must be provided in the array w.

This function computes a divided-difference representation of the interpolating Hermite polynomial for the points (x, y) stored in the arrays xa and ya of length size. Hermite interpolation constructs polynomials which also match first derivatives dy/dx which are provided in the array dya also of length size. The first derivatives can be incorported into the usual divided-difference algorithm by forming a new dataset $z = \{x_0, x_0, x_1, x_1, ...\}$, which is stored in the array za of length z-size on output. On output the divided-differences of the Hermite representation are stored in the array za of length z-size. Using the notation above, za-by-by-dd_eval and using za-by-dd_eval and using za-by-dd_eval and using za-by-dd_eval argument za-by-by-dd_eval.

6.3 Quadratic Equations

This function finds the real roots of the quadratic equation,

$$ax^2 + bx + c = 0$$

The number of real roots (either zero, one or two) is returned, and their locations are stored in x0 and x1. If no real roots are found then x0 and x1 are not modified. If one real root is found (i.e. if a=0) then it is stored in x0. When two real roots are found they are stored in x0 and x1 in ascending order. The case of coincident roots

is not considered special. For example $(x-1)^2 = 0$ will have two roots, which happen to have exactly equal values.

The number of roots found depends on the sign of the discriminant $b^2 - 4ac$. This will be subject to rounding and cancellation errors when computed in double precision, and will also be subject to errors if the coefficients of the polynomial are inexact. These errors may cause a discrete change in the number of roots. However, for polynomials with small integer coefficients the discriminant can always be computed exactly.

This function finds the complex roots of the quadratic equation,

$$az^2 + bz + c = 0$$

The number of complex roots is returned (either one or two) and the locations of the roots are stored in z0 and z1. The roots are returned in ascending order, sorted first by their real components and then by their imaginary components. If only one real root is found (i.e. if a=0) then it is stored in z0.

6.4 Cubic Equations

This function finds the real roots of the cubic equation,

$$x^3 + ax^2 + bx + c = 0$$

with a leading coefficient of unity. The number of real roots (either one or three) is returned, and their locations are stored in x0, x1 and x2. If one real root is found then only x0 is modified. When three real roots are found they are stored in x0, x1 and x2 in ascending order. The case of coincident roots is not considered special. For example, the equation $(x-1)^3=0$ will have three roots with exactly equal values. As in the quadratic case, finite precision may cause equal or closely-spaced real roots to move off the real axis into the complex plane, leading to a discrete change in the number of real roots.

int gsl_poly_complex_solve_cubic (double a, double b, double c,
gsl_complex * z0, gsl_complex * z1, gsl_complex * z2)

This function finds the complex roots of the cubic equation,

[Function]

$$z^3 + az^2 + bz + c = 0$$

The number of complex roots is returned (always three) and the locations of the roots are stored in z0, z1 and z2. The roots are returned in ascending order, sorted first by their real components and then by their imaginary components.

6.5 General Polynomial Equations

The roots of polynomial equations cannot be found analytically beyond the special cases of the quadratic, cubic and quartic equation. The algorithm described in this section uses an iterative method to find the approximate locations of roots of higher order polynomials.

This function allocates space for a $gsl_poly_complex_workspace$ struct and a workspace suitable for solving a polynomial with n coefficients using the routine $gsl_poly_complex_solve$.

The function returns a pointer to the newly allocated gsl_poly_complex_workspace if no errors were detected, and a null pointer in the case of error.

This function frees all the memory associated with the workspace w.

This function computes the roots of the general polynomial $P(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_{n-1} x^{n-1}$ using balanced-QR reduction of the companion matrix. The parameter n specifies the length of the coefficient array. The coefficient of the highest order term must be non-zero. The function requires a workspace w of the appropriate size. The n-1 roots are returned in the packed complex array z of length 2(n-1), alternating real and imaginary parts.

The function returns GSL_SUCCESS if all the roots are found. If the QR reduction does not converge, the error handler is invoked with an error code of GSL_EFAILED. Note that due to finite precision, roots of higher multiplicity are returned as a cluster of simple roots with reduced accuracy. The solution of polynomials with higher-order roots requires specialized algorithms that take the multiplicity structure into account (see e.g. Z. Zeng, Algorithm 835, ACM Transactions on Mathematical Software, Volume 30, Issue 2 (2004), pp 218–236).

6.6 Examples

To demonstrate the use of the general polynomial solver we will take the polynomial $P(x) = x^5 - 1$ which has the following roots,

$$1, e^{2\pi i/5}, e^{4\pi i/5}, e^{6\pi i/5}, e^{8\pi i/5}$$

The following program will find these roots.

```
#include <stdio.h>
#include <gsl/gsl_poly.h>

int
main (void)
{
   int i;
   /* coefficients of P(x) = -1 + x^5 */
   double a[6] = { -1, 0, 0, 0, 0, 1 };
   double z[10];

gsl_poly_complex_workspace * w
```

```
= gsl_poly_complex_workspace_alloc (6);
      gsl_poly_complex_solve (a, 6, w, z);
      gsl_poly_complex_workspace_free (w);
      for (i = 0; i < 5; i++)
          printf ("z\%d = \%+.18f \%+.18f\\n",
                  i, z[2*i], z[2*i+1]);
        }
      return 0;
The output of the program is,
    $ ./a.out
    z0 = -0.809016994374947451 + 0.587785252292473137
    z1 = -0.809016994374947451 - 0.587785252292473137
    z2 = +0.309016994374947451 +0.951056516295153642
    z3 = +0.309016994374947451 -0.951056516295153642
    which agrees with the analytic result, z_n = \exp(2\pi ni/5).
```

6.7 References and Further Reading

The balanced-QR method and its error analysis are described in the following papers,

R.S. Martin, G. Peters and J.H. Wilkinson, "The QR Algorithm for Real Hessenberg Matrices", *Numerische Mathematik*, 14 (1970), 219–231.

B.N. Parlett and C. Reinsch, "Balancing a Matrix for Calculation of Eigenvalues and Eigenvectors", *Numerische Mathematik*, 13 (1969), 293–304.

A. Edelman and H. Murakami, "Polynomial roots from companion matrix eigenvalues", *Mathematics of Computation*, Vol. 64, No. 210 (1995), 763–776.

The formulas for divided differences are given in the following texts,

Abramowitz and Stegun, Handbook of Mathematical Functions, Sections 25.1.4 and 25.2.26.

R. L. Burden and J. D. Faires, *Numerical Analysis*, 9th edition, ISBN 0-538-73351-9, 2011.

7 Special Functions

This chapter describes the GSL special function library. The library includes routines for calculating the values of Airy functions, Bessel functions, Clausen functions, Coulomb wave functions, Coupling coefficients, the Dawson function, Debye functions, Dilogarithms, Elliptic integrals, Jacobi elliptic functions, Error functions, Exponential integrals, Fermi-Dirac functions, Gamma functions, Gegenbauer functions, Hypergeometric functions, Laguerre functions, Legendre functions and Spherical Harmonics, the Psi (Digamma) Function, Synchrotron functions, Transport functions, Trigonometric functions and Zeta functions. Each routine also computes an estimate of the numerical error in the calculated value of the function.

The functions in this chapter are declared in individual header files, such as gsl_sf_airy.h, gsl_sf_bessel.h, etc. The complete set of header files can be included using the file gsl_sf.h.

7.1 Usage

The special functions are available in two calling conventions, a *natural form* which returns the numerical value of the function and an *error-handling form* which returns an error code. The two types of function provide alternative ways of accessing the same underlying code.

The natural form returns only the value of the function and can be used directly in mathematical expressions. For example, the following function call will compute the value of the Bessel function $J_0(x)$,

```
double y = gsl_sf_bessel_J0 (x);
```

There is no way to access an error code or to estimate the error using this method. To allow access to this information the alternative error-handling form stores the value and error in a modifiable argument,

```
gsl_sf_result result;
int status = gsl_sf_bessel_J0_e (x, &result);
```

The error-handling functions have the suffix _e. The returned status value indicates error conditions such as overflow, underflow or loss of precision. If there are no errors the error-handling functions return GSL_SUCCESS.

7.2 The gsl_sf_result struct

The error handling form of the special functions always calculate an error estimate along with the value of the result. Therefore, structures are provided for amalgamating a value and error estimate. These structures are declared in the header file gsl_sf_result.h.

The gsl_sf_result struct contains value and error fields.

```
typedef struct
{
  double val;
  double err;
} gsl_sf_result;
```

The field *val* contains the value and the field *err* contains an estimate of the absolute error in the value.

In some cases, an overflow or underflow can be detected and handled by a function. In this case, it may be possible to return a scaling exponent as well as an error/value pair in order to save the result from exceeding the dynamic range of the built-in types. The gsl_sf_result_e10 struct contains value and error fields as well as an exponent field such that the actual result is obtained as result * 10^(e10).

```
typedef struct
{
  double val;
  double err;
  int   e10;
} gsl_sf_result_e10;
```

7.3 Modes

The goal of the library is to achieve double precision accuracy wherever possible. However the cost of evaluating some special functions to double precision can be significant, particularly where very high order terms are required. In these cases a mode argument allows the accuracy of the function to be reduced in order to improve performance. The following precision levels are available for the mode argument,

```
GSL_PREC_DOUBLE
```

Double-precision, a relative accuracy of approximately 2×10^{-16} .

GSL_PREC_SINGLE

Single-precision, a relative accuracy of approximately 1×10^{-7} .

GSL_PREC_APPROX

Approximate values, a relative accuracy of approximately 5×10^{-4} .

The approximate mode provides the fastest evaluation at the lowest accuracy.

7.4 Airy Functions and Derivatives

The Airy functions Ai(x) and Bi(x) are defined by the integral representations,

$$Ai(x) = \frac{1}{\pi} \int_0^\infty \cos(t^3/3 + xt) dt,$$

$$Bi(x) = \frac{1}{\pi} \int_0^\infty (e^{-t^3/3 + xt} + \sin(t^3/3 + xt)) dt.$$

For further information see Abramowitz & Stegun, Section 10.4. The Airy functions are defined in the header file gsl_sf_airy.h.

7.4.1 Airy Functions

These routines compute the Airy function Ai(x) with an accuracy specified by mode.

These routines compute the Airy function Bi(x) with an accuracy specified by mode.

These routines compute a scaled version of the Airy function $S_A(x)Ai(x)$. For x > 0 the scaling factor $S_A(x)$ is $\exp(+(2/3)x^{3/2})$, and is 1 for x < 0.

These routines compute a scaled version of the Airy function $S_B(x)Bi(x)$. For x > 0 the scaling factor $S_B(x)$ is $\exp(-(2/3)x^{3/2})$, and is 1 for x < 0.

7.4.2 Derivatives of Airy Functions

These routines compute the Airy function derivative Ai'(x) with an accuracy specified by mode.

These routines compute the Airy function derivative Bi'(x) with an accuracy specified by mode.

These routines compute the scaled Airy function derivative $S_A(x)Ai'(x)$. For x > 0 the scaling factor $S_A(x)$ is $\exp(+(2/3)x^{3/2})$, and is 1 for x < 0.

These routines compute the scaled Airy function derivative $S_B(x)Bi'(x)$. For x > 0 the scaling factor $S_B(x)$ is $\exp(-(2/3)x^{3/2})$, and is 1 for x < 0.

7.4.3 Zeros of Airy Functions

double gsl_sf_airy_zero_Ai (unsigned int s) [Function] int gsl_sf_airy_zero_Ai_e (unsigned int s, gsl_sf_result * result) [Function] These routines compute the location of the s-th zero of the Airy function Ai(x).

```
double gsl_sf_airy_zero_Bi (unsigned int s) [Function] int gsl_sf_airy_zero_Bi_e (unsigned int s, gsl\_sf\_result * result) [Function] These routines compute the location of the s-th zero of the Airy function Bi(x).
```

7.4.4 Zeros of Derivatives of Airy Functions

These routines compute the location of the s-th zero of the Airy function derivative Ai'(x).

These routines compute the location of the s-th zero of the Airy function derivative Bi'(x).

7.5 Bessel Functions

The routines described in this section compute the Cylindrical Bessel functions $J_n(x)$, $Y_n(x)$, Modified cylindrical Bessel functions $I_n(x)$, $K_n(x)$, Spherical Bessel functions $j_l(x)$, $y_l(x)$, and Modified Spherical Bessel functions $i_l(x)$, $k_l(x)$. For more information see Abramowitz & Stegun, Chapters 9 and 10. The Bessel functions are defined in the header file gsl_sf_bessel.h.

7.5.1 Regular Cylindrical Bessel Functions

```
double gsl_sf_bessel_J0 (double x) [Function] int gsl_sf_bessel_J0_e (double x, gsl_sf_result * result) [Function] These routines compute the regular cylindrical Bessel function of zeroth order, J_0(x).
```

```
double gsl_sf_bessel_J1 (double x) [Function] int gsl_sf_bessel_J1_e (double x, gsl\_sf\_result * result) [Function] These routines compute the regular cylindrical Bessel function of first order, J_1(x).
```

```
double gsl_sf_bessel_Jn (int n, double x) [Function] int gsl_sf_bessel_Jn_e (int n, double x, gsl_sf_result * result) [Function] These routines compute the regular cylindrical Bessel function of order n, J_n(x).
```

This routine computes the values of the regular cylindrical Bessel functions $J_n(x)$ for n from nmin to nmax inclusive, storing the results in the array result_array. The values are computed using recurrence relations for efficiency, and therefore may differ slightly from the exact values.

7.5.2 Irregular Cylindrical Bessel Functions

double gsl_sf_bessel_Y0 (double x) [Function] int gsl_sf_bessel_Y0_e (double x, gsl_sf_result * result) [Function] These routines compute the irregular cylindrical Bessel function of zeroth order, $Y_0(x)$, for x > 0.

double gsl_sf_bessel_Y1 (double x) [Function] int gsl_sf_bessel_Y1_e (double x, gsl_sf_result * result) [Function] These routines compute the irregular cylindrical Bessel function of first order, $Y_1(x)$,

for x > 0. double gsl_sf_bessel_Yn (int n, double x) [Function]

int gsl_sf_bessel_fin (int n, double x) [Function] int gsl_sf_bessel_Yn_e (int n, double x, gsl_sf_result * result) [Function] These routines compute the irregular cylindrical Bessel function of order n, $Y_n(x)$, for x > 0.

This routine computes the values of the irregular cylindrical Bessel functions $Y_n(x)$ for n from $n\min$ to $n\max$ inclusive, storing the results in the array $result_array$. The domain of the function is x > 0. The values are computed using recurrence relations for efficiency, and therefore may differ slightly from the exact values.

7.5.3 Regular Modified Cylindrical Bessel Functions

double gsl_sf_bessel_IO (double x) [Function] int gsl_sf_bessel_IO_e (double x, gsl_sf_result * result) [Function] These routines compute the regular modified cylindrical Bessel function of zeroth order, $I_0(x)$.

double gsl_sf_bessel_I1 (double x) [Function] int gsl_sf_bessel_I1_e (double x, gsl_sf_result * result) [Function] These routines compute the regular modified cylindrical Bessel function of first order, $I_1(x)$.

double gsl_sf_bessel_In (int n, double x) [Function] int gsl_sf_bessel_In_e (int n, double x, gsl_sf_result * result) [Function] These routines compute the regular modified cylindrical Bessel function of order n, $I_n(x)$.

int gsl_sf_bessel_In_array (int nmin, int nmax, double x, double result_array[])

This routine computes the values of the regular modified cylindrical Bessel functions $I_n(x)$ for n from $n\min$ to $n\max$ inclusive, storing the results in the array $result_array$. The start of the range $n\min$ must be positive or zero. The values are computed using recurrence relations for efficiency, and therefore may differ slightly from the exact values.

double gsl_sf_bessel_I0_scaled (double x) [Function] int gsl_sf_bessel_I0_scaled_e (double x, gsl_sf_result * result) [Function] These routines compute the scaled regular modified cylindrical Bessel function of zeroth order $\exp(-|x|)I_0(x)$.

double gsl_sf_bessel_I1_scaled (double x) [Function] int gsl_sf_bessel_I1_scaled_e (double x, gsl_sf_result * result) [Function] These routines compute the scaled regular modified cylindrical Bessel function of first order $\exp(-|x|)I_1(x)$.

These routines compute the scaled regular modified cylindrical Bessel function of order n, $\exp(-|x|)I_n(x)$

This routine computes the values of the scaled regular cylindrical Bessel functions $\exp(-|x|)I_n(x)$ for n from nmin to nmax inclusive, storing the results in the array $result_array$. The start of the range nmin must be positive or zero. The values are computed using recurrence relations for efficiency, and therefore may differ slightly from the exact values.

7.5.4 Irregular Modified Cylindrical Bessel Functions

double gsl_sf_bessel_KO (double x) [Function] int gsl_sf_bessel_KO_e (double x, gsl_sf_result * result) [Function] These routines compute the irregular modified cylindrical Bessel function of zeroth order, $K_0(x)$, for x > 0.

double gsl_sf_bessel_K1 (double x) [Function] int gsl_sf_bessel_K1_e (double x, gsl_sf_result * result) [Function] These routines compute the irregular modified cylindrical Bessel function of first order, $K_1(x)$, for x > 0.

double gsl_sf_bessel_Kn (int n, double x) [Function] int gsl_sf_bessel_Kn_e (int n, double x, gsl_sf_result * result) [Function] These routines compute the irregular modified cylindrical Bessel function of order n, $K_n(x)$, for x > 0.

This routine computes the values of the irregular modified cylindrical Bessel functions $K_n(x)$ for n from $n\min$ to $n\max$ inclusive, storing the results in the array $result_array$. The start of the range $n\min$ must be positive or zero. The domain of the function is x > 0. The values are computed using recurrence relations for efficiency, and therefore may differ slightly from the exact values.

double gsl_sf_bessel_K0_scaled (double x) [Function] int gsl_sf_bessel_K0_scaled_e (double x, gsl_sf_result * result) [Function] These routines compute the scaled irregular modified cylindrical Bessel function of zeroth order $\exp(x)K_0(x)$ for x>0.

double gsl_sf_bessel_K1_scaled (double x) [Function] int gsl_sf_bessel_K1_scaled_e (double x, gsl_sf_result * result) [Function] These routines compute the scaled irregular modified cylindrical Bessel function of first order $\exp(x)K_1(x)$ for x>0.

These routines compute the scaled irregular modified cylindrical Bessel function of order n, $\exp(x)K_n(x)$, for x > 0.

This routine computes the values of the scaled irregular cylindrical Bessel functions $\exp(x)K_n(x)$ for n from $n\min$ to $n\max$ inclusive, storing the results in the array $result_array$. The start of the range $n\min$ must be positive or zero. The domain of the function is x>0. The values are computed using recurrence relations for efficiency, and therefore may differ slightly from the exact values.

7.5.5 Regular Spherical Bessel Functions

double gsl_sf_bessel_j0 (double x) [Function] int gsl_sf_bessel_j0_e (double x, gsl_sf_result * result) [Function] These routines compute the regular spherical Bessel function of zeroth order, $j_0(x) = \sin(x)/x$.

double gsl_sf_bessel_j1 (double x) [Function] int gsl_sf_bessel_j1_e (double x, gsl_sf_result * result) [Function] These routines compute the regular spherical Bessel function of first order, $j_1(x) = (\sin(x)/x - \cos(x))/x$.

double gsl_sf_bessel_j2 (double x) [Function] int gsl_sf_bessel_j2_e (double x, gsl_sf_result * result) [Function] These routines compute the regular spherical Bessel function of second order, $j_2(x) = ((3/x^2 - 1)\sin(x) - 3\cos(x)/x)/x$.

double gsl_sf_bessel_jl (int 1, double x) [Function] int gsl_sf_bessel_jl_e (int 1, double x, gsl_sf_result * result) [Function] These routines compute the regular spherical Bessel function of order l, $j_l(x)$, for $l \geq 0$ and $x \geq 0$.

This routine computes the values of the regular spherical Bessel functions $j_l(x)$ for l from 0 to lmax inclusive for lmax ≥ 0 and $x \geq 0$, storing the results in the array

result_array. The values are computed using recurrence relations for efficiency, and therefore may differ slightly from the exact values.

This routine uses Steed's method to compute the values of the regular spherical Bessel functions $j_l(x)$ for l from 0 to lmax inclusive for lmax ≥ 0 and $x \geq 0$, storing the results in the array $result_array$. The Steed/Barnett algorithm is described in Comp. Phys. Comm. 21, 297 (1981). Steed's method is more stable than the recurrence used in the other functions but is also slower.

7.5.6 Irregular Spherical Bessel Functions

```
double gsl_sf_bessel_y0 (double x) [Function] int gsl_sf_bessel_y0_e (double x, gsl_sf_result * result) [Function] These routines compute the irregular spherical Bessel function of zeroth order, y_0(x) = -\cos(x)/x.
```

```
double gsl_sf_bessel_y1 (double x) [Function] int gsl_sf_bessel_y1_e (double x, gsl_sf_result * result) [Function] These routines compute the irregular spherical Bessel function of first order, y_1(x) = -(\cos(x)/x + \sin(x))/x.
```

```
double gsl_sf_bessel_y2 (double x) [Function] int gsl_sf_bessel_y2_e (double x, gsl_sf_result * result) [Function] These routines compute the irregular spherical Bessel function of second order, y_2(x) = (-3/x^3 + 1/x)\cos(x) - (3/x^2)\sin(x).
```

```
double gsl_sf_bessel_yl (int 1, double x) [Function] int gsl_sf_bessel_yl_e (int 1, double x, gsl_sf_result * result) [Function] These routines compute the irregular spherical Bessel function of order l, y_l(x), for l > 0.
```

This routine computes the values of the irregular spherical Bessel functions $y_l(x)$ for l from 0 to lmax inclusive for $lmax \ge 0$, storing the results in the array $result_array$. The values are computed using recurrence relations for efficiency, and therefore may differ slightly from the exact values.

7.5.7 Regular Modified Spherical Bessel Functions

The regular modified spherical Bessel functions $i_l(x)$ are related to the modified Bessel functions of fractional order, $i_l(x) = \sqrt{\pi/(2x)}I_{l+1/2}(x)$

```
double gsl_sf_bessel_i0_scaled (double x) [Function] int gsl_sf_bessel_i0_scaled_e (double x, gsl_sf_result * result) [Function] These routines compute the scaled regular modified spherical Bessel function of zeroth order, \exp(-|x|)i_0(x).
```

double gsl_sf_bessel_i1_scaled (double x) [Function] int gsl_sf_bessel_i1_scaled_e (double x, gsl_sf_result * result) [Function] These routines compute the scaled regular modified spherical Bessel function of first order, $\exp(-|x|)i_1(x)$.

double gsl_sf_bessel_i2_scaled (double x) [Function] int gsl_sf_bessel_i2_scaled_e (double x, gsl_sf_result * result) [Function] These routines compute the scaled regular modified spherical Bessel function of second order, $\exp(-|x|)i_2(x)$

These routines compute the scaled regular modified spherical Bessel function of order l, $\exp(-|x|)i_l(x)$

int gsl_sf_bessel_il_scaled_array (int lmax, double x, double result_array[])

This routine computes the values of the scaled regular modified cylindrical Bessel functions $\exp(-|x|)i_l(x)$ for l from 0 to lmax inclusive for lmax \geq 0, storing the results in the array $result_array$. The values are computed using recurrence relations for efficiency, and therefore may differ slightly from the exact values.

7.5.8 Irregular Modified Spherical Bessel Functions

The irregular modified spherical Bessel functions $k_l(x)$ are related to the irregular modified Bessel functions of fractional order, $k_l(x) = \sqrt{\pi/(2x)}K_{l+1/2}(x)$.

double gsl_sf_bessel_k0_scaled (double x) [Function] int gsl_sf_bessel_k0_scaled_e (double x, gsl_sf_result * result) [Function] These routines compute the scaled irregular modified spherical Bessel function of zeroth order, $\exp(x)k_0(x)$, for x > 0.

double gsl_sf_bessel_k1_scaled (double x) [Function] int gsl_sf_bessel_k1_scaled_e (double x, gsl_sf_result * result) [Function] These routines compute the scaled irregular modified spherical Bessel function of first order, $\exp(x)k_1(x)$, for x > 0.

double gsl_sf_bessel_k2_scaled (double x) [Function] int gsl_sf_bessel_k2_scaled_e (double x, gsl_sf_result * result) [Function] These routines compute the scaled irregular modified spherical Bessel function of second order, $\exp(x)k_2(x)$, for x > 0.

These routines compute the scaled irregular modified spherical Bessel function of order l, $\exp(x)k_l(x)$, for x > 0.

This routine computes the values of the scaled irregular modified spherical Bessel functions $\exp(x)k_l(x)$ for l from 0 to lmax inclusive for lmax ≥ 0 and x > 0, storing the results in the array $result_array$. The values are computed using recurrence relations for efficiency, and therefore may differ slightly from the exact values.

7.5.9 Regular Bessel Function—Fractional Order

double gsl_sf_bessel_Jnu (double nu, double x) [Function] int gsl_sf_bessel_Jnu_e (double nu, double x, gsl_sf_result * result) [Function] These routines compute the regular cylindrical Bessel function of fractional order ν , $J_{\nu}(x)$.

int gsl_sf_bessel_sequence_Jnu_e (double nu, gsl_mode_t mode, size_t size, double v[]) [Function]

This function computes the regular cylindrical Bessel function of fractional order ν , $J_{\nu}(x)$, evaluated at a series of x values. The array ν of length size contains the x values. They are assumed to be strictly ordered and positive. The array is overwritten with the values of $J_{\nu}(x_i)$.

7.5.10 Irregular Bessel Functions—Fractional Order

double gsl_sf_bessel_Ynu (double nu, double x) [Function] int gsl_sf_bessel_Ynu_e (double nu, double x, gsl_sf_result * result) [Function] These routines compute the irregular cylindrical Bessel function of fractional order ν , $Y_{\nu}(x)$.

7.5.11 Regular Modified Bessel Functions—Fractional Order

double gsl_sf_bessel_Inu (double nu, double x) [Function] int gsl_sf_bessel_Inu_e (double nu, double x, gsl_sf_result * result) [Function] These routines compute the regular modified Bessel function of fractional order ν , $I_{\nu}(x)$ for x > 0, $\nu > 0$.

These routines compute the scaled regular modified Bessel function of fractional order ν , $\exp(-|x|)I_{\nu}(x)$ for x > 0, $\nu > 0$.

7.5.12 Irregular Modified Bessel Functions—Fractional Order

double gsl_sf_bessel_Knu (double nu, double x) [Function] int gsl_sf_bessel_Knu_e (double nu, double x, gsl_sf_result * result) [Function] These routines compute the irregular modified Bessel function of fractional order ν , $K_{\nu}(x)$ for x > 0, $\nu > 0$.

These routines compute the logarithm of the irregular modified Bessel function of fractional order ν , $\ln(K_{\nu}(x))$ for x > 0, $\nu > 0$.

These routines compute the scaled irregular modified Bessel function of fractional order ν , $\exp(+|x|)K_{\nu}(x)$ for x > 0, $\nu > 0$.

7.5.13 Zeros of Regular Bessel Functions

double gsl_sf_bessel_zero_J0 (unsigned int s) [Function] int gsl_sf_bessel_zero_J0_e (unsigned int s, gsl_sf_result * result) [Function] These routines compute the location of the s-th positive zero of the Bessel function $J_0(x)$.

double gsl_sf_bessel_zero_J1 (unsigned int s) [Function] int gsl_sf_bessel_zero_J1_e (unsigned int s, gsl_sf_result * result) [Function] These routines compute the location of the s-th positive zero of the Bessel function $J_1(x)$.

double gsl_sf_bessel_zero_Jnu (double nu, unsigned int s) [Function] int gsl_sf_bessel_zero_Jnu_e (double nu, unsigned int s, gsl_sf_result * result)

These routines compute the location of the s-th positive zero of the Bessel function $J_{\nu}(x)$. The current implementation does not support negative values of nu.

7.6 Clausen Functions

The Clausen function is defined by the following integral,

$$Cl_2(x) = -\int_0^x dt \log(2\sin(t/2))$$

It is related to the dilogarithm by $Cl_2(\theta) = \text{Im}(Li_2(e^{i\theta}))$. The Clausen functions are declared in the header file gsl_sf_clausen.h.

double gsl_sf_clausen (double x) [Function] int gsl_sf_clausen_e (double x, $gsl_sf_result * result$) [Function] These routines compute the Clausen integral $Cl_2(x)$.

7.7 Coulomb Functions

The prototypes of the Coulomb functions are declared in the header file gsl_sf_coulomb.h. Both bound state and scattering solutions are available.

7.7.1 Normalized Hydrogenic Bound States

These routines compute the lowest-order normalized hydrogenic bound state radial wavefunction $R_1 := 2Z\sqrt{Z}\exp(-Zr)$.

These routines compute the *n*-th normalized hydrogenic bound state radial wavefunction,

$$R_n := \frac{2Z^{3/2}}{n^2} \left(\frac{2Zr}{n}\right)^l \sqrt{\frac{(n-l-1)!}{(n+l)!}} \exp(-Zr/n) L_{n-l-1}^{2l+1}(2Zr/n).$$

where $L_b^a(x)$ is the generalized Laguerre polynomial (see Section 7.22 [Laguerre Functions], page 62). The normalization is chosen such that the wavefunction ψ is given by $\psi(n, l, r) = R_n Y_{lm}$.

7.7.2 Coulomb Wave Functions

The Coulomb wave functions $F_L(\eta, x)$, $G_L(\eta, x)$ are described in Abramowitz & Stegun, Chapter 14. Because there can be a large dynamic range of values for these functions, overflows are handled gracefully. If an overflow occurs, GSL_EOVRFLW is signalled and exponent(s) are returned through the modifiable parameters \exp_FF , \exp_GF . The full solution can be reconstructed from the following relations,

$$F_L(\eta, x) = fc[k_L] * \exp(exp_F)$$
$$G_L(\eta, x) = gc[k_L] * \exp(exp_G)$$

$$F'_L(\eta, x) = fcp[k_L] * \exp(exp_F)$$

$$G'_L(\eta, x) = gcp[k_L] * \exp(exp_G)$$

This function computes the Coulomb wave functions $F_L(\eta, x)$, $G_{L-k}(\eta, x)$ and their derivatives $F'_L(\eta, x)$, $G'_{L-k}(\eta, x)$ with respect to x. The parameters are restricted to L, L-k > -1/2, x > 0 and integer k. Note that L itself is not restricted to being an integer. The results are stored in the parameters F, G for the function values and Fp, Gp for the derivative values. If an overflow occurs, $GSL_EOVRFLW$ is returned and scaling exponents are stored in the modifiable parameters exp_F , exp_G .

This function computes the Coulomb wave function $F_L(\eta, x)$ for $L = Lmin \dots Lmin + kmax$, storing the results in fc-array. In the case of overflow the exponent is stored in F-exponent.

This function computes the functions $F_L(\eta, x)$, $G_L(\eta, x)$ for $L = Lmin \dots Lmin + kmax$ storing the results in f_{c_array} and g_{c_array} . In the case of overflow the exponents are stored in $F_exponent$ and $G_exponent$.

This function computes the functions $F_L(\eta, x)$, $G_L(\eta, x)$ and their derivatives $F'_L(\eta, x)$, $G'_L(\eta, x)$ for $L = Lmin \dots Lmin + kmax$ storing the results in fc-array, gc-array, fcp-array and gcp-array. In the case of overflow the exponents are stored in F-exponent and G-exponent.

This function computes the Coulomb wave function divided by the argument $F_L(\eta, x)/x$ for $L = Lmin \dots Lmin + kmax$, storing the results in fc-array. In the case of overflow the exponent is stored in F-exponent. This function reduces to spherical Bessel functions in the limit $\eta \to 0$.

7.7.3 Coulomb Wave Function Normalization Constant

The Coulomb wave function normalization constant is defined in Abramowitz 14.1.7.

This function computes the Coulomb wave function normalization constant $C_L(\eta)$ for L > -1.

This function computes the Coulomb wave function normalization constant $C_L(\eta)$ for $L = Lmin \dots Lmin + kmax$, Lmin > -1.

7.8 Coupling Coefficients

The Wigner 3-j, 6-j and 9-j symbols give the coupling coefficients for combined angular momentum vectors. Since the arguments of the standard coupling coefficient functions are integer or half-integer, the arguments of the following functions are, by convention, integers equal to twice the actual spin value. For information on the 3-j coefficients see Abramowitz & Stegun, Section 27.9. The functions described in this section are declared in the header file gsl_sf_coupling.h.

7.8.1 3-j Symbols

double gsl_sf_coupling_3j (int two_ja, int two_jb, int two_jc, int two_ma, int two_mb, int two_mc) [Function]

int gsl_sf_coupling_3j_e (int two_ja, int two_jb, int two_jc, int two_ma, int two_mb, int two_mc, gsl_sf_result * result)

These routines compute the Wigner 3-j coefficient,

$$\begin{pmatrix}
ja & jb & jc \\
ma & mb & mc
\end{pmatrix}$$

where the arguments are given in half-integer units, $ja = two_ja/2$, $ma = two_ma/2$, etc.

7.8.2 6-j Symbols

double gsl_sf_coupling_6j (int two_ja, int two_jb, int two_jc, int two_jd, int two_je, int two_jf) [Function]

These routines compute the Wigner 6-j coefficient,

$$\left\{
 \begin{array}{ccc}
 ja & jb & jc \\
 jd & je & jf
 \end{array}
\right\}$$

where the arguments are given in half-integer units, $ja = two_ja/2$, $ma = two_ma/2$, etc.

7.8.3 9-j Symbols

These routines compute the Wigner 9-j coefficient,

$$\left\{
 \begin{array}{lll}
 ja & jb & jc \\
 jd & je & jf \\
 jq & jh & ji
 \end{array}
 \right\}$$

where the arguments are given in half-integer units, $ja = two_ja/2$, $ma = two_ma/2$, etc.

7.9 Dawson Function

The Dawson integral is defined by $\exp(-x^2) \int_0^x dt \exp(t^2)$. A table of Dawson's integral can be found in Abramowitz & Stegun, Table 7.5. The Dawson functions are declared in the header file gsl_sf_dawson.h.

double gsl_sf_dawson (double x) [Function]
int gsl_sf_dawson_e (double x, gsl_sf_result * result) [Function]
These routines compute the value of Dawson's integral for x.

7.10 Debye Functions

The Debye functions $D_n(x)$ are defined by the following integral,

$$D_n(x) = \frac{n}{x^n} \int_0^x dt \frac{t^n}{e^t - 1}$$

For further information see Abramowitz & Stegun, Section 27.1. The Debye functions are declared in the header file gsl_sf_debye.h.

```
double gsl_sf_debye_1 (double x) [Function] int gsl_sf_debye_1_e (double x, gsl_sf_result * result) [Function] These routines compute the first-order Debye function D_1(x) = (1/x) \int_0^x dt (t/(e^t - 1)). [Function] int gsl_sf_debye_2 (double x, gsl_sf_result * result) [Function]
```

These routines compute the second-order Debye function $D_2(x) = (2/x^2) \int_0^x dt (t^2/(e^t - 1))$.

double gsl_sf_debye_3 (double x) [Function] int gsl_sf_debye_3_e (double x, gsl_sf_result * result) [Function] These routines compute the third-order Debye function $D_3(x) = (3/x^3) \int_0^x dt (t^3/(e^t - 1))$.

double gsl_sf_debye_4 (double x) [Function] int gsl_sf_debye_4_e (double x, gsl_sf_result * result) [Function] These routines compute the fourth-order Debye function $D_4(x) = (4/x^4) \int_0^x dt (t^4/(e^t - 1))$.

double gsl_sf_debye_5 (double x) [Function] int gsl_sf_debye_5_e (double x, gsl_sf_result * result) [Function] These routines compute the fifth-order Debye function $D_5(x) = (5/x^5) \int_0^x dt (t^5/(e^t - 1))$.

double gsl_sf_debye_6 (double x) [Function] int gsl_sf_debye_6_e (double x, gsl_sf_result * result) [Function] These routines compute the sixth-order Debye function $D_6(x) = (6/x^6) \int_0^x dt (t^6/(e^t - 1))$.

7.11 Dilogarithm

The functions described in this section are declared in the header file gsl_sf_dilog.h.

7.11.1 Real Argument

```
double gsl_sf_dilog (double x) [Function] int gsl_sf_dilog_e (double x, gsl_sf_result * result) [Function]

These routines compute the dilogarithm for a real argument. In Lewin's notation this
```

These routines compute the dilogarithm for a real argument. In Lewin's notation this is $Li_2(x)$, the real part of the dilogarithm of a real x. It is defined by the integral representation $Li_2(x) = -\text{Re} \int_0^x ds \log(1-s)/s$. Note that $\text{Im}(Li_2(x)) = 0$ for $x \leq 1$, and $-\pi \log(x)$ for x > 1.

Note that Abramowitz & Stegun refer to the Spence integral $S(x) = Li_2(1-x)$ as the dilogarithm rather than $Li_2(x)$.

7.11.2 Complex Argument

This function computes the full complex-valued dilogarithm for the complex argument $z = r \exp(i\theta)$. The real and imaginary parts of the result are returned in result_re, result_im.

7.12 Elementary Operations

The following functions allow for the propagation of errors when combining quantities by multiplication. The functions are declared in the header file gsl_sf_elementary.h.

int gsl_sf_multiply_e (double x, double y, gsl_sf_result * result) [Function]
This function multiplies x and y storing the product and its associated error in result.

This function multiplies x and y with associated absolute errors dx and dy. The product $xy \pm xy\sqrt{(dx/x)^2 + (dy/y)^2}$ is stored in result.

7.13 Elliptic Integrals

The functions described in this section are declared in the header file gsl_sf_ellint.h. Further information about the elliptic integrals can be found in Abramowitz & Stegun, Chapter 17.

7.13.1 Definition of Legendre Forms

The Legendre forms of elliptic integrals $F(\phi, k)$, $E(\phi, k)$ and $\Pi(\phi, k, n)$ are defined by,

$$F(\phi, k) = \int_0^{\phi} dt \frac{1}{\sqrt{(1 - k^2 \sin^2(t))}}$$

$$E(\phi, k) = \int_0^{\phi} dt \sqrt{(1 - k^2 \sin^2(t))}$$

$$\Pi(\phi, k, n) = \int_0^{\phi} dt \frac{1}{(1 + n \sin^2(t))\sqrt{1 - k^2 \sin^2(t)}}$$

The complete Legendre forms are denoted by $K(k) = F(\pi/2, k)$ and $E(k) = E(\pi/2, k)$.

The notation used here is based on Carlson, Numerische Mathematik 33 (1979) 1 and differs slightly from that used by Abramowitz & Stegun, where the functions are given in terms of the parameter $m = k^2$ and n is replaced by -n.

7.13.2 Definition of Carlson Forms

The Carlson symmetric forms of elliptical integrals RC(x,y), RD(x,y,z), RF(x,y,z) and RJ(x,y,z,p) are defined by,

$$RC(x,y) = 1/2 \int_0^\infty dt (t+x)^{-1/2} (t+y)^{-1}$$

$$RD(x,y,z) = 3/2 \int_0^\infty dt (t+x)^{-1/2} (t+y)^{-1/2} (t+z)^{-3/2}$$

$$RF(x,y,z) = 1/2 \int_0^\infty dt (t+x)^{-1/2} (t+y)^{-1/2} (t+z)^{-1/2}$$

$$RJ(x,y,z,p) = 3/2 \int_0^\infty dt (t+x)^{-1/2} (t+y)^{-1/2} (t+z)^{-1/2} (t+p)^{-1}$$

7.13.3 Legendre Form of Complete Elliptic Integrals

These routines compute the complete elliptic integral K(k) to the accuracy specified by the mode variable mode. Note that Abramowitz & Stegun define this function in terms of the parameter $m = k^2$.

These routines compute the complete elliptic integral E(k) to the accuracy specified by the mode variable mode. Note that Abramowitz & Stegun define this function in terms of the parameter $m = k^2$.

These routines compute the complete elliptic integral $\Pi(k,n)$ to the accuracy specified by the mode variable mode. Note that Abramowitz & Stegun define this function in terms of the parameters $m = k^2$ and $\sin^2(\alpha) = k^2$, with the change of sign $n \to -n$.

7.13.4 Legendre Form of Incomplete Elliptic Integrals

These routines compute the incomplete elliptic integral $F(\phi, k)$ to the accuracy specified by the mode variable mode. Note that Abramowitz & Stegun define this function in terms of the parameter $m = k^2$.

These routines compute the incomplete elliptic integral $E(\phi, k)$ to the accuracy specified by the mode variable mode. Note that Abramowitz & Stegun define this function in terms of the parameter $m = k^2$.

These routines compute the incomplete elliptic integral $\Pi(\phi, k, n)$ to the accuracy specified by the mode variable *mode*. Note that Abramowitz & Stegun define this function in terms of the parameters $m = k^2$ and $\sin^2(\alpha) = k^2$, with the change of sign $n \to -n$.

These functions compute the incomplete elliptic integral $D(\phi, k)$ which is defined through the Carlson form RD(x, y, z) by the following relation,

$$D(\phi, k, n) = \frac{1}{3}(\sin \phi)^3 R D(1 - \sin^2(\phi), 1 - k^2 \sin^2(\phi), 1).$$

The argument n is not used and will be removed in a future release.

7.13.5 Carlson Forms

These routines compute the incomplete elliptic integral RC(x, y) to the accuracy specified by the mode variable mode.

These routines compute the incomplete elliptic integral RD(x, y, z) to the accuracy specified by the mode variable mode.

These routines compute the incomplete elliptic integral RF(x, y, z) to the accuracy specified by the mode variable mode.

These routines compute the incomplete elliptic integral RJ(x, y, z, p) to the accuracy specified by the mode variable mode.

7.14 Elliptic Functions (Jacobi)

The Jacobian Elliptic functions are defined in Abramowitz & Stegun, Chapter 16. The functions are declared in the header file gsl_sf_elljac.h.

This function computes the Jacobian elliptic functions sn(u|m), cn(u|m), dn(u|m) by descending Landen transformations.

7.15 Error Functions

The error function is described in Abramowitz & Stegun, Chapter 7. The functions in this section are declared in the header file gsl_sf_erf.h.

7.15.1 Error Function

```
double gsl_sf_erf (double x) [Function] int gsl_sf_erf_e (double x, gsl_sf_result * result) [Function] These routines compute the error function \operatorname{erf}(x), where \operatorname{erf}(x) = (2/\sqrt{\pi}) \int_0^x dt \exp(-t^2).
```

7.15.2 Complementary Error Function

```
double gsl_sf_erfc (double x) [Function] int gsl_sf_erfc_e (double x, gsl_sf_result * result) [Function] These routines compute the complementary error function \operatorname{erfc}(x) = 1 - \operatorname{erf}(x) = (2/\sqrt{\pi}) \int_x^\infty \exp(-t^2).
```

7.15.3 Log Complementary Error Function

```
double gsl_sf_log_erfc (double x) [Function] int gsl_sf_log_erfc_e (double x, gsl_sf_result * result) [Function] These routines compute the logarithm of the complementary error function \log(\operatorname{erfc}(x)).
```

7.15.4 Probability functions

The probability functions for the Normal or Gaussian distribution are described in Abramowitz & Stegun, Section 26.2.

```
double gsl_sf_erf_Z (double x) [Function] int gsl_sf_erf_Z_e (double x, gsl_sf_result * result) [Function] These routines compute the Gaussian probability density function Z(x) = (1/\sqrt{2\pi}) \exp(-x^2/2).
```

double gsl_sf_erf_Q (double x) [Function] int gsl_sf_erf_Q_e (double x, gsl_sf_result * result) [Function] These routines compute the upper tail of the Gaussian probability function
$$Q(x) = \frac{1}{2} \int_{-\infty}^{\infty} \frac{1}{2} \int_{-$$

 $(1/\sqrt{2\pi})$ $\int_{x}^{\infty} dt \exp(-t^2/2)$.

The hazard function for the normal distribution, also known as the inverse Mills' ratio, is defined as,

$$h(x) = \frac{Z(x)}{Q(x)} = \sqrt{\frac{2}{\pi}} \frac{\exp(-x^2/2)}{\operatorname{erfc}(x/\sqrt{2})}$$

It decreases rapidly as x approaches $-\infty$ and asymptotes to $h(x) \sim x$ as x approaches $+\infty$.

double gsl_sf_hazard (double x) [Function] int gsl_sf_hazard_e (double x, gsl_sf_result * result) [Function] These routines compute the hazard function for the normal distribution.

7.16 Exponential Functions

The functions described in this section are declared in the header file gsl_sf_exp.h.

7.16.1 Exponential Function

checking.

double gsl_sf_exp (double x) [Function] int gsl_sf_exp_e (double x, gsl_sf_result * result) [Function] These routines provide an exponential function $\exp(x)$ using GSL semantics and error

int gsl_sf_exp_e10_e (double x, gsl_sf_result_e10 * result) [Function] This function computes the exponential $\exp(x)$ using the gsl_sf_result_e10 type to return a result with extended range. This function may be useful if the value of $\exp(x)$ would overflow the numeric range of double.

double gsl_sf_exp_mult (double x, double y) [Function] int gsl_sf_exp_mult_e (double x, double y, gsl_sf_result * result) [Function] These routines exponentiate x and multiply by the factor y to return the product $y \exp(x)$.

int gsl_sf_exp_mult_e10_e (const double x, const double y, [Function] gsl_sf_result_e10 * result)

This function computes the product $y \exp(x)$ using the gsl_sf_result_e10 type to return a result with extended numeric range.

7.16.2 Relative Exponential Functions

double gsl_sf_expm1 (double x) [Function] int gsl_sf_expm1_e (double x, gsl_sf_result * result) [Function]

These routines compute the quantity $\exp(x) - 1$ using an algorithm that is accurate for small x.

These routines compute the quantity $(\exp(x) - 1)/x$ using an algorithm that is accurate for small x. For small x the algorithm is based on the expansion $(\exp(x) - 1)/x = 1 + x/2 + x^2/(2*3) + x^3/(2*3*4) + \dots$

double gsl_sf_exprel_2 (double x) [Function]

int gsl_sf_exprel_2_e (double x, gsl_sf_result * result)
[Function]

These routines compute the quantity $2(\exp(x) - 1 - x)/x^2$ using an algorithm that is accurate for small x. For small x the algorithm is based on the expansion $2(\exp(x) - 1 - x)/x^2 = 1 + x/3 + x^2/(3*4) + x^3/(3*4*5) + \dots$

These routines compute the N-relative exponential, which is the n-th generalization of the functions gsl_sf_exprel and $gsl_sf_exprel_2$. The N-relative exponential is given by,

$$\operatorname{exprel}_{N}(x) = N!/x^{N} \left(\exp(x) - \sum_{k=0}^{N-1} x^{k}/k! \right)$$

$$= 1 + x/(N+1) + x^{2}/((N+1)(N+2)) + \dots$$

$$= {}_{1}F_{1}(1, 1+N, x)$$

7.16.3 Exponentiation With Error Estimate

int $gsl_sf_exp_err_e$ (double x, double dx, $gsl_sf_result * result$) [Function] This function exponentiates x with an associated absolute error dx.

This function exponentiates a quantity x with an associated absolute error dx using the $gsl_sf_result_e10$ type to return a result with extended range.

This routine computes the product $y \exp(x)$ for the quantities x, y with associated absolute errors dx, dy.

This routine computes the product $y \exp(x)$ for the quantities x, y with associated absolute errors dx, dy using the $gsl_sf_result_e10$ type to return a result with extended range.

7.17 Exponential Integrals

Information on the exponential integrals can be found in Abramowitz & Stegun, Chapter 5. These functions are declared in the header file gsl_sf_expint.h.

7.17.1 Exponential Integral

double gsl_sf_expint_E1 (double x) [Function] int gsl_sf_expint_E1_e (double x, gsl_sf_result * result) [Function] These routines compute the exponential integral $E_1(x)$,

$$E_1(x) := \operatorname{Re} \int_1^\infty dt \exp(-xt)/t.$$

double gsl_sf_expint_E2 (double x)

[Function]

int gsl_sf_expint_E2_e (double x, gsl_sf_result * result)

[Function]

These routines compute the second-order exponential integral $E_2(x)$,

$$E_2(x) := \operatorname{Re} \int_1^\infty dt \exp(-xt)/t^2.$$

double gsl_sf_expint_En (int n, double x)

[Function]

int gsl_sf_expint_En_e (int n, double x, gsl_sf_result * result)

[Function]

These routines compute the exponential integral $E_n(x)$ of order n,

$$E_n(x) := \operatorname{Re} \int_1^\infty dt \exp(-xt)/t^n.$$

$7.17.2 \, \mathrm{Ei}(x)$

double gsl_sf_expint_Ei (double x)
int gsl_sf_expint_Ei_e (double x, gsl_sf_result * result)

[Function]

[Function]

These routines compute the exponential integral Ei(x),

$$\operatorname{Ei}(x) := -PV\left(\int_{-x}^{\infty} dt \exp(-t)/t\right)$$

where PV denotes the principal value of the integral.

7.17.3 Hyperbolic Integrals

 $\verb"double gsl_sf_Shi" (double x)$

[Function]

int gsl_sf_Shi_e (double x, gsl_sf_result * result)

[Function]

These routines compute the integral $Shi(x) = \int_0^x dt \sinh(t)/t$.

double gsl_sf_Chi (double x)

[Function]

int gsl_sf_Chi_e (double x, gsl_sf_result * result)

[Function]

These routines compute the integral $Chi(x) := Re[\gamma_E + \log(x) + \int_0^x dt(\cosh(t) - 1)/t]$, where γ_E is the Euler constant (available as the macro M_EULER).

$7.17.4 \text{ Ei}_{-3}(x)$

double gsl_sf_expint_3 (double x)

[Function]

int gsl_sf_expint_3_e (double x, gsl_sf_result * result)

Function]

These routines compute the third-order exponential integral $\text{Ei}_3(x) = \int_0^x dt \exp(-t^3)$ for $x \ge 0$.

7.17.5 Trigonometric Integrals

double gsl_sf_Si (const double x) [Function] int gsl_sf_Si_e (double x, gsl_sf_result * result) [Function] These routines compute the Sine integral $Si(x) = \int_0^x dt \sin(t)/t$.

These routines compute the Sine integral $Si(x) = \int_0^x at \sin(t)/t$.

double gsl_sf_Ci (const double x) [Function]

int gsl_sf_Ci_e (double x, gsl_sf_result * result) [Function]

These routines compute the Cosine integral $Ci(x) = -\int_{x}^{\infty} dt \cos(t)/t$ for x > 0.

7.17.6 Arctangent Integral

double gsl_sf_atanint (double x) [Function] int gsl_sf_atanint_e (double x, gsl_sf_result * result) [Function] These routines compute the Arctangent integral, which is defined as AtanInt(x) = $\int_0^x dt \arctan(t)/t$.

7.18 Fermi-Dirac Function

The functions described in this section are declared in the header file gsl_sf_fermi_dirac.h.

7.18.1 Complete Fermi-Dirac Integrals

The complete Fermi-Dirac integral $F_i(x)$ is given by,

$$F_j(x) := \frac{1}{\Gamma(j+1)} \int_0^\infty dt \frac{t^j}{(\exp(t-x)+1)}$$

Note that the Fermi-Dirac integral is sometimes defined without the normalisation factor in other texts.

double gsl_sf_fermi_dirac_m1 (double x) [Function]

int gsl_sf_fermi_dirac_m1_e (double x, gsl_sf_result * result) [Function]

These routines compute the complete Fermi-Dirac integral with an index of -1. This integral is given by $F_{-1}(x) = e^x/(1 + e^x)$.

double gsl_sf_fermi_dirac_0 (double x) [Function]

int gsl_sf_fermi_dirac_0_e (double x, gsl_sf_result * result) [Function]

These routines compute the complete Fermi-Dirac integral with an index of 0. This integral is given by $F_0(x) = \ln(1 + e^x)$.

double gsl_sf_fermi_dirac_1 (double x) [Function]

int gsl_sf_fermi_dirac_1_e (double x, gsl_sf_result * result) [Function]

These routines compute the complete Fermi-Dirac integral with an index of 1, $F_1(x) = \int_0^\infty dt (t/(\exp(t-x)+1))$.

double gsl_sf_fermi_dirac_2 (double x) [Function] int gsl_sf_fermi_dirac_2_e (double x, gsl_sf_result * result)

These routines compute the complete Fermi-Dirac integral with an index of 2, $F_2(x) = (1/2) \int_0^\infty dt (t^2/(\exp(t-x)+1))$.

These routines compute the complete Fermi-Dirac integral with an integer index of j, $F_j(x) = (1/\Gamma(j+1)) \int_0^\infty dt (t^j/(\exp(t-x)+1))$.

double gsl_sf_fermi_dirac_mhalf (double x) [Function] int gsl_sf_fermi_dirac_mhalf_e (double x, gsl_sf_result * result) [Function] These routines compute the complete Fermi-Dirac integral $F_{-1/2}(x)$.

double gsl_sf_fermi_dirac_half (double x) [Function] int gsl_sf_fermi_dirac_half_e (double x, $gsl_sf_result * result$) [Function] These routines compute the complete Fermi-Dirac integral $F_{1/2}(x)$.

double gsl_sf_fermi_dirac_3half (double x) [Function] int gsl_sf_fermi_dirac_3half_e (double x, gsl_sf_result * result) [Function] These routines compute the complete Fermi-Dirac integral $F_{3/2}(x)$.

7.18.2 Incomplete Fermi-Dirac Integrals

The incomplete Fermi-Dirac integral $F_i(x, b)$ is given by,

$$F_j(x,b) := \frac{1}{\Gamma(j+1)} \int_b^\infty dt \frac{t^j}{(\exp(t-x)+1)}$$

These routines compute the incomplete Fermi-Dirac integral with an index of zero, $F_0(x,b) = \ln(1+e^{b-x}) - (b-x)$.

7.19 Gamma and Beta Functions

This following routines compute the gamma and beta functions in their full and incomplete forms, as well as various kinds of factorials. The functions described in this section are declared in the header file gsl_sf_gamma.h.

7.19.1 Gamma Functions

The Gamma function is defined by the following integral,

$$\Gamma(x) = \int_0^\infty dt \, t^{x-1} \exp(-t)$$

It is related to the factorial function by $\Gamma(n) = (n-1)!$ for positive integer n. Further information on the Gamma function can be found in Abramowitz & Stegun, Chapter 6.

double gsl_sf_gamma (double x) [Function]
int gsl_sf_gamma_e (double x, gsl_sf_result * result) [Function]
These routines compute the Common function \(\Gamma \) (which to r not being a positive

These routines compute the Gamma function $\Gamma(x)$, subject to x not being a negative integer or zero. The function is computed using the real Lanczos method. The maximum value of x such that $\Gamma(x)$ is not considered an overflow is given by the macro GSL_SF_GAMMA_XMAX and is 171.0.

These routines compute the logarithm of the Gamma function, $\log(\Gamma(x))$, subject to x not being a negative integer or zero. For x < 0 the real part of $\log(\Gamma(x))$ is returned, which is equivalent to $\log(|\Gamma(x)|)$. The function is computed using the real Lanczos method.

This routine computes the sign of the gamma function and the logarithm of its magnitude, subject to x not being a negative integer or zero. The function is computed using the real Lanczos method. The value of the gamma function and its error can be reconstructed using the relation $\Gamma(x) = sgn * \exp(result_lg)$, taking into account the two components of $result_lg$.

These routines compute the regulated Gamma Function $\Gamma^*(x)$ for x > 0. The regulated gamma function is given by,

$$\Gamma^*(x) = \Gamma(x) / (\sqrt{2\pi} x^{(x-1/2)} \exp(-x))$$
$$= \left(1 + \frac{1}{12x} + \dots\right) \quad \text{for } x \to \infty$$

and is a useful suggestion of Temme.

double gsl_sf_gammainv (double x) [Function] int gsl_sf_gammainv_e (double x, gsl_sf_result * result) [Function]

These routines compute the reciprocal of the gamma function
$$1/\Gamma(x)$$
 using the real

These routines compute the reciprocal of the gamma function, $1/\Gamma(x)$ using the real Lanczos method.

This routine computes $\log(\Gamma(z))$ for complex $z=z_r+iz_i$ and z not a negative integer or zero, using the complex Lanczos method. The returned parameters are $lnr=\log|\Gamma(z)|$ and $arg=\arg(\Gamma(z))$ in $(-\pi,\pi]$. Note that the phase part (arg) is not well-determined when |z| is very large, due to inevitable roundoff in restricting to $(-\pi,\pi]$. This will result in a GSL_ELOSS error when it occurs. The absolute value part (lnr), however, never suffers from loss of precision.

7.19.2 Factorials

Although factorials can be computed from the Gamma function, using the relation $n! = \Gamma(n+1)$ for non-negative integer n, it is usually more efficient to call the functions in this section, particularly for small values of n, whose factorial values are maintained in hardcoded tables.

symbol,

```
double gsl_sf_fact (unsigned int n)
                                                                            [Function]
int gsl_sf_fact_e (unsigned int n, gsl_sf_result * result)
                                                                            [Function]
     These routines compute the factorial n!. The factorial is related to the Gamma
     function by n! = \Gamma(n+1). The maximum value of n such that n! is not considered
     an overflow is given by the macro GSL_SF_FACT_NMAX and is 170.
double gsl_sf_doublefact (unsigned int n)
                                                                            [Function]
int gsl_sf_doublefact_e (unsigned int n, gsl_sf_result * result)
                                                                            [Function]
     These routines compute the double factorial n!! = n(n-2)(n-4)... The maximum
     value of n such that n!! is not considered an overflow is given by the macro GSL_SF_
     DOUBLEFACT_NMAX and is 297.
double gsl_sf_lnfact (unsigned int n)
                                                                            [Function]
int gsl_sf_lnfact_e (unsigned int n, gsl_sf_result * result)
                                                                            [Function]
     These routines compute the logarithm of the factorial of n, \log(n!). The algorithm is
     faster than computing \ln(\Gamma(n+1)) via gsl_sf_lngamma for n < 170, but defers for
     larger n.
double gsl_sf_lndoublefact (unsigned int n)
                                                                            [Function]
int gsl_sf_lndoublefact_e (unsigned int n, gsl_sf_result * result)
                                                                            [Function]
     These routines compute the logarithm of the double factorial of n, \log(n!!).
double gsl_sf_choose (unsigned int n, unsigned int m)
                                                                            [Function]
int gsl_sf_choose_e (unsigned int n, unsigned int m, gsl_sf_result *
                                                                            [Function]
     These routines compute the combinatorial factor n choose m = n!/(m!(n-m)!)
double gsl_sf_lnchoose (unsigned int n, unsigned int m)
                                                                            [Function]
int gsl_sf_lnchoose_e (unsigned int m, unsigned int m, gsl_sf_result *
                                                                            [Function]
         result)
     These routines compute the logarithm of n choose m. This is equivalent to the sum
     \log(n!) - \log(m!) - \log((n-m)!).
double gsl_sf_taylorcoeff (int n, double x)
                                                                            [Function]
int gsl_sf_taylorcoeff_e (int n, double x, gsl_sf_result * result)
                                                                            [Function]
     These routines compute the Taylor coefficient x^n/n! for x \geq 0, n \geq 0.
7.19.3 Pochhammer Symbol
double gsl_sf_poch (double a, double x)
                                                                            [Function]
int gsl_sf_poch_e (double a, double x, gsl_sf_result * result)
                                                                            [Function]
     These routines compute the Pochhammer symbol (a)_x = \Gamma(a+x)/\Gamma(a).
     Pochhammer symbol is also known as the Apell symbol and sometimes written as
     (a,x). When a and a+x are negative integers or zero, the limiting value of the ratio
     is returned.
double gsl_sf_lnpoch (double a, double x)
                                                                            [Function]
int gsl_sf_lnpoch_e (double a, double x, gsl_sf_result * result)
                                                                            [Function]
```

These routines compute the logarithm of the Pochhammer

 $\log((a)_x) = \log(\Gamma(a+x)/\Gamma(a)).$

These routines compute the sign of the Pochhammer symbol and the logarithm of its magnitude. The computed parameters are $result = \log(|(a)_x|)$ with a corresponding error term, and $sgn = \text{sgn}((a)_x)$ where $(a)_x = \Gamma(a+x)/\Gamma(a)$.

double gsl_sf_pochrel (double a, double x) [Function] int gsl_sf_pochrel_e (double a, double x, gsl_sf_result * result) [Function] These routines compute the relative Pochhammer symbol $((a)_x - 1)/x$ where $(a)_x = 1$

 $\Gamma(a+x)/\Gamma(a)$.

7.19.4 Incomplete Gamma Functions

double gsl_sf_gamma_inc (double a, double x) [Function]

int gsl_sf_gamma_inc_e (double a, double x, gsl_sf_result * result) [Function] These functions compute the unnormalized incomplete Gamma Function $\Gamma(a,x) = \int_x^\infty dt \, t^{(a-1)} \exp(-t)$ for a real and $x \ge 0$.

double gsl_sf_gamma_inc_Q (double a, double x)
[Function]

int gsl_sf_gamma_inc_Q_e (double a, double x, gsl_sf_result * result) [Function] These routines compute the normalized incomplete Gamma Function $Q(a,x)=1/\Gamma(a)\int_x^\infty dt\, t^{(a-1)}\exp(-t)$ for $a>0,\,x\geq0$.

double gsl_sf_gamma_inc_P (double a, double x) [Function]

int gsl_sf_gamma_inc_P_e (double a, double x, gsl_sf_result * result) [Function] These routines compute the complementary normalized incomplete Gamma Function $P(a,x) = 1 - Q(a,x) = 1/\Gamma(a) \int_0^x dt \, t^{(a-1)} \exp(-t)$ for $a > 0, x \ge 0$.

Note that Abramowitz & Stegun call P(a, x) the incomplete gamma function (section 6.5).

7.19.5 Beta Functions

double gsl_sf_beta (double a, double b) [Function]

int gsl_sf_beta_e (double a, double b, gsl_sf_result * result) [Function] These routines compute the Beta Function, $B(a,b) = \Gamma(a)\Gamma(b)/\Gamma(a+b)$ subject to a

double gsl_sf_lnbeta (double a, double b) [Function]

int gsl_sf_lnbeta_e (double a, double b, gsl_sf_result * result) [Function]

These routines compute the logarithm of the Beta Function, log(B(a,b)) subject to

a and b not being negative integers.

7.19.6 Incomplete Beta Function

and b not being negative integers.

These routines compute the normalized incomplete Beta function $I_x(a,b) = B_x(a,b)/B(a,b)$ where $B_x(a,b) = \int_0^x t^{a-1}(1-t)^{b-1}dt$ for

 $0 \le x \le 1$. For a > 0, b > 0 the value is computed using a continued fraction expansion. For all other values it is computed using the relation $I_x(a,b,x) = (1/a)x^a {}_2F_1(a,1-b,a+1,x)/B(a,b)$.

7.20 Gegenbauer Functions

tations for n = 1, 2, 3.

The Gegenbauer polynomials are defined in Abramowitz & Stegun, Chapter 22, where they are known as Ultraspherical polynomials. The functions described in this section are declared in the header file gsl_sf_gegenbauer.h.

```
double gsl_sf_gegenpoly_1 (double lambda, double x) [Function] double gsl_sf_gegenpoly_2 (double lambda, double x) [Function] double gsl_sf_gegenpoly_3 (double lambda, double x) [Function] int gsl_sf_gegenpoly_1_e (double lambda, double x, gsl_sf_result * [Function] result) [Function] result] [Function] result] [Function] result] [Function] result] [Function] result] [Function] [Funct
```

double gsl_sf_gegenpoly_n (int n, double lambda, double x) [Function]

int gsl_sf_gegenpoly_n_e (int n, double lambda, double x,
gsl_sf_result * result)

[Function]

These functions evaluate the Gegenbauer polynomial $C_n^{(\lambda)}(x)$ for a specific value of n, lambda, x subject to $\lambda > -1/2$, $n \ge 0$.

This function computes an array of Gegenbauer polynomials $C_n^{(\lambda)}(x)$ for $n = 0, 1, 2, \dots, nmax$, subject to $\lambda > -1/2, nmax \ge 0$.

7.21 Hypergeometric Functions

Hypergeometric functions are described in Abramowitz & Stegun, Chapters 13 and 15. These functions are declared in the header file gsl_sf_hyperg.h.

```
double gsl_sf_hyperg_0F1 (double c, double x) [Function] int gsl_sf_hyperg_0F1_e (double c, double x, gsl_sf_result * result) [Function] These routines compute the hypergeometric function {}_0F_1(c,x).
```

These routines compute the confluent hypergeometric function ${}_{1}F_{1}(m, n, x) = M(m, n, x)$ for integer parameters m, n.

These routines compute the confluent hypergeometric function ${}_{1}F_{1}(a,b,x) = M(a,b,x)$ for general parameters a, b.

These routines compute the confluent hypergeometric function U(m, n, x) for integer parameters m, n.

This routine computes the confluent hypergeometric function U(m, n, x) for integer parameters m, n using the $gsl_sf_result_e10$ type to return a result with extended range.

These routines compute the confluent hypergeometric function U(a, b, x).

This routine computes the confluent hypergeometric function U(a, b, x) using the $gsl_sf_result_e10$ type to return a result with extended range.

These routines compute the Gauss hypergeometric function ${}_{2}F_{1}(a,b,c,x) = F(a,b,c,x)$ for |x| < 1.

If the arguments (a, b, c, x) are too close to a singularity then the function can return the error code GSL_EMAXITER when the series approximation converges too slowly. This occurs in the region of x = 1, c - a - b = m for integer m.

These routines compute the Gauss hypergeometric function ${}_{2}F_{1}(a_{R}+ia_{I},aR-iaI,c,x)$ with complex parameters for |x|<1.

These routines compute the renormalized Gauss hypergeometric function ${}_2F_1(a,b,c,x)/\Gamma(c)$ for |x|<1.

These routines compute the renormalized Gauss hypergeometric function ${}_2F_1(a_R + ia_I, a_R - ia_I, c, x)/\Gamma(c)$ for |x| < 1.

These routines compute the hypergeometric function ${}_2F_0(a,b,x)$. The series representation is a divergent hypergeometric series. However, for x < 0 we have ${}_2F_0(a,b,x) = (-1/x)^a U(a, 1+a-b, -1/x)$

7.22 Laguerre Functions

The generalized Laguerre polynomials are defined in terms of confluent hypergeometric functions as $L_n^a(x) = ((a+1)_n/n!)_1 F_1(-n,a+1,x)$, and are sometimes referred to as the associated Laguerre polynomials. They are related to the plain Laguerre polynomials $L_n(x)$ by $L_n^0(x) = L_n(x)$ and $L_n^k(x) = (-1)^k (d^k/dx^k) L_{(n+k)}(x)$. For more information see Abramowitz & Stegun, Chapter 22.

The functions described in this section are declared in the header file gsl_sf_laguerre.h.

```
double gsl_sf_laguerre_1 (double a, double x) [Function] double gsl_sf_laguerre_2 (double a, double x) [Function] double gsl_sf_laguerre_3 (double a, double x) [Function] int gsl_sf_laguerre_1_e (double a, double x, gsl_sf_result * result) [Function] int gsl_sf_laguerre_2_e (double a, double x, gsl_sf_result * result) [Function] int gsl_sf_laguerre_3_e (double a, double x, gsl_sf_result * result) [Function] These routines evaluate the generalized Laguerre polynomials L_1^a(x), L_2^a(x), L_3^a(x) using explicit representations.
```

These routines evaluate the generalized Laguerre polynomials $L_n^a(x)$ for a > -1, $n \ge 0$.

7.23 Lambert W Functions

Lambert's W functions, W(x), are defined to be solutions of the equation $W(x) \exp(W(x)) = x$. This function has multiple branches for x < 0; however, it has only two real-valued branches. We define $W_0(x)$ to be the principal branch, where W > -1 for x < 0, and $W_{-1}(x)$ to be the other real branch, where W < -1 for x < 0. The Lambert functions are declared in the header file gsl_sf_lambert.h.

7.24 Legendre Functions and Spherical Harmonics

The Legendre Functions and Legendre Polynomials are described in Abramowitz & Stegun, Chapter 8. These functions are declared in the header file gsl_sf_legendre.h.

7.24.1 Legendre Polynomials

```
double gsl_sf_legendre_P1 (double x)
                                                                           [Function]
double gsl_sf_legendre_P2 (double x)
                                                                           [Function]
double gsl_sf_legendre_P3 (double x)
                                                                           [Function]
int gsl_sf_legendre_P1_e (double x, gsl_sf_result * result)
                                                                           [Function]
int gsl_sf_legendre_P2_e (double x, gsl_sf_result * result)
                                                                           [Function]
int gsl_sf_legendre_P3_e (double x, gsl_sf_result * result)
                                                                           [Function]
     These functions evaluate the Legendre polynomials P_l(x) using explicit representa-
     tions for l = 1, 2, 3.
double gsl_sf_legendre_Pl (int 1, double x)
                                                                           [Function]
int gsl_sf_legendre_Pl_e (int 1, double x, gsl_sf_result * result)
                                                                           [Function]
     These functions evaluate the Legendre polynomial P_l(x) for a specific value of l, x
     subject to l \geq 0, |x| \leq 1
int gsl_sf_legendre_Pl_array (int lmax, double x, double
                                                                           [Function]
         result_array[])
int gsl_sf_legendre_Pl_deriv_array (int lmax, double x, double
                                                                           [Function]
         result_array[], double result_deriv_array[])
     These functions compute arrays of Legendre polynomials P_l(x) and derivatives
     dP_l(x)/dx, for l=0,\ldots,lmax, |x|\leq 1
double gsl_sf_legendre_Q0 (double x)
                                                                           [Function]
int gsl_sf_legendre_Q0_e (double x, gsl_sf_result * result)
                                                                           [Function]
     These routines compute the Legendre function Q_0(x) for x > -1, x \neq 1.
double gsl_sf_legendre_Q1 (double x)
                                                                           [Function]
int gsl_sf_legendre_Q1_e (double x, gsl_sf_result * result)
                                                                           [Function]
     These routines compute the Legendre function Q_1(x) for x > -1, x \neq 1.
double gsl_sf_legendre_Ql (int 1, double x)
                                                                           [Function]
int gsl_sf_legendre_Ql_e (int 1, double x, gsl_sf_result * result)
                                                                           [Function]
```

These routines compute the Legendre function $Q_l(x)$ for x > -1, $x \ne 1$ and l > 0.

7.24.2 Associated Legendre Polynomials and Spherical Harmonics

The following functions compute the associated Legendre Polynomials $P_l^m(x)$. Note that this function grows combinatorially with l and can overflow for l larger than about 150. There is no trouble for small m, but overflow occurs when m and l are both large. Rather than allow overflows, these functions refuse to calculate $P_l^m(x)$ and return GSL_EOVRFLW when they can sense that l and m are too big.

If you want to calculate a spherical harmonic, then *do not* use these functions. Instead use gsl_sf_legendre_sphPlm below, which uses a similar recursion, but with the normalized functions.

These routines compute the associated Legendre polynomial $P_l^m(x)$ for $m \ge 0$, $l \ge m$, $|x| \le 1$.

These functions compute arrays of Legendre polynomials $P_l^m(x)$ and derivatives $dP_l^m(x)/dx$, for $m \ge 0$, $l = |m|, \ldots, lmax$, $|x| \le 1$.

These routines compute the normalized associated Legendre polynomial $\sqrt{(2l+1)/(4\pi)}\sqrt{(l-m)!/(l+m)!}P_l^m(x)$ suitable for use in spherical harmonics. The parameters must satisfy $m \geq 0$, $l \geq m$, $|x| \leq 1$. Theses routines avoid the overflows that occur for the standard normalization of $P_l^m(x)$.

These functions compute arrays of normalized associated Legendre functions $\sqrt{(2l+1)/(4\pi)}\sqrt{(l-m)!/(l+m)!}P_l^m(x)$, and derivatives, for $m\geq 0$, $l=|m|,\ldots,lmax,\,|x|\leq 1$

int gsl_sf_legendre_array_size (const int lmax, const int m) [Function] This function returns the size of result_array[] needed for the array versions of $P_l^m(x)$, lmax - m + 1. An inline version of this function is used when HAVE_INLINE is defined.

7.24.3 Conical Functions

The Conical Functions $P^{\mu}_{-(1/2)+i\lambda}(x)$ and $Q^{\mu}_{-(1/2)+i\lambda}$ are described in Abramowitz & Stegun, Section 8.12.

These routines compute the irregular Spherical Conical Function $P_{-1/2+i\lambda}^{1/2}(x)$ for x > -1.

These routines compute the regular Spherical Conical Function $P_{-1/2+i\lambda}^{-1/2}(x)$ for x > -1.

These routines compute the conical function $P_{-1/2+i\lambda}^0(x)$ for x > -1.

These routines compute the conical function $P^1_{-1/2+i\lambda}(x)$ for x > -1.

These routines compute the Regular Spherical Conical Function $P_{-1/2+i\lambda}^{-1/2-l}(x)$ for x > -1, $l \ge -1$.

These routines compute the Regular Cylindrical Conical Function $P_{-1/2+i\lambda}^{-m}(x)$ for x > -1, $m \ge -1$.

7.24.4 Radial Functions for Hyperbolic Space

The following spherical functions are specializations of Legendre functions which give the regular eigenfunctions of the Laplacian on a 3-dimensional hyperbolic space H3d. Of particular interest is the flat limit, $\lambda \to \infty$, $\eta \to 0$, $\lambda \eta$ fixed.

double gsl_sf_legendre_H3d_0 (double lambda, double eta) [Function] int gsl_sf_legendre_H3d_0_e (double lambda, double eta, gsl_sf_result * result)

These routines compute the zeroth radial eigenfunction of the Laplacian on the 3dimensional hyperbolic space,

$$L_0^{H3d}(\lambda, \eta) := \frac{\sin(\lambda \eta)}{\lambda \sinh(\eta)}$$

for $\eta \geq 0$. In the flat limit this takes the form $L_0^{H3d}(\lambda, \eta) = j_0(\lambda \eta)$.

These routines compute the first radial eigenfunction of the Laplacian on the 3-dimensional hyperbolic space,

$$L_1^{H3d}(\lambda, \eta) := \frac{1}{\sqrt{\lambda^2 + 1}} \left(\frac{\sin(\lambda \eta)}{\lambda \sinh(\eta)} \right) \left(\coth(\eta) - \lambda \cot(\lambda \eta) \right)$$

for $\eta \geq 0$. In the flat limit this takes the form $L_1^{H3d}(\lambda, \eta) = j_1(\lambda \eta)$.

These routines compute the *l*-th radial eigenfunction of the Laplacian on the 3-dimensional hyperbolic space $\eta \geq 0$, $l \geq 0$. In the flat limit this takes the form $L_l^{H3d}(\lambda, \eta) = j_l(\lambda \eta)$.

This function computes an array of radial eigenfunctions $L_l^{H3d}(\lambda, \eta)$ for $0 \le l \le lmax$.

7.25 Logarithm and Related Functions

Information on the properties of the Logarithm function can be found in Abramowitz & Stegun, Chapter 4. The functions described in this section are declared in the header file gsl_sf_log.h.

```
double gsl_sf_log (double x) [Function] int gsl_sf_log_e (double x, gsl_sf_result * result) [Function] These routines compute the logarithm of x, log(x), for x > 0.
```

```
double gsl_sf_log_abs (double x) [Function] int gsl_sf_log_abs_e (double x, gsl_sf_result * result) [Function] These routines compute the logarithm of the magnitude of x, \log(|x|), for x \neq 0.
```

This routine computes the complex logarithm of $z = z_r + iz_i$. The results are returned as lnr, theta such that $exp(lnr + i\theta) = z_r + iz_i$, where θ lies in the range $[-\pi, \pi]$.

```
double gsl_sf_log_1plusx (double x) [Function] int gsl_sf_log_1plusx_e (double x, gsl_sf_result * result) [Function] These routines compute \log(1+x) for x > -1 using an algorithm that is accurate for small x.
```

```
double gsl_sf_log_1plusx_mx (double x) [Function] int gsl_sf_log_1plusx_mx_e (double x, gsl_sf_result * result) [Function] These routines compute \log(1+x)-x for x>-1 using an algorithm that is accurate for small x.
```

7.26 Mathieu Functions

The routines described in this section compute the angular and radial Mathieu functions, and their characteristic values. Mathieu functions are the solutions of the following two differential equations:

$$\frac{d^2y}{dv^2} + (a - 2q\cos 2v)y = 0, \frac{d^2f}{du^2} - (a - 2q\cosh 2u)f = 0.$$

The angular Mathieu functions $ce_r(x,q)$, $se_r(x,q)$ are the even and odd periodic solutions of the first equation, which is known as Mathieu's equation. These exist only for the discrete sequence of characteristic values $a = a_r(q)$ (even-periodic) and $a = b_r(q)$ (odd-periodic).

The radial Mathieu functions $Mc_r^{(j)}(z,q)$, $Ms_r^{(j)}(z,q)$ are the solutions of the second equation, which is referred to as Mathieu's modified equation. The radial Mathieu functions of the first, second, third and fourth kind are denoted by the parameter j, which takes the value 1, 2, 3 or 4.

For more information on the Mathieu functions, see Abramowitz and Stegun, Chapter 20. These functions are defined in the header file gsl_sf_mathieu.h.

7.26.1 Mathieu Function Workspace

The Mathieu functions can be computed for a single order or for multiple orders, using array-based routines. The array-based routines require a preallocated workspace.

$$gsl_sf_mathieu_workspace * gsl_sf_mathieu_alloc (size_t n, double qmax)$$
 [Function]

This function returns a workspace for the array versions of the Mathieu routines. The arguments n and qmax specify the maximum order and q-value of Mathieu functions which can be computed with this workspace.

void gsl_sf_mathieu_free (gsl_sf_mathieu_workspace * work) [Function]
This function frees the workspace work.

7.26.2 Mathieu Function Characteristic Values

int gsl_sf_mathieu_a (int n, double q, gsl_sf_result * result) [Function] int gsl_sf_mathieu_b (int n, double q, gsl_sf_result * result) [Function] These routines compute the characteristic values $a_n(q)$, $b_n(q)$ of the Mathieu functions $ce_n(q,x)$ and $se_n(q,x)$, respectively.

These routines compute a series of Mathieu characteristic values $a_n(q)$, $b_n(q)$ for n from order_min to order_max inclusive, storing the results in the array result_array.

7.26.3 Angular Mathieu Functions

These routines compute the angular Mathieu functions $ce_n(q, x)$ and $se_n(q, x)$, respectively.

- int gsl_sf_mathieu_se_array (int nmin, int nmax, double q, double x, gsl_sf_mathieu_workspace * work, double result_array[]) [Function]

These routines compute a series of the angular Mathieu functions $ce_n(q, x)$ and $se_n(q, x)$ of order n from nmin to nmax inclusive, storing the results in the array $result_array$.

7.26.4 Radial Mathieu Functions

- int $gsl_sf_mathieu_Mc$ (int j, int n, double q, double x, $gsl_sf_result *$ [Function] result)
- int $gsl_sf_mathieu_Ms$ (int j, int n, double q, double x, gsl_sf_result * [Function] result)

These routines compute the radial j-th kind Mathieu functions $Mc_n^{(j)}(q,x)$ and $Ms_n^{(j)}(q,x)$ of order n.

The allowed values of j are 1 and 2. The functions for j=3,4 can be computed as $M_n^{(3)}=M_n^{(1)}+iM_n^{(2)}$ and $M_n^{(4)}=M_n^{(1)}-iM_n^{(2)}$, where $M_n^{(j)}=Mc_n^{(j)}$ or $Ms_n^{(j)}$.

- int gsl_sf_mathieu_Ms_array (int j, int nmin, int nmax, double q, double x, gsl_sf_mathieu_workspace * work, double result_array[])

These routines compute a series of the radial Mathieu functions of kind j, with order from nmin to nmax inclusive, storing the results in the array $result_array$.

7.27 Power Function

The following functions are equivalent to the function gsl_pow_int (see Section 4.4 [Small integer powers], page 18) with an error estimate. These functions are declared in the header file gsl_sf_pow_int.h.

These routines compute the power x^n for integer n. The power is computed using the minimum number of multiplications. For example, x^8 is computed as $((x^2)^2)^2$, requiring only 3 multiplications. For reasons of efficiency, these functions do not check for overflow or underflow conditions.

#include <gsl/gsl_sf_pow_int.h>

7.28 Psi (Digamma) Function

The polygamma functions of order n are defined by

$$\psi^{(n)}(x) = \left(\frac{d}{dx}\right)^n \psi(x) = \left(\frac{d}{dx}\right)^{n+1} \log(\Gamma(x))$$

where $\psi(x) = \Gamma'(x)/\Gamma(x)$ is known as the digamma function. These functions are declared in the header file gsl_sf_psi.h.

7.28.1 Digamma Function

double gsl_sf_psi_int (int n) [Function] int gsl_sf_psi_int_e (int n, gsl_sf_result * result) [Function] These routines compute the digamma function $\psi(n)$ for positive integer n. The digamma function is also called the Psi function.

double gsl_sf_psi (double x) [Function] int gsl_sf_psi_e (double x, gsl_sf_result * result) [Function]

These routines compute the digramma function $\psi(x)$ for general $x, x \neq 0$

These routines compute the digamma function $\psi(x)$ for general $x, x \neq 0$.

double gsl_sf_psi_1piy (double y) [Function] int gsl_sf_psi_1piy_e (double y, gsl_sf_result * result) [Function]

These routines compute the real part of the digamma function on the line 1 + iu

These routines compute the real part of the digamma function on the line 1 + iy, $\text{Re}[\psi(1+iy)]$.

7.28.2 Trigamma Function

double gsl_sf_psi_1_int (int n) [Function] int gsl_sf_psi_1_int_e (int n, gsl_sf_result * result) [Function] These routines compute the Trigamma function $\psi'(n)$ for positive integer n.

These routines compute the Trigamma function $\psi'(x)$ for general x.

7.28.3 Polygamma Function

double gsl_sf_psi_n (int n, double x) [Function] int gsl_sf_psi_n_e (int n, double x, gsl_sf_result * result) [Function] These routines compute the polygamma function $\psi^{(n)}(x)$ for $n \geq 0$, x > 0.

7.29 Synchrotron Functions

The functions described in this section are declared in the header file gsl_sf_synchrotron.h.

double gsl_sf_synchrotron_1 (double x) [Function] int gsl_sf_synchrotron_1_e (double x, gsl_sf_result * result) [Function] These routines compute the first synchrotron function $x \int_x^\infty dt K_{5/3}(t)$ for $x \ge 0$.

```
double gsl_sf_synchrotron_2 (double x) [Function] int gsl_sf_synchrotron_2_e (double x, gsl_sf_result * result) [Function] These routines compute the second synchrotron function xK_{2/3}(x) for x \ge 0.
```

7.30 Transport Functions

The transport functions J(n,x) are defined by the integral representations $J(n,x) := \int_0^x dt \, t^n e^t / (e^t - 1)^2$. They are declared in the header file gsl_sf_transport.h.

```
double gsl_sf_transport_2 (double x)
                                                                          [Function]
int gsl_sf_transport_2_e (double x, gsl_sf_result * result)
                                                                         [Function]
     These routines compute the transport function J(2,x).
double gsl_sf_transport_3 (double x)
                                                                          [Function]
int gsl_sf_transport_3_e (double x, gsl_sf_result * result)
                                                                         [Function]
     These routines compute the transport function J(3, x).
double gsl_sf_transport_4 (double x)
                                                                         [Function]
int gsl_sf_transport_4_e (double x, gsl_sf_result * result)
                                                                         [Function]
     These routines compute the transport function J(4,x).
double gsl_sf_transport_5 (double x)
                                                                         [Function]
int gsl_sf_transport_5_e (double x, gsl_sf_result * result)
                                                                         [Function]
```

7.31 Trigonometric Functions

The library includes its own trigonometric functions in order to provide consistency across platforms and reliable error estimates. These functions are declared in the header file gsl_sf_trig.h.

7.31.1 Circular Trigonometric Functions

These routines compute the transport function J(5, x).

```
double gsl_sf_sin (double x)
                                                                               [Function]
int gsl_sf_sin_e (double x, gsl_sf_result * result)
                                                                               [Function]
      These routines compute the sine function \sin(x).
double gsl_sf_cos (double x)
                                                                               [Function]
int gsl_sf_cos_e (double x, gsl_sf_result * result)
                                                                               [Function]
      These routines compute the cosine function cos(x).
double gsl_sf_hypot (double x, double y)
                                                                               [Function]
int gsl_sf_hypot_e (double x, double y, gsl_sf_result * result)
                                                                               [Function]
      These routines compute the hypotenuse function \sqrt{x^2+y^2} avoiding overflow and un-
      derflow.
double gsl_sf_sinc (double x)
                                                                               [Function]
int gsl_sf_sinc_e (double x, gsl_sf_result * result)
                                                                               [Function]
      These routines compute \operatorname{sinc}(x) = \sin(\pi x)/(\pi x) for any value of x.
```

7.31.2 Trigonometric Functions for Complex Arguments

This function computes the complex sine, $\sin(z_r + iz_i)$ storing the real and imaginary parts in szr, szi.

This function computes the complex cosine, $\cos(z_r+iz_i)$ storing the real and imaginary parts in czr, czi.

This function computes the logarithm of the complex sine, $\log(\sin(z_r + iz_i))$ storing the real and imaginary parts in lszr, lszi.

7.31.3 Hyperbolic Trigonometric Functions

double gsl_sf_lnsinh (double x) [Function] int gsl_sf_lnsinh_e (double x, gsl_sf_result * result) [Function] These routines compute $\log(\sinh(x))$ for x > 0.

7.31.4 Conversion Functions

This function converts the polar coordinates (r,theta) to rectilinear coordinates (x,y), $x = r\cos(\theta), y = r\sin(\theta)$.

This function converts the rectilinear coordinates (x,y) to polar coordinates (r,theta), such that $x = r\cos(\theta)$, $y = r\sin(\theta)$. The argument theta lies in the range $[-\pi, \pi]$.

7.31.5 Restriction Functions

double gsl_sf_angle_restrict_symm (double theta) [Function] int gsl_sf_angle_restrict_symm_e (double * theta) [Function] These routines force the angle theta to lie in the range $(-\pi, \pi]$.

Note that the mathematical value of π is slightly greater than M_PI, so the machine numbers M_PI and -M_PI are included in the range.

double gsl_sf_angle_restrict_pos (double theta) [Function] int gsl_sf_angle_restrict_pos_e (double * theta) [Function] These routines force the angle theta to lie in the range $[0, 2\pi)$.

Note that the mathematical value of 2π is slightly greater than $2*M_PI$, so the machine number $2*M_PI$ is included in the range.

7.31.6 Trigonometric Functions With Error Estimates

int gsl_sf_sin_err_e (double x, double dx, gsl_sf_result * result) [Function] This routine computes the sine of an angle x with an associated absolute error dx, $\sin(x \pm dx)$. Note that this function is provided in the error-handling form only since its purpose is to compute the propagated error.

int gsl_sf_cos_err_e (double x, double dx, gsl_sf_result * result) [Function] This routine computes the cosine of an angle x with an associated absolute error dx, $\cos(x \pm dx)$. Note that this function is provided in the error-handling form only since its purpose is to compute the propagated error.

7.32 Zeta Functions

The Riemann zeta function is defined in Abramowitz & Stegun, Section 23.2. The functions described in this section are declared in the header file gsl_sf_zeta.h.

7.32.1 Riemann Zeta Function

The Riemann zeta function is defined by the infinite sum $\zeta(s) = \sum_{k=1}^{\infty} k^{-s}$.

```
double gsl_sf_zeta_int (int n) [Function] int gsl_sf_zeta_int_e (int n, gsl_sf_result * result) [Function] These routines compute the Riemann zeta function \zeta(n) for integer n, n \neq 1.
```

```
double gsl_sf_zeta (double s) [Function] int gsl_sf_zeta_e (double s, gsl_sf_result * result) [Function] These routines compute the Riemann zeta function \zeta(s) for arbitrary s, s \neq 1.
```

7.32.2 Riemann Zeta Function Minus One

For large positive argument, the Riemann zeta function approaches one. In this region the fractional part is interesting, and therefore we need a function to evaluate it explicitly.

```
double gsl_sf_zetam1_int (int n) [Function] int gsl_sf_zetam1_int_e (int n, gsl_sf_result * result) [Function] These routines compute \zeta(n) - 1 for integer n, n \neq 1.

double gsl_sf_zetam1 (double s) [Function] int gsl_sf_zetam1_e (double s, gsl_sf_result * result) [Function] These routines compute \zeta(s) - 1 for arbitrary s, s \neq 1.
```

7.32.3 Hurwitz Zeta Function

The Hurwitz zeta function is defined by $\zeta(s,q) = \sum_{0}^{\infty} (k+q)^{-s}$.

```
double gsl_sf_hzeta (double s, double q) [Function] int gsl_sf_hzeta_e (double s, double q, gsl_sf_result * result) [Function] These routines compute the Hurwitz zeta function \zeta(s,q) for s>1, q>0.
```

7.32.4 Eta Function

```
The eta function is defined by \eta(s) = (1 - 2^{1-s})\zeta(s).
double gsl_sf_eta_int (int n)
                                                                                [Function]
int gsl_sf_eta_int_e (int n, gsl_sf_result * result)
                                                                                [Function]
      These routines compute the eta function \eta(n) for integer n.
double gsl_sf_eta (double s)
                                                                                [Function]
int gsl_sf_eta_e (double s, gsl_sf_result * result)
                                                                                [Function]
      These routines compute the eta function \eta(s) for arbitrary s.
```

7.33 Examples

The following example demonstrates the use of the error handling form of the special functions, in this case to compute the Bessel function $J_0(5.0)$,

```
#include <stdio.h>
     #include <gsl/gsl_errno.h>
     #include <gsl/gsl_sf_bessel.h>
     int
     main (void)
       double x = 5.0;
       gsl_sf_result result;
       double expected = -0.17759677131433830434739701;
       int status = gsl_sf_bessel_J0_e (x, &result);
       printf ("status = %s\n", gsl_strerror(status));
       printf ("J0(5.0) = %.18f\n"
                      +/- \% .18f\n'',
               result.val, result.err);
       printf ("exact = %.18f\n", expected);
       return status;
Here are the results of running the program,
     $ ./a.out
```

```
status = success
J0(5.0) = -0.177596771314338292
     +/- 0.00000000000000193
       = -0.177596771314338292
```

The next program computes the same quantity using the natural form of the function. In this case the error term result.err and return status are not accessible.

```
#include <stdio.h>
#include <gsl/gsl_sf_bessel.h>
```

```
int
main (void)
{
    double x = 5.0;
    double expected = -0.17759677131433830434739701;

    double y = gsl_sf_bessel_J0 (x);

    printf ("J0(5.0) = %.18f\n", y);
    printf ("exact = %.18f\n", expected);
    return 0;
}

The results of the function are the same,
    $ ./a.out
    J0(5.0) = -0.177596771314338292
    exact = -0.177596771314338292
```

7.34 References and Further Reading

The library follows the conventions of Abramowitz & Stegun where possible,

Abramowitz & Stegun (eds.), Handbook of Mathematical Functions

The following papers contain information on the algorithms used to compute the special functions,

- Allan J. MacLeod, MISCFUN: A software package to compute uncommon special functions. ACM Trans. Math. Soft., vol. 22, 1996, 288–301
- G.N. Watson, A Treatise on the Theory of Bessel Functions, 2nd Edition (Cambridge University Press, 1944).
- G. Nemeth, Mathematical Approximations of Special Functions, Nova Science Publishers, ISBN 1-56072-052-2
- B.C. Carlson, Special Functions of Applied Mathematics (1977)
- N. M. Temme, Special Functions: An Introduction to the Classical Functions of Mathematical Physics (1996), ISBN 978-0471113133.
- W.J. Thompson, Atlas for Computing Mathematical Functions, John Wiley & Sons, New York (1997).
- Y.Y. Luke, Algorithms for the Computation of Mathematical Functions, Academic Press, New York (1977).

8 Vectors and Matrices

The functions described in this chapter provide a simple vector and matrix interface to ordinary C arrays. The memory management of these arrays is implemented using a single underlying type, known as a block. By writing your functions in terms of vectors and matrices you can pass a single structure containing both data and dimensions as an argument without needing additional function parameters. The structures are compatible with the vector and matrix formats used by BLAS routines.

8.1 Data types

All the functions are available for each of the standard data-types. The versions for double have the prefix gsl_block, gsl_vector and gsl_matrix. Similarly the versions for single-precision float arrays have the prefix gsl_block_float, gsl_vector_float and gsl_matrix_float. The full list of available types is given below,

```
gsl_block
                                 double
gsl_block_float
                                 float
gsl_block_long_double
                                 long double
gsl_block_int
                                 int
gsl_block_uint
                                 unsigned int
gsl_block_long
                                 long
gsl_block_ulong
                                 unsigned long
gsl_block_short
                                 short
gsl_block_ushort
                                 unsigned short
gsl_block_char
                                 char
gsl_block_uchar
                                 unsigned char
gsl_block_complex
                                 complex double
gsl_block_complex_float
                                 complex float
gsl_block_complex_long_double
                                 complex long double
```

Corresponding types exist for the gsl_vector and gsl_matrix functions.

8.2 Blocks

For consistency all memory is allocated through a gsl_block structure. The structure contains two components, the size of an area of memory and a pointer to the memory. The gsl_block structure looks like this,

```
typedef struct
{
   size_t size;
   double * data;
} gsl_block;
```

Vectors and matrices are made by *slicing* an underlying block. A slice is a set of elements formed from an initial offset and a combination of indices and step-sizes. In the case of a matrix the step-size for the column index represents the row-length. The step-size for a vector is known as the *stride*.

The functions for allocating and deallocating blocks are defined in gsl_block.h

8.2.1 Block allocation

The functions for allocating memory to a block follow the style of malloc and free. In addition they also perform their own error checking. If there is insufficient memory available to allocate a block then the functions call the GSL error handler (with an error number of GSL_ENOMEM) in addition to returning a null pointer. Thus if you use the library error handler to abort your program then it isn't necessary to check every alloc.

gsl_block * gsl_block_alloc (size_t n)

[Function]

This function allocates memory for a block of *n* double-precision elements, returning a pointer to the block struct. The block is not initialized and so the values of its elements are undefined. Use the function gsl_block_calloc if you want to ensure that all the elements are initialized to zero.

A null pointer is returned if insufficient memory is available to create the block.

gsl_block * gsl_block_calloc (size_t n)

[Function]

This function allocates memory for a block and initializes all the elements of the block to zero.

void gsl_block_free (gsl_block * b)

[Function]

This function frees the memory used by a block b previously allocated with gsl_block_alloc or gsl_block_calloc.

8.2.2 Reading and writing blocks

The library provides functions for reading and writing blocks to a file as binary data or formatted text.

int gsl_block_fwrite (FILE * stream, const gsl_block * b)

[Function]

This function writes the elements of the block b to the stream stream in binary format. The return value is 0 for success and $GSL_EFAILED$ if there was a problem writing to the file. Since the data is written in the native binary format it may not be portable between different architectures.

int gsl_block_fread (FILE * stream, gsl_block * b)

[Function]

This function reads into the block b from the open stream stream in binary format. The block b must be preallocated with the correct length since the function uses the size of b to determine how many bytes to read. The return value is 0 for success and GSL_EFAILED if there was a problem reading from the file. The data is assumed to have been written in the native binary format on the same architecture.

This function writes the elements of the block b line-by-line to the stream stream using the format specifier format, which should be one of the %g, %e or %f formats for floating point numbers and %d for integers. The function returns 0 for success and GSL_EFAILED if there was a problem writing to the file.

int gsl_block_fscanf (FILE * stream, gsl_block * b)

[Function]

This function reads formatted data from the stream stream into the block b. The block b must be preallocated with the correct length since the function uses the size

of b to determine how many numbers to read. The function returns 0 for success and $GSL_EFAILED$ if there was a problem reading from the file.

8.2.3 Example programs for blocks

The following program shows how to allocate a block,

```
#include <stdio.h>
#include <gsl/gsl_block.h>

int
main (void)
{
    gsl_block * b = gsl_block_alloc (100);

    printf ("length of block = %u\n", b->size);
    printf ("block data address = %#x\n", b->data);

    gsl_block_free (b);
    return 0;
}

Here is the output from the program,
    length of block = 100
    block data address = 0x804b0d8
```

8.3 Vectors

Vectors are defined by a gsl_vector structure which describes a slice of a block. Different vectors can be created which point to the same block. A vector slice is a set of equally-spaced elements of an area of memory.

The gsl_vector structure contains five components, the *size*, the *stride*, a pointer to the memory where the elements are stored, *data*, a pointer to the block owned by the vector, *block*, if any, and an ownership flag, *owner*. The structure is very simple and looks like this,

```
typedef struct
{
   size_t size;
   size_t stride;
   double * data;
   gsl_block * block;
   int owner;
} gsl_vector;
```

The size is simply the number of vector elements. The range of valid indices runs from 0 to size-1. The stride is the step-size from one element to the next in physical memory, measured in units of the appropriate datatype. The pointer data gives the location of the first element of the vector in memory. The pointer block stores the location of the memory block in which the vector elements are located (if any). If the vector owns this block then the owner field is set to one and the block will be deallocated when the vector is freed. If

the vector points to a block owned by another object then the *owner* field is zero and any underlying block will not be deallocated with the vector.

The functions for allocating and accessing vectors are defined in gsl_vector.h

8.3.1 Vector allocation

The functions for allocating memory to a vector follow the style of malloc and free. In addition they also perform their own error checking. If there is insufficient memory available to allocate a vector then the functions call the GSL error handler (with an error number of GSL_ENOMEM) in addition to returning a null pointer. Thus if you use the library error handler to abort your program then it isn't necessary to check every alloc.

gsl_vector * gsl_vector_alloc (size_t n)

[Function]

This function creates a vector of length n, returning a pointer to a newly initialized vector struct. A new block is allocated for the elements of the vector, and stored in the *block* component of the vector struct. The block is "owned" by the vector, and will be deallocated when the vector is deallocated.

gsl_vector * gsl_vector_calloc (size_t n)

[Function]

This function allocates memory for a vector of length n and initializes all the elements of the vector to zero.

void gsl_vector_free (gsl_vector * v)

[Function]

This function frees a previously allocated vector v. If the vector was created using gsl_vector_alloc then the block underlying the vector will also be deallocated. If the vector has been created from another object then the memory is still owned by that object and will not be deallocated.

8.3.2 Accessing vector elements

Unlike FORTRAN compilers, C compilers do not usually provide support for range checking of vectors and matrices.¹ The functions gsl_vector_get and gsl_vector_set can perform portable range checking for you and report an error if you attempt to access elements outside the allowed range.

The functions for accessing the elements of a vector or matrix are defined in gsl_vector.h and declared extern inline to eliminate function-call overhead. You must compile your program with the preprocessor macro HAVE_INLINE defined to use these functions.

If necessary you can turn off range checking completely without modifying any source files by recompiling your program with the preprocessor definition GSL_RANGE_CHECK_OFF. Provided your compiler supports inline functions the effect of turning off range checking is to replace calls to gsl_vector_get(v,i) by v->data[i*v->stride] and calls to gsl_vector_set(v,i,x) by v->data[i*v->stride]=x. Thus there should be no performance penalty for using the range checking functions when range checking is turned off.

If you use a C99 compiler which requires inline functions in header files to be declared inline instead of extern inline, define the macro GSL_C99_INLINE (see Section 2.5 [Inline

Range checking is available in the GNU C Compiler bounds-checking extension, but it is not part of the default installation of GCC. Memory accesses can also be checked with Valgrind or the gcc -fmudflap memory protection option.

functions], page 6). With GCC this is selected automatically when compiling in C99 mode (-std=c99).

If inline functions are not used, calls to the functions gsl_vector_get and gsl_vector_set will link to the compiled versions of these functions in the library itself. The range checking in these functions is controlled by the global integer variable gsl_check_range. It is enabled by default—to disable it, set gsl_check_range to zero. Due to function-call overhead, there is less benefit in disabling range checking here than for inline functions.

double gsl_vector_get (const gsl_vector * v, size_t i) [Function]

This function returns the i-th element of a vector v. If i lies outside the allowed range

This function returns the *i*-th element of a vector v. If i lies outside the allowed range of 0 to n-1 then the error handler is invoked and 0 is returned. An inline version of this function is used when HAVE_INLINE is defined.

void gsl_vector_set (gsl_vector * v, size_t i, double x) [Function]

This function sets the value of the *i*-th element of a vector v to x. If i lies outside the allowed range of 0 to n-1 then the error handler is invoked. An inline version of this function is used when HAVE_INLINE is defined.

These functions return a pointer to the *i*-th element of a vector v. If i lies outside the allowed range of 0 to n-1 then the error handler is invoked and a null pointer is returned. Inline versions of these functions are used when HAVE_INLINE is defined.

8.3.3 Initializing vector elements

void $gsl_vector_set_all\ (gsl_vector * v, double x)$ [Function] This function sets all the elements of the vector v to the value x.

void gsl_vector_set_zero (gsl_vector * v) [Function]
This function sets all the elements of the vector v to zero.

int gsl_vector_set_basis (gsl_vector * v, size_t i) [Function]
This function makes a basis vector by setting all the elements of the vector v to zero

This function makes a basis vector by setting all the elements of the vector v to zero except for the i-th element which is set to one.

8.3.4 Reading and writing vectors

The library provides functions for reading and writing vectors to a file as binary data or formatted text.

int gsl_vector_fwrite (FILE * stream, const gsl_vector * v) [Function]

This function writes the elements of the vector v to the stream stream in binary format. The return value is 0 for success and GSL_EFAILED if there was a problem writing to the file. Since the data is written in the native binary format it may not be portable between different architectures.

int gsl_vector_fread (FILE * stream, gsl_vector * v) [Function]

This function reads into the vector v from the open stream in binary format.

The vector v must be preallocated with the correct length since the function uses the

size of v to determine how many bytes to read. The return value is 0 for success and $GSL_EFAILED$ if there was a problem reading from the file. The data is assumed to have been written in the native binary format on the same architecture.

This function writes the elements of the vector v line-by-line to the stream stream using the format specifier format, which should be one of the %g, %e or %f formats for floating point numbers and %d for integers. The function returns 0 for success and $GSL_EFAILED$ if there was a problem writing to the file.

int gsl_vector_fscanf (FILE * stream, gsl_vector * v) [Function]

This function reads formatted data from the stream stream into the vector v. The

vector v must be preallocated with the correct length since the function uses the size

vector v must be preallocated with the correct length since the function uses the size of v to determine how many numbers to read. The function returns 0 for success and $GSL_EFAILED$ if there was a problem reading from the file.

8.3.5 Vector views

In addition to creating vectors from slices of blocks it is also possible to slice vectors and create vector views. For example, a subvector of another vector can be described with a view, or two views can be made which provide access to the even and odd elements of a vector.

A vector view is a temporary object, stored on the stack, which can be used to operate on a subset of vector elements. Vector views can be defined for both constant and non-constant vectors, using separate types that preserve constness. A vector view has the type gsl_vector_view and a constant vector view has the type gsl_vector_const_view. In both cases the elements of the view can be accessed as a gsl_vector using the vector component of the view object. A pointer to a vector of type gsl_vector * or const gsl_vector * can be obtained by taking the address of this component with the & operator.

When using this pointer it is important to ensure that the view itself remains in scope—the simplest way to do so is by always writing the pointer as &view.vector, and never storing this value in another variable.

gsl_vector_const_view gsl_vector_const_subvector (const gsl_vector * v, size_t offset, size_t n) [Function]

These functions return a vector view of a subvector of another vector v. The start of the new vector is offset by offset elements from the start of the original vector. The new vector has n elements. Mathematically, the i-th element of the new vector v is given by,

v'(i) = v->data[(offset + i)*v->stride]

where the index i runs from 0 to n-1.

The data pointer of the returned vector struct is set to null if the combined parameters (offset,n) overrun the end of the original vector.

The new vector is only a view of the block underlying the original vector, v. The block containing the elements of v is not owned by the new vector. When the view goes

out of scope the original vector v and its block will continue to exist. The original memory can only be deallocated by freeing the original vector. Of course, the original vector should not be deallocated while the view is still in use.

The function gsl_vector_const_subvector is equivalent to gsl_vector_subvector but can be used for vectors which are declared const.

gsl_vector_const_subvector_with_stride (const gsl_vector * v, size_t
offset, size_t stride, size_t n)

These functions return a vector view of a subvector of another vector v with an additional stride argument. The subvector is formed in the same way as for $gsl_vector_subvector$ but the new vector has n elements with a step-size of stride from one element to the next in the original vector. Mathematically, the i-th element of the new vector v is given by,

```
v'(i) = v->data[(offset + i*stride)*v->stride]
```

where the index i runs from 0 to n-1.

Note that subvector views give direct access to the underlying elements of the original vector. For example, the following code will zero the even elements of the vector v of length n, while leaving the odd elements untouched,

A vector view can be passed to any subroutine which takes a vector argument just as a directly allocated vector would be, using &view.vector. For example, the following code computes the norm of the odd elements of v using the BLAS routine DNRM2,

The function gsl_vector_const_subvector_with_stride is equivalent to gsl_vector_subvector_with_stride but can be used for vectors which are declared const.

These functions return a vector view of the real parts of the complex vector v.

The function gsl_vector_complex_const_real is equivalent to gsl_vector_complex_real but can be used for vectors which are declared const.

```
gsl_vector_const_view gsl_vector_complex_const_imag (const gsl_vector_complex * v)
```

These functions return a vector view of the imaginary parts of the complex vector v.

The function gsl_vector_complex_const_imag is equivalent to gsl_vector_complex_imag but can be used for vectors which are declared const.

These functions return a vector view of an array. The start of the new vector is given by base and has n elements. Mathematically, the i-th element of the new vector v' is given by,

$$v'(i) = base[i]$$

where the index i runs from 0 to n-1.

The array containing the elements of v is not owned by the new vector view. When the view goes out of scope the original array will continue to exist. The original memory can only be deallocated by freeing the original pointer base. Of course, the original array should not be deallocated while the view is still in use.

The function gsl_vector_const_view_array is equivalent to gsl_vector_view_array but can be used for arrays which are declared const.

These functions return a vector view of an array base with an additional stride argument. The subvector is formed in the same way as for gsl_vector_view_array but the new vector has n elements with a step-size of stride from one element to the next in the original array. Mathematically, the i-th element of the new vector v' is given by,

```
v'(i) = base[i*stride]
```

where the index i runs from 0 to n-1.

Note that the view gives direct access to the underlying elements of the original array. A vector view can be passed to any subroutine which takes a vector argument just as a directly allocated vector would be, using &view.vector.

The function gsl_vector_const_view_array_with_stride is equivalent to gsl_vector_view_array_with_stride but can be used for arrays which are declared const.

8.3.6 Copying vectors

Common operations on vectors such as addition and multiplication are available in the BLAS part of the library (see Chapter 13 [BLAS Support], page 121). However, it is useful to have a small number of utility functions which do not require the full BLAS code. The following functions fall into this category.

```
int gsl_vector_memcpy (gsl_vector * dest, const gsl_vector * src) [Function]

This function copies the elements of the vector src into the vector dest. The two vectors must have the same length.
```

int gsl_vector_swap (gsl_vector * v, gsl_vector * w) [Function]

This function exchanges the elements of the vectors v and w by copying. The two vectors must have the same length.

8.3.7 Exchanging elements

The following function can be used to exchange, or permute, the elements of a vector.

- int gsl_vector_swap_elements (gsl_vector * v, size_t i, size_t j) [Function]
 This function exchanges the i-th and j-th elements of the vector v in-place.
- int gsl_vector_reverse (gsl_vector * v) [Function]

 This function reverses the order of the elements of the vector v.

8.3.8 Vector operations

- int gsl_vector_add (gsl_vector * a, const gsl_vector * b) [Function] This function adds the elements of vector b to the elements of vector a. The result $a_i \leftarrow a_i + b_i$ is stored in a and b remains unchanged. The two vectors must have the same length.
- int gsl_vector_sub (gsl_vector * a, const gsl_vector * b) [Function] This function subtracts the elements of vector b from the elements of vector a. The result $a_i \leftarrow a_i b_i$ is stored in a and b remains unchanged. The two vectors must have the same length.
- int gsl_vector_mul (gsl_vector * a, const gsl_vector * b) [Function] This function multiplies the elements of vector a by the elements of vector b. The result $a_i \leftarrow a_i * b_i$ is stored in a and b remains unchanged. The two vectors must have the same length.
- int gsl_vector_div (gsl_vector * a, const gsl_vector * b) [Function] This function divides the elements of vector a by the elements of vector b. The result $a_i \leftarrow a_i/b_i$ is stored in a and b remains unchanged. The two vectors must have the same length.
- int gsl_vector_scale (gsl_vector * a, const double x) [Function]

 This function multiplies the elements of vector a by the constant factor x. The result $a_i \leftarrow xa_i$ is stored in a.
- int gsl_vector_add_constant (gsl_vector * a, const double x) [Function]

 This function adds the constant value x to the elements of the vector a. The result $a_i \leftarrow a_i + x$ is stored in a.

8.3.9 Finding maximum and minimum elements of vectors

The following operations are only defined for real vectors.

double gsl_vector_max (const gsl_vector * v) [Function]
This function returns the maximum value in the vector v.

double gsl_vector_min (const gsl_vector * v) [Function]
This function returns the minimum value in the vector v.

This function returns the minimum and maximum values in the vector v, storing them in min_out and max_out .

```
size_t gsl_vector_max_index (const gsl_vector * v) [Function]
```

This function returns the index of the maximum value in the vector v. When there are several equal maximum elements then the lowest index is returned.

```
size_t gsl_vector_min_index (const gsl_vector * v) [Function]

This function returns the index of the minimum value in the vector v. When there are several equal minimum elements then the lowest index is returned.
```

This function returns the indices of the minimum and maximum values in the vector v, storing them in imin and imax. When there are several equal minimum or maximum elements then the lowest indices are returned.

8.3.10 Vector properties

The following functions are defined for real and complex vectors. For complex vectors both the real and imaginary parts must satisfy the conditions.

int gsl_vector_equal ($const \ gsl_vector * u$, $const \ gsl_vector * v$) [Function] This function returns 1 if the vectors u and v are equal (by comparison of element values) and 0 otherwise.

8.3.11 Example programs for vectors

This program shows how to allocate, initialize and read from a vector using the functions gsl_vector_alloc, gsl_vector_set and gsl_vector_get.

```
#include <stdio.h>
#include <gsl/gsl_vector.h>

int
main (void)
{
   int i;
   gsl_vector * v = gsl_vector_alloc (3);
   for (i = 0; i < 3; i++)
   {</pre>
```

Here is the output from the program. The final loop attempts to read outside the range of the vector v, and the error is trapped by the range-checking code in gsl_vector_get.

```
$ ./a.out
v_0 = 1.23
v_1 = 2.23
v_2 = 3.23
gsl: vector_source.c:12: ERROR: index out of range
Default GSL error handler invoked.
Aborted (core dumped)
```

The next program shows how to write a vector to a file.

```
#include <stdio.h>
#include <gsl/gsl_vector.h>
int
main (void)
{
  int i;
 gsl_vector * v = gsl_vector_alloc (100);
 for (i = 0; i < 100; i++)
      gsl\_vector\_set(v, i, 1.23 + i);
    }
  {
     FILE * f = fopen ("test.dat", "w");
     gsl_vector_fprintf (f, v, "%.5g");
     fclose (f);
  }
  gsl_vector_free (v);
 return 0;
}
```

After running this program the file test.dat should contain the elements of v, written using the format specifier %.5g. The vector could then be read back in using the function gsl_vector_fscanf (f, v) as follows:

```
#include <stdio.h>
#include <gsl/gsl_vector.h>
int
main (void)
  int i;
  gsl_vector * v = gsl_vector_alloc (10);
  {
     FILE * f = fopen ("test.dat", "r");
     gsl_vector_fscanf (f, v);
     fclose (f);
  }
  for (i = 0; i < 10; i++)
      printf ("%g\n", gsl_vector_get(v, i));
  gsl_vector_free (v);
  return 0;
}
```

8.4 Matrices

Matrices are defined by a gsl_matrix structure which describes a generalized slice of a block. Like a vector it represents a set of elements in an area of memory, but uses two indices instead of one.

The gsl_matrix structure contains six components, the two dimensions of the matrix, a physical dimension, a pointer to the memory where the elements of the matrix are stored, data, a pointer to the block owned by the matrix block, if any, and an ownership flag, owner. The physical dimension determines the memory layout and can differ from the matrix dimension to allow the use of submatrices. The gsl_matrix structure is very simple and looks like this,

```
typedef struct
{
    size_t size1;
    size_t size2;
    size_t tda;
    double * data;
    gsl_block * block;
    int owner;
} gsl_matrix;
```

Matrices are stored in row-major order, meaning that each row of elements forms a contiguous block in memory. This is the standard "C-language ordering" of two-dimensional arrays. Note that FORTRAN stores arrays in column-major order. The number of rows is size1. The range of valid row indices runs from 0 to size1-1. Similarly size2 is the number of columns. The range of valid column indices runs from 0 to size2-1. The physical row dimension tda, or trailing dimension, specifies the size of a row of the matrix as laid out in memory.

For example, in the following matrix size1 is 3, size2 is 4, and tda is 8. The physical memory layout of the matrix begins in the top left hand-corner and proceeds from left to right along each row in turn.

```
00 01 02 03 XX XX XX XX
10 11 12 13 XX XX XX XX
20 21 22 23 XX XX XX XX
```

Each unused memory location is represented by "XX". The pointer data gives the location of the first element of the matrix in memory. The pointer block stores the location of the memory block in which the elements of the matrix are located (if any). If the matrix owns this block then the owner field is set to one and the block will be deallocated when the matrix is freed. If the matrix is only a slice of a block owned by another object then the owner field is zero and any underlying block will not be freed.

The functions for allocating and accessing matrices are defined in gsl_matrix.h

8.4.1 Matrix allocation

The functions for allocating memory to a matrix follow the style of malloc and free. They also perform their own error checking. If there is insufficient memory available to allocate a matrix then the functions call the GSL error handler (with an error number of GSL_ENOMEM) in addition to returning a null pointer. Thus if you use the library error handler to abort your program then it isn't necessary to check every alloc.

```
gsl_matrix * gsl_matrix_alloc (size_t n1, size_t n2) [Function]

This function creates a matrix of size n1 rows by n2 columns, returning a pointer to a newly initialized matrix struct. A new block is allocated for the elements of the matrix, and stored in the block component of the matrix struct. The block is "owned" by the matrix, and will be deallocated when the matrix is deallocated.
```

```
gsl_matrix * gsl_matrix_calloc (size_t n1, size_t n2) [Function]

This function allocates memory for a matrix of size n1 rows by n2 columns and initializes all the elements of the matrix to zero.
```

```
void gsl_matrix_free (gsl_matrix * m) [Function]

This function frees a previously allocated matrix m. If the matrix was created using gsl_matrix_alloc then the block underlying the matrix will also be deallocated. If the matrix has been created from another object then the memory is still owned by that object and will not be deallocated.
```

8.4.2 Accessing matrix elements

The functions for accessing the elements of a matrix use the same range checking system as vectors. You can turn off range checking by recompiling your program with the preprocessor definition GSL_RANGE_CHECK_OFF.

The elements of the matrix are stored in "C-order", where the second index moves continuously through memory. More precisely, the element accessed by the function gsl_matrix_get(m,i,j) and gsl_matrix_set(m,i,j,x) is

```
m->data[i * m->tda + j]
```

where tda is the physical row-length of the matrix.

- double gsl_matrix_get (const gsl_matrix * m, size_t i, size_t j) [Function] This function returns the (i, j)-th element of a matrix m. If i or j lie outside the allowed range of 0 to n1 1 and 0 to n2 1 then the error handler is invoked and 0 is returned. An inline version of this function is used when HAVE_INLINE is defined.
- void gsl_matrix_set (gsl_matrix * m, size_t i, size_t j, double x) [Function] This function sets the value of the (i, j)-th element of a matrix m to x. If i or j lies outside the allowed range of 0 to n1 1 and 0 to n2 1 then the error handler is invoked. An inline version of this function is used when HAVE_INLINE is defined.

```
double * gsl_matrix_ptr (gsl_matrix * m, size_t i, size_t j) [Function] const double * gsl_matrix_const_ptr (const gsl_matrix * m, size_t [Function] i, size_t j)
```

These functions return a pointer to the (i, j)-th element of a matrix m. If i or j lie outside the allowed range of 0 to n1 - 1 and 0 to n2 - 1 then the error handler is invoked and a null pointer is returned. Inline versions of these functions are used when HAVE_INLINE is defined.

8.4.3 Initializing matrix elements

void $gsl_matrix_set_all\ (gsl_matrix * m, double x)$ [Function]
This function sets all the elements of the matrix m to the value x.

```
void gsl_matrix_set_zero (gsl_matrix * m) [Function]

This function sets all the elements of the matrix m to zero.
```

void gsl_matrix_set_identity ($gsl_matrix * m$) [Function] This function sets the elements of the matrix m to the corresponding elements of the identity matrix, $m(i,j) = \delta(i,j)$, i.e. a unit diagonal with all off-diagonal elements zero. This applies to both square and rectangular matrices.

8.4.4 Reading and writing matrices

The library provides functions for reading and writing matrices to a file as binary data or formatted text.

int gsl_matrix_fwrite (FILE * stream, const gsl_matrix * m) [Function]

This function writes the elements of the matrix m to the stream stream in binary format. The return value is 0 for success and GSL_EFAILED if there was a problem writing to the file. Since the data is written in the native binary format it may not be portable between different architectures.

int gsl_matrix_fread (FILE * stream, gsl_matrix * m) [Function]

This function reads into the matrix m from the open stream in binary format.

The matrix m must be preallocated with the correct dimensions since the function

uses the size of m to determine how many bytes to read. The return value is 0 for success and $GSL_EFAILED$ if there was a problem reading from the file. The data is assumed to have been written in the native binary format on the same architecture.

This function writes the elements of the matrix m line-by-line to the stream stream using the format specifier format, which should be one of the %g, %e or %f formats for floating point numbers and %d for integers. The function returns 0 for success and $GSL_EFAILED$ if there was a problem writing to the file.

int gsl_matrix_fscanf (FILE * stream, gsl_matrix * m) [Function]

This function reads formatted data from the stream stream into the matrix m. The matrix m must be preallocated with the correct dimensions since the function uses the size of m to determine how many numbers to read. The function returns 0 for success and GSL_EFAILED if there was a problem reading from the file.

8.4.5 Matrix views

A matrix view is a temporary object, stored on the stack, which can be used to operate on a subset of matrix elements. Matrix views can be defined for both constant and non-constant matrices using separate types that preserve constness. A matrix view has the type gsl_matrix_view and a constant matrix view has the type gsl_matrix_const_view. In both cases the elements of the view can by accessed using the matrix component of the view object. A pointer gsl_matrix * or const gsl_matrix * can be obtained by taking the address of the matrix component with the & operator. In addition to matrix views it is also possible to create vector views of a matrix, such as row or column views.

gsl_matrix_view gsl_matrix_submatrix (gsl_matrix * m, size_t k1, s

gsl_matrix_const_view gsl_matrix_const_submatrix (const gsl_matrix * m, size_t k1, size_t k2, size_t n1, size_t n2) [Function]

These functions return a matrix view of a submatrix of the matrix m. The upper-left element of the submatrix is the element (k1,k2) of the original matrix. The submatrix has n1 rows and n2 columns. The physical number of columns in memory given by tda is unchanged. Mathematically, the (i,j)-th element of the new matrix is given by,

$$m'(i,j) = m-data[(k1*m-da+k2) + i*m-da+j]$$

where the index i runs from 0 to n1-1 and the index j runs from 0 to n2-1.

The data pointer of the returned matrix struct is set to null if the combined parameters (i,j,n1,n2,tda) overrun the ends of the original matrix.

The new matrix view is only a view of the block underlying the existing matrix, m. The block containing the elements of m is not owned by the new matrix view. When the view goes out of scope the original matrix m and its block will continue to exist. The original memory can only be deallocated by freeing the original matrix. Of course, the original matrix should not be deallocated while the view is still in use.

The function gsl_matrix_const_submatrix is equivalent to gsl_matrix_submatrix but can be used for matrices which are declared const.

gsl_matrix_const_view gsl_matrix_const_view_array (const double * base, size_t n1, size_t n2) [Function]

These functions return a matrix view of the array base. The matrix has n1 rows and n2 columns. The physical number of columns in memory is also given by n2. Mathematically, the (i, j)-th element of the new matrix is given by,

$$m'(i,j) = base[i*n2 + j]$$

where the index i runs from 0 to n1-1 and the index j runs from 0 to n2-1.

The new matrix is only a view of the array base. When the view goes out of scope the original array base will continue to exist. The original memory can only be deallocated by freeing the original array. Of course, the original array should not be deallocated while the view is still in use.

The function gsl_matrix_const_view_array is equivalent to gsl_matrix_view_array but can be used for matrices which are declared const.

gsl_matrix_view gsl_matrix_view_array_with_tda (double * [Function] base, size_t n1, size_t n2, size_t tda)

gsl_matrix_const_view gsl_matrix_const_view_array_with_tda [Function] (const double * base, size_t n1, size_t n2, size_t tda)

These functions return a matrix view of the array base with a physical number of columns tda which may differ from the corresponding dimension of the matrix. The matrix has n1 rows and n2 columns, and the physical number of columns in memory is given by tda. Mathematically, the (i, j)-th element of the new matrix is given by,

$$m'(i,j) = base[i*tda + j]$$

where the index i runs from 0 to n1-1 and the index j runs from 0 to n2-1.

The new matrix is only a view of the array base. When the view goes out of scope the original array base will continue to exist. The original memory can only be deallocated by freeing the original array. Of course, the original array should not be deallocated while the view is still in use.

The function gsl_matrix_const_view_array_with_tda is equivalent to gsl_matrix_view_array_with_tda but can be used for matrices which are declared const.

gsl_matrix_const_view gsl_matrix_const_view_vector (const gsl_vector * v, size_t n1, size_t n2) [Function]

These functions return a matrix view of the vector v. The matrix has n1 rows and n2 columns. The vector must have unit stride. The physical number of columns in memory is also given by n2. Mathematically, the (i,j)-th element of the new matrix is given by,

$$m'(i,j) = v \rightarrow data[i*n2 + j]$$

where the index i runs from 0 to n1-1 and the index j runs from 0 to n2-1.

The new matrix is only a view of the vector v. When the view goes out of scope the original vector v will continue to exist. The original memory can only be deallocated by freeing the original vector. Of course, the original vector should not be deallocated while the view is still in use.

The function gsl_matrix_const_view_vector is equivalent to gsl_matrix_view_vector but can be used for matrices which are declared const.

```
gsl_matrix_view gsl_matrix_view_vector_with_tda (gsl_vector * [Function] v, size_t n1, size_t n2, size_t tda)
gsl_matrix_const_view [Function] gsl_matrix_const_view_vector_with_tda (const gsl_vector * v, size_t n1, size_t n2, size_t tda)
```

These functions return a matrix view of the vector v with a physical number of columns tda which may differ from the corresponding matrix dimension. The vector must have unit stride. The matrix has n1 rows and n2 columns, and the physical number of columns in memory is given by tda. Mathematically, the (i, j)-th element of the new matrix is given by,

```
m'(i,j) = v \rightarrow data[i*tda + j]
```

where the index i runs from 0 to n1-1 and the index j runs from 0 to n2-1.

The new matrix is only a view of the vector v. When the view goes out of scope the original vector v will continue to exist. The original memory can only be deallocated by freeing the original vector. Of course, the original vector should not be deallocated while the view is still in use.

The function gsl_matrix_const_view_vector_with_tda is equivalent to gsl_matrix_view_vector_with_tda but can be used for matrices which are declared const.

8.4.6 Creating row and column views

In general there are two ways to access an object, by reference or by copying. The functions described in this section create vector views which allow access to a row or column of a matrix by reference. Modifying elements of the view is equivalent to modifying the matrix, since both the vector view and the matrix point to the same memory block.

These functions return a vector view of the i-th row of the matrix m. The data pointer of the new vector is set to null if i is out of range.

The function gsl_vector_const_row is equivalent to gsl_matrix_row but can be used for matrices which are declared const.

```
 \begin{array}{lll} {\tt gsl\_vector\_view} & {\tt gsl\_matrix\_column} & ({\tt gsl\_matrix} * {\tt m}, {\tt size\_t} \; {\tt j}) & [{\tt Function}] \\ {\tt gsl\_vector\_const\_view} & {\tt gsl\_matrix\_const\_column} & ({\tt const} \\ & {\tt gsl\_matrix} * {\tt m}, {\tt size\_t} \; {\tt j}) & [{\tt Function}] \\ \end{array}
```

These functions return a vector view of the j-th column of the matrix m. The data pointer of the new vector is set to null if j is out of range.

The function gsl_vector_const_column is equivalent to gsl_matrix_column but can be used for matrices which are declared const.

gsl_vector_const_view gsl_matrix_const_subrow (const gsl_matrix * m, size_t i, size_t offset, size_t n) [Function]

These functions return a vector view of the i-th row of the matrix m beginning at offset elements past the first column and containing n elements. The data pointer of the new vector is set to null if i, offset, or n are out of range.

The function gsl_vector_const_subrow is equivalent to gsl_matrix_subrow but can be used for matrices which are declared const.

gsl_vector_view gsl_matrix_subcolumn (gsl_matrix * m, size_t j, size_t offset, size_t n) [Function]

gsl_vector_const_view gsl_matrix_const_subcolumn (const gsl_matrix * m, size_t j, size_t offset, size_t n) [Function]

These functions return a vector view of the j-th column of the matrix m beginning at offset elements past the first row and containing n elements. The data pointer of the new vector is set to null if j, offset, or n are out of range.

The function gsl_vector_const_subcolumn is equivalent to gsl_matrix_subcolumn but can be used for matrices which are declared const.

gsl_vector_view gsl_matrix_diagonal (gsl_matrix * m) [Function] gsl_vector_const_view gsl_matrix_const_diagonal (const gsl_matrix * m)

These functions return a vector view of the diagonal of the matrix m. The matrix m is not required to be square. For a rectangular matrix the length of the diagonal is the same as the smaller dimension of the matrix.

The function gsl_matrix_const_diagonal is equivalent to gsl_matrix_diagonal but can be used for matrices which are declared const.

These functions return a vector view of the k-th subdiagonal of the matrix m. The matrix m is not required to be square. The diagonal of the matrix corresponds to k = 0.

The function gsl_matrix_const_subdiagonal is equivalent to gsl_matrix_subdiagonal but can be used for matrices which are declared const.

gsl_vector_const_view gsl_matrix_const_superdiagonal (const [Function] gsl_matrix * m, size_t k)

These functions return a vector view of the k-th superdiagonal of the matrix m. The matrix m is not required to be square. The diagonal of the matrix corresponds to k=0.

The function gsl_matrix_const_superdiagonal is equivalent to gsl_matrix_superdiagonal but can be used for matrices which are declared const.

8.4.7 Copying matrices

- int gsl_matrix_memcpy (gsl_matrix * dest, const gsl_matrix * src) [Function]

 This function copies the elements of the matrix src into the matrix dest. The two matrices must have the same size.
- int gsl_matrix_swap (gsl_matrix * m1, gsl_matrix * m2) [Function]

 This function exchanges the elements of the matrices m1 and m2 by copying. The two matrices must have the same size.

8.4.8 Copying rows and columns

The functions described in this section copy a row or column of a matrix into a vector. This allows the elements of the vector and the matrix to be modified independently. Note that if the matrix and the vector point to overlapping regions of memory then the result will be undefined. The same effect can be achieved with more generality using gsl_vector_memcpy with vector views of rows and columns.

- int gsl_matrix_get_row (gsl_vector * v, const gsl_matrix * m, size_t i) [Function]

 This function copies the elements of the *i*-th row of the matrix *m* into the vector *v*.

 The length of the vector must be the same as the length of the row.
- int gsl_matrix_get_col (gsl_vector * v, const gsl_matrix * m, size_t j) [Function] This function copies the elements of the j-th column of the matrix m into the vector v. The length of the vector must be the same as the length of the column.
- int gsl_matrix_set_row (gsl_matrix * m, size_t i, const gsl_vector * v) [Function]
 This function copies the elements of the vector v into the i-th row of the matrix m.
 The length of the vector must be the same as the length of the row.
- int gsl_matrix_set_col (gsl_matrix * m, size_t j, const gsl_vector * v) [Function]
 This function copies the elements of the vector v into the j-th column of the matrix
 m. The length of the vector must be the same as the length of the column.

8.4.9 Exchanging rows and columns

The following functions can be used to exchange the rows and columns of a matrix.

- int gsl_matrix_swap_rows (gsl_matrix * m, size_t i, size_t j) [Function]
 This function exchanges the i-th and j-th rows of the matrix m in-place.
- int gsl_matrix_swap_columns ($gsl_matrix * m$, $size_t i$, $size_t j$) [Function] This function exchanges the i-th and j-th columns of the matrix m in-place.
- int gsl_matrix_swap_rowcol (gsl_matrix * m, size_t i, size_t j) [Function]

 This function exchanges the i-th row and j-th column of the matrix m in-place. The matrix must be square for this operation to be possible.

int gsl_matrix_transpose_memcpy (gsl_matrix * dest, const gsl_matrix * src) [Function]

This function makes the matrix dest the transpose of the matrix src by copying the elements of src into dest. This function works for all matrices provided that the dimensions of the matrix dest match the transposed dimensions of the matrix src.

int gsl_matrix_transpose (gsl_matrix * m) [Function]

This function replaces the matrix m by its transpose by copying the elements of the matrix in-place. The matrix must be square for this operation to be possible.

8.4.10 Matrix operations

The following operations are defined for real and complex matrices.

- int gsl_matrix_add (gsl_matrix * a, const gsl_matrix * b) [Function] This function adds the elements of matrix b to the elements of matrix a. The result $a(i,j) \leftarrow a(i,j) + b(i,j)$ is stored in a and b remains unchanged. The two matrices must have the same dimensions.
- int gsl_matrix_sub (gsl_matrix * a, const gsl_matrix * b) [Function] This function subtracts the elements of matrix b from the elements of matrix a. The result $a(i,j) \leftarrow a(i,j) b(i,j)$ is stored in a and b remains unchanged. The two matrices must have the same dimensions.
- int gsl_matrix_mul_elements (gsl_matrix * a, const gsl_matrix * b) [Function] This function multiplies the elements of matrix a by the elements of matrix b. The result $a(i,j) \leftarrow a(i,j) * b(i,j)$ is stored in a and b remains unchanged. The two matrices must have the same dimensions.
- int gsl_matrix_div_elements (gsl_matrix * a, const gsl_matrix * b) [Function] This function divides the elements of matrix a by the elements of matrix b. The result $a(i,j) \leftarrow a(i,j)/b(i,j)$ is stored in a and b remains unchanged. The two matrices must have the same dimensions.
- int gsl_matrix_scale (gsl_matrix * a, const double x) [Function] This function multiplies the elements of matrix a by the constant factor x. The result $a(i,j) \leftarrow xa(i,j)$ is stored in a.
- int gsl_matrix_add_constant (gsl_matrix * a, const double x) [Function] This function adds the constant value x to the elements of the matrix a. The result $a(i,j) \leftarrow a(i,j) + x$ is stored in a.

8.4.11 Finding maximum and minimum elements of matrices

The following operations are only defined for real matrices.

double gsl_matrix_max ($const \ gsl_matrix * m$) [Function] This function returns the maximum value in the matrix m.

double gsl_matrix_min ($const \ gsl_matrix * m$) [Function] This function returns the minimum value in the matrix m.

This function returns the minimum and maximum values in the matrix m, storing them in min_out and max_out .

This function returns the indices of the maximum value in the matrix m, storing them in imax and jmax. When there are several equal maximum elements then the first element found is returned, searching in row-major order.

This function returns the indices of the minimum value in the matrix m, storing them in imin and jmin. When there are several equal minimum elements then the first element found is returned, searching in row-major order.

```
void gsl_matrix_minmax_index (const gsl_matrix * m, size_t * imin, size_t * imax, size_t * jmax)
[Function]
```

This function returns the indices of the minimum and maximum values in the matrix m, storing them in (imin,jmin) and (imax,jmax). When there are several equal minimum or maximum elements then the first elements found are returned, searching in row-major order.

8.4.12 Matrix properties

The following functions are defined for real and complex matrices. For complex matrices both the real and imaginary parts must satisfy the conditions.

```
int gsl_matrix_ispos (const gsl_matrix * m) [Function]
int gsl_matrix_ispos (const gsl_matrix * m) [Function]
int gsl_matrix_isneg (const gsl_matrix * m) [Function]
int gsl_matrix_isnonneg (const gsl_matrix * m) [Function]
These functions return 1 if all the elements of the matrix m are zero, strictly positive, strictly negative, or non-negative respectively, and 0 otherwise. To test whether a
```

strictly negative, or non-negative respectively, and 0 otherwise. To test whether a matrix is positive-definite, use the Cholesky decomposition (see Section 14.5 [Cholesky Decomposition], page 139).

```
int gsl_matrix_equal (const gsl_matrix * a, const gsl_matrix * b) [Function]
This function returns 1 if the matrices a and b are equal (by comparison of element values) and 0 otherwise.
```

8.4.13 Example programs for matrices

The program below shows how to allocate, initialize and read from a matrix using the functions gsl_matrix_alloc, gsl_matrix_set and gsl_matrix_get.

```
#include <stdio.h>
#include <gsl/gsl_matrix.h>
```

int i, j, k = 0;

for (i = 0; i < 100; i++)for (j = 0; j < 100; j++)

gsl_matrix * m = gsl_matrix_alloc (100, 100);
gsl_matrix * a = gsl_matrix_alloc (100, 100);

gsl_matrix_set (m, i, j, 0.23 + i + j);

```
main (void)
     {
       int i, j;
       gsl_matrix * m = gsl_matrix_alloc (10, 3);
       for (i = 0; i < 10; i++)
         for (j = 0; j < 3; j++)
           gsl_matrix_set (m, i, j, 0.23 + 100*i + j);
       for (i = 0; i < 100; i++) /* OUT OF RANGE ERROR */
         for (j = 0; j < 3; j++)
           printf (m(%d,%d) = %g\n'', i, j,
                    gsl_matrix_get (m, i, j));
       gsl_matrix_free (m);
       return 0;
     }
Here is the output from the program. The final loop attempts to read outside the range of
the matrix m, and the error is trapped by the range-checking code in gsl_matrix_get.
     $ ./a.out
     m(0,0) = 0.23
     m(0,1) = 1.23
     m(0,2) = 2.23
     m(1,0) = 100.23
     m(1,1) = 101.23
     m(1,2) = 102.23
     m(9,2) = 902.23
     gsl: matrix_source.c:13: ERROR: first index out of range
     Default GSL error handler invoked.
     Aborted (core dumped)
The next program shows how to write a matrix to a file.
     #include <stdio.h>
     #include <gsl/gsl_matrix.h>
     int
     main (void)
```

```
{
   FILE * f = fopen ("test.dat", "wb");
   gsl_matrix_fwrite (f, m);
   fclose (f);
}
{
   FILE * f = fopen ("test.dat", "rb");
   gsl_matrix_fread (f, a);
   fclose (f);
}
for (i = 0; i < 100; i++)
  for (j = 0; j < 100; j++)
      double mij = gsl_matrix_get (m, i, j);
      double aij = gsl_matrix_get (a, i, j);
      if (mij != aij) k++;
    }
gsl_matrix_free (m);
gsl_matrix_free (a);
printf ("differences = %d (should be zero)\n", k);
return (k > 0);
```

After running this program the file test.dat should contain the elements of m, written in binary format. The matrix which is read back in using the function gsl_matrix_fread should be exactly equal to the original matrix.

The following program demonstrates the use of vector views. The program computes the column norms of a matrix.

```
#include <math.h>
#include <stdio.h>
#include <gsl/gsl_matrix.h>
#include <gsl/gsl_blas.h>

int
main (void)
{
    size_t i,j;
    gsl_matrix *m = gsl_matrix_alloc (10, 10);
    for (i = 0; i < 10; i++)
        for (j = 0; j < 10; j++)</pre>
```

```
gsl_matrix_set (m, i, j, sin (i) + cos (j));
      for (j = 0; j < 10; j++)
         {
           gsl_vector_view column = gsl_matrix_column (m, j);
           double d;
           d = gsl_blas_dnrm2 (&column.vector);
          printf ("matrix column %d, norm = %g\n", j, d);
         }
      gsl_matrix_free (m);
      return 0;
Here is the output of the program,
     $ ./a.out
    matrix column 0, norm = 4.31461
    matrix column 1, norm = 3.1205
    matrix column 2, norm = 2.19316
    matrix column 3, norm = 3.26114
    matrix column 4, norm = 2.53416
    matrix column 5, norm = 2.57281
    matrix column 6, norm = 4.20469
    matrix column 7, norm = 3.65202
    matrix column 8, norm = 2.08524
    matrix column 9, norm = 3.07313
The results can be confirmed using GNU OCTAVE,
     $ octave
     GNU Octave, version 2.0.16.92
     octave> m = sin(0:9), * ones(1,10)
                    + ones(10,1) * cos(0:9);
     octave> sqrt(sum(m.^2))
     ans =
       4.3146 3.1205 2.1932 3.2611
                                       2.5342 2.5728
      4.2047 3.6520 2.0852 3.0731
```

8.5 References and Further Reading

The block, vector and matrix objects in GSL follow the valarray model of C++. A description of this model can be found in the following reference,

B. Stroustrup, *The C++ Programming Language* (3rd Ed), Section 22.4 Vector Arithmetic. Addison-Wesley 1997, ISBN 0-201-88954-4.

9 Permutations

This chapter describes functions for creating and manipulating permutations. A permutation p is represented by an array of n integers in the range 0 to n-1, where each value p_i occurs once and only once. The application of a permutation p to a vector v yields a new vector v' where $v'_i = v_{p_i}$. For example, the array (0, 1, 3, 2) represents a permutation which exchanges the last two elements of a four element vector. The corresponding identity permutation is (0, 1, 2, 3).

Note that the permutations produced by the linear algebra routines correspond to the exchange of matrix columns, and so should be considered as applying to row-vectors in the form v' = vP rather than column-vectors, when permuting the elements of a vector.

The functions described in this chapter are defined in the header file gsl_permutation.h.

9.1 The Permutation struct

A permutation is defined by a structure containing two components, the size of the permutation and a pointer to the permutation array. The elements of the permutation array are all of type size_t. The gsl_permutation structure looks like this,

```
typedef struct
{
   size_t size;
   size_t * data;
} gsl_permutation;
```

9.2 Permutation allocation

```
gsl_permutation * gsl_permutation_alloc (size_t n) [Function]
This function allocates memory for a new permutation of size n. The permutation is not initialized and its elements are undefined. Use the function gsl_permutation_calloc if you want to create a permutation which is initialized to the identity. A null pointer is returned if insufficient memory is available to create the permutation.
```

```
gsl_permutation * gsl_permutation_calloc (size_t n) [Function]

This function allocates memory for a new permutation of size n and initializes it to the identity. A null pointer is returned if insufficient memory is available to create the permutation.
```

```
void gsl_permutation_init (gsl_permutation * p) [Function] This function initializes the permutation p to the identity, i.e. (0, 1, 2, ..., n-1).
```

```
void gsl_permutation_free (gsl_permutation * p) [Function]
This function frees all the memory used by the permutation p.
```

This function copies the elements of the permutation src into the permutation dest. The two permutations must have the same size.

9.3 Accessing permutation elements

The following functions can be used to access and manipulate permutations.

This function returns the value of the *i*-th element of the permutation p. If i lies outside the allowed range of 0 to n-1 then the error handler is invoked and 0 is returned. An inline version of this function is used when HAVE_INLINE is defined.

int gsl_permutation_swap ($gsl_permutation * p, const size_t i, const$ [Function] $size_t j$)

This function exchanges the i-th and j-th elements of the permutation p.

9.4 Permutation properties

size_t gsl_permutation_size ($const gsl_permutation * p$) [Function] This function returns the size of the permutation p.

size_t * gsl_permutation_data ($const\ gsl_permutation\ * p$) [Function] This function returns a pointer to the array of elements in the permutation p.

int gsl_permutation_valid (const gsl_permutation * p) [Function] This function checks that the permutation p is valid. The n elements should contain each of the numbers 0 to n-1 once and only once.

9.5 Permutation functions

void $gsl_permutation_reverse$ ($gsl_permutation * p$) [Function] This function reverses the elements of the permutation p.

This function computes the inverse of the permutation p, storing the result in inv.

int gsl_permutation_next (gsl_permutation * p) [Function]
This function advances the permutation p to the next permutation in lexicographic order and returns GSL_SUCCESS. If no further permutations are available it returns GSL_FAILURE and leaves p unmodified. Starting with the identity permutation and repeatedly applying this function will iterate through all possible permutations of a given order.

int gsl_permutation_prev (gsl_permutation * p) [Function]
This function steps backwards from the permutation p to the previous permutation in lexicographic order, returning GSL_SUCCESS. If no previous permutation is available it returns GSL_FAILURE and leaves p unmodified.

9.6 Applying Permutations

This function applies the permutation p to the array data of size n with stride stride.

This function applies the inverse of the permutation p to the array data of size n with stride stride.

int gsl_permute_vector (const gsl_permutation * p, gsl_vector * v) [Function] This function applies the permutation p to the elements of the vector v, considered as a row-vector acted on by a permutation matrix from the right, v' = vP. The j-th column of the permutation matrix P is given by the p_j -th column of the identity matrix. The permutation p and the vector v must have the same length.

This function applies the inverse of the permutation p to the elements of the vector v, considered as a row-vector acted on by an inverse permutation matrix from the right, $v' = vP^T$. Note that for permutation matrices the inverse is the same as the transpose. The j-th column of the permutation matrix P is given by the p_j -th column of the identity matrix. The permutation p and the vector v must have the same length.

This function combines the two permutations pa and pb into a single permutation p, where p = pa * pb. The permutation p is equivalent to applying pb first and then pa.

9.7 Reading and writing permutations

The library provides functions for reading and writing permutations to a file as binary data or formatted text.

This function writes the elements of the permutation p to the stream stream in binary format. The function returns $\texttt{GSL_EFAILED}$ if there was a problem writing to the file. Since the data is written in the native binary format it may not be portable between different architectures.

int gsl_permutation_fread (FILE * stream, gsl_permutation * p) [Function]

This function reads into the permutation p from the open stream in binary format. The permutation p must be preallocated with the correct length since the function uses the size of p to determine how many bytes to read. The function returns GSL_EFAILED if there was a problem reading from the file. The data is assumed to have been written in the native binary format on the same architecture.

This function writes the elements of the permutation p line-by-line to the stream stream using the format specifier format, which should be suitable for a type of $size_t$. In ISO C99 the type modifier z represents $size_t$, so "%zu\n" is a suitable format. The function returns $GSL_EFAILED$ if there was a problem writing to the file.

int gsl_permutation_fscanf ($FILE * stream, gsl_permutation * p$) [Function] This function reads formatted data from the stream stream into the permutation p. The permutation p must be preallocated with the correct length since the function uses the size of p to determine how many numbers to read. The function returns $GSL_EFAILED$ if there was a problem reading from the file.

9.8 Permutations in cyclic form

A permutation can be represented in both *linear* and *cyclic* notations. The functions described in this section convert between the two forms. The linear notation is an index mapping, and has already been described above. The cyclic notation expresses a permutation as a series of circular rearrangements of groups of elements, or *cycles*.

For example, under the cycle (1 2 3), 1 is replaced by 2, 2 is replaced by 3 and 3 is replaced by 1 in a circular fashion. Cycles of different sets of elements can be combined independently, for example (1 2 3) (4 5) combines the cycle (1 2 3) with the cycle (4 5), which is an exchange of elements 4 and 5. A cycle of length one represents an element which is unchanged by the permutation and is referred to as a *singleton*.

It can be shown that every permutation can be decomposed into combinations of cycles. The decomposition is not unique, but can always be rearranged into a standard *canonical* form by a reordering of elements. The library uses the canonical form defined in Knuth's Art of Computer Programming (Vol 1, 3rd Ed, 1997) Section 1.3.3, p.178.

The procedure for obtaining the canonical form given by Knuth is,

- 1. Write all singleton cycles explicitly
- 2. Within each cycle, put the smallest number first
- 3. Order the cycles in decreasing order of the first number in the cycle.

For example, the linear representation (2 4 3 0 1) is represented as (1 4) (0 2 3) in canonical form. The permutation corresponds to an exchange of elements 1 and 4, and rotation of elements 0, 2 and 3.

The important property of the canonical form is that it can be reconstructed from the contents of each cycle without the brackets. In addition, by removing the brackets it can be considered as a linear representation of a different permutation. In the example given above the permutation $(2\ 4\ 3\ 0\ 1)$ would become $(1\ 4\ 0\ 2\ 3)$. This mapping has many applications in the theory of permutations.

This function computes the canonical form of the permutation p and stores it in the output argument q.

 $^{^{1}}$ In versions of the GNU C library prior to the ISO C99 standard, the type modifier Z was used instead.

This function converts a permutation q in canonical form back into linear form storing it in the output argument p.

- size_t gsl_permutation_inversions (const gsl_permutation * p) [Function] This function counts the number of inversions in the permutation p. An inversion is any pair of elements that are not in order. For example, the permutation 2031 has three inversions, corresponding to the pairs (2,0) (2,1) and (3,1). The identity permutation has no inversions.
- size_t gsl_permutation_linear_cycles ($const gsl_permutation * p$) [Function] This function counts the number of cycles in the permutation p, given in linear form.

This function counts the number of cycles in the permutation q, given in canonical form.

9.9 Examples

The example program below creates a random permutation (by shuffling the elements of the identity) and finds its inverse.

```
#include <stdio.h>
#include <gsl/gsl_rng.h>
#include <gsl/gsl_randist.h>
#include <gsl/gsl_permutation.h>
int
main (void)
{
  const size_t N = 10;
  const gsl_rng_type * T;
  gsl_rng * r;
  gsl_permutation * p = gsl_permutation_alloc (N);
  gsl_permutation * q = gsl_permutation_alloc (N);
  gsl_rng_env_setup();
  T = gsl_rng_default;
  r = gsl_rng_alloc (T);
  printf ("initial permutation:");
  gsl_permutation_init (p);
  gsl_permutation_fprintf (stdout, p, " %u");
  printf ("\n");
  printf (" random permutation:");
```

```
gsl_ran_shuffle (r, p->data, N, sizeof(size_t));
       gsl_permutation_fprintf (stdout, p, " %u");
       printf ("\n");
       printf ("inverse permutation:");
       gsl_permutation_inverse (q, p);
       gsl_permutation_fprintf (stdout, q, " %u");
       printf ("\n");
       gsl_permutation_free (p);
       gsl_permutation_free (q);
       gsl_rng_free (r);
       return 0;
Here is the output from the program,
     $ ./a.out
     initial permutation: 0 1 2 3 4 5 6 7 8 9
      random permutation: 1 3 5 2 7 6 0 4 9 8
     inverse permutation: 6 0 3 1 7 2 5 4 9 8
```

The random permutation p[i] and its inverse q[i] are related through the identity p[q[i]] = i, which can be verified from the output.

The next example program steps forwards through all possible third order permutations, starting from the identity,

```
#include <stdio.h>
#include <gsl/gsl_permutation.h>

int
main (void)
{
    gsl_permutation * p = gsl_permutation_alloc (3);
    gsl_permutation_init (p);

    do
     {
        gsl_permutation_fprintf (stdout, p, " %u");
        printf ("\n");
    }
    while (gsl_permutation_next(p) == GSL_SUCCESS);
    gsl_permutation_free (p);
    return 0;
}
```

Here is the output from the program,

\$./a	a.out
0	1	2
0	2	1
1	0	2
1	2	0
2	0	1
2	1	0

The permutations are generated in lexicographic order. To reverse the sequence, begin with the final permutation (which is the reverse of the identity) and replace gsl_permutation_next with gsl_permutation_prev.

9.10 References and Further Reading

The subject of permutations is covered extensively in Knuth's Sorting and Searching, Donald E. Knuth, The Art of Computer Programming: Sorting and Searching (Vol 3, 3rd Ed, 1997), Addison-Wesley, ISBN 0201896850.

For the definition of the canonical form see,

Donald E. Knuth, *The Art of Computer Programming: Fundamental Algorithms* (Vol 1, 3rd Ed, 1997), Addison-Wesley, ISBN 0201896850. Section 1.3.3, *An Unusual Correspondence*, p.178–179.

10 Combinations

This chapter describes functions for creating and manipulating combinations. A combination c is represented by an array of k integers in the range 0 to n-1, where each value c_i occurs at most once. The combination c corresponds to indices of k elements chosen from an n element vector. Combinations are useful for iterating over all k-element subsets of a set.

The functions described in this chapter are defined in the header file gsl_combination.h.

10.1 The Combination struct

A combination is defined by a structure containing three components, the values of n and k, and a pointer to the combination array. The elements of the combination array are all of type $size_t$, and are stored in increasing order. The $gsl_combination$ structure looks like this,

```
typedef struct
{
   size_t n;
   size_t k;
   size_t *data;
} gsl_combination;
```

10.2 Combination allocation

```
gsl_combination * gsl_combination_alloc (size_t n, size_t k) [Function]
This function allocates memory for a new combination with parameters n, k. The combination is not initialized and its elements are undefined. Use the function gsl_combination_calloc if you want to create a combination which is initialized to the lexicographically first combination. A null pointer is returned if insufficient memory is available to create the combination.
```

gsl_combination * gsl_combination_calloc (size_t n, size_t k) [Function] This function allocates memory for a new combination with parameters n, k and initializes it to the lexicographically first combination. A null pointer is returned if insufficient memory is available to create the combination.

```
void gsl_combination_init_first (gsl_combination * c) [Function] This function initializes the combination c to the lexicographically first combination, i.e. (0, 1, 2, \ldots, k-1).
```

```
void gsl_combination_init_last (gsl_combination * c) [Function] This function initializes the combination c to the lexicographically last combination, i.e. (n-k, n-k+1, \ldots, n-1).
```

```
void gsl\_combination\_free (gsl\_combination * c) [Function]

This function frees all the memory used by the combination c.
```

This function copies the elements of the combination src into the combination dest. The two combinations must have the same size.

10.3 Accessing combination elements

The following function can be used to access the elements of a combination.

This function returns the value of the *i*-th element of the combination c. If i lies outside the allowed range of 0 to k-1 then the error handler is invoked and 0 is returned. An inline version of this function is used when HAVE_INLINE is defined.

10.4 Combination properties

- size_t gsl_combination_n ($const gsl_combination * c$) [Function] This function returns the range (n) of the combination c.
- size_t gsl_combination_k ($const\ gsl_combination\ ^*\ c$) [Function] This function returns the number of elements (k) in the combination c.
- size_t * gsl_combination_data ($const\ gsl_combination\ * c$) [Function] This function returns a pointer to the array of elements in the combination c.
- int gsl_combination_valid (gsl_combination * c) [Function] This function checks that the combination c is valid. The k elements should lie in the range 0 to n-1, with each value occurring once at most and in increasing order.

10.5 Combination functions

int gsl_combination_next (gsl_combination * c) [Function]

This function advances the combination c to the next combination in lexicographic order and returns GSL_SUCCESS. If no further combinations are available it returns GSL_FAILURE and leaves c unmodified. Starting with the first combination and repeatedly applying this function will iterate through all possible combinations of a given order.

int gsl_combination_prev (gsl_combination * c) [Function]

This function steps backwards from the combination c to the previous combination in lexicographic order, returning GSL_SUCCESS. If no previous combination is available it returns GSL_FAILURE and leaves c unmodified.

10.6 Reading and writing combinations

The library provides functions for reading and writing combinations to a file as binary data or formatted text.

This function writes the elements of the combination c to the stream stream in binary format. The function returns $\texttt{GSL_EFAILED}$ if there was a problem writing to the file. Since the data is written in the native binary format it may not be portable between different architectures.

int gsl_combination_fread (FILE * stream, gsl_combination * c) [Function] This function reads elements from the open stream stream into the combination c in binary format. The combination c must be preallocated with correct values of n and k since the function uses the size of c to determine how many bytes to read. The function returns GSL_EFAILED if there was a problem reading from the file. The data is assumed to have been written in the native binary format on the same architecture.

This function writes the elements of the combination c line-by-line to the stream stream using the format specifier format, which should be suitable for a type of $size_t$. In ISO C99 the type modifier z represents $size_t$, so "%zu\n" is a suitable format. The function returns GSL_EFAILED if there was a problem writing to the file.

int gsl_combination_fscanf (FILE * stream, gsl_combination * c) [Function] This function reads formatted data from the stream stream into the combination c. The combination c must be preallocated with correct values of n and k since the function uses the size of c to determine how many numbers to read. The function returns GSL_EFAILED if there was a problem reading from the file.

10.7 Examples

The example program below prints all subsets of the set $\{0, 1, 2, 3\}$ ordered by size. Subsets of the same size are ordered lexicographically.

¹ In versions of the GNU C library prior to the ISO C99 standard, the type modifier Z was used instead.

```
printf ("{");
               gsl_combination_fprintf (stdout, c, " %u");
               printf (" }\n");
             }
           while (gsl_combination_next (c) == GSL_SUCCESS);
           gsl_combination_free (c);
         }
       return 0;
Here is the output from the program,
     $ ./a.out
     All subsets of \{0,1,2,3\} by size:
     { }
     { 0 }
     { 1 }
     { 2 }
     { 3 }
     { 0 1 }
     { 0 2 }
     { 0 3 }
     {12}
     {13}
     { 2 3 }
     { 0 1 2 }
     { 0 1 3 }
     { 0 2 3 }
     { 1 2 3 }
     { 0 1 2 3 }
```

All 16 subsets are generated, and the subsets of each size are sorted lexicographically.

10.8 References and Further Reading

Further information on combinations can be found in,

Donald L. Kreher, Douglas R. Stinson, Combinatorial Algorithms: Generation, Enumeration and Search, 1998, CRC Press LLC, ISBN 084933988X

11 Multisets

This chapter describes functions for creating and manipulating multisets. A multiset c is represented by an array of k integers in the range 0 to n-1, where each value c_i may occur more than once. The multiset c corresponds to indices of k elements chosen from an n element vector with replacement. In mathematical terms, n is the cardinality of the multiset while k is the maximum multiplicity of any value. Multisets are useful, for example, when iterating over the indices of a k-th order symmetric tensor in n-space.

The functions described in this chapter are defined in the header file gsl_multiset.h.

11.1 The Multiset struct

A multiset is defined by a structure containing three components, the values of n and k, and a pointer to the multiset array. The elements of the multiset array are all of type $\mathtt{size_t}$, and are stored in increasing order. The $\mathtt{gsl_multiset}$ structure looks like this,

```
typedef struct
{
   size_t n;
   size_t k;
   size_t *data;
} gsl_multiset;
```

11.2 Multiset allocation

```
gsl_multiset * gsl_multiset_alloc (size_t n, size_t k)
```

[Function]

This function allocates memory for a new multiset with parameters n, k. The multiset is not initialized and its elements are undefined. Use the function ${\tt gsl_multiset_calloc}$ if you want to create a multiset which is initialized to the lexicographically first multiset element. A null pointer is returned if insufficient memory is available to create the multiset.

```
gsl_multiset * gsl_multiset_calloc (size_t n, size_t k)
```

[Function]

This function allocates memory for a new multiset with parameters n, k and initializes it to the lexicographically first multiset element. A null pointer is returned if insufficient memory is available to create the multiset.

```
void gsl_multiset_init_first (gsl_multiset * c)
```

[Function]

This function initializes the multiset c to the lexicographically first multiset element, i.e. 0 repeated k times.

```
void gsl_multiset_init_last (gsl_multiset * c)
```

[Function]

This function initializes the multiset c to the lexicographically last multiset element, i.e. n-1 repeated k times.

```
void gsl_multiset_free (gsl_multiset * c)
```

[Function]

This function frees all the memory used by the multiset c.

This function copies the elements of the multiset *src* into the multiset *dest*. The two multisets must have the same size.

11.3 Accessing multiset elements

The following function can be used to access the elements of a multiset.

size_t gsl_multiset_get (const gsl_multiset * c, const size_t i) [Function] This function returns the value of the *i*-th element of the multiset c. If i lies outside the allowed range of 0 to k-1 then the error handler is invoked and 0 is returned. An inline version of this function is used when HAVE_INLINE is defined.

11.4 Multiset properties

size_t gsl_multiset_n ($const gsl_multiset * c$) [Function]
This function returns the range (n) of the multiset c.

size_t gsl_multiset_k ($const gsl_multiset * c$) [Function]
This function returns the number of elements (k) in the multiset c.

size_t * gsl_multiset_data ($const \ gsl_multiset * c$) [Function]

This function returns a pointer to the array of elements in the multiset c.

int gsl_multiset_valid (gsl_multiset * c) [Function] This function checks that the multiset c is valid. The k elements should lie in the range 0 to n-1, with each value occurring in nondecreasing order.

11.5 Multiset functions

order.

int $gsl_multiset_next$ ($gsl_multiset * c$) [Function] This function advances the multiset c to the next multiset element in lexicographic order and returns $GSL_SUCCESS$. If no further multisets elements are available it returns $GSL_FAILURE$ and leaves c unmodified. Starting with the first multiset and repeatedly applying this function will iterate through all possible multisets of a given

int gsl_multiset_prev (gsl_multiset * c) [Function]

This function steps backwards from the multiset c to the previous multiset element in lexicographic order, returning GSL_SUCCESS. If no previous multiset is available it returns GSL_FAILURE and leaves c unmodified.

11.6 Reading and writing multisets

The library provides functions for reading and writing multisets to a file as binary data or formatted text.

int gsl_multiset_fwrite (FILE * stream, const gsl_multiset * c) [Function]

This function writes the elements of the multiset c to the stream stream in binary format. The function returns GSL_EFAILED if there was a problem writing to the file. Since the data is written in the native binary format it may not be portable between different architectures.

int gsl_multiset_fread (FILE * stream, gsl_multiset * c) [Function] This function reads elements from the open stream stream into the multiset c in binary format. The multiset c must be preallocated with correct values of n and k since the function uses the size of c to determine how many bytes to read. The function returns GSL_EFAILED if there was a problem reading from the file. The data is assumed to have been written in the native binary format on the same architecture.

This function writes the elements of the multiset c line-by-line to the stream stream using the format specifier format, which should be suitable for a type of size_t. In ISO C99 the type modifier z represents size_t, so "%zu\n" is a suitable format. The function returns GSL_EFAILED if there was a problem writing to the file.

```
int gsl_multiset_fscanf (FILE * stream, gsl_multiset * c) [Function] This function reads formatted data from the stream stream into the multiset c. The multiset c must be preallocated with correct values of n and k since the function uses the size of c to determine how many numbers to read. The function returns GSL_EFAILED if there was a problem reading from the file.
```

11.7 Examples

The example program below prints all multisets elements containing the values $\{0, 1, 2, 3\}$ ordered by size. Multiset elements of the same size are ordered lexicographically.

¹ In versions of the GNU C library prior to the ISO C99 standard, the type modifier Z was used instead.

```
gsl_multiset_fprintf (stdout, c, " %u");
               printf (" }\n");
           while (gsl_multiset_next (c) == GSL_SUCCESS);
           gsl_multiset_free (c);
         }
       return 0;
     }
Here is the output from the program,
     $ ./a.out
     all multisets of \{0,1,2,3\} by size:
     { }
     { 0 }
     { 1 }
     { 2 }
     { 3 }
     { 0 0 }
     { 0 1 }
     { 0 2 }
     { 0 3 }
     { 1 1 }
     { 1 2 }
     { 1 3 }
     { 2 2 }
     { 2 3 }
     { 3 3 }
     { 0 0 0 }
     { 0 0 1 }
     { 0 0 2 }
     { 0 0 3 }
     { 0 1 1 }
     { 0 1 2 }
     { 0 1 3 }
     { 0 2 2 }
     { 0 2 3 }
     { 0 3 3 }
     { 1 1 1 }
     {112}
     { 1 1 3 }
     { 1 2 2 }
     { 1 2 3 }
     { 1 3 3 }
     { 2 2 2 }
     { 2 2 3 }
     { 2 3 3 }
```

```
{ 3 3 3 }
{0000}
{ 0 0 0 1 }
{0002}
{ 0 0 0 3 }
{ 0 0 1 1 }
{ 0 0 1 2 }
{ 0 0 1 3 }
{ 0 0 2 2 }
{ 0 0 2 3 }
{ 0 0 3 3 }
{ 0 1 1 1 }
{ 0 1 1 2 }
{ 0 1 1 3 }
{ 0 1 2 2 }
{ 0 1 2 3 }
{ 0 1 3 3 }
{ 0 2 2 2 }
{ 0 2 2 3 }
{ 0 2 3 3 }
{ 0 3 3 3 }
{1111}
{1112}
{1113}
{1122}
{1123}
{1133}
{ 1 2 2 2 }
{1223}
{ 1 2 3 3 }
{ 1 3 3 3 }
{ 2 2 2 2 }
{ 2 2 2 3 }
{ 2 2 3 3 }
{ 2 3 3 3 }
{ 3 3 3 3 }
```

All 70 multisets are generated and sorted lexicographically.

12 Sorting

This chapter describes functions for sorting data, both directly and indirectly (using an index). All the functions use the heapsort algorithm. Heapsort is an $O(N \log N)$ algorithm which operates in-place and does not require any additional storage. It also provides consistent performance, the running time for its worst-case (ordered data) being not significantly longer than the average and best cases. Note that the heapsort algorithm does not preserve the relative ordering of equal elements—it is an unstable sort. However the resulting order of equal elements will be consistent across different platforms when using these functions.

12.1 Sorting objects

The following function provides a simple alternative to the standard library function qsort. It is intended for systems lacking qsort, not as a replacement for it. The function qsort should be used whenever possible, as it will be faster and can provide stable ordering of equal elements. Documentation for qsort is available in the GNU C Library Reference Manual.

The functions described in this section are defined in the header file gsl_heapsort.h.

This function sorts the *count* elements of the array *array*, each of size *size*, into ascending order using the comparison function *compare*. The type of the comparison function is defined by,

A comparison function should return a negative integer if the first argument is less than the second argument, 0 if the two arguments are equal and a positive integer if the first argument is greater than the second argument.

For example, the following function can be used to sort doubles into ascending numerical order.

The appropriate function call to perform the sort is,

Note that unlike qsort the heapsort algorithm cannot be made into a stable sort by pointer arithmetic. The trick of comparing pointers for equal elements in the comparison function does not work for the heapsort algorithm. The heapsort algorithm performs an internal rearrangement of the data which destroys its initial ordering.

int gsl_heapsort_index (size_t * p, const void * array, size_t count, size_t size, gsl_comparison_fn_t compare) [Function]

This function indirectly sorts the *count* elements of the array array, each of size size, into ascending order using the comparison function compare. The resulting permutation is stored in p, an array of length n. The elements of p give the index of the array element which would have been stored in that position if the array had been sorted in place. The first element of p gives the index of the least element in array, and the last element of p gives the index of the greatest element in array. The array itself is not changed.

12.2 Sorting vectors

The following functions will sort the elements of an array or vector, either directly or indirectly. They are defined for all real and integer types using the normal suffix rules. For example, the float versions of the array functions are gsl_sort_float and gsl_sort_index. The corresponding vector functions are gsl_sort_vector_float and gsl_sort_vector_float_index. The prototypes are available in the header files gsl_sort_float.h gsl_sort_vector_float.h. The complete set of prototypes can be included using the header files gsl_sort.h and gsl_sort_vector.h.

There are no functions for sorting complex arrays or vectors, since the ordering of complex numbers is not uniquely defined. To sort a complex vector by magnitude compute a real vector containing the magnitudes of the complex elements, and sort this vector indirectly. The resulting index gives the appropriate ordering of the original complex vector.

- void gsl_sort (double * data, const size_t stride, size_t n) [Function]
 This function sorts the n elements of the array data with stride stride into ascending numerical order.

This function sorts the n elements of the array data1 with stride stride1 into ascending numerical order, while making the same rearrangement of the array data2 with stride stride2, also of size n.

- void gsl_sort_vector (gsl_vector * v) [Function]
 This function sorts the elements of the vector v into ascending numerical order.
- void gsl_sort_vector2 (gsl_vector * v1, gsl_vector * v2) [Function]
 This function sorts the elements of the vector v1 into ascending numerical order, while making the same rearrangement of the vector v2.

This function indirectly sorts the n elements of the array data with stride stride into ascending order, storing the resulting permutation in p. The array p must be

allocated with a sufficient length to store the n elements of the permutation. The elements of p give the index of the array element which would have been stored in that position if the array had been sorted in place. The array data is not changed.

int gsl_sort_vector_index (gsl_permutation * p, const gsl_vector * v) [Function]

This function indirectly sorts the elements of the vector v into ascending order, storing the resulting permutation in p. The elements of p give the index of the vector element which would have been stored in that position if the vector had been sorted in place. The first element of p gives the index of the least element in v, and the last element of p gives the index of the greatest element in v. The vector v is not changed.

12.3 Selecting the k smallest or largest elements

The functions described in this section select the k smallest or largest elements of a data set of size N. The routines use an O(kN) direct insertion algorithm which is suited to subsets that are small compared with the total size of the dataset. For example, the routines are useful for selecting the 10 largest values from one million data points, but not for selecting the largest 100,000 values. If the subset is a significant part of the total dataset it may be faster to sort all the elements of the dataset directly with an $O(N \log N)$ algorithm and obtain the smallest or largest values that way.

This function copies the k smallest elements of the array src, of size n and stride stride, in ascending numerical order into the array dest. The size k of the subset must be less than or equal to n. The data src is not modified by this operation.

This function copies the k largest elements of the array src, of size n and stride stride, in descending numerical order into the array dest. k must be less than or equal to n. The data src is not modified by this operation.

These functions copy the k smallest or largest elements of the vector v into the array dest. k must be less than or equal to the length of the vector v.

The following functions find the indices of the k smallest or largest elements of a dataset,

This function stores the indices of the k smallest elements of the array src, of size n and stride stride, in the array p. The indices are chosen so that the corresponding data is in ascending numerical order. k must be less than or equal to n. The data src is not modified by this operation.

This function stores the indices of the k largest elements of the array src, of size n and stride stride, in the array p. The indices are chosen so that the corresponding data is in descending numerical order. k must be less than or equal to n. The data src is not modified by this operation.

These functions store the indices of the k smallest or largest elements of the vector v in the array p. k must be less than or equal to the length of the vector v.

12.4 Computing the rank

The rank of an element is its order in the sorted data. The rank is the inverse of the index permutation, p. It can be computed using the following algorithm,

```
for (i = 0; i < p->size; i++)
{
    size_t pi = p->data[i];
    rank->data[pi] = i;
}
```

This can be computed directly from the function gsl_permutation_inverse(rank,p).

The following function will print the rank of each element of the vector v,

```
print_rank (gsl_vector * v)
{
  size_t i;
  size_t n = v->size;
  gsl_permutation * perm = gsl_permutation_alloc(n);
  gsl_permutation * rank = gsl_permutation_alloc(n);
  gsl_sort_vector_index (perm, v);
  gsl_permutation_inverse (rank, perm);
  for (i = 0; i < n; i++)
    double vi = gsl_vector_get(v, i);
    printf ("element = %d, value = %g, rank = %d\n",
             i, vi, rank->data[i]);
  }
  gsl_permutation_free (perm);
  gsl_permutation_free (rank);
}
```

12.5 Examples

The following example shows how to use the permutation p to print the elements of the vector v in ascending order,

```
gsl_sort_vector_index (p, v);
for (i = 0; i < v->size; i++)
{
    double vpi = gsl_vector_get (v, p->data[i]);
    printf ("order = %d, value = %g\n", i, vpi);
}
```

The next example uses the function gsl_sort_smallest to select the 5 smallest numbers from 100000 uniform random variates stored in an array,

```
#include <gsl/gsl_rng.h>
#include <gsl/gsl_sort_double.h>
int
main (void)
{
  const gsl_rng_type * T;
 gsl_rng * r;
 size_t i, k = 5, N = 100000;
  double * x = malloc (N * sizeof(double));
  double * small = malloc (k * sizeof(double));
 gsl_rng_env_setup();
 T = gsl_rng_default;
  r = gsl_rng_alloc (T);
  for (i = 0; i < N; i++)
    {
      x[i] = gsl_rng_uniform(r);
 gsl_sort_smallest (small, k, x, 1, N);
 printf ("%d smallest values from %d\n", k, N);
  for (i = 0; i < k; i++)
     printf ("%d: %.18f\n", i, small[i]);
  free (x);
```

```
free (small);
  gsl_rng_free (r);
  return 0;
}
The output lists the 5 smallest values, in ascending order,
  $ ./a.out
  5 smallest values from 100000
  0:  0.000003489200025797
  1:  0.000008199829608202
  2:  0.000008953968062997
  3:  0.000010712770745158
  4:  0.000033531803637743
```

12.6 References and Further Reading

The subject of sorting is covered extensively in Knuth's Sorting and Searching,

Donald E. Knuth, The Art of Computer Programming: Sorting and Searching (Vol 3, 3rd Ed, 1997), Addison-Wesley, ISBN 0201896850.

The Heapsort algorithm is described in the following book,

Robert Sedgewick, Algorithms in C, Addison-Wesley, ISBN 0201514257.

13 BLAS Support

The Basic Linear Algebra Subprograms (BLAS) define a set of fundamental operations on vectors and matrices which can be used to create optimized higher-level linear algebra functionality.

The library provides a low-level layer which corresponds directly to the C-language BLAS standard, referred to here as "CBLAS", and a higher-level interface for operations on GSL vectors and matrices. Users who are interested in simple operations on GSL vector and matrix objects should use the high-level layer described in this chapter. The functions are declared in the file gsl_blas.h and should satisfy the needs of most users.

Note that GSL matrices are implemented using dense-storage so the interface only includes the corresponding dense-storage BLAS functions. The full BLAS functionality for band-format and packed-format matrices is available through the low-level CBLAS interface. Similarly, GSL vectors are restricted to positive strides, whereas the low-level CBLAS interface supports negative strides as specified in the BLAS standard.¹

The interface for the gsl_cblas layer is specified in the file gsl_cblas.h. This interface corresponds to the BLAS Technical Forum's standard for the C interface to legacy BLAS implementations. Users who have access to other conforming CBLAS implementations can use these in place of the version provided by the library. Note that users who have only a Fortran BLAS library can use a CBLAS conformant wrapper to convert it into a CBLAS library. A reference CBLAS wrapper for legacy Fortran implementations exists as part of the CBLAS standard and can be obtained from Netlib. The complete set of CBLAS functions is listed in an appendix (see Appendix D [GSL CBLAS Library], page 463).

There are three levels of BLAS operations,

```
Level 1 Vector operations, e.g. y = \alpha x + y
```

Level 2 Matrix-vector operations, e.g. $y = \alpha Ax + \beta y$

Level 3 Matrix-matrix operations, e.g. $C = \alpha AB + C$

Each routine has a name which specifies the operation, the type of matrices involved and their precisions. Some of the most common operations and their names are given below,

DOT scalar product, $x^T y$ **AXPY** vector sum, $\alpha x + y$ **MV** matrix-vector product, Ax

SV matrix-vector solve, inv(A)xMM matrix-matrix product, AB

SM matrix-matrix solve, inv(A)B

The types of matrices are,

GE general

GB general band

¹ In the low-level CBLAS interface, a negative stride accesses the vector elements in reverse order, i.e. the *i*-th element is given by (N-i)*|incx| for incx < 0.

\mathbf{SY}	symmetric
SB	symmetric band
SP	symmetric packed
\mathbf{HE}	hermitian
нв	hermitian band
HP	hermitian packed
\mathbf{TR}	triangular
тв	triangular band

Each operation is defined for four precisions,

triangular packed

 $egin{array}{lll} \mathbf{S} & & & & & & & & & \\ \mathbf{D} & & & & & & & & & \\ \mathbf{C} & & & & & & & & & \\ \mathbf{Z} & & & & & & & & & \\ \end{array} \qquad \qquad \begin{array}{ll} & & & & & & & \\ \mathbf{S} & & & & & & & \\ \mathbf{Z} & & & & & & & \\ \end{array} \qquad \qquad \begin{array}{ll} & & & & & & \\ \mathbf{S} & & & & & & \\ \end{array} \qquad \qquad \begin{array}{ll} \mathbf{S} & & & & & \\ \mathbf{S} & & & & & \\ \mathbf{S} & & & & & \\ \end{array} \qquad \qquad \begin{array}{ll} \mathbf{S} & & & & \\ \mathbf{S} & & & & \\ \mathbf{S} & & & & \\ \end{array} \qquad \qquad \begin{array}{ll} \mathbf{S} & & & & \\ \mathbf{S} & & & & \\ \mathbf{S} & & & \\ \end{array} \qquad \qquad \begin{array}{ll} \mathbf{S} & & & \\ \mathbf{S} & & & \\ \mathbf{S} & & & \\ \end{array} \qquad \qquad \begin{array}{ll} \mathbf{S} & & & \\ \mathbf{S} & & & \\ \mathbf{S} & & \\ \mathbf{S} & & & \\ \end{array} \qquad \qquad \begin{array}{ll} \mathbf{S} & & & \\ \mathbf{S} & & \\ \mathbf{S} & & \\ \mathbf{S} & & \\ \end{array} \qquad \qquad \begin{array}{ll} \mathbf{S} & & \\ \mathbf{S} & & \\ \mathbf{S} & & \\ \mathbf{S} & & \\ \end{array} \qquad \qquad \begin{array}{ll} \mathbf{S} & & \\ \mathbf{S} & & \\$

 \mathbf{TP}

Thus, for example, the name SGEMM stands for "single-precision general matrix-matrix multiply" and ZGEMM stands for "double-precision complex matrix-matrix multiply".

Note that the vector and matrix arguments to BLAS functions must not be aliased, as the results are undefined when the underlying arrays overlap (see Section 2.11 [Aliasing of arrays], page 9).

13.1 GSL BLAS Interface

GSL provides dense vector and matrix objects, based on the relevant built-in types. The library provides an interface to the BLAS operations which apply to these objects. The interface to this functionality is given in the file gsl_blas.h.

13.1.1 Level 1

This function computes the sum $\alpha + x^T y$ for the vectors x and y, returning the result in result.

These functions compute the scalar product x^Ty for the vectors x and y, returning the result in result.

int gsl_blas_cdotu (const gsl_vector_complex_float * x, const gsl_vector_complex_float * y, gsl_complex_float * dotu) [Function]

int gsl_blas_zdotu (const gsl_vector_complex * x, const gsl_vector_complex * y, gsl_complex * dotu) [Function]

These functions compute the complex scalar product x^Ty for the vectors x and y, returning the result in dotu

int gsl_blas_zdotc (const gsl_vector_complex * x, const gsl_vector_complex * y, gsl_complex * dotc) [Function]

These functions compute the complex conjugate scalar product x^Hy for the vectors x and y, returning the result in dotc

float gsl_blas_snrm2 (const gsl_vector_float * x) [Function] double gsl_blas_dnrm2 (const gsl_vector * x) [Function] These functions compute the Euclidean norm $||x||_2 = \sqrt{\sum x_i^2}$ of the vector x.

float gsl_blas_scnrm2 (const gsl_vector_complex_float * x) [Function] double gsl_blas_dznrm2 (const gsl_vector_complex * x) [Function]

These functions compute the Euclidean norm of the complex vector x,

$$||x||_2 = \sqrt{\sum (\text{Re}(x_i)^2 + \text{Im}(x_i)^2)}.$$

These functions compute the absolute sum $\sum |x_i|$ of the elements of the vector x.

These functions compute the sum of the magnitudes of the real and imaginary parts of the complex vector x, $\sum (|\text{Re}(x_i)| + |\text{Im}(x_i)|)$.

CBLAS_INDEX_t gsl_blas_isamax (const gsl_vector_float * x) [Function]
CBLAS_INDEX_t gsl_blas_idamax (const gsl_vector * x) [Function]

CBLAS_INDEX_t gsl_blas_icamax (const gsl_vector_complex_float * x) [Function]
CBLAS_INDEX_t gsl_blas_izamax (const gsl_vector_complex * x) [Function]

These functions return the index of the largest element of the vector x. The largest element is determined by its absolute magnitude for real vectors and by the sum of the magnitudes of the real and imaginary parts $|\text{Re}(x_i)| + |\text{Im}(x_i)|$ for complex vectors. If the largest value occurs several times then the index of the first occurrence is returned.

int gsl_blas_sswap (gsl_vector_float * x, gsl_vector_float * y) [Function] int gsl_blas_dswap (gsl_vector * x, gsl_vector * y) [Function]

int gsl_blas_cswap (gsl_vector_complex_float * x, gsl_vector_complex_float * y)

gsl_vector_complex_float * y)

[Function]

int gsl_blas_zswap (gsl_vector_complex * x, gsl_vector_complex * y) [Function]

These functions exchange the elements of the vectors x and y.

```
int gsl_blas_scopy (const gsl_vector_float * x, gsl_vector_float * y)
                                                                              [Function]
int gsl_blas_dcopy (const gsl_vector * x, gsl_vector * y)
                                                                              [Function]
int gsl_blas_ccopy (const gsl_vector_complex_float * x,
                                                                              [Function]
          gsl_vector_complex_float * y)
int gsl_blas_zcopy (const gsl_vector_complex * x, gsl_vector_complex
                                                                              [Function]
     These functions copy the elements of the vector x into the vector y.
int gsl_blas_saxpy (float alpha, const gsl_vector_float * x,
                                                                              [Function]
          gsl_vector_float * y)
int gsl_blas_daxpy (double alpha, const gsl_vector * x, gsl_vector * y)
                                                                              [Function]
int gsl_blas_caxpy (const gsl_complex_float alpha, const
                                                                              [Function]
          gsl_vector_complex_float * x, gsl_vector_complex_float * y)
int gsl_blas_zaxpy (const gsl_complex alpha, const
                                                                              [Function]
          gsl_vector_complex * x, gsl_vector_complex * y)
     These functions compute the sum y = \alpha x + y for the vectors x and y.
void gsl_blas_sscal (float alpha, gsl_vector_float * x)
                                                                              [Function]
void gsl_blas_dscal (double alpha, gsl_vector * x)
                                                                              [Function]
void gsl_blas_cscal (const gsl_complex_float alpha,
                                                                              [Function]
          gsl\_vector\_complex\_float * x)
void gsl_blas_zscal (const gsl_complex alpha, gsl_vector_complex *
                                                                              [Function]
          x)
void gsl_blas_csscal (float alpha, gsl_vector_complex_float * x)
                                                                              [Function]
void gsl_blas_zdscal (double alpha, gsl_vector_complex * x)
                                                                              [Function]
     These functions rescale the vector x by the multiplicative factor alpha.
int gsl_blas_srotg (float a[], float b[], float c[], float s[])
                                                                              [Function]
int gsl_blas_drotg (double a[], double b[], double c[], double s[])
                                                                              [Function]
     These functions compute a Givens rotation (c, s) which zeroes the vector (a, b),
```

$$\begin{pmatrix} c & s \\ -s & c \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} r' \\ 0 \end{pmatrix}$$

The variables a and b are overwritten by the routine.

These functions apply a Givens rotation (x', y') = (cx + sy, -sx + cy) to the vectors x, y.

These functions compute a modified Givens transformation. The modified Givens transformation is defined in the original Level-1 BLAS specification, given in the references.

- int gsl_blas_drotm (gsl_vector * x, gsl_vector * y, const double P[]) [Function] These functions apply a modified Givens transformation.

13.1.2 Level 2

- int gsl_blas_dgemv (CBLAS_TRANSPOSE_t TransA, double alpha, [Function] const gsl_matrix * A, const gsl_vector * x, double beta, gsl_vector * y)
- int gsl_blas_zgemv (CBLAS_TRANSPOSE_t TransA, const [Function] gsl_complex alpha, const gsl_matrix_complex * A, const gsl_vector_complex * x, const gsl_complex beta, gsl_vector_complex * y)

 These functions compute the matrix-vector product and sum $y = \alpha op(A)x + \beta y$, where
 - $op(A) = A, A^T, A^H$ for TransA = CblasNoTrans, CblasTrans, CblasConjTrans.
- int gsl_blas_strmv (CBLAS_UPLO_t Uplo, CBLAS_TRANSPOSE_t [Function] TransA, CBLAS_DIAG_t Diag, const gsl_matrix_float * A, gsl_vector_float * x)
- int gsl_blas_dtrmv (CBLAS_UPLO_t Uplo, CBLAS_TRANSPOSE_t [Function] TransA, CBLAS_DIAG_t Diag, const gsl_matrix * A, gsl_vector * x)
- - These functions compute the matrix-vector product x = op(A)x for the triangular matrix A, where op(A) = A, A^T , A^H for TransA = CblasNoTrans, CblasConjTrans. When Uplo is CblasUpper then the upper triangle of A is used, and when Uplo is CblasLower then the lower triangle of A is used. If Diag is CblasNonUnit then the diagonal of the matrix is used, but if Diag is CblasNonUnit then the diagonal elements of the matrix A are taken as unity and are not referenced.
- int gsl_blas_strsv (CBLAS_UPLO_t Uplo, CBLAS_TRANSPOSE_t [Function]
 TransA, CBLAS_DIAG_t Diag, const gsl_matrix_float * A, gsl_vector_float * x)
- int gsl_blas_dtrsv (CBLAS_UPLO_t Uplo, CBLAS_TRANSPOSE_t [Function]

 **TransA, CBLAS_DIAG_t Diag, const gsl_matrix * A, gsl_vector * x)

These functions compute inv(op(A))x for x, where op(A) = A, A^T , A^H for TransA = CblasNoTrans, CblasTrans, CblasConjTrans. When Uplo is CblasUpper then the upper triangle of A is used, and when Uplo is CblasLower then the lower triangle of A is used. If Diag is CblasNonUnit then the diagonal of the matrix is used, but if Diag is CblasUnit then the diagonal elements of the matrix A are taken as unity and are not referenced.

- int gsl_blas_ssymv (CBLAS_UPLO_t Uplo, float alpha, const gsl_matrix_float * A, const gsl_vector_float * x, float beta, gsl_vector_float * y)
- int gsl_blas_dsymv (CBLAS_UPLO_t Uplo, double alpha, const gsl_matrix * A, const gsl_vector * x, double beta, gsl_vector * y) [Function]

These functions compute the matrix-vector product and sum $y = \alpha Ax + \beta y$ for the symmetric matrix A. Since the matrix A is symmetric only its upper half or lower half need to be stored. When Uplo is CblasUpper then the upper triangle and diagonal of A are used, and when Uplo is CblasLower then the lower triangle and diagonal of A are used.

- int gsl_blas_chemv (CBLAS_UPLO_t Uplo, const gsl_complex_float [Function] alpha, const gsl_matrix_complex_float * A, const gsl_vector_complex_float * x, const gsl_complex_float beta, gsl_vector_complex_float * y)

These functions compute the matrix-vector product and sum $y = \alpha Ax + \beta y$ for the hermitian matrix A. Since the matrix A is hermitian only its upper half or lower half need to be stored. When Uplo is CblasUpper then the upper triangle and diagonal of A are used, and when Uplo is CblasLower then the lower triangle and diagonal of A are used. The imaginary elements of the diagonal are automatically assumed to be zero and are not referenced.

- int gsl_blas_zgeru (const gsl_complex alpha, const gsl_vector_complex * x, const gsl_vector_complex * y, gsl_matrix_complex * A) These functions compute the rank-1 update $A = \alpha xy^T + A$ of the matrix A.

- int gsl_blas_zgerc (const gsl_complex alpha, const gsl_vector_complex * x, const gsl_vector_complex * y, gsl_matrix_complex * A) These functions compute the conjugate rank-1 update $A = \alpha xy^H + A$ of the matrix A.
- int gsl_blas_ssyr (CBLAS_UPLO_t Uplo, float alpha, const gsl_vector_float * x, gsl_matrix_float * A) [Function]
- int gsl_blas_dsyr (CBLAS_UPLO_t Uplo, double alpha, const gsl_vector * x, gsl_matrix * A) [Function]

These functions compute the symmetric rank-1 update $A = \alpha x x^T + A$ of the symmetric matrix A. Since the matrix A is symmetric only its upper half or lower half need to be stored. When Uplo is CblasUpper then the upper triangle and diagonal of A are used, and when Uplo is CblasLower then the lower triangle and diagonal of A are used.

- int gsl_blas_cher (CBLAS_UPLO_t Uplo, float alpha, const gsl_vector_complex_float * x, gsl_matrix_complex_float * A) [Function]
- int gsl_blas_zher (CBLAS_UPLO_t Uplo, double alpha, const gsl_vector_complex * x, gsl_matrix_complex * A) [Function]

These functions compute the hermitian rank-1 update $A = \alpha x x^H + A$ of the hermitian matrix A. Since the matrix A is hermitian only its upper half or lower half need to be stored. When Uplo is CblasUpper then the upper triangle and diagonal of A are used, and when Uplo is CblasLower then the lower triangle and diagonal of A are used. The imaginary elements of the diagonal are automatically set to zero.

- int gsl_blas_ssyr2 (CBLAS_UPLO_t Uplo, float alpha, const gsl_vector_float * x, const gsl_vector_float * y, gsl_matrix_float * A) [Function]
- int gsl_blas_dsyr2 (CBLAS_UPLO_t Uplo, double alpha, const gsl_vector * x, const gsl_vector * y, gsl_matrix * A) [Function]

These functions compute the symmetric rank-2 update $A = \alpha xy^T + \alpha yx^T + A$ of the symmetric matrix A. Since the matrix A is symmetric only its upper half or lower half need to be stored. When Uplo is CblasUpper then the upper triangle and diagonal of A are used, and when Uplo is CblasLower then the lower triangle and diagonal of A are used.

- int gsl_blas_cher2 (CBLAS_UPLO_t Uplo, const gsl_complex_float [Function] alpha, const gsl_vector_complex_float * x, const gsl_vector_complex_float * y, gsl_matrix_complex_float * A)

These functions compute the hermitian rank-2 update $A = \alpha x y^H + \alpha^* y x^H + A$ of the hermitian matrix A. Since the matrix A is hermitian only its upper half or lower half need to be stored. When Uplo is CblasUpper then the upper triangle and diagonal of A are used, and when Uplo is CblasLower then the lower triangle and diagonal of A are used. The imaginary elements of the diagonal are automatically set to zero.

13.1.3 Level 3

- int gsl_blas_sgemm (CBLAS_TRANSPOSE_t TransA, [Function] CBLAS_TRANSPOSE_t TransB, float alpha, const gsl_matrix_float * A, const gsl_matrix_float * B, float beta, gsl_matrix_float * C)
- int gsl_blas_dgemm (CBLAS_TRANSPOSE_t TransA, [Function]

 CBLAS_TRANSPOSE_t TransB, double alpha, const gsl_matrix * A, const

 gsl_matrix * B, double beta, gsl_matrix * C)
- - These functions compute the matrix-matrix product and sum $C = \alpha op(A)op(B) + \beta C$ where op(A) = A, A^T , A^H for TransA = CblasNoTrans, CblasConjTrans and similarly for the parameter TransB.

- int gsl_blas_zsymm (CBLAS_SIDE_t Side, CBLAS_UPLO_t Uplo, const gsl_complex alpha, const gsl_matrix_complex * A, const gsl_matrix_complex * B, const gsl_complex beta, gsl_matrix_complex * C) These functions compute the matrix-matrix product and sum $C = \alpha AB + \beta C$ for Side is CblasLeft and $C = \alpha BA + \beta C$ for Side is CblasRight, where the matrix A is symmetric. When Uplo is CblasUpper then the upper triangle and diagonal of A are used, and when Uplo is CblasLower then the lower triangle and diagonal of A are used.
- int gsl_blas_zhemm (CBLAS_SIDE_t Side, CBLAS_UPLO_t Uplo, const gsl_complex alpha, const gsl_matrix_complex * A, const gsl_matrix_complex * B, const gsl_complex beta, gsl_matrix_complex * C) These functions compute the matrix-matrix product and sum $C = \alpha AB + \beta C$ for Side is CblasLeft and $C = \alpha BA + \beta C$ for Side is CblasRight, where the matrix A

is hermitian. When Uplo is CblasUpper then the upper triangle and diagonal of A are used, and when Uplo is CblasLower then the lower triangle and diagonal of A are used. The imaginary elements of the diagonal are automatically set to zero.

- int gsl_blas_strmm (CBLAS_SIDE_t Side, CBLAS_UPLO_t Uplo, [Function] CBLAS_TRANSPOSE_t TransA, CBLAS_DIAG_t Diag, float alpha, const gsl_matrix_float * A, gsl_matrix_float * B)

- int gsl_blas_ztrmm (CBLAS_SIDE_t Side, CBLAS_UPLO_t Uplo, [Function] CBLAS_TRANSPOSE_t TransA, CBLAS_DIAG_t Diag, const gsl_complex alpha, const gsl_matrix_complex * A, gsl_matrix_complex * B) These functions compute the matrix-matrix product $B = \alpha op(A)B$ for Side is CblasLeft and $B = \alpha Bop(A)$ for Side is CblasRight. The matrix A is triangular and op(A) = A, A^T , A^H for TransA = CblasNoTrans, CblasTrans, CblasConjTrans. When Uplo is CblasUpper then the upper triangle of A is used, and when Uplo is CblasLower then the lower triangle of A is used. If Diag is CblasNonUnit then the diagonal of A is used, but if Diag is CblasUnit then the diagonal elements of the
- int gsl_blas_strsm (CBLAS_SIDE_t Side, CBLAS_UPLO_t Uplo, [Function] CBLAS_TRANSPOSE_t TransA, CBLAS_DIAG_t Diag, float alpha, const gsl_matrix_float * A, gsl_matrix_float * B)

matrix A are taken as unity and are not referenced.

- int gsl_blas_ztrsm (CBLAS_SIDE_t Side, CBLAS_UPLO_t Uplo, [Function]

 CBLAS_TRANSPOSE_t TransA, CBLAS_DIAG_t Diag, const gsl_complex

 alpha, const gsl_matrix_complex * A, gsl_matrix_complex * B)

These functions compute the inverse-matrix matrix product $B = \alpha op(inv(A))B$ for Side is CblasLeft and $B = \alpha Bop(inv(A))$ for Side is CblasRight. The matrix A is triangular and op(A) = A, A^T , A^H for TransA = CblasNoTrans, CblasConjTrans. When Uplo is CblasUpper then the upper triangle of A is used, and when Uplo is CblasLower then the lower triangle of A is used. If Diag is CblasNonUnit then the diagonal of A is used, but if Diag is CblasUnit then the diagonal elements of the matrix A are taken as unity and are not referenced.

- int gsl_blas_ssyrk (CBLAS_UPLO_t Uplo, CBLAS_TRANSPOSE_t [Function]

 Trans, float alpha, const gsl_matrix_float * A, float beta, gsl_matrix_float * C)
- int gsl_blas_dsyrk (CBLAS_UPLO_t Uplo, CBLAS_TRANSPOSE_t [Function]

 Trans, double alpha, const gsl_matrix * A, double beta, gsl_matrix * C)
- int gsl_blas_zsyrk (CBLAS_UPLO_t Uplo, CBLAS_TRANSPOSE_t [Function]
 Trans, const gsl_complex alpha, const gsl_matrix_complex * A, const
 gsl_complex beta, gsl_matrix_complex * C)
 - These functions compute a rank-k update of the symmetric matrix C, $C = \alpha AA^T + \beta C$ when Trans is CblasNoTrans and $C = \alpha A^T A + \beta C$ when Trans is CblasTrans. Since the matrix C is symmetric only its upper half or lower half need to be stored. When Uplo is CblasUpper then the upper triangle and diagonal of C are used, and when Uplo is CblasLower then the lower triangle and diagonal of C are used.
- - These functions compute a rank-k update of the hermitian matrix C, $C = \alpha AA^H + \beta C$ when Trans is CblasNoTrans and $C = \alpha A^H A + \beta C$ when Trans is CblasConjTrans. Since the matrix C is hermitian only its upper half or lower half need to be stored. When Uplo is CblasUpper then the upper triangle and diagonal of C are used, and when Uplo is CblasLower then the lower triangle and diagonal of C are used. The imaginary elements of the diagonal are automatically set to zero.

- int gsl_blas_zsyr2k (CBLAS_UPLO_t Uplo, CBLAS_TRANSPOSE_t [Function] Trans, const gsl_complex alpha, const gsl_matrix_complex * A, const gsl_matrix_complex * B, const gsl_complex beta, gsl_matrix_complex * C) These functions compute a rank-2k update of the symmetric matrix C, $C = \alpha AB^T + \alpha BA^T + \beta C$ when Trans is CblasNoTrans and $C = \alpha A^TB + \alpha B^TA + \beta C$ when Trans is CblasTrans. Since the matrix C is symmetric only its upper half or lower half need to be stored. When Uplo is CblasUpper then the upper triangle and diagonal of C are used, and when Uplo is CblasLower then the lower triangle and diagonal of C are used.

int gsl_blas_cher2k (CBLAS_UPLO_t Uplo, CBLAS_TRANSPOSE_t [Function] Trans, const gsl_complex_float alpha, const gsl_matrix_complex_float * A, const gsl_matrix_complex_float * B, float beta, gsl_matrix_complex_float * C) int gsl_blas_zher2k (CBLAS_UPLO_t Uplo, CBLAS_TRANSPOSE_t [Function] Trans, const gsl_complex alpha, const gsl_matrix_complex * A, const gsl_matrix_complex * B, double beta, gsl_matrix_complex * C) These functions compute a rank-2k update of the hermitian matrix C, $C = \alpha AB^H + \alpha^*BA^H + \beta C$ when Trans is CblasNoTrans and $C = \alpha A^HB + \alpha^*B^HA + \beta C$ when Trans is CblasConjTrans. Since the matrix C is hermitian only its upper half or lower half need to be stored. When Uplo is CblasUpper then the upper triangle and

These functions compute a rank-2k update of the hermitian matrix C, $C = \alpha AB^{AB} + \alpha^*BA^H + \beta C$ when Trans is CblasNoTrans and $C = \alpha A^HB + \alpha^*B^HA + \beta C$ when Trans is CblasConjTrans. Since the matrix C is hermitian only its upper half or lower half need to be stored. When Uplo is CblasUpper then the upper triangle and diagonal of C are used, and when Uplo is CblasLower then the lower triangle and diagonal of C are used. The imaginary elements of the diagonal are automatically set to zero.

13.2 Examples

The following program computes the product of two matrices using the Level-3 BLAS function DGEMM,

$$\begin{pmatrix} 0.11 & 0.12 & 0.13 \\ 0.21 & 0.22 & 0.23 \end{pmatrix} \begin{pmatrix} 1011 & 1012 \\ 1021 & 1022 \\ 1031 & 1031 \end{pmatrix} = \begin{pmatrix} 367.76 & 368.12 \\ 674.06 & 674.72 \end{pmatrix}$$

The matrices are stored in row major order, according to the C convention for arrays.

```
#include <stdio.h>
#include <gsl/gsl_blas.h>
int
main (void)
{
  double a[] = \{ 0.11, 0.12, 0.13, 
                 0.21, 0.22, 0.23;
  double b[] = \{ 1011, 1012, \}
                 1021, 1022,
                 1031, 1032 };
  double c[] = { 0.00, 0.00, }
                 0.00. 0.00 h:
  gsl_matrix_view A = gsl_matrix_view_array(a, 2, 3);
  gsl_matrix_view B = gsl_matrix_view_array(b, 3, 2);
  gsl_matrix_view C = gsl_matrix_view_array(c, 2, 2);
  /* Compute C = A B */
  gsl_blas_dgemm (CblasNoTrans, CblasNoTrans,
                  1.0, &A.matrix, &B.matrix,
```

```
0.0, &C.matrix);
printf ("[ %g, %g\n", c[0], c[1]);
printf (" %g, %g ]\n", c[2], c[3]);

return 0;
}
Here is the output from the program,
$ ./a.out
[ 367.76, 368.12
674.06, 674.72 ]
```

13.3 References and Further Reading

Information on the BLAS standards, including both the legacy and updated interface standards, is available online from the BLAS Homepage and BLAS Technical Forum web-site.

```
BLAS Homepage
http://www.netlib.org/blas/
BLAS Technical Forum
http://www.netlib.org/blas/blast-forum/
```

The following papers contain the specifications for Level 1, Level 2 and Level 3 BLAS.

- C. Lawson, R. Hanson, D. Kincaid, F. Krogh, "Basic Linear Algebra Subprograms for Fortran Usage", *ACM Transactions on Mathematical Software*, Vol. 5 (1979), Pages 308–325.
- J.J. Dongarra, J. DuCroz, S. Hammarling, R. Hanson, "An Extended Set of Fortran Basic Linear Algebra Subprograms", *ACM Transactions on Mathematical Software*, Vol. 14, No. 1 (1988), Pages 1–32.
- J.J. Dongarra, I. Duff, J. DuCroz, S. Hammarling, "A Set of Level 3 Basic Linear Algebra Subprograms", *ACM Transactions on Mathematical Software*, Vol. 16 (1990), Pages 1–28.

Postscript versions of the latter two papers are available from http://www.netlib.org/blas/. A CBLAS wrapper for Fortran BLAS libraries is available from the same location.

14 Linear Algebra

This chapter describes functions for solving linear systems. The library provides linear algebra operations which operate directly on the gsl_vector and gsl_matrix objects. These routines use the standard algorithms from Golub & Van Loan's *Matrix Computations* with Level-1 and Level-2 BLAS calls for efficiency.

The functions described in this chapter are declared in the header file gsl_linalg.h.

14.1 LU Decomposition

A general square matrix A has an LU decomposition into upper and lower triangular matrices,

$$PA = LU$$

where P is a permutation matrix, L is unit lower triangular matrix and U is upper triangular matrix. For square matrices this decomposition can be used to convert the linear system Ax = b into a pair of triangular systems (Ly = Pb, Ux = y), which can be solved by forward and back-substitution. Note that the LU decomposition is valid for singular matrices.

These functions factorize the square matrix A into the LU decomposition PA = LU. On output the diagonal and upper triangular part of the input matrix A contain the matrix U. The lower triangular part of the input matrix (excluding the diagonal) contains L. The diagonal elements of L are unity, and are not stored.

The permutation matrix P is encoded in the permutation p. The j-th column of the matrix P is given by the k-th column of the identity matrix, where $k = p_j$ the j-th element of the permutation vector. The sign of the permutation is given by signum. It has the value $(-1)^n$, where n is the number of interchanges in the permutation.

The algorithm used in the decomposition is Gaussian Elimination with partial pivoting (Golub & Van Loan, *Matrix Computations*, Algorithm 3.4.1).

int gsl_linalg_LU_solve (const gsl_matrix * LU, const gsl_permutation * p, const gsl_vector * b, gsl_vector * x) [Function]

These functions solve the square system Ax = b using the LU decomposition of A into (LU, p) given by gsl_linalg_LU_decomp or gsl_linalg_complex_LU_decomp as input.

int gsl_linalg_complex_LU_svx (const gsl_matrix_complex * LU, const gsl_permutation * p, gsl_vector_complex * x) [Function]

These functions solve the square system Ax = b in-place using the precomputed LU decomposition of A into (LU,p). On input x should contain the right-hand side b, which is replaced by the solution on output.

These functions apply an iterative improvement to x, the solution of Ax = b, from the precomputed LU decomposition of A into (LU,p). The initial residual r = Ax - b is also computed and stored in residual.

These functions compute the inverse of a matrix A from its LU decomposition (LU,p), storing the result in the matrix *inverse*. The inverse is computed by solving the system Ax = b for each column of the identity matrix. It is preferable to avoid direct use of the inverse whenever possible, as the linear solver functions can obtain the same result more efficiently and reliably (consult any introductory textbook on numerical linear algebra for details).

These functions compute the determinant of a matrix A from its LU decomposition, LU. The determinant is computed as the product of the diagonal elements of U and the sign of the row permutation signum.

double gsl_linalg_LU_lndet ($gsl_matrix * LU$) [Function] double gsl_linalg_complex_LU_lndet ($gsl_matrix_complex * LU$) [Function] These functions compute the logarithm of the absolute value of the determinant of a matrix A, $\ln |\det(A)|$, from its LU decomposition, LU. This function may be useful if the direct computation of the determinant would overflow or underflow.

These functions compute the sign or phase factor of the determinant of a matrix A, det(A)/|det(A)|, from its LU decomposition, LU.

14.2 QR Decomposition

A general rectangular M-by-N matrix A has a QR decomposition into the product of an orthogonal M-by-M square matrix Q (where $Q^TQ = I$) and an M-by-N right-triangular matrix R,

$$A = QR$$

This decomposition can be used to convert the linear system Ax = b into the triangular system $Rx = Q^Tb$, which can be solved by back-substitution. Another use of the QR decomposition is to compute an orthonormal basis for a set of vectors. The first N columns of Q form an orthonormal basis for the range of A, ran(A), when A has full column rank.

int gsl_linalg_QR_decomp (gsl_matrix * A, gsl_vector * tau) [Function] This function factorizes the M-by-N matrix A into the QR decomposition A = QR. On output the diagonal and upper triangular part of the input matrix contain the matrix R. The vector tau and the columns of the lower triangular part of the matrix A contain the Householder coefficients and Householder vectors which encode the orthogonal matrix Q. The vector tau must be of length $k = \min(M, N)$. The matrix Q is related to these components by, $Q = Q_k...Q_2Q_1$ where $Q_i = I - \tau_i v_i v_i^T$ and v_i is the Householder vector $v_i = (0, ..., 1, A(i+1,i), A(i+2,i), ..., A(m,i))$. This is the same storage scheme as used by LAPACK.

The algorithm used to perform the decomposition is Householder QR (Golub & Van Loan, *Matrix Computations*, Algorithm 5.2.1).

This function solves the square system Ax = b using the QR decomposition of A held in (QR, tau) which must have been computed previously with $gsl_linalg_QR_decomp$. The least-squares solution for rectangular systems can be found using $gsl_linalg_QR_lssolve$.

This function solves the square system Ax = b in-place using the QR decomposition of A held in (QR,tau) which must have been computed previously by gsl_linalg_R QR_decomp. On input x should contain the right-hand side b, which is replaced by the solution on output.

int gsl_linalg_QR_lssolve (const gsl_matrix * QR, const gsl_vector * tau, const gsl_vector * b, gsl_vector * x, gsl_vector * residual) [Function]

This function finds the least squares solution to the overdetermined system Ax = b where the matrix A has more rows than columns. The least squares solution minimizes the Euclidean norm of the residual, ||Ax - b||. The routine requires as input the QR decomposition of A into (QR, tau) given by $gsl_linalg_QR_decomp$. The solution is returned in x. The residual is computed as a by-product and stored in residual.

This function applies the matrix Q^T encoded in the decomposition (QR,tau) to the vector v, storing the result Q^Tv in v. The matrix multiplication is carried out directly

using the encoding of the Householder vectors without needing to form the full matrix Q^T .

This function applies the matrix Q encoded in the decomposition (QR,tau) to the vector v, storing the result Qv in v. The matrix multiplication is carried out directly using the encoding of the Householder vectors without needing to form the full matrix Q.

This function applies the matrix Q^T encoded in the decomposition (QR,tau) to the matrix A, storing the result Q^TA in A. The matrix multiplication is carried out directly using the encoding of the Householder vectors without needing to form the full matrix Q^T .

- - This function solves the triangular system Rx = b for x. It may be useful if the product $b' = Q^T b$ has already been computed using gsl_linalg_QR_QTvec.
- int gsl_linalg_QR_Rsvx (const gsl_matrix * QR, gsl_vector * x) [Function] This function solves the triangular system Rx = b for x in-place. On input x should contain the right-hand side b and is replaced by the solution on output. This function may be useful if the product $b' = Q^T b$ has already been computed using gsl_linalg_QR_QTvec.
- int gsl_linalg_QR_unpack (const gsl_matrix * QR, const gsl_vector * tau, gsl_matrix * Q, gsl_matrix * R) [Function]

This function unpacks the encoded QR decomposition (QR,tau) into the matrices Q and R, where Q is M-by-M and R is M-by-N.

- - This function solves the system $Rx = Q^T b$ for x. It can be used when the QR decomposition of a matrix is available in unpacked form as (Q, R).
- int gsl_linalg_QR_update (gsl_matrix * Q, gsl_matrix * R, gsl_vector *
 w, const gsl_vector * v)
 [Function]

This function performs a rank-1 update wv^T of the QR decomposition (Q, R). The update is given by $Q'R' = Q(R + wv^T)$ where the output matrices Q' and R' are also orthogonal and right triangular. Note that w is destroyed by the update.

This function solves the triangular system Rx = b for the N-by-N matrix R.

int gsl_linalg_R_svx (const gsl_matrix * R, gsl_vector * x) [Function] This function solves the triangular system Rx = b in-place. On input x should contain the right-hand side b, which is replaced by the solution on output.

14.3 QR Decomposition with Column Pivoting

The QR decomposition can be extended to the rank deficient case by introducing a column permutation P,

$$AP = QR$$

The first r columns of Q form an orthonormal basis for the range of A for a matrix with column rank r. This decomposition can also be used to convert the linear system Ax = b into the triangular system $Ry = Q^Tb, x = Py$, which can be solved by back-substitution and permutation. We denote the QR decomposition with column pivoting by QRP^T since $A = QRP^T$.

int gsl_linalg_QRPT_decomp (gsl_matrix * A, gsl_vector * tau, gsl_permutation * p, int * signum, gsl_vector * norm) [Function]

This function factorizes the M-by-N matrix A into the QRP^T decomposition $A = QRP^T$. On output the diagonal and upper triangular part of the input matrix contain the matrix R. The permutation matrix P is stored in the permutation p. The sign of the permutation is given by signum. It has the value $(-1)^n$, where n is the number of interchanges in the permutation. The vector tau and the columns of the lower triangular part of the matrix A contain the Householder coefficients and vectors which encode the orthogonal matrix Q. The vector tau must be of length $k = \min(M, N)$. The matrix Q is related to these components by, $Q = Q_k ... Q_2 Q_1$ where $Q_i = I - \tau_i v_i v_i^T$ and v_i is the Householder vector $v_i = (0, ..., 1, A(i+1, i), A(i+2, i), ..., A(m, i))$. This is the same storage scheme as used by LAPACK. The vector norm is a workspace of length N used for column pivoting.

The algorithm used to perform the decomposition is Householder QR with column pivoting (Golub & Van Loan, *Matrix Computations*, Algorithm 5.4.1).

This function factorizes the matrix A into the decomposition $A = QRP^T$ without modifying A itself and storing the output in the separate matrices q and r.

- int gsl_linalg_QRPT_solve (const gsl_matrix * QR, const gsl_vector * [Function] tau, const gsl_permutation * p, const gsl_vector * b, gsl_vector * x)

 This function solves the square system Ax = b using the QRP^T decomposition of A held in (QR, tau, p) which must have been computed previously by gsl_linalg_QRPT_decomp.
- int gsl_linalg_QRPT_svx (const gsl_matrix * QR, const gsl_vector * [Function] tau, const gsl_permutation * p, gsl_vector * x)

 This function solves the square system Ax = b in-place using the QRP^T decomposition of A held in (QR, tau, p). On input x should contain the right-hand side b, which is replaced by the solution on output.
- int gsl_linalg_QRPT_QRsolve (const gsl_matrix * Q, const gsl_matrix [Function] * R, const gsl_permutation * p, const gsl_vector * b, gsl_vector * x)

 This function solves the square system $RP^Tx = Q^Tb$ for x. It can be used when the QR decomposition of a matrix is available in unpacked form as (Q, R).

int gsl_linalg_QRPT_update (gsl_matrix * Q, gsl_matrix * R, const gsl_permutation * p, gsl_vector * w, const gsl_vector * v) [Function]

This function performs a rank-1 update wv^T of the QRP^T decomposition (Q, R, p). The update is given by $Q'R' = Q(R + wv^TP)$ where the output matrices Q' and R' are also orthogonal and right triangular. Note that w is destroyed by the update. The permutation p is not changed.

int gsl_linalg_QRPT_Rsolve (const gsl_matrix * QR, const gsl_permutation * p, const gsl_vector * b, gsl_vector * x) [Function]

This function solves the triangular system $RP^Tx = b$ for the N-by-N matrix R contained in QR.

This function solves the triangular system $RP^Tx = b$ in-place for the N-by-N matrix R contained in QR. On input x should contain the right-hand side b, which is replaced by the solution on output.

14.4 Singular Value Decomposition

A general rectangular M-by-N matrix A has a singular value decomposition (SVD) into the product of an M-by-N orthogonal matrix U, an N-by-N diagonal matrix of singular values S and the transpose of an N-by-N orthogonal square matrix V,

$$A = USV^T$$

The singular values $\sigma_i = S_{ii}$ are all non-negative and are generally chosen to form a non-increasing sequence $\sigma_1 \geq \sigma_2 \geq ... \geq \sigma_N \geq 0$.

The singular value decomposition of a matrix has many practical uses. The condition number of the matrix is given by the ratio of the largest singular value to the smallest singular value. The presence of a zero singular value indicates that the matrix is singular. The number of non-zero singular values indicates the rank of the matrix. In practice singular value decomposition of a rank-deficient matrix will not produce exact zeroes for singular values, due to finite numerical precision. Small singular values should be edited by choosing a suitable tolerance.

For a rank-deficient matrix, the null space of A is given by the columns of V corresponding to the zero singular values. Similarly, the range of A is given by columns of U corresponding to the non-zero singular values.

Note that the routines here compute the "thin" version of the SVD with U as M-by-N orthogonal matrix. This allows in-place computation and is the most commonly-used form in practice. Mathematically, the "full" SVD is defined with U as an M-by-M orthogonal matrix and S as an M-by-N diagonal matrix (with additional rows of zeros).

This function factorizes the M-by-N matrix A into the singular value decomposition $A = USV^T$ for $M \geq N$. On output the matrix A is replaced by U. The diagonal elements of the singular value matrix S are stored in the vector S. The singular values are non-negative and form a non-increasing sequence from S_1 to S_N . The matrix V

contains the elements of V in untransposed form. To form the product USV^T it is necessary to take the transpose of V. A workspace of length N is required in work. This routine uses the Golub-Reinsch SVD algorithm.

This function computes the SVD using the modified Golub-Reinsch algorithm, which is faster for $M \gg N$. It requires the vector work of length N and the N-by-N matrix X as additional working space.

This function computes the SVD of the M-by-N matrix A using one-sided Jacobi orthogonalization for $M \geq N$. The Jacobi method can compute singular values to higher relative accuracy than Golub-Reinsch algorithms (see references for details).

int gsl_linalg_SV_solve (const gsl_matrix * U, const gsl_matrix * V, const gsl_vector * S, const gsl_vector * b, gsl_vector * x) [Function]

This function solves the system Ax = b using the singular value decomposition (U, S, V) of A which must have been computed previously with gsl_linalg_SV_decomp.

Only non-zero singular values are used in computing the solution. The parts of the solution corresponding to singular values of zero are ignored. Other singular values can be edited out by setting them to zero before calling this function.

In the over-determined case where A has more rows than columns the system is solved in the least squares sense, returning the solution x which minimizes $||Ax - b||_2$.

int gsl_linalg_SV_leverage (const gsl_matrix * U, gsl_vector * h) [Function] This function computes the statistical leverage values h_i of a matrix A using its singular value decomposition (U, S, V) previously computed with gsl_linalg_SV_decomp. h_i are the diagonal values of the matrix $A(A^TA)^{-1}A^T$ and depend only on the matrix U which is the input to this function.

14.5 Cholesky Decomposition

A symmetric, positive definite square matrix A has a Cholesky decomposition into a product of a lower triangular matrix L and its transpose L^T ,

$$A = LL^T$$

This is sometimes referred to as taking the square-root of a matrix. The Cholesky decomposition can only be carried out when all the eigenvalues of the matrix are positive. This decomposition can be used to convert the linear system Ax = b into a pair of triangular systems $(Ly = b, L^Tx = y)$, which can be solved by forward and back-substitution.

int gsl_linalg_cholesky_decomp ($gsl_matrix * A$) [Function] int gsl_linalg_complex_cholesky_decomp ($gsl_matrix_complex * A$) [Function] These functions factorize the symmetric, positive-definite square matrix A into the Cholesky decomposition $A = LL^T$ (or $A = LL^\dagger$ for the complex case). On input, the values from the diagonal and lower-triangular part of the matrix A are used (the

upper triangular part is ignored). On output the diagonal and lower triangular part of the input matrix A contain the matrix L, while the upper triangular part of the input matrix is overwritten with L^T (the diagonal terms being identical for both L and L^T). If the matrix is not positive-definite then the decomposition will fail, returning the error code GSL_EDOM .

When testing whether a matrix is positive-definite, disable the error handler first to avoid triggering an error.

These functions solve the system Ax = b using the Cholesky decomposition of A held in the matrix *cholesky* which must have been previously computed by gsl_linalg_cholesky_decomp or gsl_linalg_complex_cholesky_decomp.

These functions solve the system Ax = b in-place using the Cholesky decomposition of A held in the matrix *cholesky* which must have been previously computed by $gsl_linalg_cholesky_decomp$ or $gsl_linalg_complex_cholesky_decomp$. On input x should contain the right-hand side b, which is replaced by the solution on output.

These functions compute the inverse of a matrix from its Cholesky decomposition *cholesky*, which must have been previously computed by gsl_linalg_cholesky_decomp or gsl_linalg_complex_cholesky_decomp. On output, the inverse is stored in-place in *cholesky*.

14.6 Tridiagonal Decomposition of Real Symmetric Matrices

A symmetric matrix A can be factorized by similarity transformations into the form,

$$A = QTQ^T$$

where Q is an orthogonal matrix and T is a symmetric tridiagonal matrix.

int gsl_linalg_symmtd_decomp ($gsl_matrix * A$, $gsl_vector * tau$) [Function] This function factorizes the symmetric square matrix A into the symmetric tridiagonal decomposition QTQ^T . On output the diagonal and subdiagonal part of the input matrix A contain the tridiagonal matrix T. The remaining lower triangular part of the input matrix contains the Householder vectors which, together with the Householder coefficients tau, encode the orthogonal matrix Q. This storage scheme is the same as used by LAPACK. The upper triangular part of A is not referenced.

int gsl_linalg_symmtd_unpack (const gsl_matrix * A, const gsl_vector [Function]
* tau, gsl_matrix * Q, gsl_vector * diag, gsl_vector * subdiag)

This function unpacks the encoded symmetric tridiagonal decomposition (A, tau) obtained from gsl_linalg_symmtd_decomp into the orthogonal matrix Q, the vector of diagonal elements diag and the vector of subdiagonal elements subdiag.

This function unpacks the diagonal and subdiagonal of the encoded symmetric tridiagonal decomposition (A, tau) obtained from gsl_linalg_symmtd_decomp into the vectors diag and subdiag.

14.7 Tridiagonal Decomposition of Hermitian Matrices

A hermitian matrix A can be factorized by similarity transformations into the form,

$$A = UTU^T$$

where U is a unitary matrix and T is a real symmetric tridiagonal matrix.

int gsl_linalg_hermtd_decomp (gsl_matrix_complex * A, gsl_vector_complex * tau) [Function]

This function factorizes the hermitian matrix A into the symmetric tridiagonal decomposition UTU^T . On output the real parts of the diagonal and subdiagonal part of the input matrix A contain the tridiagonal matrix T. The remaining lower triangular part of the input matrix contains the Householder vectors which, together with the Householder coefficients tau, encode the unitary matrix U. This storage scheme is the same as used by LAPACK. The upper triangular part of A and imaginary parts of the diagonal are not referenced.

This function unpacks the encoded tridiagonal decomposition (A, tau) obtained from $gsl_linalg_hermtd_decomp$ into the unitary matrix U, the real vector of diagonal elements diag and the real vector of subdiagonal elements subdiag.

int gsl_linalg_hermtd_unpack_T (const gsl_matrix_complex * A, gsl_vector * diag, gsl_vector * subdiag) [Function]

This function unpacks the diagonal and subdiagonal of the encoded tridiagonal decomposition (A, tau) obtained from the gsl_linalg_hermtd_decomp into the real vectors diag and subdiag.

14.8 Hessenberg Decomposition of Real Matrices

A general real matrix A can be decomposed by orthogonal similarity transformations into the form

$$A = UHU^T$$

where U is orthogonal and H is an upper Hessenberg matrix, meaning that it has zeros below the first subdiagonal. The Hessenberg reduction is the first step in the Schur decomposition for the nonsymmetric eigenvalue problem, but has applications in other areas as well.

- int gsl_linalg_hessenberg_decomp ($gsl_matrix * A, gsl_vector * tau$) [Function] This function computes the Hessenberg decomposition of the matrix A by applying the similarity transformation $H = U^T AU$. On output, H is stored in the upper portion of A. The information required to construct the matrix U is stored in the lower triangular portion of A. U is a product of N-2 Householder matrices. The Householder vectors are stored in the lower portion of A (below the subdiagonal) and the Householder coefficients are stored in the vector tau. tau must be of length N.
- int gsl_linalg_hessenberg_unpack (gsl_matrix * H, gsl_vector * tau, [Function] gsl_matrix * U)

This function constructs the orthogonal matrix U from the information stored in the Hessenberg matrix H along with the vector tau. H and tau are outputs from $gsl_linalg_hessenberg_decomp$.

This function is similar to $gsl_linalg_hessenberg_unpack$, except it accumulates the matrix U into V, so that V' = VU. The matrix V must be initialized prior to calling this function. Setting V to the identity matrix provides the same result as $gsl_linalg_hessenberg_unpack$. If H is order N, then V must have N columns but may have any number of rows.

int gsl_linalg_hessenberg_set_zero (gsl_matrix * H) [Function]
This function sets the lower triangular portion of H, below the subdiagonal, to zero. It is useful for clearing out the Householder vectors after calling gsl_linalg_hessenberg_decomp.

14.9 Hessenberg-Triangular Decomposition of Real Matrices

A general real matrix pair (A, B) can be decomposed by orthogonal similarity transformations into the form

$$A = UHV^T$$

$$B = URV^T$$

where U and V are orthogonal, H is an upper Hessenberg matrix, and R is upper triangular. The Hessenberg-Triangular reduction is the first step in the generalized Schur decomposition for the generalized eigenvalue problem.

This function computes the Hessenberg-Triangular decomposition of the matrix pair (A, B). On output, H is stored in A, and R is stored in B. If U and V are provided (they may be null), the similarity transformations are stored in them. Additional workspace of length N is needed in work.

14.10 Bidiagonalization

A general matrix A can be factorized by similarity transformations into the form,

$$A = UBV^T$$

where U and V are orthogonal matrices and B is a N-by-N bidiagonal matrix with non-zero entries only on the diagonal and superdiagonal. The size of U is M-by-N and the size of V is N-by-N.

This function factorizes the M-by-N matrix A into bidiagonal form UBV^T . The diagonal and superdiagonal of the matrix B are stored in the diagonal and superdiagonal of A. The orthogonal matrices U and V are stored as compressed Householder vectors in the remaining elements of A. The Householder coefficients are stored in the vectors $tau_{-}U$ and $tau_{-}V$. The length of $tau_{-}U$ must equal the number of elements in the diagonal of A and the length of $tau_{-}V$ should be one element shorter.

This function unpacks the bidiagonal decomposition of A produced by $gsl_linalg_bidiag_decomp$, (A, tau_U, tau_V) into the separate orthogonal matrices U, V and the diagonal vector diag and superdiagonal superdiag. Note that U is stored as a compact M-by-N orthogonal matrix satisfying $U^TU = I$ for efficiency.

int gsl_linalg_bidiag_unpack2 (gsl_matrix * A, gsl_vector * tau_U, [Function] gsl_vector * tau_V, gsl_matrix * V)

This function unpacks the bidiagonal decomposition of A produced by $gsl_linalg_bidiag_decomp$, (A, tau_U, tau_V) into the separate orthogonal matrices U, V and the diagonal vector diag and superdiagonal superdiag. The matrix U is stored in-place in A.

This function unpacks the diagonal and superdiagonal of the bidiagonal decomposition of A from gsl_linalg_bidiag_decomp, into the diagonal vector diag and superdiagonal vector superdiag.

14.11 Householder Transformations

A Householder transformation is a rank-1 modification of the identity matrix which can be used to zero out selected elements of a vector. A Householder matrix P takes the form,

$$P = I - \tau v v^T$$

where v is a vector (called the Householder vector) and $\tau = 2/(v^T v)$. The functions described in this section use the rank-1 structure of the Householder matrix to create and apply Householder transformations efficiently.

 $\begin{array}{lll} \mbox{double gsl_linalg_householder_transform } (gsl_vector * v) & [Function] \\ \mbox{gsl_complex gsl_linalg_complex_householder_transform} & [Function] \\ \mbox{} (gsl_vector_complex * v) & [Function] \\ \end{array}$

This function prepares a Householder transformation $P = I - \tau v v^T$ which can be used to zero all the elements of the input vector except the first. On output the transformation is stored in the vector v and the scalar τ is returned.

- int gsl_linalg_complex_householder_hm (gsl_complex tau, const gsl_vector_complex * v, gsl_matrix_complex * A) [Function]

This function applies the Householder matrix P defined by the scalar tau and the vector v to the left-hand side of the matrix A. On output the result PA is stored in A.

This function applies the Householder matrix P defined by the scalar tau and the vector v to the right-hand side of the matrix A. On output the result AP is stored in A.

This function applies the Householder transformation P defined by the scalar tau and the vector v to the vector w. On output the result Pw is stored in w.

14.12 Householder solver for linear systems

This function solves the system Ax = b directly using Householder transformations. On output the solution is stored in x and b is not modified. The matrix A is destroyed by the Householder transformations.

int gsl_linalg_HH_svx ($gsl_matrix * A$, $gsl_vector * x$) [Function] This function solves the system Ax = b in-place using Householder transformations. On input x should contain the right-hand side b, which is replaced by the solution on output. The matrix A is destroyed by the Householder transformations.

14.13 Tridiagonal Systems

The functions described in this section efficiently solve symmetric, non-symmetric and cyclic tridiagonal systems with minimal storage. Note that the current implementations of these functions use a variant of Cholesky decomposition, so the tridiagonal matrix must be positive definite. For non-positive definite matrices, the functions return the error code GSL_ESING.

int gsl_linalg_solve_tridiag (const gsl_vector * diag, const gsl_vector * e, const gsl_vector * f, const gsl_vector * b, gsl_vector * x)

This function solves the general N-by-N system Ax = b where A is tridiagonal ($N \ge 2$). The super-diagonal and sub-diagonal vectors e and f must be one element shorter than the diagonal vector diag. The form of A for the 4-by-4 case is shown below,

$$A = \begin{pmatrix} d_0 & e_0 & 0 & 0 \\ f_0 & d_1 & e_1 & 0 \\ 0 & f_1 & d_2 & e_2 \\ 0 & 0 & f_2 & d_3 \end{pmatrix}$$

int gsl_linalg_solve_symm_tridiag (const gsl_vector * diag, const gsl_vector * e, const gsl_vector * b, gsl_vector * x) [Function]

This function solves the general N-by-N system Ax = b where A is symmetric tridiagonal $(N \ge 2)$. The off-diagonal vector e must be one element shorter than the diagonal vector diag. The form of A for the 4-by-4 case is shown below,

$$A = \begin{pmatrix} d_0 & e_0 & 0 & 0 \\ e_0 & d_1 & e_1 & 0 \\ 0 & e_1 & d_2 & e_2 \\ 0 & 0 & e_2 & d_3 \end{pmatrix}$$

This function solves the general N-by-N system Ax = b where A is cyclic tridiagonal $(N \ge 3)$. The cyclic super-diagonal and sub-diagonal vectors e and f must have the same number of elements as the diagonal vector diag. The form of A for the 4-by-4 case is shown below,

$$A = \begin{pmatrix} d_0 & e_0 & 0 & f_3 \\ f_0 & d_1 & e_1 & 0 \\ 0 & f_1 & d_2 & e_2 \\ e_3 & 0 & f_2 & d_3 \end{pmatrix}$$

This function solves the general N-by-N system Ax = b where A is symmetric cyclic tridiagonal $(N \ge 3)$. The cyclic off-diagonal vector e must have the same number of elements as the diagonal vector diag. The form of A for the 4-by-4 case is shown below,

$$A = \begin{pmatrix} d_0 & e_0 & 0 & e_3 \\ e_0 & d_1 & e_1 & 0 \\ 0 & e_1 & d_2 & e_2 \\ e_3 & 0 & e_2 & d_3 \end{pmatrix}$$

14.14 Balancing

The process of balancing a matrix applies similarity transformations to make the rows and columns have comparable norms. This is useful, for example, to reduce roundoff errors in

the solution of eigenvalue problems. Balancing a matrix A consists of replacing A with a similar matrix

$$A' = D^{-1}AD$$

where D is a diagonal matrix whose entries are powers of the floating point radix.

int gsl_linalg_balance_matrix ($gsl_matrix * A$, $gsl_vector * D$) [Function] This function replaces the matrix A with its balanced counterpart and stores the diagonal elements of the similarity transformation into the vector D.

14.15 Examples

The following program solves the linear system Ax = b. The system to be solved is,

$$\begin{pmatrix} 0.18 & 0.60 & 0.57 & 0.96 \\ 0.41 & 0.24 & 0.99 & 0.58 \\ 0.14 & 0.30 & 0.97 & 0.66 \\ 0.51 & 0.13 & 0.19 & 0.85 \end{pmatrix} \begin{pmatrix} x_0 \\ x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 1.0 \\ 2.0 \\ 3.0 \\ 4.0 \end{pmatrix}$$

and the solution is found using LU decomposition of the matrix A.

```
#include <stdio.h>
#include <gsl/gsl_linalg.h>
int
main (void)
{
  double a_data[] = \{ 0.18, 0.60, 0.57, 0.96, 
                      0.41, 0.24, 0.99, 0.58,
                      0.14, 0.30, 0.97, 0.66,
                      0.51, 0.13, 0.19, 0.85;
  double b_data[] = { 1.0, 2.0, 3.0, 4.0 };
  gsl_matrix_view m
    = gsl_matrix_view_array (a_data, 4, 4);
  gsl_vector_view b
    = gsl_vector_view_array (b_data, 4);
  gsl_vector *x = gsl_vector_alloc (4);
  int s;
  gsl_permutation * p = gsl_permutation_alloc (4);
  gsl_linalg_LU_decomp (&m.matrix, p, &s);
  gsl_linalg_LU_solve (&m.matrix, p, &b.vector, x);
```

```
printf ("x = \n");
    gsl_vector_fprintf (stdout, x, "%g");

    gsl_permutation_free (p);
    gsl_vector_free (x);
    return 0;
}

Here is the output from the program,
    x = -4.05205
    -12.6056
    1.66091
    8.69377
```

This can be verified by multiplying the solution x by the original matrix A using GNU OCTAVE,

This reproduces the original right-hand side vector, b, in accordance with the equation Ax = b.

14.16 References and Further Reading

Further information on the algorithms described in this section can be found in the following book,

G. H. Golub, C. F. Van Loan, *Matrix Computations* (3rd Ed, 1996), Johns Hopkins University Press, ISBN 0-8018-5414-8.

The LAPACK library is described in the following manual,

 $LAPACK\ Users'$ Guide (Third Edition, 1999), Published by SIAM, ISBN 0-89871-447-8.

```
http://www.netlib.org/lapack
```

The LAPACK source code can be found at the website above, along with an online copy of the users guide.

The Modified Golub-Reinsch algorithm is described in the following paper,

T.F. Chan, "An Improved Algorithm for Computing the Singular Value Decomposition", ACM Transactions on Mathematical Software, 8 (1982), pp 72–83.

The Jacobi algorithm for singular value decomposition is described in the following papers,

- J.C. Nash, "A one-sided transformation method for the singular value decomposition and algebraic eigenproblem", $Computer\ Journal$, Volume 18, Number 1 (1975), p 74–76
- J.C. Nash and S. Shlien "Simple algorithms for the partial singular value decomposition", *Computer Journal*, Volume 30 (1987), p 268–275.

James Demmel, Krešimir Veselić, "Jacobi's Method is more accurate than QR", Lapack Working Note 15 (LAWN-15), October 1989. Available from netlib, http://www.netlib.org/lapack/ in the lawns or lawnspdf directories.

15 Eigensystems

This chapter describes functions for computing eigenvalues and eigenvectors of matrices. There are routines for real symmetric, real nonsymmetric, complex hermitian, real generalized symmetric-definite, complex generalized hermitian-definite, and real generalized nonsymmetric eigensystems. Eigenvalues can be computed with or without eigenvectors. The hermitian and real symmetric matrix algorithms are symmetric bidiagonalization followed by QR reduction. The nonsymmetric algorithm is the Francis QR double-shift. The generalized nonsymmetric algorithm is the QZ method due to Moler and Stewart.

The functions described in this chapter are declared in the header file gsl_eigen.h.

15.1 Real Symmetric Matrices

For real symmetric matrices, the library uses the symmetric bidiagonalization and QR reduction method. This is described in Golub & van Loan, section 8.3. The computed eigenvalues are accurate to an absolute accuracy of $\epsilon ||A||_2$, where ϵ is the machine precision.

This function allocates a workspace for computing eigenvalues of n-by-n real symmetric matrices. The size of the workspace is O(2n).

void gsl_eigen_symm_free (gsl_eigen_symm_workspace * w) [Function]
This function frees the memory associated with the workspace w.

This function computes the eigenvalues of the real symmetric matrix A. Additional workspace of the appropriate size must be provided in w. The diagonal and lower triangular part of A are destroyed during the computation, but the strict upper triangular part is not referenced. The eigenvalues are stored in the vector eval and are unordered.

This function allocates a workspace for computing eigenvalues and eigenvectors of n-by-n real symmetric matrices. The size of the workspace is O(4n).

void gsl_eigen_symmv_free (gsl_eigen_symmv_workspace * w) [Function]
This function frees the memory associated with the workspace w.

This function computes the eigenvalues and eigenvectors of the real symmetric matrix A. Additional workspace of the appropriate size must be provided in w. The diagonal and lower triangular part of A are destroyed during the computation, but the strict upper triangular part is not referenced. The eigenvalues are stored in the vector eval and are unordered. The corresponding eigenvectors are stored in the columns of the matrix evec. For example, the eigenvector in the first column corresponds to

the first eigenvalue. The eigenvectors are guaranteed to be mutually orthogonal and normalised to unit magnitude.

15.2 Complex Hermitian Matrices

For hermitian matrices, the library uses the complex form of the symmetric bidiagonalization and QR reduction method.

This function allocates a workspace for computing eigenvalues of n-by-n complex hermitian matrices. The size of the workspace is O(3n).

void gsl_eigen_herm_free (gsl_eigen_herm_workspace * w) [Function]
This function frees the memory associated with the workspace w.

This function computes the eigenvalues of the complex hermitian matrix A. Additional workspace of the appropriate size must be provided in w. The diagonal and lower triangular part of A are destroyed during the computation, but the strict upper triangular part is not referenced. The imaginary parts of the diagonal are assumed to be zero and are not referenced. The eigenvalues are stored in the vector eval and are unordered.

This function allocates a workspace for computing eigenvalues and eigenvectors of n-by-n complex hermitian matrices. The size of the workspace is O(5n).

void gsl_eigen_hermv_free (gsl_eigen_hermv_workspace * w) [Function]
This function frees the memory associated with the workspace w.

This function computes the eigenvalues and eigenvectors of the complex hermitian matrix A. Additional workspace of the appropriate size must be provided in w. The diagonal and lower triangular part of A are destroyed during the computation, but the strict upper triangular part is not referenced. The imaginary parts of the diagonal are assumed to be zero and are not referenced. The eigenvalues are stored in the vector eval and are unordered. The corresponding complex eigenvectors are stored in the columns of the matrix evec. For example, the eigenvector in the first column corresponds to the first eigenvalue. The eigenvectors are guaranteed to be mutually orthogonal and normalised to unit magnitude.

15.3 Real Nonsymmetric Matrices

The solution of the real nonsymmetric eigensystem problem for a matrix A involves computing the Schur decomposition

$$A = ZTZ^T$$

where Z is an orthogonal matrix of Schur vectors and T, the Schur form, is quasi upper triangular with diagonal 1-by-1 blocks which are real eigenvalues of A, and diagonal 2-by-2 blocks whose eigenvalues are complex conjugate eigenvalues of A. The algorithm used is the double-shift Francis method.

This function allocates a workspace for computing eigenvalues of n-by-n real nonsymmetric matrices. The size of the workspace is O(2n).

void gsl_eigen_nonsymm_free (gsl_eigen_nonsymm_workspace * w) [Function]
This function frees the memory associated with the workspace w.

This function sets some parameters which determine how the eigenvalue problem is solved in subsequent calls to gsl_eigen_nonsymm.

If $compute_t$ is set to 1, the full Schur form T will be computed by $gsl_eigen_nonsymm$. If it is set to 0, T will not be computed (this is the default setting). Computing the full Schur form T requires approximately 1.5–2 times the number of flops.

If balance is set to 1, a balancing transformation is applied to the matrix prior to computing eigenvalues. This transformation is designed to make the rows and columns of the matrix have comparable norms, and can result in more accurate eigenvalues for matrices whose entries vary widely in magnitude. See Section 14.14 [Balancing], page 145 for more information. Note that the balancing transformation does not preserve the orthogonality of the Schur vectors, so if you wish to compute the Schur vectors with gsl_eigen_nonsymm_Z you will obtain the Schur vectors of the balanced matrix instead of the original matrix. The relationship will be

$$T = Q^t D^{-1} A D Q$$

where Q is the matrix of Schur vectors for the balanced matrix, and D is the balancing transformation. Then $gsl_eigen_nonsymm_Z$ will compute a matrix Z which satisfies

$$T = Z^{-1}AZ$$

with Z = DQ. Note that Z will not be orthogonal. For this reason, balancing is not performed by default.

This function computes the eigenvalues of the real nonsymmetric matrix A and stores them in the vector eval. If T is desired, it is stored in the upper portion of A on output. Otherwise, on output, the diagonal of A will contain the 1-by-1 real eigenvalues and 2-by-2 complex conjugate eigenvalue systems, and the rest of A is destroyed. In rare cases, this function may fail to find all eigenvalues. If this happens, an error code is returned and the number of converged eigenvalues is stored in $w->n_evals$. The converged eigenvalues are stored in the beginning of eval.

int gsl_eigen_nonsymm_Z (gsl_matrix * A, gsl_vector_complex * eval, [Function] gsl_matrix * Z, gsl_eigen_nonsymm_workspace * w)

This function is identical to $gsl_eigen_nonsymm$ except that it also computes the Schur vectors and stores them into Z.

gsl_eigen_nonsymmv_workspace * gsl_eigen_nonsymmv_alloc [Function] (const size_t n)

This function allocates a workspace for computing eigenvalues and eigenvectors of n-by-n real nonsymmetric matrices. The size of the workspace is O(5n).

void gsl_eigen_nonsymmv_free (gsl_eigen_nonsymmv_workspace * w) [Function]
This function frees the memory associated with the workspace w.

void gsl_eigen_nonsymmv_params (const int balance, gsl_eigen_nonsymm_workspace * w) [Function]

This function sets parameters which determine how the eigenvalue problem is solved in subsequent calls to gsl_eigen_nonsymmv. If balance is set to 1, a balancing transformation is applied to the matrix. See gsl_eigen_nonsymm_params for more information. Balancing is turned off by default since it does not preserve the orthogonality of the Schur vectors.

int gsl_eigen_nonsymmv (gsl_matrix * A, gsl_vector_complex * eval, gsl_matrix_complex * evec, gsl_eigen_nonsymmv_workspace * w) [Function]

This function computes eigenvalues and right eigenvectors of the n-by-n real nonsymmetric matrix A. It first calls $\mathtt{gsl_eigen_nonsymm}$ to compute the eigenvalues, Schur form T, and Schur vectors. Then it finds eigenvectors of T and backtransforms them using the Schur vectors. The Schur vectors are destroyed in the process, but can be saved by using $\mathtt{gsl_eigen_nonsymmv_Z}$. The computed eigenvectors are normalized to have unit magnitude. On output, the upper portion of A contains the Schur form T. If $\mathtt{gsl_eigen_nonsymm}$ fails, no eigenvectors are computed, and an error code is returned.

int gsl_eigen_nonsymmv_Z (gsl_matrix * A, gsl_vector_complex * eval, [Function] gsl_matrix_complex * evec, gsl_matrix * Z, gsl_eigen_nonsymmv_workspace * w)

This function is identical to $gsl_eigen_nonsymmv$ except that it also saves the Schur vectors into Z.

15.4 Real Generalized Symmetric-Definite Eigensystems

The real generalized symmetric-definite eigenvalue problem is to find eigenvalues λ and eigenvectors x such that

$$Ax = \lambda Bx$$

where A and B are symmetric matrices, and B is positive-definite. This problem reduces to the standard symmetric eigenvalue problem by applying the Cholesky decomposition to

B:

$$Ax = \lambda Bx$$
$$Ax = \lambda LL^{t}x$$
$$(L^{-1}AL^{-t}) L^{t}x = \lambda L^{t}x$$

Therefore, the problem becomes $Cy = \lambda y$ where $C = L^{-1}AL^{-t}$ is symmetric, and $y = L^t x$. The standard symmetric eigensolver can be applied to the matrix C. The resulting eigenvectors are backtransformed to find the vectors of the original problem. The eigenvalues and eigenvectors of the generalized symmetric-definite eigenproblem are always real.

This function allocates a workspace for computing eigenvalues of n-by-n real generalized symmetric-definite eigensystems. The size of the workspace is O(2n).

void gsl_eigen_gensymm_free (gsl_eigen_gensymm_workspace * w) [Function]
This function frees the memory associated with the workspace w.

This function computes the eigenvalues of the real generalized symmetric-definite matrix pair (A, B), and stores them in eval, using the method outlined above. On output, B contains its Cholesky decomposition and A is destroyed.

gsl_eigen_gensymmv_workspace * gsl_eigen_gensymmv_alloc [Function] (const size_t n)

This function allocates a workspace for computing eigenvalues and eigenvectors of n-by-n real generalized symmetric-definite eigensystems. The size of the workspace is O(4n).

void gsl_eigen_gensymmv_free (gsl_eigen_gensymmv_workspace * w) [Function]
This function frees the memory associated with the workspace w.

int gsl_eigen_gensymmv (gsl_matrix * A, gsl_matrix * B, gsl_vector * eval, gsl_matrix * evec, gsl_eigen_gensymmv_workspace * w) [Function]

This function computes the eigenvalues and eigenvectors of the real generalized symmetric-definite matrix pair (A, B), and stores them in eval and evec respectively. The computed eigenvectors are normalized to have unit magnitude. On output, B contains its Cholesky decomposition and A is destroyed.

15.5 Complex Generalized Hermitian-Definite Eigensystems

The complex generalized hermitian-definite eigenvalue problem is to find eigenvalues λ and eigenvectors x such that

$$Ax = \lambda Bx$$

where A and B are hermitian matrices, and B is positive-definite. Similarly to the real case, this can be reduced to $Cy = \lambda y$ where $C = L^{-1}AL^{-\dagger}$ is hermitian, and $y = L^{\dagger}x$. The standard hermitian eigensolver can be applied to the matrix C. The resulting eigenvectors

are backtransformed to find the vectors of the original problem. The eigenvalues of the generalized hermitian-definite eigenproblem are always real.

This function allocates a workspace for computing eigenvalues of n-by-n complex generalized hermitian-definite eigensystems. The size of the workspace is O(3n).

- void gsl_eigen_genherm_free (gsl_eigen_genherm_workspace * w) [Function]
 This function frees the memory associated with the workspace w.
- int gsl_eigen_genherm (gsl_matrix_complex * A, gsl_matrix_complex * [Function]

 B, gsl_vector * eval, gsl_eigen_genherm_workspace * w)

 This function computes the eigenvalues of the complex generalized hermitian-definite

This function computes the eigenvalues of the complex generalized hermitian-definite matrix pair (A, B), and stores them in *eval*, using the method outlined above. On output, B contains its Cholesky decomposition and A is destroyed.

This function allocates a workspace for computing eigenvalues and eigenvectors of n-by-n complex generalized hermitian-definite eigensystems. The size of the workspace is O(5n).

- void gsl_eigen_genhermv_free (gsl_eigen_genhermv_workspace * w) [Function]
 This function frees the memory associated with the workspace w.

This function computes the eigenvalues and eigenvectors of the complex generalized hermitian-definite matrix pair (A, B), and stores them in eval and evec respectively. The computed eigenvectors are normalized to have unit magnitude. On output, B contains its Cholesky decomposition and A is destroyed.

15.6 Real Generalized Nonsymmetric Eigensystems

Given two square matrices (A, B), the generalized nonsymmetric eigenvalue problem is to find eigenvalues λ and eigenvectors x such that

$$Ax = \lambda Bx$$

We may also define the problem as finding eigenvalues μ and eigenvectors y such that

$$\mu Ay = By$$

Note that these two problems are equivalent (with $\lambda=1/\mu$) if neither λ nor μ is zero. If say, λ is zero, then it is still a well defined eigenproblem, but its alternate problem involving μ is not. Therefore, to allow for zero (and infinite) eigenvalues, the problem which is actually solved is

$$\beta Ax = \alpha Bx$$

The eigensolver routines below will return two values α and β and leave it to the user to perform the divisions $\lambda = \alpha/\beta$ and $\mu = \beta/\alpha$.

If the determinant of the matrix pencil $A - \lambda B$ is zero for all λ , the problem is said to be singular; otherwise it is called regular. Singularity normally leads to some $\alpha = \beta = 0$ which means the eigenproblem is ill-conditioned and generally does not have well defined eigenvalue solutions. The routines below are intended for regular matrix pencils and could yield unpredictable results when applied to singular pencils.

The solution of the real generalized nonsymmetric eigensystem problem for a matrix pair (A, B) involves computing the generalized Schur decomposition

$$A = QSZ^T$$

$$B = QTZ^T$$

where Q and Z are orthogonal matrices of left and right Schur vectors respectively, and (S,T) is the generalized Schur form whose diagonal elements give the α and β values. The algorithm used is the QZ method due to Moler and Stewart (see references).

- gsl_eigen_gen_workspace * gsl_eigen_gen_alloc ($const\ size_t\ n$) [Function] This function allocates a workspace for computing eigenvalues of n-by-n real generalized nonsymmetric eigensystems. The size of the workspace is O(n).
- void gsl_eigen_gen_free (gsl_eigen_gen_workspace * w) [Function]
 This function frees the memory associated with the workspace w.

This function sets some parameters which determine how the eigenvalue problem is solved in subsequent calls to gsl_eigen_gen.

If $compute_s$ is set to 1, the full Schur form S will be computed by gsl_eigen_gen . If it is set to 0, S will not be computed (this is the default setting). S is a quasi upper triangular matrix with 1-by-1 and 2-by-2 blocks on its diagonal. 1-by-1 blocks correspond to real eigenvalues, and 2-by-2 blocks correspond to complex eigenvalues.

If $compute_t$ is set to 1, the full Schur form T will be computed by gsl_eigen_gen . If it is set to 0, T will not be computed (this is the default setting). T is an upper triangular matrix with non-negative elements on its diagonal. Any 2-by-2 blocks in S will correspond to a 2-by-2 diagonal block in T.

The balance parameter is currently ignored, since generalized balancing is not yet implemented.

int gsl_eigen_gen (gsl_matrix * A, gsl_matrix * B, gsl_vector_complex * [Function] alpha, gsl_vector * beta, gsl_eigen_gen_workspace * w)

This function computes the eigenvalues of the real generalized nonsymmetric matrix pair (A, B), and stores them as pairs in (alpha, beta), where alpha is complex and beta is real. If β_i is non-zero, then $\lambda = \alpha_i/\beta_i$ is an eigenvalue. Likewise, if α_i is non-zero, then $\mu = \beta_i/\alpha_i$ is an eigenvalue of the alternate problem $\mu Ay = By$. The elements of beta are normalized to be non-negative.

If S is desired, it is stored in A on output. If T is desired, it is stored in B on output. The ordering of eigenvalues in (alpha, beta) follows the ordering of the diagonal blocks in the Schur forms S and T. In rare cases, this function may fail to find all eigenvalues. If this occurs, an error code is returned.

This function is identical to gsl_eigen_gen except that it also computes the left and right Schur vectors and stores them into Q and Z respectively.

This function allocates a workspace for computing eigenvalues and eigenvectors of n-by-n real generalized nonsymmetric eigensystems. The size of the workspace is O(7n).

void gsl_eigen_genv_free (gsl_eigen_genv_workspace * w) [Function]
This function frees the memory associated with the workspace w.

This function computes eigenvalues and right eigenvectors of the n-by-n real generalized nonsymmetric matrix pair (A, B). The eigenvalues are stored in (alpha, beta) and the eigenvectors are stored in evec. It first calls gsl_eigen_gen to compute the eigenvalues, Schur forms, and Schur vectors. Then it finds eigenvectors of the Schur forms and backtransforms them using the Schur vectors. The Schur vectors are destroyed in the process, but can be saved by using $gsl_eigen_genv_QZ$. The computed eigenvectors are normalized to have unit magnitude. On output, (A, B) contains the generalized Schur form (S, T). If gsl_eigen_gen fails, no eigenvectors are computed, and an error code is returned.

This function is identical to gsl_eigen_genv except that it also computes the left and right Schur vectors and stores them into Q and Z respectively.

15.7 Sorting Eigenvalues and Eigenvectors

int gsl_eigen_symmv_sort (gsl_vector * eval, gsl_matrix * evec, gsl_eigen_sort_t sort_type) [Function]

This function simultaneously sorts the eigenvalues stored in the vector eval and the corresponding real eigenvectors stored in the columns of the matrix evec into ascending or descending order according to the value of the parameter sort_type,

GSL_EIGEN_SORT_VAL_ASC
ascending order in numerical value

GSL_EIGEN_SORT_VAL_DESC

descending order in numerical value

GSL_EIGEN_SORT_ABS_ASC

ascending order in magnitude

GSL_EIGEN_SORT_ABS_DESC

descending order in magnitude

This function simultaneously sorts the eigenvalues stored in the vector eval and the corresponding complex eigenvectors stored in the columns of the matrix evec into ascending or descending order according to the value of the parameter sort_type as shown above.

This function simultaneously sorts the eigenvalues stored in the vector *eval* and the corresponding complex eigenvectors stored in the columns of the matrix *evec* into ascending or descending order according to the value of the parameter *sort_type* as shown above. Only GSL_EIGEN_SORT_ABS_ASC and GSL_EIGEN_SORT_ABS_DESC are supported due to the eigenvalues being complex.

int gsl_eigen_gensymmv_sort (gsl_vector * eval, gsl_matrix * evec, [Function] gsl_eigen_sort_t sort_type)

This function simultaneously sorts the eigenvalues stored in the vector eval and the corresponding real eigenvectors stored in the columns of the matrix evec into ascending or descending order according to the value of the parameter sort_type as shown above.

This function simultaneously sorts the eigenvalues stored in the vector eval and the corresponding complex eigenvectors stored in the columns of the matrix evec into ascending or descending order according to the value of the parameter sort_type as shown above.

int gsl_eigen_genv_sort (gsl_vector_complex * alpha, gsl_vector * [Function] beta, gsl_matrix_complex * evec, gsl_eigen_sort_t sort_type)

This function simultaneously sorts the eigenvalues stored in the vectors (alpha, beta) and the corresponding complex eigenvectors stored in the columns of the matrix evec into ascending or descending order according to the value of the parameter sort_type as shown above. Only GSL_EIGEN_SORT_ABS_ASC and GSL_EIGEN_SORT_ABS_DESC are supported due to the eigenvalues being complex.

15.8 Examples

The following program computes the eigenvalues and eigenvectors of the 4-th order Hilbert matrix, H(i, j) = 1/(i + j + 1).

```
#include <stdio.h>
#include <gsl/gsl_math.h>
#include <gsl/gsl_eigen.h>
int
main (void)
{
  double data[] = \{ 1.0 , 1/2.0, 1/3.0, 1/4.0, 
                    1/2.0, 1/3.0, 1/4.0, 1/5.0,
                    1/3.0, 1/4.0, 1/5.0, 1/6.0,
                    1/4.0, 1/5.0, 1/6.0, 1/7.0 };
  gsl_matrix_view m
    = gsl_matrix_view_array (data, 4, 4);
  gsl_vector *eval = gsl_vector_alloc (4);
  gsl_matrix *evec = gsl_matrix_alloc (4, 4);
  gsl_eigen_symmv_workspace * w =
    gsl_eigen_symmv_alloc (4);
  gsl_eigen_symmv (&m.matrix, eval, evec, w);
  gsl_eigen_symmv_free (w);
  gsl_eigen_symmv_sort (eval, evec,
                        GSL_EIGEN_SORT_ABS_ASC);
  {
    int i;
    for (i = 0; i < 4; i++)
      {
        double eval_i
           = gsl_vector_get (eval, i);
        gsl_vector_view evec_i
           = gsl_matrix_column (evec, i);
        printf ("eigenvalue = %g\n", eval_i);
        printf ("eigenvector = \n");
        gsl_vector_fprintf (stdout,
                            &evec_i.vector, "%g");
     }
  }
  gsl_vector_free (eval);
  gsl_matrix_free (evec);
```

```
return 0;
Here is the beginning of the output from the program,
     $ ./a.out
     eigenvalue = 9.67023e-05
     eigenvector =
     -0.0291933
     0.328712
     -0.791411
     0.514553
This can be compared with the corresponding output from GNU OCTAVE,
     octave> [v,d] = eig(hilb(4));
     octave> diag(d)
     ans =
        9.6702e-05
        6.7383e-03
        1.6914e-01
        1.5002e+00
     octave> v
     v =
                  0.179186 -0.582076
                                           0.792608
        0.029193
       -0.328712 -0.741918
                               0.370502
                                           0.451923
        0.791411
                   0.100228
                               0.509579
                                           0.322416
       -0.514553
                   0.638283
                               0.514048
                                           0.252161
```

Note that the eigenvectors can differ by a change of sign, since the sign of an eigenvector is arbitrary.

The following program illustrates the use of the nonsymmetric eigensolver, by computing the eigenvalues and eigenvectors of the Vandermonde matrix $V(x;i,j) = x_i^{n-j}$ with x = (-1, -2, 3, 4).

```
gsl_matrix_view m
         = gsl_matrix_view_array (data, 4, 4);
       gsl_vector_complex *eval = gsl_vector_complex_alloc (4);
       gsl_matrix_complex *evec = gsl_matrix_complex_alloc (4, 4);
       gsl_eigen_nonsymmv_workspace * w =
         gsl_eigen_nonsymmv_alloc (4);
       gsl_eigen_nonsymmv (&m.matrix, eval, evec, w);
       gsl_eigen_nonsymmv_free (w);
       gsl_eigen_nonsymmv_sort (eval, evec,
                                 GSL_EIGEN_SORT_ABS_DESC);
       {
         int i, j;
         for (i = 0; i < 4; i++)
             gsl_complex eval_i
                = gsl_vector_complex_get (eval, i);
             gsl_vector_complex_view evec_i
                = gsl_matrix_complex_column (evec, i);
             printf ("eigenvalue = %g + %gi\n",
                     GSL_REAL(eval_i), GSL_IMAG(eval_i));
             printf ("eigenvector = \n");
             for (j = 0; j < 4; ++j)
               {
                 gsl\_complex z =
                   gsl_vector_complex_get(&evec_i.vector, j);
                 printf("%g + %gi\n", GSL_REAL(z), GSL_IMAG(z));
           }
       }
       gsl_vector_complex_free(eval);
       gsl_matrix_complex_free(evec);
       return 0;
     }
Here is the beginning of the output from the program,
     $ ./a.out
```

```
eigenvalue = -6.41391 + 0i

eigenvector =

-0.0998822 + 0i

-0.111251 + 0i

0.292501 + 0i

0.944505 + 0i

eigenvalue = 5.54555 + 3.08545i

eigenvector =

-0.043487 + -0.0076308i

0.0642377 + -0.142127i

-0.515253 + 0.0405118i

-0.840592 + -0.00148565i

...
```

This can be compared with the corresponding output from GNU OCTAVE,

```
octave> [v,d] = eig(vander([-1 -2 3 4]));
octave > diag(d)
ans =
 -6.4139 + 0.0000i
  5.5456 + 3.0854i
  5.5456 - 3.0854i
  2.3228 + 0.0000i
octave> v
v =
Columns 1 through 3:
 -0.09988 + 0.00000i -0.04350 - 0.00755i -0.04350 + 0.00755i
 -0.11125 + 0.00000i 0.06399 - 0.14224i 0.06399 + 0.14224i
  0.29250 + 0.00000i -0.51518 + 0.04142i -0.51518 - 0.04142i
  0.94451 + 0.00000i -0.84059 + 0.00000i -0.84059 - 0.00000i
Column 4:
 -0.14493 + 0.00000i
  0.35660 + 0.00000i
  0.91937 + 0.00000i
  0.08118 + 0.00000i
```

Note that the eigenvectors corresponding to the eigenvalue 5.54555 + 3.08545i differ by the multiplicative constant 0.9999984 + 0.0017674i which is an arbitrary phase factor of magnitude 1.

15.9 References and Further Reading

Further information on the algorithms described in this section can be found in the following book,

G. H. Golub, C. F. Van Loan, *Matrix Computations* (3rd Ed, 1996), Johns Hopkins University Press, ISBN 0-8018-5414-8.

Further information on the generalized eigensystems QZ algorithm can be found in this paper,

C. Moler, G. Stewart, "An Algorithm for Generalized Matrix Eigenvalue Problems", SIAM J. Numer. Anal., Vol 10, No 2, 1973.

Eigensystem routines for very large matrices can be found in the Fortran library LAPACK. The LAPACK library is described in,

 $LAPACK\ Users'$ Guide (Third Edition, 1999), Published by SIAM, ISBN 0-89871-447-8.

http://www.netlib.org/lapack

The LAPACK source code can be found at the website above along with an online copy of the users guide.

16 Fast Fourier Transforms (FFTs)

This chapter describes functions for performing Fast Fourier Transforms (FFTs). The library includes radix-2 routines (for lengths which are a power of two) and mixed-radix routines (which work for any length). For efficiency there are separate versions of the routines for real data and for complex data. The mixed-radix routines are a reimplementation of the FFTPACK library of Paul Swarztrauber. Fortran code for FFTPACK is available on Netlib (FFTPACK also includes some routines for sine and cosine transforms but these are currently not available in GSL). For details and derivations of the underlying algorithms consult the document GSL FFT Algorithms (see Section 16.8 [FFT References and Further Reading], page 179)

16.1 Mathematical Definitions

Fast Fourier Transforms are efficient algorithms for calculating the discrete Fourier transform (DFT),

$$x_j = \sum_{k=0}^{n-1} z_k \exp(-2\pi i j k/n)$$

The DFT usually arises as an approximation to the continuous Fourier transform when functions are sampled at discrete intervals in space or time. The naive evaluation of the discrete Fourier transform is a matrix-vector multiplication $W\vec{z}$. A general matrix-vector multiplication takes $O(n^2)$ operations for n data-points. Fast Fourier transform algorithms use a divide-and-conquer strategy to factorize the matrix W into smaller sub-matrices, corresponding to the integer factors of the length n. If n can be factorized into a product of integers $f_1 f_2 \dots f_m$ then the DFT can be computed in $O(n \sum f_i)$ operations. For a radix-2 FFT this gives an operation count of $O(n \log_2 n)$.

All the FFT functions offer three types of transform: forwards, inverse and backwards, based on the same mathematical definitions. The definition of the forward Fourier transform, x = FFT(z), is,

$$x_j = \sum_{k=0}^{n-1} z_k \exp(-2\pi i j k/n)$$

and the definition of the inverse Fourier transform, x = IFFT(z), is,

$$z_j = \frac{1}{n} \sum_{k=0}^{n-1} x_k \exp(2\pi i j k/n).$$

The factor of 1/n makes this a true inverse. For example, a call to $gsl_fft_complex_forward$ followed by a call to $gsl_fft_complex_inverse$ should return the original data (within numerical errors).

In general there are two possible choices for the sign of the exponential in the transform/inverse-transform pair. GSL follows the same convention as FFTPACK, using a negative exponential for the forward transform. The advantage of this convention is that the inverse transform recreates the original function with simple Fourier synthesis. Numerical Recipes uses the opposite convention, a positive exponential in the forward transform.

The $backwards\ FFT$ is simply our terminology for an unscaled version of the inverse FFT,

$$z_j^{backwards} = \sum_{k=0}^{n-1} x_k \exp(2\pi i j k/n).$$

When the overall scale of the result is unimportant it is often convenient to use the backwards FFT instead of the inverse to save unnecessary divisions.

16.2 Overview of complex data FFTs

The inputs and outputs for the complex FFT routines are *packed arrays* of floating point numbers. In a packed array the real and imaginary parts of each complex number are placed in alternate neighboring elements. For example, the following definition of a packed array of length 6,

```
double x[3*2];
gsl_complex_packed_array data = x;
```

can be used to hold an array of three complex numbers, z[3], in the following way,

```
data[0] = Re(z[0])
data[1] = Im(z[0])
data[2] = Re(z[1])
data[3] = Im(z[1])
data[4] = Re(z[2])
data[5] = Im(z[2])
```

The array indices for the data have the same ordering as those in the definition of the DFT—i.e. there are no index transformations or permutations of the data.

A stride parameter allows the user to perform transforms on the elements z[stride*i] instead of z[i]. A stride greater than 1 can be used to take an in-place FFT of the column of a matrix. A stride of 1 accesses the array without any additional spacing between elements.

To perform an FFT on a vector argument, such as gsl_vector_complex * v, use the following definitions (or their equivalents) when calling the functions described in this chapter:

```
gsl_complex_packed_array data = v->data;
size_t stride = v->stride;
size_t n = v->size;
```

For physical applications it is important to remember that the index appearing in the DFT does not correspond directly to a physical frequency. If the time-step of the DFT is Δ then the frequency-domain includes both positive and negative frequencies, ranging from $-1/(2\Delta)$ through 0 to $+1/(2\Delta)$. The positive frequencies are stored from the beginning of the array up to the middle, and the negative frequencies are stored backwards from the end of the array.

Here is a table which shows the layout of the array data, and the correspondence between the time-domain data z, and the frequency-domain data x.

index z
$$x = FFT(z)$$

0 $z(t = 0)$ $x(f = 0)$

```
1
        z(t = 1)
                        x(f = 1/(n Delta))
2
        z(t = 2)
                        x(f = 2/(n Delta))
                        . . . . . . . .
                        x(f = +1/(2 Delta),
n/2
        z(t = n/2)
                              -1/(2 Delta))
                        . . . . . . . .
                        x(f = -3/(n Delta))
n-3
        z(t = n-3)
                        x(f = -2/(n Delta))
n-2
        z(t = n-2)
        z(t = n-1)
                        x(f = -1/(n Delta))
n-1
```

When n is even the location n/2 contains the most positive and negative frequencies $(+1/(2\Delta), -1/(2\Delta))$ which are equivalent. If n is odd then general structure of the table above still applies, but n/2 does not appear.

16.3 Radix-2 FFT routines for complex data

The radix-2 algorithms described in this section are simple and compact, although not necessarily the most efficient. They use the Cooley-Tukey algorithm to compute in-place complex FFTs for lengths which are a power of 2—no additional storage is required. The corresponding self-sorting mixed-radix routines offer better performance at the expense of requiring additional working space.

All the functions described in this section are declared in the header file gsl_fft_complex.h.

```
int gsl_fft_complex_radix2_forward (gsl_complex_packed_array data, size_t stride, size_t n)
int gsl_fft_complex_radix2_transform (gsl_complex_packed_array data, size_t stride, size_t n, gsl_fft_direction sign)
int gsl_fft_complex_radix2_backward (gsl_complex_packed_array data, size_t stride, size_t n)
[Function]
```

These functions compute forward, backward and inverse FFTs of length n with stride stride, on the packed complex array data using an in-place radix-2 decimation-in-time algorithm. The length of the transform n is restricted to powers of two. For the transform version of the function the sign argument can be either forward (-1) or backward (+1).

The functions return a value of GSL_SUCCESS if no errors were detected, or GSL_EDOM if the length of the data n is not a power of two.

Here is an example program which computes the FFT of a short pulse in a sample of length 128. To make the resulting Fourier transform real the pulse is defined for equal positive and negative times $(-10 \dots 10)$, where the negative times wrap around the end of the array.

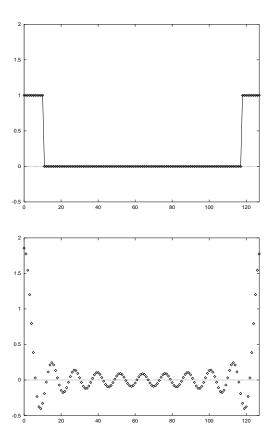
```
#include <stdio.h>
#include <math.h>
#include <gsl/gsl_errno.h>
#include <gsl/gsl_fft_complex.h>
#define REAL(z,i) ((z)[2*(i)])
#define IMAG(z,i) ((z)[2*(i)+1])
int
main (void)
{
  int i; double data[2*128];
  for (i = 0; i < 128; i++)
    {
       REAL(data,i) = 0.0; IMAG(data,i) = 0.0;
    }
 REAL(data,0) = 1.0;
  for (i = 1; i \le 10; i++)
    {
       REAL(data,i) = REAL(data,128-i) = 1.0;
    }
  for (i = 0; i < 128; i++)
      printf ("%d %e %e\n", i,
              REAL(data,i), IMAG(data,i));
    }
 printf ("\n");
  gsl_fft_complex_radix2_forward (data, 1, 128);
  for (i = 0; i < 128; i++)
    {
      printf ("%d %e %e\n", i,
              REAL(data,i)/sqrt(128),
              IMAG(data,i)/sqrt(128));
```

```
}
return 0;
```

Note that we have assumed that the program is using the default error handler (which calls abort for any errors). If you are not using a safe error handler you would need to check the return status of gsl_fft_complex_radix2_forward.

The transformed data is rescaled by $1/\sqrt{n}$ so that it fits on the same plot as the input. Only the real part is shown, by the choice of the input data the imaginary part is zero. Allowing for the wrap-around of negative times at t=128, and working in units of k/n, the DFT approximates the continuum Fourier transform, giving a modulated sine function.

$$\int_{-a}^{+a} e^{-2\pi i k x} dx = \frac{\sin(2\pi k a)}{\pi k}$$



A pulse and its discrete Fourier transform, output from the example program.

16.4 Mixed-radix FFT routines for complex data

This section describes mixed-radix FFT algorithms for complex data. The mixed-radix functions work for FFTs of any length. They are a reimplementation of Paul Swarztrauber's Fortran FFTPACK library. The theory is explained in the review article Self-sorting Mixed-

radix FFTs by Clive Temperton. The routines here use the same indexing scheme and basic algorithms as FFTPACK.

The mixed-radix algorithm is based on sub-transform modules—highly optimized small length FFTs which are combined to create larger FFTs. There are efficient modules for factors of 2, 3, 4, 5, 6 and 7. The modules for the composite factors of 4 and 6 are faster than combining the modules for 2 * 2 and 2 * 3.

For factors which are not implemented as modules there is a fall-back to a general length-n module which uses Singleton's method for efficiently computing a DFT. This module is $O(n^2)$, and slower than a dedicated module would be but works for any length n. Of course, lengths which use the general length-n module will still be factorized as much as possible. For example, a length of 143 will be factorized into 11*13. Large prime factors are the worst case scenario, e.g. as found in n = 2*3*99991, and should be avoided because their $O(n^2)$ scaling will dominate the run-time (consult the document GSL FFT Algorithms included in the GSL distribution if you encounter this problem).

The mixed-radix initialization function $\mathtt{gsl_fft_complex_wavetable_alloc}$ returns the list of factors chosen by the library for a given length n. It can be used to check how well the length has been factorized, and estimate the run-time. To a first approximation the run-time scales as $n \sum f_i$, where the f_i are the factors of n. For programs under user control you may wish to issue a warning that the transform will be slow when the length is poorly factorized. If you frequently encounter data lengths which cannot be factorized using the existing small-prime modules consult GSL FFT Algorithms for details on adding support for other factors.

All the functions described in this section are declared in the header file gsl_fft_complex.h.

This function prepares a trigonometric lookup table for a complex FFT of length n. The function returns a pointer to the newly allocated $\mathtt{gsl_fft_complex_wavetable}$ if no errors were detected, and a null pointer in the case of error. The length n is factorized into a product of subtransforms, and the factors and their trigonometric coefficients are stored in the wavetable. The trigonometric coefficients are computed using direct calls to \mathtt{sin} and \mathtt{cos} , for accuracy. Recursion relations could be used to compute the lookup table faster, but if an application performs many FFTs of the same length then this computation is a one-off overhead which does not affect the final throughput.

The wavetable structure can be used repeatedly for any transform of the same length. The table is not modified by calls to any of the other FFT functions. The same wavetable can be used for both forward and backward (or inverse) transforms of a given length.

This function frees the memory associated with the wavetable wavetable. The wavetable can be freed if no further FFTs of the same length will be needed.

These functions operate on a gsl_fft_complex_wavetable structure which contains internal parameters for the FFT. It is not necessary to set any of the components directly but it can sometimes be useful to examine them. For example, the chosen factorization of the FFT length is given and can be used to provide an estimate of the run-time or numerical error. The wavetable structure is declared in the header file gsl_fft_complex.h.

gsl_fft_complex_wavetable

[Data Type]

This is a structure that holds the factorization and trigonometric lookup tables for the mixed radix fft algorithm. It has the following components:

size_t n This is the number of complex data points

size_t nf This is the number of factors that the length n was decomposed into.

size_t factor[64]

This is the array of factors. Only the first **nf** elements are used.

gsl_complex * trig

This is a pointer to a preallocated trigonometric lookup table of ${\tt n}$ complex elements.

gsl_complex * twiddle[64]

This is an array of pointers into trig, giving the twiddle factors for each pass.

The mixed radix algorithms require additional working space to hold the intermediate steps of the transform.

 $gsl_fft_complex_workspace *$

[Function]

gsl_fft_complex_workspace_alloc (size_t n)

This function allocates a workspace for a complex transform of length n.

This function frees the memory associated with the workspace workspace. The workspace can be freed if no further FFTs of the same length will be needed.

The following functions compute the transform,

These functions compute forward, backward and inverse FFTs of length n with stride stride, on the packed complex array data, using a mixed radix decimation-in-frequency

algorithm. There is no restriction on the length n. Efficient modules are provided for subtransforms of length 2, 3, 4, 5, 6 and 7. Any remaining factors are computed with a slow, $O(n^2)$, general-n module. The caller must supply a wavetable containing the trigonometric lookup tables and a workspace work. For the transform version of the function the sign argument can be either forward (-1) or backward (+1).

The functions return a value of 0 if no errors were detected. The following gsl_errno conditions are defined for these functions:

 GSL_EDOM The length of the data n is not a positive integer (i.e. n is zero).

GSL_EINVAL

The length of the data n and the length used to compute the given wavetable do not match.

Here is an example program which computes the FFT of a short pulse in a sample of length $630 \ (= 2 * 3 * 3 * 5 * 7)$ using the mixed-radix algorithm.

```
#include <stdio.h>
#include <math.h>
#include <gsl/gsl_errno.h>
#include <gsl/gsl_fft_complex.h>
#define REAL(z,i) ((z)[2*(i)])
#define IMAG(z,i) ((z)[2*(i)+1])
int
main (void)
{
  int i;
  const int n = 630;
  double data[2*n];
  gsl_fft_complex_wavetable * wavetable;
  gsl_fft_complex_workspace * workspace;
  for (i = 0; i < n; i++)
    {
      REAL(data,i) = 0.0;
      IMAG(data,i) = 0.0;
    }
  data[0] = 1.0;
  for (i = 1; i \le 10; i++)
      REAL(data,i) = REAL(data,n-i) = 1.0;
    }
  for (i = 0; i < n; i++)
```

```
{
     printf ("%d: %e %e\n", i, REAL(data,i),
                                IMAG(data,i));
    }
 printf ("\n");
 wavetable = gsl_fft_complex_wavetable_alloc (n);
  workspace = gsl_fft_complex_workspace_alloc (n);
  for (i = 0; i < wavetable -> nf; i++)
    {
       printf ("# factor %d: %d\n", i,
               wavetable->factor[i]);
    }
 gsl_fft_complex_forward (data, 1, n,
                           wavetable, workspace);
  for (i = 0; i < n; i++)
    {
     printf ("%d: %e %e\n", i, REAL(data,i),
                                IMAG(data,i));
    }
 gsl_fft_complex_wavetable_free (wavetable);
 gsl_fft_complex_workspace_free (workspace);
 return 0;
}
```

Note that we have assumed that the program is using the default gsl error handler (which calls abort for any errors). If you are not using a safe error handler you would need to check the return status of all the gsl routines.

16.5 Overview of real data FFTs

The functions for real data are similar to those for complex data. However, there is an important difference between forward and inverse transforms. The Fourier transform of a real sequence is not real. It is a complex sequence with a special symmetry:

$$z_k = z_{n-k}^*$$

A sequence with this symmetry is called *conjugate-complex* or *half-complex*. This different structure requires different storage layouts for the forward transform (from real to half-complex) and inverse transform (from half-complex back to real). As a consequence the routines are divided into two sets: functions in gsl_fft_real which operate on real sequences and functions in gsl_fft_halfcomplex which operate on half-complex sequences.

Functions in gsl_fft_real compute the frequency coefficients of a real sequence. The half-complex coefficients c of a real sequence x are given by Fourier analysis,

$$c_k = \sum_{j=0}^{n-1} x_j \exp(-2\pi i j k/n)$$

Functions in $gsl_fft_halfcomplex$ compute inverse or backwards transforms. They reconstruct real sequences by Fourier synthesis from their half-complex frequency coefficients, c,

$$x_j = \frac{1}{n} \sum_{k=0}^{n-1} c_k \exp(2\pi i j k/n)$$

The symmetry of the half-complex sequence implies that only half of the complex numbers in the output need to be stored. The remaining half can be reconstructed using the half-complex symmetry condition. This works for all lengths, even and odd—when the length is even the middle value where k=n/2 is also real. Thus only n real numbers are required to store the half-complex sequence, and the transform of a real sequence can be stored in the same size array as the original data.

The precise storage arrangements depend on the algorithm, and are different for radix-2 and mixed-radix routines. The radix-2 function operates in-place, which constrains the locations where each element can be stored. The restriction forces real and imaginary parts to be stored far apart. The mixed-radix algorithm does not have this restriction, and it stores the real and imaginary parts of a given term in neighboring locations (which is desirable for better locality of memory accesses).

16.6 Radix-2 FFT routines for real data

This section describes radix-2 FFT algorithms for real data. They use the Cooley-Tukey algorithm to compute in-place FFTs for lengths which are a power of 2.

The radix-2 FFT functions for real data are declared in the header files gsl_fft_real.h

This function computes an in-place radix-2 FFT of length n and stride stride on the real array data. The output is a half-complex sequence, which is stored in-place. The arrangement of the half-complex terms uses the following scheme: for k < n/2 the real part of the k-th term is stored in location k, and the corresponding imaginary part is stored in location n - k. Terms with k > n/2 can be reconstructed using the symmetry $z_k = z_{n-k}^*$. The terms for k = 0 and k = n/2 are both purely real, and count as a special case. Their real parts are stored in locations 0 and n/2 respectively, while their imaginary parts which are zero are not stored.

The following table shows the correspondence between the output *data* and the equivalent results obtained by considering the input data as a complex sequence with zero imaginary part (assuming *stride=1*),

```
complex[0].real = data[0]
complex[0].imag = 0
complex[1].real = data[1]
```

```
complex[1].imag
                               data[n-1]
. . . . . . . . . . . . . . . .
                                . . . . . . . . . . . . . . . .
complex[k].real
                               data[k]
complex[k].imag
                               data[n-k]
. . . . . . . . . . . . . . .
                                . . . . . . . . . . . . . . . .
complex[n/2].real =
                               data[n/2]
complex[n/2].imag =
. . . . . . . . . . . . . . . .
                                              k' = n - k
complex[k'].real
                               data[k]
complex[k'].imag
                              -data[n-k]
. . . . . . . . . . . . . . .
                                . . . . . . . . . . . . . . . .
complex[n-1].real =
                               data[1]
complex[n-1].imag =
                              -data[n-1]
```

Note that the output data can be converted into the full complex sequence using the function gsl_fft_halfcomplex_radix2_unpack described below.

The radix-2 FFT functions for halfcomplex data are declared in the header file gsl_fft_halfcomplex.h.

These functions compute the inverse or backwards in-place radix-2 FFT of length n and stride stride on the half-complex sequence data stored according the output scheme used by $gsl_fft_real_radix2$. The result is a real array stored in natural order.

This function converts halfcomplex_coefficient, an array of half-complex coefficients as returned by gsl_fft_real_radix2_transform, into an ordinary complex array, complex_coefficient. It fills in the complex array using the symmetry $z_k = z_{n-k}^*$ to reconstruct the redundant elements. The algorithm for the conversion is,

16.7 Mixed-radix FFT routines for real data

This section describes mixed-radix FFT algorithms for real data. The mixed-radix functions work for FFTs of any length. They are a reimplementation of the real-FFT routines in the Fortran FFTPACK library by Paul Swarztrauber. The theory behind the algorithm is explained in the article Fast Mixed-Radix Real Fourier Transforms by Clive Temperton. The routines here use the same indexing scheme and basic algorithms as FFTPACK.

The functions use the FFTPACK storage convention for half-complex sequences. In this convention the half-complex transform of a real sequence is stored with frequencies in increasing order, starting at zero, with the real and imaginary parts of each frequency in neighboring locations. When a value is known to be real the imaginary part is not stored. The imaginary part of the zero-frequency component is never stored. It is known to be zero (since the zero frequency component is simply the sum of the input data (all real)). For a sequence of even length the imaginary part of the frequency n/2 is not stored either, since the symmetry $z_k = z_{n-k}^*$ implies that this is purely real too.

The storage scheme is best shown by some examples. The table below shows the output for an odd-length sequence, n = 5. The two columns give the correspondence between the 5 values in the half-complex sequence returned by $gsl_fft_real_transform$, halfcomplex[] and the values complex[] that would be returned if the same real input sequence were passed to $gsl_fft_complex_backward$ as a complex sequence (with imaginary parts set to 0),

```
complex[0].real = halfcomplex[0]
complex[0].imag = 0
complex[1].real = halfcomplex[1]
complex[1].imag = halfcomplex[2]
complex[2].real = halfcomplex[3]
complex[2].imag = halfcomplex[4]
complex[3].real = halfcomplex[4]
complex[3].imag = -halfcomplex[4]
complex[4].real = halfcomplex[1]
```

The upper elements of the *complex* array, <code>complex[3]</code> and <code>complex[4]</code> are filled in using the symmetry condition. The imaginary part of the zero-frequency term <code>complex[0].imag</code> is known to be zero by the symmetry.

The next table shows the output for an even-length sequence, n = 6. In the even case there are two values which are purely real,

```
complex[0].real =
                   halfcomplex[0]
complex[0].imag
                    0
complex[1].real =
                   halfcomplex[1]
complex[1].imag
                   halfcomplex[2]
complex[2].real
                   halfcomplex[3]
complex[2].imag
                   halfcomplex[4]
complex[3].real
                   halfcomplex[5]
complex[3].imag
complex[4].real = halfcomplex[3]
complex[4].imag = -halfcomplex[4]
complex[5].real
                   halfcomplex[1]
complex[5].imag = -halfcomplex[2]
```

The upper elements of the *complex* array, <code>complex[4]</code> and <code>complex[5]</code> are filled in using the symmetry condition. Both <code>complex[0].imag</code> and <code>complex[3].imag</code> are known to be zero.

All these functions are declared in the header files gsl_fft_real.h and gsl_fft_halfcomplex.h.

These functions prepare trigonometric lookup tables for an FFT of size n real elements. The functions return a pointer to the newly allocated struct if no errors were detected, and a null pointer in the case of error. The length n is factorized into a product of subtransforms, and the factors and their trigonometric coefficients are stored in the wavetable. The trigonometric coefficients are computed using direct calls to \sin and \cos , for accuracy. Recursion relations could be used to compute the lookup table faster, but if an application performs many FFTs of the same length then computing the wavetable is a one-off overhead which does not affect the final throughput.

The wavetable structure can be used repeatedly for any transform of the same length. The table is not modified by calls to any of the other FFT functions. The appropriate type of wavetable must be used for forward real or inverse half-complex transforms.

wavetable can be freed if no further FFTs of the same length will be needed.

The mixed radix algorithms require additional working space to hold the intermediate steps of the transform,

This function allocates a workspace for a real transform of length n. The same workspace can be used for both forward real and inverse halfcomplex transforms.

This function frees the memory associated with the workspace workspace. The workspace can be freed if no further FFTs of the same length will be needed.

The following functions compute the transforms of real and half-complex data,

```
int gsl_fft_real_transform (double data[], size_t stride, size_t n, [Function] const gsl_fft_real_wavetable * wavetable, gsl_fft_real_workspace * work)
```

These functions compute the FFT of data, a real or half-complex array of length n, using a mixed radix decimation-in-frequency algorithm. For $\mathtt{gsl_fft_real_transform}$ data is an array of time-ordered real data. For $\mathtt{gsl_fft_halfcomplex_transform}$ data contains Fourier coefficients in the half-complex ordering described above. There is no restriction on the length n. Efficient modules are provided for subtransforms of length 2, 3, 4 and 5. Any remaining factors are computed with a slow, $O(n^2)$, generaln module. The caller must supply a wavetable containing trigonometric lookup tables and a workspace work.

This function converts $halfcomplex_coefficient$, an array of half-complex coefficients as returned by $gsl_fft_real_transform$, into an ordinary complex array, $complex_coefficient$. It fills in the complex array using the symmetry $z_k = z_{n-k}^*$ to reconstruct the redundant elements. The algorithm for the conversion is,

```
complex_coefficient[0].imag
  = 0.0;
for (i = 1; i < n - i; i++)
  {
    double hc_real
      = halfcomplex_coefficient[(2 * i - 1)*stride];
    double hc_imag
      = halfcomplex_coefficient[(2 * i)*stride];
    complex_coefficient[i*stride].real = hc_real;
    complex_coefficient[i*stride].imag = hc_imag;
    complex_coefficient[(n - i)*stride].real = hc_real;
    complex_coefficient[(n - i)*stride].imag = -hc_imag;
  }
if (i == n - i)
  {
    complex_coefficient[i*stride].real
      = halfcomplex_coefficient[(n - 1)*stride];
    complex_coefficient[i*stride].imag
      = 0.0;
  }
```

Here is an example program using gsl_fft_real_transform and gsl_fft_halfcomplex_inverse. It generates a real signal in the shape of a square pulse. The pulse is Fourier transformed to frequency space, and all but the lowest ten frequency components are removed from the array of Fourier coefficients returned by gsl_fft_real_transform.

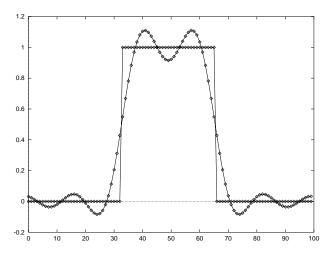
The remaining Fourier coefficients are transformed back to the time-domain, to give a filtered version of the square pulse. Since Fourier coefficients are stored using the half-complex symmetry both positive and negative frequencies are removed and the final filtered signal is also real.

```
#include <stdio.h>
#include <math.h>
#include <gsl/gsl_errno.h>
#include <gsl/gsl_fft_real.h>
#include <gsl/gsl_fft_halfcomplex.h>

int
main (void)
{
   int i, n = 100;
   double data[n];

   gsl_fft_real_wavetable * real;
   gsl_fft_halfcomplex_wavetable * hc;
   gsl_fft_real_workspace * work;
```

```
for (i = 0; i < n; i++)
    {
      data[i] = 0.0;
    }
  for (i = n / 3; i < 2 * n / 3; i++)
    {
      data[i] = 1.0;
    }
  for (i = 0; i < n; i++)
    {
     printf ("%d: %e\n", i, data[i]);
    }
 printf ("\n");
 work = gsl_fft_real_workspace_alloc (n);
 real = gsl_fft_real_wavetable_alloc (n);
 gsl_fft_real_transform (data, 1, n,
                          real, work);
 gsl_fft_real_wavetable_free (real);
 for (i = 11; i < n; i++)
    {
      data[i] = 0;
 hc = gsl_fft_halfcomplex_wavetable_alloc (n);
 gsl_fft_halfcomplex_inverse (data, 1, n,
                               hc, work);
 gsl_fft_halfcomplex_wavetable_free (hc);
 for (i = 0; i < n; i++)
      printf ("%d: %e\n", i, data[i]);
 gsl_fft_real_workspace_free (work);
 return 0;
}
```



Low-pass filtered version of a real pulse, output from the example program.

16.8 References and Further Reading

A good starting point for learning more about the FFT is the review article Fast Fourier Transforms: A Tutorial Review and A State of the Art by Duhamel and Vetterli,

P. Duhamel and M. Vetterli. Fast Fourier transforms: A tutorial review and a state of the art. Signal Processing, 19:259–299, 1990.

To find out about the algorithms used in the GSL routines you may want to consult the document GSL FFT Algorithms (it is included in GSL, as doc/fftalgorithms.tex). This has general information on FFTs and explicit derivations of the implementation for each routine. There are also references to the relevant literature. For convenience some of the more important references are reproduced below.

There are several introductory books on the FFT with example programs, such as *The Fast Fourier Transform* by Brigham and *DFT/FFT and Convolution Algorithms* by Burrus and Parks,

- E. Oran Brigham. The Fast Fourier Transform. Prentice Hall, 1974.
- C. S. Burrus and T. W. Parks. DFT/FFT and Convolution Algorithms. Wiley, 1984.

Both these introductory books cover the radix-2 FFT in some detail. The mixed-radix algorithm at the heart of the FFTPACK routines is reviewed in Clive Temperton's paper,

Clive Temperton. Self-sorting mixed-radix fast Fourier transforms. Journal of Computational Physics, 52(1):1–23, 1983.

The derivation of FFTs for real-valued data is explained in the following two articles,

Henrik V. Sorenson, Douglas L. Jones, Michael T. Heideman, and C. Sidney Burrus. Real-valued fast Fourier transform algorithms. *IEEE Transactions on Acoustics*, Speech, and Signal Processing, ASSP-35(6):849–863, 1987.

Clive Temperton. Fast mixed-radix real Fourier transforms. *Journal of Computational Physics*, 52:340–350, 1983.

In 1979 the IEEE published a compendium of carefully-reviewed Fortran FFT programs in *Programs for Digital Signal Processing*. It is a useful reference for implementations of many different FFT algorithms,

Digital Signal Processing Committee and IEEE Acoustics, Speech, and Signal Processing Committee, editors. *Programs for Digital Signal Processing*. IEEE Press, 1979.

For large-scale FFT work we recommend the use of the dedicated FFTW library by Frigo and Johnson. The FFTW library is self-optimizing—it automatically tunes itself for each hardware platform in order to achieve maximum performance. It is available under the GNU GPL.

FFTW Website, http://www.fftw.org/

The source code for FFTPACK is available from Netlib,

FFTPACK, http://www.netlib.org/fftpack/

17 Numerical Integration

This chapter describes routines for performing numerical integration (quadrature) of a function in one dimension. There are routines for adaptive and non-adaptive integration of general functions, with specialised routines for specific cases. These include integration over infinite and semi-infinite ranges, singular integrals, including logarithmic singularities, computation of Cauchy principal values and oscillatory integrals. The library reimplements the algorithms used in QUADPACK, a numerical integration package written by Piessens, de Doncker-Kapenga, Ueberhuber and Kahaner. Fortran code for QUADPACK is available on Netlib. Also included are non-adaptive, fixed-order Gauss-Legendre integration routines with high precision coefficients by Pavel Holoborodko.

The functions described in this chapter are declared in the header file gsl_integration.h.

17.1 Introduction

Each algorithm computes an approximation to a definite integral of the form,

$$I = \int_{a}^{b} f(x)w(x) \, dx$$

where w(x) is a weight function (for general integrands w(x) = 1). The user provides absolute and relative error bounds (*epsabs*, *epsrel*) which specify the following accuracy requirement,

$$|RESULT - I| \le \max(epsabs, epsrel |I|)$$

where RESULT is the numerical approximation obtained by the algorithm. The algorithms attempt to estimate the absolute error ABSERR = |RESULT - I| in such a way that the following inequality holds,

$$|RESULT - I| \le ABSERR \le \max(epsabs, epsrel |I|)$$

In short, the routines return the first approximation which has an absolute error smaller than *epsabs* or a relative error smaller than *epsrel*.

Note that this is an *either-or* constraint, not simultaneous. To compute to a specified absolute error, set *epsrel* to zero. To compute to a specified relative error, set *epsabs* to zero. The routines will fail to converge if the error bounds are too stringent, but always return the best approximation obtained up to that stage.

The algorithms in QUADPACK use a naming convention based on the following letters,

- Q quadrature routine
- N non-adaptive integrator
- A adaptive integrator
- G general integrand (user-defined)
- W weight function with integrand
- S singularities can be more readily integrated

- P points of special difficulty can be supplied
- I infinite range of integration
- O oscillatory weight function, cos or sin
- F Fourier integral
- C Cauchy principal value

The algorithms are built on pairs of quadrature rules, a higher order rule and a lower order rule. The higher order rule is used to compute the best approximation to an integral over a small range. The difference between the results of the higher order rule and the lower order rule gives an estimate of the error in the approximation.

17.1.1 Integrands without weight functions

The algorithms for general functions (without a weight function) are based on Gauss-Kronrod rules.

A Gauss-Kronrod rule begins with a classical Gaussian quadrature rule of order m. This is extended with additional points between each of the abscissae to give a higher order Kronrod rule of order 2m + 1. The Kronrod rule is efficient because it reuses existing function evaluations from the Gaussian rule.

The higher order Kronrod rule is used as the best approximation to the integral, and the difference between the two rules is used as an estimate of the error in the approximation.

17.1.2 Integrands with weight functions

For integrands with weight functions the algorithms use Clenshaw-Curtis quadrature rules.

A Clenshaw-Curtis rule begins with an n-th order Chebyshev polynomial approximation to the integrand. This polynomial can be integrated exactly to give an approximation to the integral of the original function. The Chebyshev expansion can be extended to higher orders to improve the approximation and provide an estimate of the error.

17.1.3 Integrands with singular weight functions

The presence of singularities (or other behavior) in the integrand can cause slow convergence in the Chebyshev approximation. The modified Clenshaw-Curtis rules used in QUADPACK separate out several common weight functions which cause slow convergence.

These weight functions are integrated analytically against the Chebyshev polynomials to precompute *modified Chebyshev moments*. Combining the moments with the Chebyshev approximation to the function gives the desired integral. The use of analytic integration for the singular part of the function allows exact cancellations and substantially improves the overall convergence behavior of the integration.

17.2 QNG non-adaptive Gauss-Kronrod integration

The QNG algorithm is a non-adaptive procedure which uses fixed Gauss-Kronrod-Patterson abscissae to sample the integrand at a maximum of 87 points. It is provided for fast integration of smooth functions.

This function applies the Gauss-Kronrod 10-point, 21-point, 43-point and 87-point integration rules in succession until an estimate of the integral of f over (a,b) is achieved within the desired absolute and relative error limits, epsabs and epsrel. The function returns the final approximation, result, an estimate of the absolute error, abserr and the number of function evaluations used, neval. The Gauss-Kronrod rules are designed in such a way that each rule uses all the results of its predecessors, in order to minimize the total number of function evaluations.

17.3 QAG adaptive integration

The QAG algorithm is a simple adaptive integration procedure. The integration region is divided into subintervals, and on each iteration the subinterval with the largest estimated error is bisected. This reduces the overall error rapidly, as the subintervals become concentrated around local difficulties in the integrand. These subintervals are managed by a gsl_integration_workspace struct, which handles the memory for the subinterval ranges, results and error estimates.

This function allocates a workspace sufficient to hold n double precision intervals, their integration results and error estimates. One workspace may be used multiple times as all necessary reinitialization is performed automatically by the integration routines.

```
void gsl_integration_workspace_free (gsl_integration_workspace * [Function]
w)
```

This function frees the memory associated with the workspace w.

```
int gsl_integration_qag (const gsl_function * f, double a, double b, double epsabs, double epsrel, size_t limit, int key,
```

gsl_integration_workspace * workspace, double * result, double * abserr) This function applies an integration rule adaptively until an estimate of the integral of f over (a,b) is achieved within the desired absolute and relative error limits, epsabs and epsrel. The function returns the final approximation, result, and an estimate of the absolute error, abserr. The integration rule is determined by the value of key, which should be chosen from the following symbolic names,

```
GSL_INTEG_GAUSS15 (key = 1)
GSL_INTEG_GAUSS21 (key = 2)
GSL_INTEG_GAUSS31 (key = 3)
GSL_INTEG_GAUSS41 (key = 4)
GSL_INTEG_GAUSS51 (key = 5)
GSL_INTEG_GAUSS61 (key = 6)
```

corresponding to the 15, 21, 31, 41, 51 and 61 point Gauss-Kronrod rules. The higher-order rules give better accuracy for smooth functions, while lower-order rules save time when the function contains local difficulties, such as discontinuities.

On each iteration the adaptive integration strategy bisects the interval with the largest error estimate. The subintervals and their results are stored in the memory provided by workspace. The maximum number of subintervals is given by limit, which may not exceed the allocated size of the workspace.

17.4 QAGS adaptive integration with singularities

The presence of an integrable singularity in the integration region causes an adaptive routine to concentrate new subintervals around the singularity. As the subintervals decrease in size the successive approximations to the integral converge in a limiting fashion. This approach to the limit can be accelerated using an extrapolation procedure. The QAGS algorithm combines adaptive bisection with the Wynn epsilon-algorithm to speed up the integration of many types of integrable singularities.

This function applies the Gauss-Kronrod 21-point integration rule adaptively until an estimate of the integral of f over (a,b) is achieved within the desired absolute and relative error limits, epsabs and epsrel. The results are extrapolated using the epsilon-algorithm, which accelerates the convergence of the integral in the presence of discontinuities and integrable singularities. The function returns the final approximation from the extrapolation, result, and an estimate of the absolute error, abserr. The subintervals and their results are stored in the memory provided by workspace. The maximum number of subintervals is given by limit, which may not exceed the allocated size of the workspace.

17.5 QAGP adaptive integration with known singular points

This function applies the adaptive integration algorithm QAGS taking account of the user-supplied locations of singular points. The array pts of length npts should contain the endpoints of the integration ranges defined by the integration region and locations of the singularities. For example, to integrate over the region (a, b) with break-points at x_1, x_2, x_3 (where $a < x_1 < x_2 < x_3 < b$) the following pts array should be used

```
pts[0] = a
pts[1] = x_1
pts[2] = x_2
pts[3] = x_3
pts[4] = b
```

with npts = 5.

If you know the locations of the singular points in the integration region then this routine will be faster than QAGS.

17.6 QAGI adaptive integration on infinite intervals

This function computes the integral of the function f over the infinite interval $(-\infty, +\infty)$. The integral is mapped onto the semi-open interval (0,1] using the transformation x = (1-t)/t,

$$\int_{-\infty}^{+\infty} dx \, f(x) = \int_{0}^{1} dt \, (f((1-t)/t) + f(-(1-t)/t))/t^{2}.$$

It is then integrated using the QAGS algorithm. The normal 21-point Gauss-Kronrod rule of QAGS is replaced by a 15-point rule, because the transformation can generate an integrable singularity at the origin. In this case a lower-order rule is more efficient.

This function computes the integral of the function f over the semi-infinite interval $(a, +\infty)$. The integral is mapped onto the semi-open interval (0, 1] using the transformation x = a + (1 - t)/t,

$$\int_{a}^{+\infty} dx \, f(x) = \int_{0}^{1} dt \, f(a + (1 - t)/t)/t^{2}$$

and then integrated using the QAGS algorithm.

This function computes the integral of the function f over the semi-infinite interval $(-\infty, b)$. The integral is mapped onto the semi-open interval (0, 1] using the transformation x = b - (1 - t)/t,

$$\int_{-\infty}^{b} dx \, f(x) = \int_{0}^{1} dt \, f(b - (1 - t)/t)/t^{2}$$

and then integrated using the QAGS algorithm.

17.7 QAWC adaptive integration for Cauchy principal values

int gsl_integration_qawc (gsl_function * f, double a, double b, double c, double epsabs, double epsrel, size_t limit,

gsl_integration_workspace * workspace, double * result, double * abserr) This function computes the Cauchy principal value of the integral of f over (a,b), with a singularity at c,

$$I = \int_a^b dx \, \frac{f(x)}{x - c} = \lim_{\epsilon \to 0} \left\{ \int_a^{c - \epsilon} dx \, \frac{f(x)}{x - c} + \int_{c + \epsilon}^b dx \, \frac{f(x)}{x - c} \right\}$$

The adaptive bisection algorithm of QAG is used, with modifications to ensure that subdivisions do not occur at the singular point x = c. When a subinterval contains the point x = c or is close to it then a special 25-point modified Clenshaw-Curtis rule is used to control the singularity. Further away from the singularity the algorithm uses an ordinary 15-point Gauss-Kronrod integration rule.

17.8 QAWS adaptive integration for singular functions

The QAWS algorithm is designed for integrands with algebraic-logarithmic singularities at the end-points of an integration region. In order to work efficiently the algorithm requires a precomputed table of Chebyshev moments.

This function allocates space for a gsl_integration_qaws_table struct describing a singular weight function W(x) with the parameters $(\alpha, \beta, \mu, \nu)$,

$$W(x) = (x - a)^{\alpha} (b - x)^{\beta} \log^{\mu} (x - a) \log^{\nu} (b - x)$$

where $\alpha > -1$, $\beta > -1$, and $\mu = 0, 1$, $\nu = 0, 1$. The weight function can take four different forms depending on the values of μ and ν ,

$$\begin{split} W(x) &= (x-a)^{\alpha}(b-x)^{\beta} & (\mu=0,\nu=0) \\ W(x) &= (x-a)^{\alpha}(b-x)^{\beta}\log(x-a) & (\mu=1,\nu=0) \\ W(x) &= (x-a)^{\alpha}(b-x)^{\beta}\log(b-x) & (\mu=0,\nu=1) \\ W(x) &= (x-a)^{\alpha}(b-x)^{\beta}\log(x-a)\log(b-x) & (\mu=1,\nu=1) \end{split}$$

The singular points (a, b) do not have to be specified until the integral is computed, where they are the endpoints of the integration range.

The function returns a pointer to the newly allocated table gsl_integration_qaws_table if no errors were detected, and 0 in the case of error.

int gsl_integration_qaws_table_set (gsl_integration_qaws_table * [Function] t, double alpha, double beta, int mu, int nu)

This function modifies the parameters $(\alpha, \beta, \mu, \nu)$ of an existing gsl_integration_qaws_table struct t.

This function frees all the memory associated with the $gsl_integration_qaws_table$ struct t.

This function computes the integral of the function f(x) over the interval (a, b) with the singular weight function $(x - a)^{\alpha}(b - x)^{\beta} \log^{\mu}(x - a) \log^{\nu}(b - x)$. The parameters

of the weight function $(\alpha, \beta, \mu, \nu)$ are taken from the table t. The integral is,

$$I = \int_{a}^{b} dx \, f(x)(x-a)^{\alpha} (b-x)^{\beta} \log^{\mu}(x-a) \log^{\nu}(b-x).$$

The adaptive bisection algorithm of QAG is used. When a subinterval contains one of the endpoints then a special 25-point modified Clenshaw-Curtis rule is used to control the singularities. For subintervals which do not include the endpoints an ordinary 15-point Gauss-Kronrod integration rule is used.

17.9 QAWO adaptive integration for oscillatory functions

The QAWO algorithm is designed for integrands with an oscillatory factor, $\sin(\omega x)$ or $\cos(\omega x)$. In order to work efficiently the algorithm requires a table of Chebyshev moments which must be pre-computed with calls to the functions below.

gsl_integration_qawo_table *

[Function]

gsl_integration_qawo_table_alloc (double omega, double L, enum gsl_integration_qawo_enum sine, size_t n)

This function allocates space for a gsl_integration_qawo_table struct and its associated workspace describing a sine or cosine weight function W(x) with the parameters (ω, L) ,

$$W(x) = \left\{ \frac{\sin(\omega x)}{\cos(\omega x)} \right\}$$

The parameter L must be the length of the interval over which the function will be integrated L = b - a. The choice of sine or cosine is made with the parameter sine which should be chosen from one of the two following symbolic values:

GSL_INTEG_COSINE
GSL_INTEG_SINE

The gsl_integration_qawo_table is a table of the trigonometric coefficients required in the integration process. The parameter n determines the number of levels of coefficients that are computed. Each level corresponds to one bisection of the interval L, so that n levels are sufficient for subintervals down to the length $L/2^n$. The integration routine gsl_integration_qawo returns the error GSL_ETABLE if the number of levels is insufficient for the requested accuracy.

int gsl_integration_qawo_table_set (gsl_integration_qawo_table * [Function] t, double omega, double L, enum gsl_integration_qawo_enum sine)

This function changes the parameters omega, L and sine of the existing workspace t.

int gsl_integration_qawo_table_set_length

[Function]

(gsl_integration_gawo_table * t, double L)

This function allows the length parameter L of the workspace t to be changed.

This function frees all the memory associated with the workspace t.

This function uses an adaptive algorithm to compute the integral of f over (a, b) with the weight function $\sin(\omega x)$ or $\cos(\omega x)$ defined by the table wf,

$$I = \int_{a}^{b} dx \, f(x) \left\{ \frac{\sin(\omega x)}{\cos(\omega x)} \right\}$$

The results are extrapolated using the epsilon-algorithm to accelerate the convergence of the integral. The function returns the final approximation from the extrapolation, result, and an estimate of the absolute error, abserr. The subintervals and their results are stored in the memory provided by workspace. The maximum number of subintervals is given by limit, which may not exceed the allocated size of the workspace.

Those subintervals with "large" widths d where $d\omega > 4$ are computed using a 25-point Clenshaw-Curtis integration rule, which handles the oscillatory behavior. Subintervals with a "small" widths where $d\omega < 4$ are computed using a 15-point Gauss-Kronrod integration.

17.10 QAWF adaptive integration for Fourier integrals

This function attempts to compute a Fourier integral of the function f over the semi-infinite interval $[a, +\infty)$.

$$I = \int_{a}^{+\infty} dx \, f(x) \left\{ \frac{\sin(\omega x)}{\cos(\omega x)} \right\}$$

The parameter ω and choice of sin or cos is taken from the table wf (the length L can take any value, since it is overridden by this function to a value appropriate for the Fourier integration). The integral is computed using the QAWO algorithm over each of the subintervals,

$$C_1 = [a, a + c]$$

 $C_2 = [a + c, a + 2c]$
... = ...
 $C_k = [a + (k - 1)c, a + kc]$

where $c = (2 \operatorname{floor}(|\omega|) + 1)\pi/|\omega|$. The width c is chosen to cover an odd number of periods so that the contributions from the intervals alternate in sign and are monotonically decreasing when f is positive and monotonically decreasing. The sum of this sequence of contributions is accelerated using the epsilon-algorithm.

This function works to an overall absolute tolerance of abserr. The following strategy is used: on each interval C_k the algorithm tries to achieve the tolerance

$$TOL_k = u_k abserr$$

where $u_k = (1-p)p^{k-1}$ and p = 9/10. The sum of the geometric series of contributions from each interval gives an overall tolerance of abserr.

If the integration of a subinterval leads to difficulties then the accuracy requirement for subsequent intervals is relaxed,

$$TOL_k = u_k \max(abserr, \max_{i < k} \{E_i\})$$

where E_k is the estimated error on the interval C_k .

The subintervals and their results are stored in the memory provided by workspace. The maximum number of subintervals is given by limit, which may not exceed the allocated size of the workspace. The integration over each subinterval uses the memory provided by cycle_workspace as workspace for the QAWO algorithm.

17.11 CQUAD doubly-adaptive integration

CQUAD is a new doubly-adaptive general-purpose quadrature routine which can handle most types of singularities, non-numerical function values such as Inf or NaN, as well as some divergent integrals. It generally requires more function evaluations than the integration routines in QUADPACK, yet fails less often for difficult integrands.

The underlying algorithm uses a doubly-adaptive scheme in which Clenshaw-Curtis quadrature rules of increasing degree are used to compute the integral in each interval. The L_2 -norm of the difference between the underlying interpolatory polynomials of two successive rules is used as an error estimate. The interval is subdivided if the difference between two successive rules is too large or a rule of maximum degree has been reached.

This function allocates a workspace sufficient to hold the data for n intervals. The number n is not the maximum number of intervals that will be evaluated. If the workspace is full, intervals with smaller error estimates will be discarded. A minimum of 3 intervals is required and for most functions, a workspace of size 100 is sufficient.

This function frees the memory associated with the workspace w.

This function computes the integral of f over (a,b) within the desired absolute and relative error limits, epsabs and epsrel using the CQUAD algorithm. The function returns the final approximation, result, an estimate of the absolute error, abserr, and the number of function evaluations required, nevals.

The CQUAD algorithm divides the integration region into subintervals, and in each iteration, the subinterval with the largest estimated error is processed. The algorithm uses Clenshaw-Curits quadrature rules of degree 4, 8, 16 and 32 over 5, 9, 17 and 33 nodes respectively. Each interval is initialized with the lowest-degree rule. When an interval is processed, the next-higher degree rule is evaluated and an error estimate is computed based on the L_2 -norm of the difference between the underlying interpolating polynomials of both rules. If the highest-degree rule has already been used, or the interpolatory polynomials differ significantly, the interval is bisected.

The subintervals and their results are stored in the memory provided by workspace. If the error estimate or the number of function evaluations is not needed, the pointers absert and nevals can be set to NULL.

17.12 Gauss-Legendre integration

The fixed-order Gauss-Legendre integration routines are provided for fast integration of smooth functions with known polynomial order. The n-point Gauss-Legendre rule is exact for polynomials of order 2*n-1 or less. For example, these rules are useful when integrating basis functions to form mass matrices for the Galerkin method. Unlike other numerical integration routines within the library, these routines do not accept absolute or relative error bounds.

```
gsl_integration_glfixed_table * [Function]
gsl_integration_glfixed_table_alloc (size_t n)
```

This function determines the Gauss-Legendre abscissae and weights necessary for an n-point fixed order integration scheme. If possible, high precision precomputed coefficients are used. If precomputed weights are not available, lower precision coefficients are computed on the fly.

This function applies the Gauss-Legendre integration rule contained in table t and returns the result.

For i in [0, ..., t-> n-1], this function obtains the i-th Gauss-Legendre point xi and weight wi on the interval [a,b]. The points and weights are ordered by increasing point value. A function f may be integrated on [a,b] by summing wi * f(xi) over i.

```
void gsl_integration_glfixed_table_free [Function]

(gsl_integration_glfixed_table * t)

This function frees the memory associated with the table t.
```

v

In addition to the standard error codes for invalid arguments the functions can return the

```
following values, GSL_EMAXITER
```

17.13 Error codes

the maximum number of subdivisions was exceeded.

GSL_EROUND

cannot reach tolerance because of roundoff error, or roundoff error was detected in the extrapolation table.

GSL_ESING

a non-integrable singularity or other bad integrand behavior was found in the integration interval.

GSL_EDIVERGE

the integral is divergent, or too slowly convergent to be integrated numerically.

17.14 Examples

The integrator QAGS will handle a large class of definite integrals. For example, consider the following integral, which has an algebraic-logarithmic singularity at the origin,

$$\int_0^1 x^{-1/2} \log(x) \, dx = -4$$

The program below computes this integral to a relative accuracy bound of 1e-7.

```
#include <stdio.h>
#include <math.h>
#include <gsl/gsl_integration.h>
double f (double x, void * params) {
  double alpha = *(double *) params;
 double f = log(alpha*x) / sqrt(x);
 return f;
}
int
main (void)
{
 gsl_integration_workspace * w
    = gsl_integration_workspace_alloc (1000);
  double result, error;
  double expected = -4.0;
  double alpha = 1.0;
  gsl_function F;
 F.function = &f;
 F.params = α
  gsl_integration_qags (&F, 0, 1, 0, 1e-7, 1000,
                       w, &result, &error);
                   = \% .18f\n", result);
 printf ("result
 printf ("exact result = % .18f\n", expected);
```

```
printf ("estimated error = % .18f\n", error);
printf ("actual error = % .18f\n", result - expected);
printf ("intervals = %d\n", w->size);

gsl_integration_workspace_free (w);
return 0;
}
```

The results below show that the desired accuracy is achieved after 8 subdivisions.

In fact, the extrapolation procedure used by QAGS produces an accuracy of almost twice as many digits. The error estimate returned by the extrapolation procedure is larger than the actual error, giving a margin of safety of one order of magnitude.

17.15 References and Further Reading

The following book is the definitive reference for QUADPACK, and was written by the original authors. It provides descriptions of the algorithms, program listings, test programs and examples. It also includes useful advice on numerical integration and many references to the numerical integration literature used in developing QUADPACK.

R. Piessens, E. de Doncker-Kapenga, C.W. Ueberhuber, D.K. Kahaner. QUADPACK A subroutine package for automatic integration Springer Verlag, 1983.

The CQUAD integration algorithm is described in the following paper:

P. Gonnet, "Increasing the Reliability of Adaptive Quadrature Using Explicit Interpolants", ACM Transactions on Mathematical Software, Volume 37 (2010), Issue 3, Article 26.

18 Random Number Generation

The library provides a large collection of random number generators which can be accessed through a uniform interface. Environment variables allow you to select different generators and seeds at runtime, so that you can easily switch between generators without needing to recompile your program. Each instance of a generator keeps track of its own state, allowing the generators to be used in multi-threaded programs. Additional functions are available for transforming uniform random numbers into samples from continuous or discrete probability distributions such as the Gaussian, log-normal or Poisson distributions.

These functions are declared in the header file gsl_rng.h.

18.1 General comments on random numbers

In 1988, Park and Miller wrote a paper entitled "Random number generators: good ones are hard to find." [Commun. ACM, 31, 1192–1201]. Fortunately, some excellent random number generators are available, though poor ones are still in common use. You may be happy with the system-supplied random number generator on your computer, but you should be aware that as computers get faster, requirements on random number generators increase. Nowadays, a simulation that calls a random number generator millions of times can often finish before you can make it down the hall to the coffee machine and back.

A very nice review of random number generators was written by Pierre L'Ecuyer, as Chapter 4 of the book: Handbook on Simulation, Jerry Banks, ed. (Wiley, 1997). The chapter is available in postscript from L'Ecuyer's ftp site (see references). Knuth's volume on Seminumerical Algorithms (originally published in 1968) devotes 170 pages to random number generators, and has recently been updated in its 3rd edition (1997). It is brilliant, a classic. If you don't own it, you should stop reading right now, run to the nearest bookstore, and buy it.

A good random number generator will satisfy both theoretical and statistical properties. Theoretical properties are often hard to obtain (they require real math!), but one prefers a random number generator with a long period, low serial correlation, and a tendency *not* to "fall mainly on the planes." Statistical tests are performed with numerical simulations. Generally, a random number generator is used to estimate some quantity for which the theory of probability provides an exact answer. Comparison to this exact answer provides a measure of "randomness".

18.2 The Random Number Generator Interface

It is important to remember that a random number generator is not a "real" function like sine or cosine. Unlike real functions, successive calls to a random number generator yield different return values. Of course that is just what you want for a random number generator, but to achieve this effect, the generator must keep track of some kind of "state" variable. Sometimes this state is just an integer (sometimes just the value of the previously generated random number), but often it is more complicated than that and may involve a whole array of numbers, possibly with some indices thrown in. To use the random number generators, you do not need to know the details of what comprises the state, and besides that varies from algorithm to algorithm.

The random number generator library uses two special structs, gsl_rng_type which holds static information about each type of generator and gsl_rng which describes an instance of a generator created from a given gsl_rng_type.

The functions described in this section are declared in the header file gsl_rng.h.

18.3 Random number generator initialization

gsl_rng * gsl_rng_alloc (const gsl_rng_type * T)

[Function]

This function returns a pointer to a newly-created instance of a random number generator of type T. For example, the following code creates an instance of the Tausworthe generator,

```
gsl_rng * r = gsl_rng_alloc (gsl_rng_taus);
```

If there is insufficient memory to create the generator then the function returns a null pointer and the error handler is invoked with an error code of GSL_ENOMEM.

The generator is automatically initialized with the default seed, gsl_rng_default_seed. This is zero by default but can be changed either directly or by using the environment variable GSL_RNG_SEED (see Section 18.6 [Random number environment variables], page 196).

The details of the available generator types are described later in this chapter.

void gsl_rng_set (const gsl_rng * r, unsigned long int s)

[Function]

This function initializes (or 'seeds') the random number generator. If the generator is seeded with the same value of s on two different runs, the same stream of random numbers will be generated by successive calls to the routines below. If different values of $s \ge 1$ are supplied, then the generated streams of random numbers should be completely different. If the seed s is zero then the standard seed from the original implementation is used instead. For example, the original Fortran source code for the ranlux generator used a seed of 314159265, and so choosing s equal to zero reproduces this when using s_1 _rng_ranlux.

When using multiple seeds with the same generator, choose seed values greater than zero to avoid collisions with the default setting.

Note that the most generators only accept 32-bit seeds, with higher values being reduced modulo 2^{32} . For generators with smaller ranges the maximum seed value will typically be lower.

void gsl_rng_free (gsl_rng * r)

[Function]

This function frees all the memory associated with the generator r.

18.4 Sampling from a random number generator

The following functions return uniformly distributed random numbers, either as integers or double precision floating point numbers. Inline versions of these functions are used when HAVE_INLINE is defined. To obtain non-uniform distributions see Chapter 20 [Random Number Distributions], page 213.

unsigned long int gsl_rng_get (const gsl_rng * r)

[Function]

This function returns a random integer from the generator r. The minimum and maximum values depend on the algorithm used, but all integers in the range [min,max] are equally likely. The values of min and max can be determined using the auxiliary functions gsl_rg_max (r) and gsl_rg_min (r).

double gsl_rng_uniform (const gsl_rng * r)

[Function]

This function returns a double precision floating point number uniformly distributed in the range [0,1). The range includes 0.0 but excludes 1.0. The value is typically obtained by dividing the result of gsl_rng_get(r) by gsl_rng_max(r) + 1.0 in double precision. Some generators compute this ratio internally so that they can provide floating point numbers with more than 32 bits of randomness (the maximum number of bits that can be portably represented in a single unsigned long int).

double gsl_rng_uniform_pos (const gsl_rng * r)

Function]

This function returns a positive double precision floating point number uniformly distributed in the range (0,1), excluding both 0.0 and 1.0. The number is obtained by sampling the generator with the algorithm of gsl_rng_uniform until a non-zero value is obtained. You can use this function if you need to avoid a singularity at 0.0.

```
unsigned long int gsl_rng_uniform_int (const gsl_rng * r, unsigned long int n) [Function]
```

This function returns a random integer from 0 to n-1 inclusive by scaling down and/or discarding samples from the generator r. All integers in the range [0, n-1] are produced with equal probability. For generators with a non-zero minimum value an offset is applied so that zero is returned with the correct probability.

Note that this function is designed for sampling from ranges smaller than the range of the underlying generator. The parameter n must be less than or equal to the range of the generator r. If n is larger than the range of the generator then the function calls the error handler with an error code of GSL_EINVAL and returns zero.

In particular, this function is not intended for generating the full range of unsigned integer values $[0, 2^{32} - 1]$. Instead choose a generator with the maximal integer range and zero minimum value, such as $gsl_rng_ranlxd1$, $gsl_rng_mt19937$ or gsl_rng_taus , and sample it directly using gsl_rng_get . The range of each generator can be found using the auxiliary functions described in the next section.

18.5 Auxiliary random number generator functions

The following functions provide information about an existing generator. You should use them in preference to hard-coding the generator parameters into your own code.

```
const char * gsl_rng_name (const gsl_rng * r)
```

[Function]

This function returns a pointer to the name of the generator. For example,

would print something like r is a 'taus' generator.

```
unsigned long int gsl_rng_max (const gsl_rng * r) [Function] gsl_rng_max returns the largest value that gsl_rng_get can return.
```

```
unsigned long int gsl_rng_min (const gsl_rng * r) [Function] gsl_rng_min returns the smallest value that gsl_rng_get can return. Usually this value is zero. There are some generators with algorithms that cannot return zero, and for these generators the minimum value is 1.
```

These functions return a pointer to the state of generator r and its size. You can use this information to access the state directly. For example, the following code will write the state of a generator to a stream,

```
void * state = gsl_rng_state (r);
size_t n = gsl_rng_size (r);
fwrite (state, n, 1, stream);
```

```
const gsl_rng_type ** gsl_rng_types_setup (void) [Function]
```

This function returns a pointer to an array of all the available generator types, terminated by a null pointer. The function should be called once at the start of the program, if needed. The following code fragment shows how to iterate over the array of generator types to print the names of the available algorithms,

```
const gsl_rng_type **t, **t0;

t0 = gsl_rng_types_setup ();

printf ("Available generators:\n");

for (t = t0; *t != 0; t++)
   {
    printf ("%s\n", (*t)->name);
   }
```

18.6 Random number environment variables

The library allows you to choose a default generator and seed from the environment variables GSL_RNG_TYPE and GSL_RNG_SEED and the function gsl_rng_env_setup. This makes it easy try out different generators and seeds without having to recompile your program.

```
const gsl_rng_type * gsl_rng_env_setup (void) [Function]

This function reads the environment variables GSL_RNG_TYPE and GSL_RNG_SEED and
```

uses their values to set the corresponding library variables gsl_rng_default and gsl_rng_default_seed. These global variables are defined as follows,

```
extern const gsl_rng_type *gsl_rng_default
extern unsigned long int gsl_rng_default_seed
```

The environment variable GSL_RNG_TYPE should be the name of a generator, such as taus or mt19937. The environment variable GSL_RNG_SEED should contain the desired seed value. It is converted to an unsigned long int using the C library function strtoul.

If you don't specify a generator for GSL_RNG_TYPE then gsl_rng_mt19937 is used as the default. The initial value of gsl_rng_default_seed is zero.

Here is a short program which shows how to create a global generator using the environment variables GSL_RNG_TYPE and GSL_RNG_SEED,

```
#include <stdio.h>
#include <gsl/gsl_rng.h>

gsl_rng * r; /* global generator */

int
main (void)
{
    const gsl_rng_type * T;
    gsl_rng_env_setup();

    T = gsl_rng_default;
    r = gsl_rng_alloc (T);

    printf ("generator type: %s\n", gsl_rng_name (r));
    printf ("seed = %lu\n", gsl_rng_default_seed);
    printf ("first value = %lu\n", gsl_rng_get (r));

    gsl_rng_free (r);
    return 0;
}
```

Running the program without any environment variables uses the initial defaults, an mt19937 generator with a seed of 0,

```
$ ./a.out
generator type: mt19937
seed = 0
first value = 4293858116
```

By setting the two variables on the command line we can change the default generator and the seed,

```
$ GSL_RNG_TYPE="taus" GSL_RNG_SEED=123 ./a.out
GSL_RNG_TYPE=taus
GSL_RNG_SEED=123
generator type: taus
seed = 123
first value = 2720986350
```

18.7 Copying random number generator state

The above methods do not expose the random number 'state' which changes from call to call. It is often useful to be able to save and restore the state. To permit these practices, a few somewhat more advanced functions are supplied. These include:

int gsl_rng_memcpy (gsl_rng * dest, const gsl_rng * src)

[Function]

This function copies the random number generator src into the pre-existing generator dest, making dest into an exact copy of src. The two generators must be of the same type.

gsl_rng * gsl_rng_clone (const gsl_rng * r)

[Function]

This function returns a pointer to a newly created generator which is an exact copy of the generator r.

18.8 Reading and writing random number generator state

The library provides functions for reading and writing the random number state to a file as binary data.

int gsl_rng_fwrite (FILE * stream, const gsl_rng * r)

[Function]

This function writes the random number state of the random number generator r to the stream stream in binary format. The return value is 0 for success and $GSL_EFAILED$ if there was a problem writing to the file. Since the data is written in the native binary format it may not be portable between different architectures.

int gsl_rng_fread (FILE * stream, gsl_rng * r)

[Function]

This function reads the random number state into the random number generator r from the open stream stream in binary format. The random number generator r must be preinitialized with the correct random number generator type since type information is not saved. The return value is 0 for success and $GSL_EFAILED$ if there was a problem reading from the file. The data is assumed to have been written in the native binary format on the same architecture.

18.9 Random number generator algorithms

The functions described above make no reference to the actual algorithm used. This is deliberate so that you can switch algorithms without having to change any of your application source code. The library provides a large number of generators of different types, including simulation quality generators, generators provided for compatibility with other libraries and historical generators from the past.

The following generators are recommended for use in simulation. They have extremely long periods, low correlation and pass most statistical tests. For the most reliable source of uncorrelated numbers, the second-generation RANLUX generators have the strongest proof of randomness.

gsl_rng_mt19937

[Generator]

The MT19937 generator of Makoto Matsumoto and Takuji Nishimura is a variant of the twisted generalized feedback shift-register algorithm, and is known as the "Mersenne Twister" generator. It has a Mersenne prime period of $2^{19937} - 1$ (about 10^{6000}) and is equi-distributed in 623 dimensions. It has passed the DIEHARD statistical tests. It uses 624 words of state per generator and is comparable in speed to the other generators. The original generator used a default seed of 4357 and choosing s equal to zero in gsl_rng_set reproduces this. Later versions switched to 5489 as the default seed, you can choose this explicitly via gsl_rng_set instead if you require it.

For more information see,

Makoto Matsumoto and Takuji Nishimura, "Mersenne Twister: A 623-dimensionally equidistributed uniform pseudorandom number generator". ACM Transactions on Modeling and Computer Simulation, Vol. 8, No. 1 (Jan. 1998), Pages 3–30

The generator gsl_rng_mt19937 uses the second revision of the seeding procedure published by the two authors above in 2002. The original seeding procedures could cause spurious artifacts for some seed values. They are still available through the alternative generators gsl_rng_mt19937_1999 and gsl_rng_mt19937_1998.

gsl_rng_ranlxs0[Generator]gsl_rng_ranlxs1[Generator]gsl_rng_ranlxs2[Generator]

The generator ranlxs0 is a second-generation version of the RANLUX algorithm of Lüscher, which produces "luxury random numbers". This generator provides single precision output (24 bits) at three luxury levels ranlxs0, ranlxs1 and ranlxs2, in increasing order of strength. It uses double-precision floating point arithmetic internally and can be significantly faster than the integer version of ranlux, particularly on 64-bit architectures. The period of the generator is about 10¹⁷¹. The algorithm has mathematically proven properties and can provide truly decorrelated numbers at a known level of randomness. The higher luxury levels provide increased decorrelation between samples as an additional safety margin.

Note that the range of allowed seeds for this generator is $[0, 2^{31} - 1]$. Higher seed values are wrapped modulo 2^{31} .

gsl_rng_ranlxd1 [Generator]
gsl_rng_ranlxd2 [Generator]

These generators produce double precision output (48 bits) from the RANLXS generator. The library provides two luxury levels ranlxd1 and ranlxd2, in increasing order of strength.

gsl_rng_ranlux [Generator]
gsl_rng_ranlux389 [Generator]

The ranlux generator is an implementation of the original algorithm developed by Lüscher. It uses a lagged-fibonacci-with-skipping algorithm to produce "luxury random numbers". It is a 24-bit generator, originally designed for single-precision IEEE floating point numbers. This implementation is based on integer arithmetic, while the second-generation versions RANLXS and RANLXD described above provide floating-point implementations which will be faster on many platforms. The period of the generator is about 10¹⁷¹. The algorithm has mathematically proven properties and it can provide truly decorrelated numbers at a known level of randomness. The default level of decorrelation recommended by Lüscher is provided by gsl_rng_ranlux, while gsl_rng_ranlux389 gives the highest level of randomness, with all 24 bits decorrelated. Both types of generator use 24 words of state per generator.

For more information see,

M. Lüscher, "A portable high-quality random number generator for lattice field theory calculations", Computer Physics Communications, 79 (1994) 100–110.

F. James, "RANLUX: A Fortran implementation of the high-quality pseudorandom number generator of Lüscher", Computer Physics Communications, 79 (1994) 111–114

gsl_rng_cmrg [Generator]

This is a combined multiple recursive generator by L'Ecuyer. Its sequence is,

$$z_n = (x_n - y_n) \operatorname{mod} m_1$$

where the two underlying generators x_n and y_n are,

$$x_n = (a_1 x_{n-1} + a_2 x_{n-2} + a_3 x_{n-3}) \bmod m_1$$

$$y_n = (b_1 y_{n-1} + b_2 y_{n-2} + b_3 y_{n-3}) \bmod m_2$$

with coefficients $a_1 = 0$, $a_2 = 63308$, $a_3 = -183326$, $b_1 = 86098$, $b_2 = 0$, $b_3 = -539608$, and moduli $m_1 = 2^{31} - 1 = 2147483647$ and $m_2 = 2145483479$.

The period of this generator is $lcm(m_1^3-1, m_2^3-1)$, which is approximately 2^{185} (about 10^{56}). It uses 6 words of state per generator. For more information see,

P. L'Ecuyer, "Combined Multiple Recursive Random Number Generators", Operations Research, 44, 5 (1996), 816–822.

gsl_rng_mrg [Generator]

This is a fifth-order multiple recursive generator by L'Ecuyer, Blouin and Coutre. Its sequence is,

$$x_n = (a_1 x_{n-1} + a_5 x_{n-5}) \mod m$$

with $a_1 = 107374182$, $a_2 = a_3 = a_4 = 0$, $a_5 = 104480$ and $m = 2^{31} - 1$.

The period of this generator is about 10^{46} . It uses 5 words of state per generator. More information can be found in the following paper,

P. L'Ecuyer, F. Blouin, and R. Coutre, "A search for good multiple recursive random number generators", *ACM Transactions on Modeling and Computer Simulation* 3, 87–98 (1993).

gsl_rng_taus [Generator]
gsl_rng_taus2 [Generator]

This is a maximally equidistributed combined Tausworthe generator by L'Ecuyer. The sequence is,

$$x_n = (s_n^1 \oplus s_n^2 \oplus s_n^3)$$

where,

$$\begin{split} s_{n+1}^1 &= (((s_n^1 \& 4294967294) \ll 12) \oplus (((s_n^1 \ll 13) \oplus s_n^1) \gg 19)) \\ s_{n+1}^2 &= (((s_n^2 \& 4294967288) \ll 4) \oplus (((s_n^2 \ll 2) \oplus s_n^2) \gg 25)) \\ s_{n+1}^3 &= (((s_n^3 \& 4294967280) \ll 17) \oplus (((s_n^3 \ll 3) \oplus s_n^3) \gg 11)) \end{split}$$

The period of this generator is 2^{88} (about 10^{26}). It uses 3 words of state per generator. For more information see,

P. L'Ecuyer, "Maximally Equidistributed Combined Tausworthe Generators", Mathematics of Computation, 65, 213 (1996), 203–213.

The generator gsl_rng_taus2 uses the same algorithm as gsl_rng_taus but with an improved seeding procedure described in the paper,

P. L'Ecuyer, "Tables of Maximally Equidistributed Combined LFSR Generators", Mathematics of Computation, 68, 225 (1999), 261–269

The generator gsl_rng_taus2 should now be used in preference to gsl_rng_taus.

gsl_rng_gfsr4 [Generator

The gfsr4 generator is like a lagged-fibonacci generator, and produces each number as an xor'd sum of four previous values.

$$r_n = r_{n-A} \oplus r_{n-B} \oplus r_{n-C} \oplus r_{n-D}$$

Ziff (ref below) notes that "it is now widely known" that two-tap registers (such as R250, which is described below) have serious flaws, the most obvious one being the three-point correlation that comes from the definition of the generator. Nice mathematical properties can be derived for GFSR's, and numerics bears out the claim that 4-tap GFSR's with appropriately chosen offsets are as random as can be measured, using the author's test.

This implementation uses the values suggested the example on p392 of Ziff's article: A = 471, B = 1586, C = 6988, D = 9689.

If the offsets are appropriately chosen (such as the one ones in this implementation), then the sequence is said to be maximal; that means that the period is $2^D - 1$, where D is the longest lag. (It is one less than 2^D because it is not permitted to have all zeros in the ra[] array.) For this implementation with D = 9689 that works out to about 10^{2917} .

Note that the implementation of this generator using a 32-bit integer amounts to 32 parallel implementations of one-bit generators. One consequence of this is that the period of this 32-bit generator is the same as for the one-bit generator. Moreover, this independence means that all 32-bit patterns are equally likely, and in particular that 0 is an allowed random value. (We are grateful to Heiko Bauke for clarifying for us these properties of GFSR random number generators.)

For more information see,

Robert M. Ziff, "Four-tap shift-register-sequence random-number generators", Computers in Physics, 12(4), Jul/Aug 1998, pp 385–392.

18.10 Unix random number generators

The standard Unix random number generators rand, random and rand48 are provided as part of GSL. Although these generators are widely available individually often they aren't all available on the same platform. This makes it difficult to write portable code using them and so we have included the complete set of Unix generators in GSL for convenience. Note that these generators don't produce high-quality randomness and aren't suitable for work requiring accurate statistics. However, if you won't be measuring statistical quantities and just want to introduce some variation into your program then these generators are quite acceptable.

gsl_rng_rand [Generator]

This is the BSD rand generator. Its sequence is

$$x_{n+1} = (ax_n + c) \bmod m$$

with a = 1103515245, c = 12345 and $m = 2^{31}$. The seed specifies the initial value, x_1 . The period of this generator is 2^{31} , and it uses 1 word of storage per generator.

These generators implement the random family of functions, a set of linear feedback shift register generators originally used in BSD Unix. There are several versions of random in use today: the original BSD version (e.g. on SunOS4), a libc5 version (found on older GNU/Linux systems) and a glibc2 version. Each version uses a different seeding procedure, and thus produces different sequences.

The original BSD routines accepted a variable length buffer for the generator state, with longer buffers providing higher-quality randomness. The random function implemented algorithms for buffer lengths of 8, 32, 64, 128 and 256 bytes, and the algorithm with the largest length that would fit into the user-supplied buffer was used. To support these algorithms additional generators are available with the following names,

```
gsl_rng_random8_bsd
gsl_rng_random32_bsd
gsl_rng_random64_bsd
gsl_rng_random128_bsd
gsl_rng_random256_bsd
```

where the numeric suffix indicates the buffer length. The original BSD random function used a 128-byte default buffer and so gsl_rng_random_bsd has been made equivalent to gsl_rng_random128_bsd. Corresponding versions of the libc5 and glibc2 generators are also available, with the names gsl_rng_random8_libc5, gsl_rng_random8_glibc2, etc.

gsl_rng_rand48 [Generator]

This is the Unix rand48 generator. Its sequence is

$$x_{n+1} = (ax_n + c) \bmod m$$

defined on 48-bit unsigned integers with a=25214903917, c=11 and $m=2^{48}$. The seed specifies the upper 32 bits of the initial value, x_1 , with the lower 16 bits set to 0x330E. The function gsl_rng_get returns the upper 32 bits from each term of the sequence. This does not have a direct parallel in the original rand48 functions, but forcing the result to type long int reproduces the output of mrand48. The function gsl_rng_uniform uses the full 48 bits of internal state to return the double precision number x_n/m , which is equivalent to the function drand48. Note that some versions of the GNU C Library contained a bug in mrand48 function which caused it to produce different results (only the lower 16-bits of the return value were set).

18.11 Other random number generators

The generators in this section are provided for compatibility with existing libraries. If you are converting an existing program to use GSL then you can select these generators to check your new implementation against the original one, using the same random number generator. After verifying that your new program reproduces the original results you can then switch to a higher-quality generator.

Note that most of the generators in this section are based on single linear congruence relations, which are the least sophisticated type of generator. In particular, linear congruences have poor properties when used with a non-prime modulus, as several of these routines do (e.g. with a power of two modulus, 2^{31} or 2^{32}). This leads to periodicity in the least significant bits of each number, with only the higher bits having any randomness. Thus if you want to produce a random bitstream it is best to avoid using the least significant bits.

gsl_rng_ranf [Generator]

This is the CRAY random number generator RANF. Its sequence is

$$x_{n+1} = (ax_n) \operatorname{mod} m$$

defined on 48-bit unsigned integers with a=44485709377909 and $m=2^{48}$. The seed specifies the lower 32 bits of the initial value, x_1 , with the lowest bit set to prevent the seed taking an even value. The upper 16 bits of x_1 are set to 0. A consequence of this procedure is that the pairs of seeds 2 and 3, 4 and 5, etc. produce the same sequences.

The generator compatible with the CRAY MATHLIB routine RANF. It produces double precision floating point numbers which should be identical to those from the original RANF.

There is a subtlety in the implementation of the seeding. The initial state is reversed through one step, by multiplying by the modular inverse of $a \mod m$. This is done for compatibility with the original CRAY implementation.

Note that you can only seed the generator with integers up to 2^{32} , while the original CRAY implementation uses non-portable wide integers which can cover all 2^{48} states of the generator.

The function gsl_rng_get returns the upper 32 bits from each term of the sequence. The function gsl_rng_uniform uses the full 48 bits to return the double precision number x_n/m .

The period of this generator is 2^{46} .

gsl_rng_ranmar [Generator]

This is the RANMAR lagged-fibonacci generator of Marsaglia, Zaman and Tsang. It is a 24-bit generator, originally designed for single-precision IEEE floating point numbers. It was included in the CERNLIB high-energy physics library.

gsl_rng_r250 [Generator]

This is the shift-register generator of Kirkpatrick and Stoll. The sequence is based on the recurrence

$$x_n = x_{n-103} \oplus x_{n-250}$$

[Generator]

where \oplus denotes "exclusive-or", defined on 32-bit words. The period of this generator is about 2^{250} and it uses 250 words of state per generator.

For more information see,

S. Kirkpatrick and E. Stoll, "A very fast shift-register sequence random number generator", Journal of Computational Physics, 40, 517–526 (1981)

gsl_rng_tt800 [Generator]

This is an earlier version of the twisted generalized feedback shift-register generator, and has been superseded by the development of MT19937. However, it is still an acceptable generator in its own right. It has a period of 2^{800} and uses 33 words of storage per generator.

For more information see,

Makoto Matsumoto and Yoshiharu Kurita, "Twisted GFSR Generators II", ACM Transactions on Modelling and Computer Simulation, Vol. 4, No. 3, 1994, pages 254–266.

gsl_rng_vax [Generator]

This is the VAX generator MTH\$RANDOM. Its sequence is,

$$x_{n+1} = (ax_n + c) \bmod m$$

with a = 69069, c = 1 and $m = 2^{32}$. The seed specifies the initial value, x_1 . The period of this generator is 2^{32} and it uses 1 word of storage per generator.

gsl_rng_transputer

This is the random number generator from the INMOS Transputer Development system. Its sequence is,

$$x_{n+1} = (ax_n) \bmod m$$

with a = 1664525 and $m = 2^{32}$. The seed specifies the initial value, x_1 .

gsl_rng_randu [Generator]

This is the IBM RANDU generator. Its sequence is

$$x_{n+1} = (ax_n) \bmod m$$

with a = 65539 and $m = 2^{31}$. The seed specifies the initial value, x_1 . The period of this generator was only 2^{29} . It has become a textbook example of a poor generator.

gsl_rng_minstd [Generator]

This is Park and Miller's "minimal standard" MINSTD generator, a simple linear congruence which takes care to avoid the major pitfalls of such algorithms. Its sequence is,

$$x_{n+1} = (ax_n) \operatorname{mod} m$$

with a = 16807 and $m = 2^{31} - 1 = 2147483647$. The seed specifies the initial value, x_1 . The period of this generator is about 2^{31} .

This generator was used in the IMSL Library (subroutine RNUN) and in MATLAB (the RAND function) in the past. It is also sometimes known by the acronym "GGL" (I'm not sure what that stands for).

For more information see,

Park and Miller, "Random Number Generators: Good ones are hard to find", Communications of the ACM, October 1988, Volume 31, No 10, pages 1192–1201.

gsl_rng_uni

[Generator]

gsl_rng_uni32

[Generator]

This is a reimplementation of the 16-bit SLATEC random number generator RUNIF. A generalization of the generator to 32 bits is provided by gsl_rng_uni32. The original source code is available from NETLIB.

gsl_rng_slatec

[Generator]

This is the SLATEC random number generator RAND. It is ancient. The original source code is available from NETLIB.

gsl_rng_zuf

[Generator]

This is the ZUFALL lagged Fibonacci series generator of Peterson. Its sequence is,

$$t = u_{n-273} + u_{n-607}$$
$$u_n = t - \text{floor}(t)$$

The original source code is available from NETLIB. For more information see,

W. Petersen, "Lagged Fibonacci Random Number Generators for the NEC SX-3", International Journal of High Speed Computing (1994).

gsl_rng_knuthran2

[Generator]

This is a second-order multiple recursive generator described by Knuth in Seminumerical Algorithms, 3rd Ed., page 108. Its sequence is,

$$x_n = (a_1 x_{n-1} + a_2 x_{n-2}) \bmod m$$

with $a_1 = 271828183$, $a_2 = 314159269$, and $m = 2^{31} - 1$.

gsl_rng_knuthran2002

[Generator]

gsl_rng_knuthran

[Generator]

This is a second-order multiple recursive generator described by Knuth in Seminumerical Algorithms, 3rd Ed., Section 3.6. Knuth provides its C code. The updated routine gsl_rng_knuthran2002 is from the revised 9th printing and corrects some weaknesses in the earlier version, which is implemented as gsl_rng_knuthran.

gsl_rng_borosh13

[Generator]

gsl_rng_fishman18

[Generator]

gsl_rng_fishman20

[Generator]

gsl_rng_lecuyer21

[Generator]

gsl_rng_waterman14

[Generator]

These multiplicative generators are taken from Knuth's Seminumerical Algorithms, 3rd Ed., pages 106–108. Their sequence is,

$$x_{n+1} = (ax_n) \operatorname{mod} m$$

where the seed specifies the initial value, x_1 . The parameters a and m are as follows, Borosh-Niederreiter: a = 1812433253, $m = 2^{32}$, Fishman 18: a = 62089911, m = 62089911 $2^{31} - 1$, Fishman 20: a = 48271, $m = 2^{31} - 1$, L'Ecuyer: a = 40692, $m = 2^{31} - 249$, Waterman: a = 1566083941, $m = 2^{32}$.

gsl_rng_fishman2x

[Generator]

This is the L'Ecuyer-Fishman random number generator. It is taken from Knuth's Seminumerical Algorithms, 3rd Ed., page 108. Its sequence is,

$$z_{n+1} = (x_n - y_n) \bmod m$$

with $m = 2^{31} - 1$. x_n and y_n are given by the fishman20 and lecuyer21 algorithms. The seed specifies the initial value, x_1 .

gsl_rng_coveyou

[Generator]

This is the Coveyou random number generator. It is taken from Knuth's Seminumerical Algorithms, 3rd Ed., Section 3.2.2. Its sequence is,

$$x_{n+1} = (x_n(x_n+1)) \bmod m$$

with $m=2^{32}$. The seed specifies the initial value, x_1 .

18.12 Performance

The following table shows the relative performance of a selection the available random number generators. The fastest simulation quality generators are taus, gfsr4 and mt19937. The generators which offer the best mathematically-proven quality are those based on the RANLUX algorithm.

```
1754 k ints/sec,
                    870 k doubles/sec, taus
1613 k ints/sec,
                    855 k doubles/sec, gfsr4
1370 k ints/sec.
                    769 k doubles/sec, mt19937
565 k ints/sec,
                    571 k doubles/sec, ranlxs0
400 k ints/sec,
                    405 k doubles/sec, ranlxs1
                    389 k doubles/sec, mrg
490 k ints/sec,
407 k ints/sec,
                    297 k doubles/sec, ranlux
243 k ints/sec,
                    254 k doubles/sec, ranlxd1
 251 k ints/sec,
                    253 k doubles/sec, ranlxs2
 238 k ints/sec,
                    215 k doubles/sec, cmrg
 247 k ints/sec.
                    198 k doubles/sec, ranlux389
 141 k ints/sec,
                    140 k doubles/sec, ranlxd2
```

18.13 Examples

The following program demonstrates the use of a random number generator to produce uniform random numbers in the range [0.0, 1.0),

```
#include <stdio.h>
#include <gsl/gsl_rng.h>
int
main (void)
{
   const gsl_rng_type * T;
```

```
gsl_rng * r;
       int i, n = 10;
       gsl_rng_env_setup();
       T = gsl_rng_default;
       r = gsl_rng_alloc (T);
       for (i = 0; i < n; i++)
         {
           double u = gsl_rng_uniform (r);
           printf ("\%.5f\n", u);
         }
       gsl_rng_free (r);
       return 0;
Here is the output of the program,
     $ ./a.out
     0.99974
     0.16291
     0.28262
     0.94720
     0.23166
     0.48497
     0.95748
     0.74431
     0.54004
     0.73995
```

The numbers depend on the seed used by the generator. The default seed can be changed with the GSL_RNG_SEED environment variable to produce a different stream of numbers. The generator itself can be changed using the environment variable GSL_RNG_TYPE. Here is the output of the program using a seed value of 123 and the multiple-recursive generator mrg,

```
$ GSL_RNG_SEED=123 GSL_RNG_TYPE=mrg ./a.out
GSL_RNG_TYPE=mrg
GSL_RNG_SEED=123
0.33050
0.86631
0.32982
0.67620
0.53391
0.06457
0.16847
```

- 0.70229
- 0.04371
- 0.86374

18.14 References and Further Reading

The subject of random number generation and testing is reviewed extensively in Knuth's Seminumerical Algorithms.

Donald E. Knuth, The Art of Computer Programming: Seminumerical Algorithms (Vol 2, 3rd Ed, 1997), Addison-Wesley, ISBN 0201896842.

Further information is available in the review paper written by Pierre L'Ecuyer,

P. L'Ecuyer, "Random Number Generation", Chapter 4 of the Handbook on Simulation, Jerry Banks Ed., Wiley, 1998, 93–137.

http://www.iro.umontreal.ca/~lecuyer/papers.html in the file handsim.ps.

The source code for the DIEHARD random number generator tests is also available online,

DIEHARD source code G. Marsaglia,

http://stat.fsu.edu/pub/diehard/

A comprehensive set of random number generator tests is available from NIST,

NIST Special Publication 800-22, "A Statistical Test Suite for the Validation of Random Number Generators and Pseudo Random Number Generators for Cryptographic Applications".

http://csrc.nist.gov/rng/

18.15 Acknowledgements

Thanks to Makoto Matsumoto, Takuji Nishimura and Yoshiharu Kurita for making the source code to their generators (MT19937, MM&TN; TT800, MM&YK) available under the GNU General Public License. Thanks to Martin Lüscher for providing notes and source code for the RANLXS and RANLXD generators.

19 Quasi-Random Sequences

This chapter describes functions for generating quasi-random sequences in arbitrary dimensions. A quasi-random sequence progressively covers a d-dimensional space with a set of points that are uniformly distributed. Quasi-random sequences are also known as low-discrepancy sequences. The quasi-random sequence generators use an interface that is similar to the interface for random number generators, except that seeding is not required—each generator produces a single sequence.

The functions described in this section are declared in the header file gsl_qrng.h.

19.1 Quasi-random number generator initialization

gsl_qrng * gsl_qrng_alloc (const gsl_qrng_type * T, unsigned int d) [Function]
This function returns a pointer to a newly-created instance of a quasi-random sequence generator of type T and dimension d. If there is insufficient memory to create the generator then the function returns a null pointer and the error handler is invoked with an error code of GSL_ENOMEM.

```
void gsl\_qrng\_free (gsl\_qrng * q) [Function] This function frees all the memory associated with the generator q.
```

```
void gsl\_qrng\_init (gsl\_qrng * q) [Function] This function reinitializes the generator q to its starting point. Note that quasi-random sequences do not use a seed and always produce the same set of values.
```

19.2 Sampling from a quasi-random number generator

```
int gsl_qrng_get (const gsl_qrng * q, double x[]) [Function] This function stores the next point from the sequence generator q in the array x. The space available for x must match the dimension of the generator. The point x will lie in the range 0 < x_i < 1 for each x_i. An inline version of this function is used when HAVE_INLINE is defined.
```

19.3 Auxiliary quasi-random number generator functions

```
const char * gsl_qrng_name (const gsl_qrng * q) [Function]

This function returns a pointer to the name of the generator.
```

```
size_t gsl_qrng_size (const gsl_qrng * q) [Function]
void * gsl_qrng_state (const gsl_qrng * q) [Function]
These functions return a pointer to the state of generator r and its size. You can
```

These functions return a pointer to the state of generator r and its size. You can use this information to access the state directly. For example, the following code will write the state of a generator to a stream,

```
void * state = gsl_qrng_state (q);
size_t n = gsl_qrng_size (q);
fwrite (state, n, 1, stream);
```

19.4 Saving and resorting quasi-random number generator state

```
int gsl_qrng_memcpy (gsl_qrng * dest, const gsl_qrng * src) [Function]

This function copies the quasi-random sequence generator src into the pre-existing generator dest, making dest into an exact copy of src. The two generators must be of the same type.
```

```
gsl_qrng * gsl_qrng_clone (const gsl_qrng * q) [Function]
This function returns a pointer to a newly created generator which is an exact copy of the generator q.
```

19.5 Quasi-random number generator algorithms

The following quasi-random sequence algorithms are available,

```
gsl_qrng_niederreiter_2
```

[Generator]

This generator uses the algorithm described in Bratley, Fox, Niederreiter, ACM Trans. Model. Comp. Sim. 2, 195 (1992). It is valid up to 12 dimensions.

```
gsl_qrng_sobol [Generator]
```

This generator uses the Sobol sequence described in Antonov, Saleev, USSR Comput. Maths. Math. Phys. 19, 252 (1980). It is valid up to 40 dimensions.

```
gsl_qrng_halton [Generator]
gsl_qrng_reversehalton [Generator]
These generators use the Halton and reverse Halton sequences described in LH Hallon
```

These generators use the Halton and reverse Halton sequences described in J.H. Halton, Numerische Mathematik 2, 84-90 (1960) and B. Vandewoestyne and R. Cools Computational and Applied Mathematics 189, 1&2, 341-361 (2006). They are valid up to 1229 dimensions.

19.6 Examples

The following program prints the first 1024 points of the 2-dimensional Sobol sequence.

```
#include <stdio.h>
#include <gsl/gsl_qrng.h>
int
main (void)
{
   int i;
   gsl_qrng * q = gsl_qrng_alloc (gsl_qrng_sobol, 2);
   for (i = 0; i < 1024; i++)
      {
       double v[2];
       gsl_qrng_get (q, v);
       printf ("%.5f %.5f\n", v[0], v[1]);
   }</pre>
```

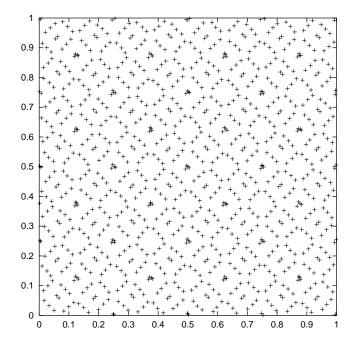
```
gsl_qrng_free (q);
return 0;
}
```

Here is the output from the program,

```
$ ./a.out
0.50000 0.50000
0.75000 0.25000
0.25000 0.75000
0.37500 0.37500
0.87500 0.87500
0.62500 0.12500
0.12500 0.62500
```

It can be seen that successive points progressively fill-in the spaces between previous points.

The following plot shows the distribution in the x-y plane of the first 1024 points from the Sobol sequence,



Distribution of the first 1024 points from the quasi-random Sobol sequence

19.7 References

The implementations of the quasi-random sequence routines are based on the algorithms described in the following paper,

P. Bratley and B.L. Fox and H. Niederreiter, "Algorithm 738: Programs to Generate Niederreiter's Low-discrepancy Sequences", *ACM Transactions on Mathematical Software*, Vol. 20, No. 4, December, 1994, p. 494–495.

20 Random Number Distributions

This chapter describes functions for generating random variates and computing their probability distributions. Samples from the distributions described in this chapter can be obtained using any of the random number generators in the library as an underlying source of randomness.

In the simplest cases a non-uniform distribution can be obtained analytically from the uniform distribution of a random number generator by applying an appropriate transformation. This method uses one call to the random number generator. More complicated distributions are created by the *acceptance-rejection* method, which compares the desired distribution against a distribution which is similar and known analytically. This usually requires several samples from the generator.

The library also provides cumulative distribution functions and inverse cumulative distribution functions, sometimes referred to as quantile functions. The cumulative distribution functions and their inverses are computed separately for the upper and lower tails of the distribution, allowing full accuracy to be retained for small results.

The functions for random variates and probability density functions described in this section are declared in gsl_randist.h. The corresponding cumulative distribution functions are declared in gsl_cdf.h.

Note that the discrete random variate functions always return a value of type unsigned int, and on most platforms this has a maximum value of $2^{32} - 1 \approx 4.29 \times 10^9$. They should only be called with a safe range of parameters (where there is a negligible probability of a variate exceeding this limit) to prevent incorrect results due to overflow.

20.1 Introduction

Continuous random number distributions are defined by a probability density function, p(x), such that the probability of x occurring in the infinitesimal range x to x + dx is p dx.

The cumulative distribution function for the lower tail P(x) is defined by the integral,

$$P(x) = \int_{-\infty}^{x} dx' p(x')$$

and gives the probability of a variate taking a value less than x.

The cumulative distribution function for the upper tail Q(x) is defined by the integral,

$$Q(x) = \int_{x}^{+\infty} dx' p(x')$$

and gives the probability of a variate taking a value greater than x.

The upper and lower cumulative distribution functions are related by P(x) + Q(x) = 1 and satisfy $0 \le P(x) \le 1$, $0 \le Q(x) \le 1$.

The inverse cumulative distributions, $x = P^{-1}(P)$ and $x = Q^{-1}(Q)$ give the values of x which correspond to a specific value of P or Q. They can be used to find confidence limits from probability values.

For discrete distributions the probability of sampling the integer value k is given by p(k), where $\sum_{k} p(k) = 1$. The cumulative distribution for the lower tail P(k) of a discrete

distribution is defined as,

$$P(k) = \sum_{i \le k} p(i)$$

where the sum is over the allowed range of the distribution less than or equal to k.

The cumulative distribution for the upper tail of a discrete distribution $\mathcal{Q}(k)$ is defined as

$$Q(k) = \sum_{i>k} p(i)$$

giving the sum of probabilities for all values greater than k. These two definitions satisfy the identity P(k) + Q(k) = 1.

If the range of the distribution is 1 to n inclusive then P(n) = 1, Q(n) = 0 while P(1) = p(1), Q(1) = 1 - p(1).

20.2 The Gaussian Distribution

double gsl_ran_gaussian (const gsl_rng * r, double sigma) [Function]

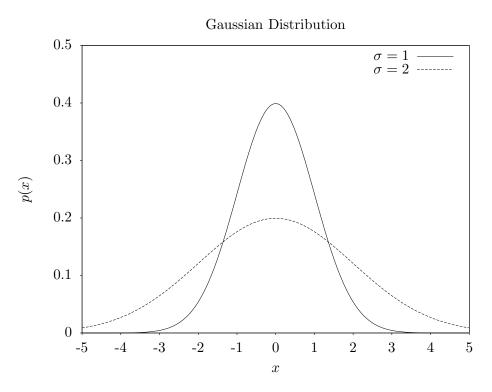
This function returns a Gaussian random variate, with mean zero and standard de-

viation sigma. The probability distribution for Gaussian random variates is,

$$p(x)dx = \frac{1}{\sqrt{2\pi\sigma^2}} \exp(-x^2/2\sigma^2)dx$$

for x in the range $-\infty$ to $+\infty$. Use the transformation $z=\mu+x$ on the numbers returned by gsl_ran_gaussian to obtain a Gaussian distribution with mean μ . This function uses the Box-Muller algorithm which requires two calls to the random number generator r.

double gsl_ran_gaussian_pdf (double x, double sigma) [Function] This function computes the probability density p(x) at x for a Gaussian distribution with standard deviation sigma, using the formula given above.



This function computes a Gaussian random variate using the alternative Marsaglia-Tsang ziggurat and Kinderman-Monahan-Leva ratio methods. The Ziggurat algorithm is the fastest available algorithm in most cases.

```
double gsl_ran_ugaussian (const\ gsl\_rng\ ^*r) [Function] double gsl_ran_ugaussian_pdf (double\ x) [Function] double gsl_ran_ugaussian_ratio_method (const\ gsl\_rng\ ^*r) [Function] These functions compute results for the unit Gaussian distribution. They are equivalent to the functions above with a standard deviation of one, sigma = 1.
```

```
double gsl_cdf_gaussian_P (double x, double sigma)[Function]double gsl_cdf_gaussian_Q (double x, double sigma)[Function]double gsl_cdf_gaussian_Pinv (double P, double sigma)[Function]double gsl_cdf_gaussian_Qinv (double Q, double sigma)[Function]
```

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the Gaussian distribution with standard deviation sigma.

```
\begin{array}{lll} \mbox{double gsl\_cdf\_ugaussian\_P } & (double \ x) & [Function] \\ \mbox{double gsl\_cdf\_ugaussian\_Q } & (double \ x) & [Function] \\ \mbox{double gsl\_cdf\_ugaussian\_Pinv } & (double \ P) & [Function] \\ \mbox{double gsl\_cdf\_ugaussian\_Qinv } & (double \ Q) & [Function] \\ \end{array}
```

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the unit Gaussian distribution.

20.3 The Gaussian Tail Distribution

This function provides random variates from the upper tail of a Gaussian distribution with standard deviation *sigma*. The values returned are larger than the lower limit a, which must be positive. The method is based on Marsaglia's famous rectangle-wedge-tail algorithm (Ann. Math. Stat. 32, 894–899 (1961)), with this aspect explained in Knuth, v2, 3rd ed, p139,586 (exercise 11).

The probability distribution for Gaussian tail random variates is,

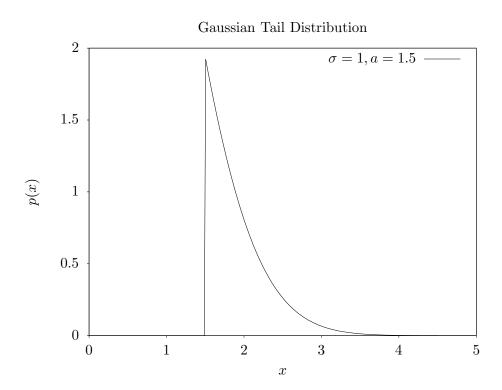
$$p(x)dx = \frac{1}{N(a;\sigma)\sqrt{2\pi\sigma^2}} \exp(-x^2/2\sigma^2)dx$$

for x > a where $N(a; \sigma)$ is the normalization constant,

$$N(a; \sigma) = \frac{1}{2} \operatorname{erfc}\left(\frac{a}{\sqrt{2\sigma^2}}\right).$$

double gsl_ran_gaussian_tail_pdf (double x, double a, double sigma)
[Function]

This function computes the probability density p(x) at x for a Gaussian tail distribution with standard deviation sigma and lower limit a, using the formula given above.



double gsl_ran_ugaussian_tail (const gsl_rng * r, double a) [Function]
double gsl_ran_ugaussian_tail_pdf (double x, double a) [Function]

These functions compute results for the tail of a unit Conscion distribution. There

These functions compute results for the tail of a unit Gaussian distribution. They are equivalent to the functions above with a standard deviation of one, sigma = 1.

20.4 The Bivariate Gaussian Distribution

This function generates a pair of correlated Gaussian variates, with mean zero, correlation coefficient rho and standard deviations $sigma_x$ and $sigma_y$ in the x and y directions. The probability distribution for bivariate Gaussian random variates is,

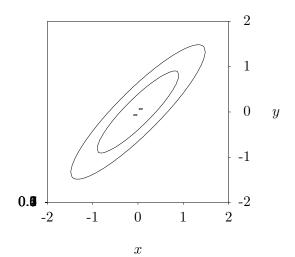
$$p(x,y)dxdy = \frac{1}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \exp\left(-\frac{(x^2/\sigma_x^2 + y^2/\sigma_y^2 - 2\rho xy/(\sigma_x\sigma_y))}{2(1-\rho^2)}\right) dxdy$$

for x, y in the range $-\infty$ to $+\infty$. The correlation coefficient *rho* should lie between 1 and -1.

This function computes the probability density p(x, y) at (x, y) for a bivariate Gaussian distribution with standard deviations $sigma_x$, $sigma_y$ and correlation coefficient rho, using the formula given above.

Bivariate Gaussian Distribution

$$\sigma_x = 1, \sigma_y = 1, \rho = 0.9$$



20.5 The Exponential Distribution

double gsl_ran_exponential (const gsl_rng * r, double mu)

[Function]

This function returns a random variate from the exponential distribution with mean mu. The distribution is,

$$p(x)dx = \frac{1}{\mu} \exp(-x/\mu)dx$$

for $x \geq 0$.

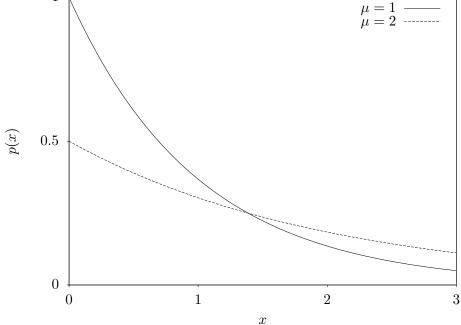
double gsl_ran_exponential_pdf (double x, double mu)

[Function]

This function computes the probability density p(x) at x for an exponential distribution with mean mu, using the formula given above.

Exponential Distribution





double gsl_cdf_exponential_P (double x, double mu) [Function] double gsl_cdf_exponential_Q (double x, double mu) [Function] double gsl_cdf_exponential_Pinv (double P, double mu) [Function] double gsl_cdf_exponential_Qinv (double Q, double mu) [Function]

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the exponential distribution with mean mu.

20.6 The Laplace Distribution

double gsl_ran_laplace (const gsl_rng * r, double a)

[Function]

This function returns a random variate from the Laplace distribution with width a. The distribution is,

$$p(x)dx = \frac{1}{2a}\exp(-|x/a|)dx$$

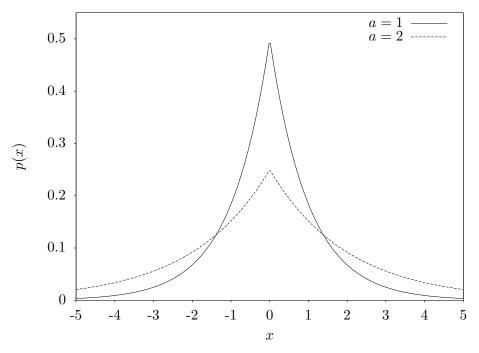
for $-\infty < x < \infty$.

double gsl_ran_laplace_pdf (double x, double a)

[Function]

This function computes the probability density p(x) at x for a Laplace distribution with width a, using the formula given above.

Laplace Distribution (Two-sided Exponential)



double gsl_cdf_laplace_P (double x, double a)[Function]double gsl_cdf_laplace_Q (double x, double a)[Function]double gsl_cdf_laplace_Pinv (double P, double a)[Function]double gsl_cdf_laplace_Qinv (double Q, double a)[Function]

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the Laplace distribution with width a.

20.7 The Exponential Power Distribution

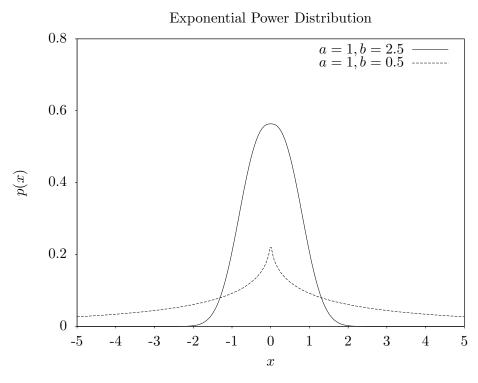
double gsl_ran_exppow (const gsl_rng * r, double a, double b) [Function]

This function returns a random variate from the exponential power distribution with scale parameter a and exponent b. The distribution is,

$$p(x)dx = \frac{1}{2a\Gamma(1+1/b)} \exp(-|x/a|^b)dx$$

for $x \ge 0$. For b=1 this reduces to the Laplace distribution. For b=2 it has the same form as a Gaussian distribution, but with $a=\sqrt{2}\sigma$.

double gsl_ran_exppow_pdf (double x, double a, double b) [Function] This function computes the probability density p(x) at x for an exponential power distribution with scale parameter a and exponent b, using the formula given above.



double gsl_cdf_exppow_P (double x, double a, double b)
double gsl_cdf_exppow_Q (double x, double a, double b)
[Function]
[Function]

These functions compute the cumulative distribution functions P(x), Q(x) for the exponential power distribution with parameters a and b.

20.8 The Cauchy Distribution

double gsl_ran_cauchy (const gsl_rng * r, double a)

[Function]

This function returns a random variate from the Cauchy distribution with scale parameter a. The probability distribution for Cauchy random variates is,

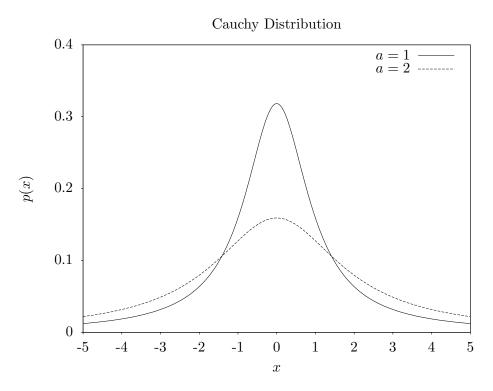
$$p(x)dx = \frac{1}{a\pi(1 + (x/a)^2)}dx$$

for x in the range $-\infty$ to $+\infty$. The Cauchy distribution is also known as the Lorentz distribution.

double gsl_ran_cauchy_pdf (double x, double a)

[Function]

This function computes the probability density p(x) at x for a Cauchy distribution with scale parameter a, using the formula given above.



double gsl_cdf_cauchy_P (double x, double a)[Function]double gsl_cdf_cauchy_Q (double x, double a)[Function]double gsl_cdf_cauchy_Pinv (double P, double a)[Function]double gsl_cdf_cauchy_Qinv (double Q, double a)[Function]

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the Cauchy distribution with scale parameter a.

20.9 The Rayleigh Distribution

double gsl_ran_rayleigh (const gsl_rng * r, double sigma) [Function]

This function returns a random variate from the Rayleigh distribution with scale

parameter sigma. The distribution is,

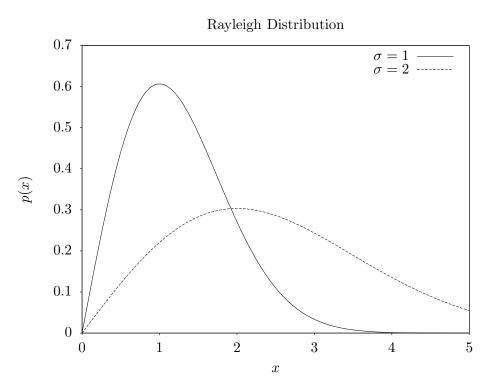
$$p(x)dx = \frac{x}{\sigma^2} \exp(-x^2/(2\sigma^2))dx$$

for x > 0.

double gsl_ran_rayleigh_pdf (double x, double sigma)

[Function]

This function computes the probability density p(x) at x for a Rayleigh distribution with scale parameter sigma, using the formula given above.



double gsl_cdf_rayleigh_P (double x, double sigma)[Function]double gsl_cdf_rayleigh_Q (double x, double sigma)[Function]double gsl_cdf_rayleigh_Pinv (double P, double sigma)[Function]double gsl_cdf_rayleigh_Qinv (double Q, double sigma)[Function]

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the Rayleigh distribution with scale parameter sigma.

20.10 The Rayleigh Tail Distribution

double gsl_ran_rayleigh_tail (const gsl_rng * r, double a, double sigma)
[Function]

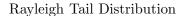
This function returns a random variate from the tail of the Rayleigh distribution with scale parameter *sigma* and a lower limit of a. The distribution is,

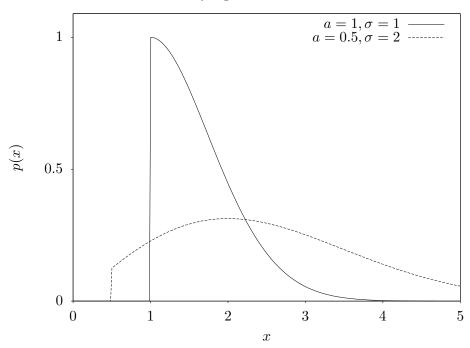
$$p(x)dx = \frac{x}{\sigma^2} \exp((a^2 - x^2)/(2\sigma^2))dx$$

for x > a.

double gsl_ran_rayleigh_tail_pdf (double x, double a, double sigma)
[Function]

This function computes the probability density p(x) at x for a Rayleigh tail distribution with scale parameter sigma and lower limit a, using the formula given above.





20.11 The Landau Distribution

double gsl_ran_landau (const gsl_rng * r)

[Function]

This function returns a random variate from the Landau distribution. The probability distribution for Landau random variates is defined analytically by the complex integral,

$$p(x) = \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} ds \, \exp(s \log(s) + xs)$$

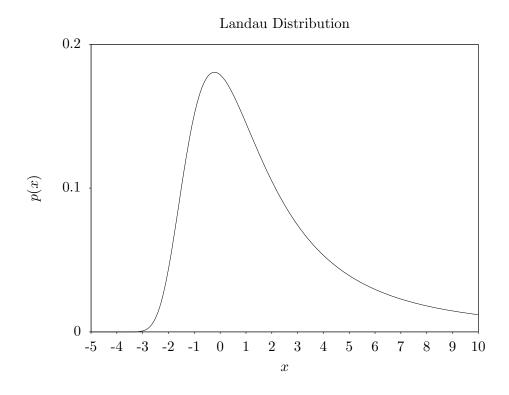
For numerical purposes it is more convenient to use the following equivalent form of the integral,

$$p(x) = (1/\pi) \int_0^\infty dt \exp(-t \log(t) - xt) \sin(\pi t).$$

double gsl_ran_landau_pdf (double x)

[Function]

This function computes the probability density p(x) at x for the Landau distribution using an approximation to the formula given above.



20.12 The Levy alpha-Stable Distributions

double gsl_ran_levy (const gsl_rng * r, double c, double alpha) [Function]

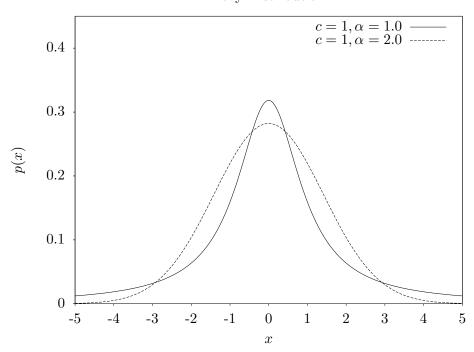
This function returns a random variate from the Levy symmetric stable distribution with scale c and exponent alpha. The symmetric stable probability distribution is defined by a Fourier transform,

$$p(x) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} dt \exp(-itx - |ct|^{\alpha})$$

There is no explicit solution for the form of p(x) and the library does not define a corresponding pdf function. For $\alpha=1$ the distribution reduces to the Cauchy distribution. For $\alpha=2$ it is a Gaussian distribution with $\sigma=\sqrt{2}c$. For $\alpha<1$ the tails of the distribution become extremely wide.

The algorithm only works for $0 < \alpha \le 2$.





20.13 The Levy skew alpha-Stable Distribution

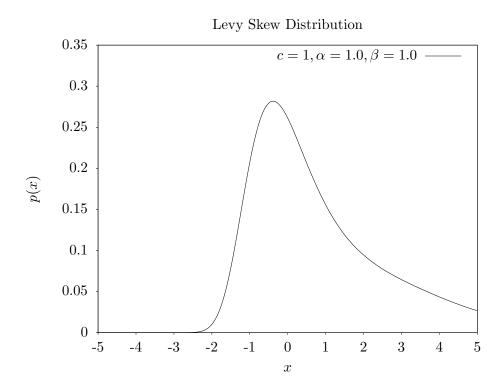
This function returns a random variate from the Levy skew stable distribution with scale c, exponent alpha and skewness parameter beta. The skewness parameter must lie in the range [-1,1]. The Levy skew stable probability distribution is defined by a Fourier transform.

$$p(x) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} dt \exp(-itx - |ct|^{\alpha} (1 - i\beta \operatorname{sign}(t) \tan(\pi\alpha/2)))$$

When $\alpha=1$ the term $\tan(\pi\alpha/2)$ is replaced by $-(2/\pi)\log|t|$. There is no explicit solution for the form of p(x) and the library does not define a corresponding pdf function. For $\alpha=2$ the distribution reduces to a Gaussian distribution with $\sigma=\sqrt{2}c$ and the skewness parameter has no effect. For $\alpha<1$ the tails of the distribution become extremely wide. The symmetric distribution corresponds to $\beta=0$.

The algorithm only works for $0 < \alpha \le 2$.

The Levy alpha-stable distributions have the property that if N alpha-stable variates are drawn from the distribution $p(c, \alpha, \beta)$ then the sum $Y = X_1 + X_2 + \ldots + X_N$ will also be distributed as an alpha-stable variate, $p(N^{1/\alpha}c, \alpha, \beta)$.



20.14 The Gamma Distribution

double gsl_ran_gamma (const gsl_rng * r, double a, double b) [Function]

This function returns a random variate from the gamma distribution. The distribution function is,

$$p(x)dx = \frac{1}{\Gamma(a)b^a}x^{a-1}e^{-x/b}dx$$

for x > 0.

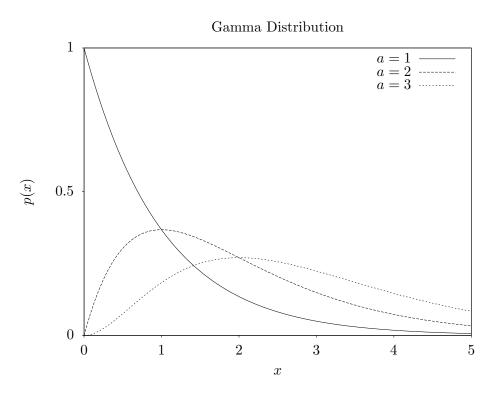
The gamma distribution with an integer parameter a is known as the Erlang distribution.

The variates are computed using the Marsaglia-Tsang fast gamma method. This function for this method was previously called gsl_ran_gamma_mt and can still be accessed using this name.

double gsl_ran_gamma_knuth (const gsl_rng * r, double a, double b) [Function]

This function returns a gamma variate using the algorithms from Knuth (vol 2).

double gsl_ran_gamma_pdf (double x, double a, double b) [Function] This function computes the probability density p(x) at x for a gamma distribution with parameters a and b, using the formula given above.



double gsl_cdf_gamma_P (double x, double a, double b)[Function]double gsl_cdf_gamma_Q (double x, double a, double b)[Function]double gsl_cdf_gamma_Pinv (double P, double a, double b)[Function]

double gsl_cdf_gamma_Qinv (double Q, double a, double b) [Function] These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the gamma distribution with parameters a and b.

20.15 The Flat (Uniform) Distribution

double gsl_ran_flat (const gsl_rng * r, double a, double b) [Function]

This function returns a random variate from the flat (uniform) distribution from a to b. The distribution is,

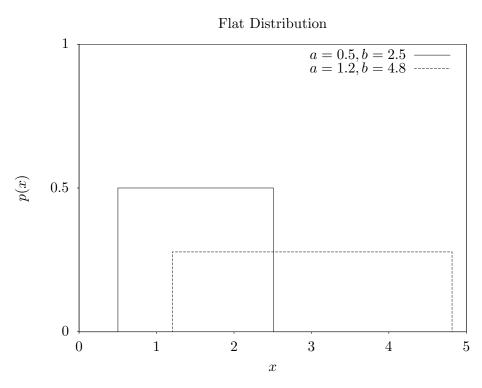
$$p(x)dx = \frac{1}{(b-a)}dx$$

if $a \le x < b$ and 0 otherwise.

double gsl_ran_flat_pdf (double x, double a, double b)

[Function]

This function computes the probability density p(x) at x for a uniform distribution from a to b, using the formula given above.



These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for a uniform distribution from a to b.

20.16 The Lognormal Distribution

double gsl_ran_lognormal (const gsl_rng * r, double zeta, double sigma)
[Function]

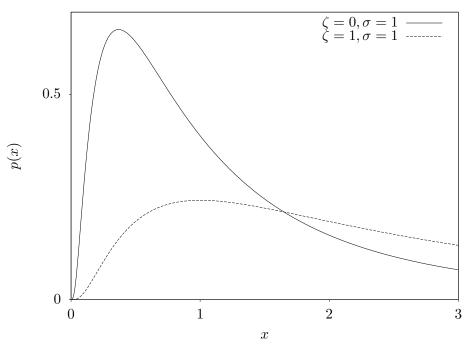
This function returns a random variate from the lognormal distribution. The distribution function is,

$$p(x)dx = \frac{1}{x\sqrt{2\pi\sigma^2}}\exp(-(\ln(x) - \zeta)^2/2\sigma^2)dx$$

for x > 0.

This function computes the probability density p(x) at x for a lognormal distribution with parameters zeta and sigma, using the formula given above.

Lognormal Distribution



These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the lognormal distribution with parameters zeta and sigma.

20.17 The Chi-squared Distribution

The chi-squared distribution arises in statistics. If Y_i are n independent Gaussian random variates with unit variance then the sum-of-squares,

$$X_i = \sum_i Y_i^2$$

has a chi-squared distribution with n degrees of freedom.

[Function]

This function returns a random variate from the chi-squared distribution with nu degrees of freedom. The distribution function is,

$$p(x)dx = \frac{1}{2\Gamma(\nu/2)}(x/2)^{\nu/2-1}\exp(-x/2)dx$$

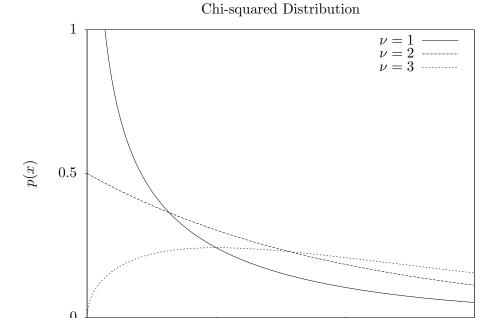
for $x \geq 0$.

double gsl_ran_chisq_pdf (double x, double nu)

[Function]

3

This function computes the probability density p(x) at x for a chi-squared distribution with nu degrees of freedom, using the formula given above.



1

0

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the chi-squared distribution with nu degrees of freedom.

x

2

20.18 The F-distribution

The F-distribution arises in statistics. If Y_1 and Y_2 are chi-squared deviates with ν_1 and ν_2 degrees of freedom then the ratio,

$$X = \frac{(Y_1/\nu_1)}{(Y_2/\nu_2)}$$

has an F-distribution $F(x; \nu_1, \nu_2)$.

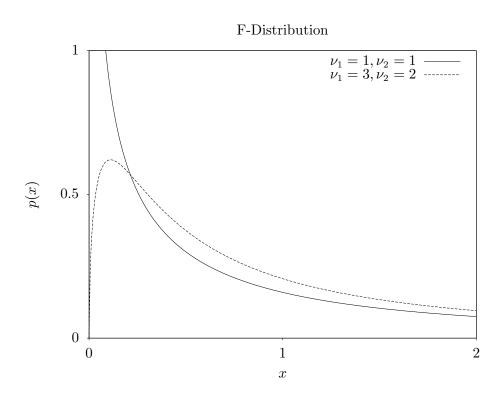
double gsl_ran_fdist (const gsl_rng * r, double nu1, double nu2) [Function]

This function returns a random variate from the F-distribution with degrees of freedom nu1 and nu2. The distribution function is,

$$p(x)dx = \frac{\Gamma((\nu_1 + \nu_2)/2)}{\Gamma(\nu_1/2)\Gamma(\nu_2/2)} \nu_1^{\nu_1/2} \nu_2^{\nu_2/2} x^{\nu_1/2 - 1} (\nu_2 + \nu_1 x)^{-\nu_1/2 - \nu_2/2}$$

for $x \geq 0$.

double gsl_ran_fdist_pdf (double x, double nu1, double nu2) [Function] This function computes the probability density p(x) at x for an F-distribution with nu1 and nu2 degrees of freedom, using the formula given above.



double gsl_cdf_fdist_P (double x, double nu1, double nu2)
double gsl_cdf_fdist_Q (double x, double nu1, double nu2)
double gsl_cdf_fdist_Pinv (double P, double nu1, double nu2)

[Function]
[Function]

double gsl_cdf_fdist_Qinv (double Q, double nu1, double nu2) [Function] These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the F-distribution with nu1 and nu2 degrees of freedom.

20.19 The t-distribution

The t-distribution arises in statistics. If Y_1 has a normal distribution and Y_2 has a chi-squared distribution with ν degrees of freedom then the ratio,

$$X = \frac{Y_1}{\sqrt{Y_2/\nu}}$$

has a t-distribution $t(x; \nu)$ with ν degrees of freedom.

double gsl_ran_tdist (const gsl_rng * r, double nu)

[Function]

This function returns a random variate from the t-distribution. The distribution function is,

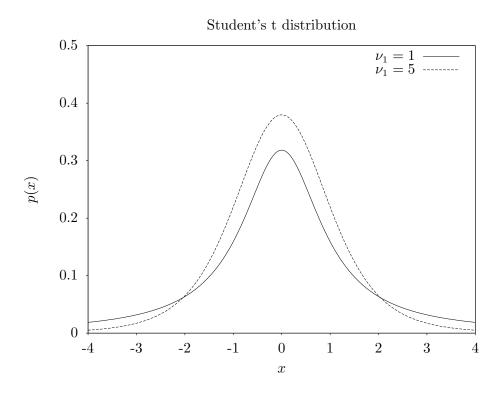
$$p(x)dx = \frac{\Gamma((\nu+1)/2)}{\sqrt{\pi\nu}\Gamma(\nu/2)} (1 + x^2/\nu)^{-(\nu+1)/2} dx$$

for $-\infty < x < +\infty$.

double gsl_ran_tdist_pdf (double x, double nu)

[Function]

This function computes the probability density p(x) at x for a t-distribution with nu degrees of freedom, using the formula given above.



double gsl_cdf_tdist_P (double x, double nu)
double gsl_cdf_tdist_Q (double x, double nu)
double gsl_cdf_tdist_Pinv (double P, double nu)

[Function]

[Function]

[Function]

double gsl_cdf_tdist_Qinv (double Q, double nu)

[Function]

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the t-distribution with nu degrees of freedom.

20.20 The Beta Distribution

double gsl_ran_beta (const gsl_rng * r, double a, double b) [Function]

This function returns a random variate from the beta distribution. The distribution function is,

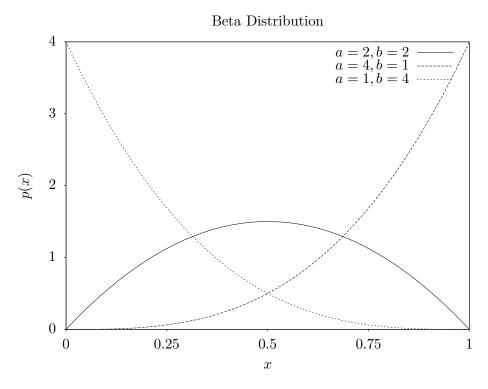
$$p(x)dx = \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)}x^{a-1}(1-x)^{b-1}dx$$

for $0 \le x \le 1$.

double gsl_ran_beta_pdf (double x, double a, double b)

[Function]

This function computes the probability density p(x) at x for a beta distribution with parameters a and b, using the formula given above.



double gsl_cdf_beta_P (double x, double a, double b)[Function]double gsl_cdf_beta_Q (double x, double a, double b)[Function]double gsl_cdf_beta_Pinv (double P, double a, double b)[Function]double gsl_cdf_beta_Qinv (double Q, double a, double b)[Function]

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the beta distribution with parameters a and b.

20.21 The Logistic Distribution

double gsl_ran_logistic (const gsl_rng * r, double a)

[Function]

This function returns a random variate from the logistic distribution. The distribution function is,

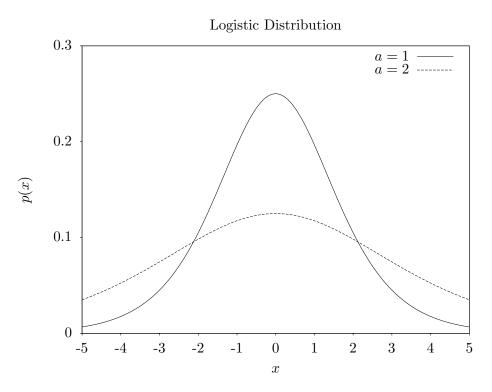
$$p(x)dx = \frac{\exp(-x/a)}{a(1 + \exp(-x/a))^2}dx$$

for $-\infty < x < +\infty$.

double gsl_ran_logistic_pdf (double x, double a)

[Function]

This function computes the probability density p(x) at x for a logistic distribution with scale parameter a, using the formula given above.



double gsl_cdf_logistic_P (double x, double a)[Function]double gsl_cdf_logistic_Q (double x, double a)[Function]double gsl_cdf_logistic_Pinv (double P, double a)[Function]double gsl_cdf_logistic_Qinv (double Q, double a)[Function]

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the logistic distribution with scale parameter a.

20.22 The Pareto Distribution

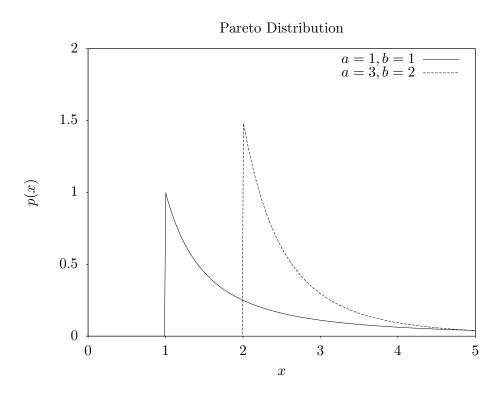
double gsl_ran_pareto (const gsl_rng * r, double a, double b) [Function]

This function returns a random variate from the Pareto distribution of order a. The distribution function is.

$$p(x)dx = (a/b)/(x/b)^{a+1}dx$$

for $x \geq b$.

double gsl_ran_pareto_pdf (double x, double a, double b) [Function] This function computes the probability density p(x) at x for a Pareto distribution with exponent a and scale b, using the formula given above.



double gsl_cdf_pareto_P (double x, double a, double b)[Function]double gsl_cdf_pareto_Q (double x, double a, double b)[Function]double gsl_cdf_pareto_Pinv (double P, double a, double b)[Function]double gsl_cdf_pareto_Qinv (double Q, double a, double b)[Function]

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the Pareto distribution with exponent a and scale b.

20.23 Spherical Vector Distributions

The spherical distributions generate random vectors, located on a spherical surface. They can be used as random directions, for example in the steps of a random walk.

This function returns a random direction vector v = (x,y) in two dimensions. The vector is normalized such that $|v|^2 = x^2 + y^2 = 1$. The obvious way to do this is to take a uniform random number between 0 and 2π and let x and y be the sine and cosine respectively. Two trig functions would have been expensive in the old days, but with modern hardware implementations, this is sometimes the fastest way to go. This is the case for the Pentium (but not the case for the Sun Sparcstation). One can avoid the trig evaluations by choosing x and y in the interior of a unit circle (choose them at random from the interior of the enclosing square, and then reject those that are outside the unit circle), and then dividing by $\sqrt{x^2 + y^2}$. A much cleverer approach, attributed to von Neumann (See Knuth, v2, 3rd ed, p140, exercise 23), requires neither trig nor a square root. In this approach, u and v are chosen at random from the interior of a unit circle, and then $x = (u^2 - v^2)/(u^2 + v^2)$ and $y = 2uv/(u^2 + v^2)$.

This function returns a random direction vector v = (x,y,z) in three dimensions. The vector is normalized such that $|v|^2 = x^2 + y^2 + z^2 = 1$. The method employed is due to Robert E. Knop (CACM 13, 326 (1970)), and explained in Knuth, v2, 3rd ed, p136. It uses the surprising fact that the distribution projected along any axis is actually uniform (this is only true for 3 dimensions).

void gsl_ran_dir_nd (const gsl_rng * r, size_t n, double * x) [Function] This function returns a random direction vector $v = (x_1, x_2, \ldots, x_n)$ in n dimensions. The vector is normalized such that $|v|^2 = x_1^2 + x_2^2 + \cdots + x_n^2 = 1$. The method uses the fact that a multivariate Gaussian distribution is spherically symmetric. Each component is generated to have a Gaussian distribution, and then the components are normalized. The method is described by Knuth, v2, 3rd ed, p135–136, and attributed to G. W. Brown, Modern Mathematics for the Engineer (1956).

20.24 The Weibull Distribution

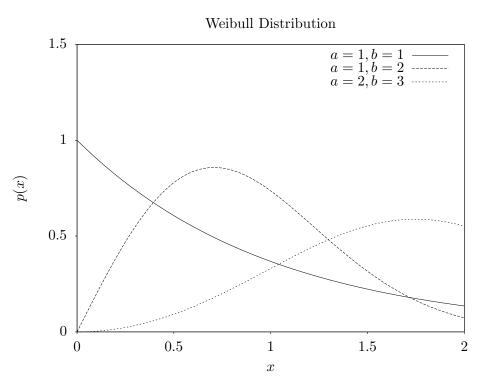
double gsl_ran_weibull (const gsl_rng * r, double a, double b) [Function]

This function returns a random variate from the Weibull distribution. The distribution function is,

$$p(x)dx = \frac{b}{a^b}x^{b-1}\exp(-(x/a)^b)dx$$

for $x \geq 0$.

double gsl_ran_weibull_pdf (double x, double a, double b) [Function] This function computes the probability density p(x) at x for a Weibull distribution with scale a and exponent b, using the formula given above.



double gsl_cdf_weibull_P (double x, double a, double b)[Function]double gsl_cdf_weibull_Q (double x, double a, double b)[Function]double gsl_cdf_weibull_Pinv (double P, double a, double b)[Function]double gsl_cdf_weibull_Qinv (double Q, double a, double b)[Function]

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the Weibull distribution with scale a and exponent b.

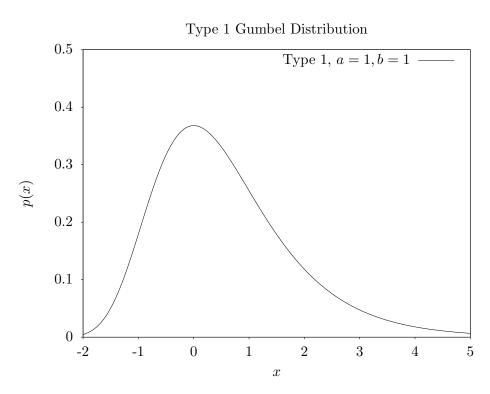
20.25 The Type-1 Gumbel Distribution

double gsl_ran_gumbel1 (const gsl_rng * r, double a, double b) [Function]
This function returns a random variate from the Type-1 Gumbel distribution. The
Type-1 Gumbel distribution function is,

$$p(x)dx = ab\exp(-(b\exp(-ax) + ax))dx$$

for $-\infty < x < \infty$.

double gsl_ran_gumbel1_pdf (double x, double a, double b) [Function] This function computes the probability density p(x) at x for a Type-1 Gumbel distribution with parameters a and b, using the formula given above.



double gsl_cdf_gumbel1_P (double x, double a, double b)[Function]double gsl_cdf_gumbel1_Q (double x, double a, double b)[Function]double gsl_cdf_gumbel1_Pinv (double P, double a, double b)[Function]double gsl_cdf_gumbel1_Qinv (double Q, double a, double b)[Function]

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the Type-1 Gumbel distribution with parameters a and b.

20.26 The Type-2 Gumbel Distribution

double gsl_ran_gumbel2 (const gsl_rng * r, double a, double b) [Function]

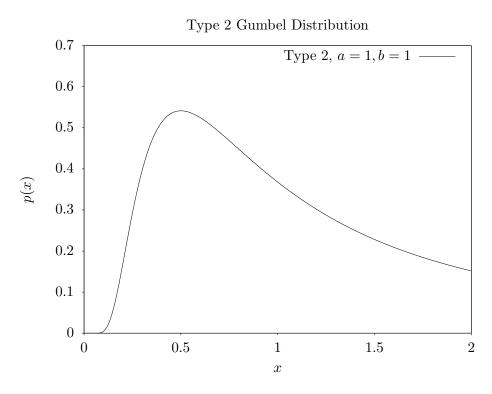
This function returns a random variate from the Type-2 Gumbel distribution. The

Type-2 Gumbel distribution function is,

$$p(x)dx = abx^{-a-1}\exp(-bx^{-a})dx$$

for $0 < x < \infty$.

double gsl_ran_gumbel2_pdf (double x, double a, double b) [Function] This function computes the probability density p(x) at x for a Type-2 Gumbel distribution with parameters a and b, using the formula given above.



double gsl_cdf_gumbel2_P (double x, double a, double b)[Function]double gsl_cdf_gumbel2_Q (double x, double a, double b)[Function]double gsl_cdf_gumbel2_Pinv (double P, double a, double b)[Function]double gsl_cdf_gumbel2_Qinv (double Q, double a, double b)[Function]

These functions compute the cumulative distribution functions P(x), Q(x) and their inverses for the Type-2 Gumbel distribution with parameters a and b.

20.27 The Dirichlet Distribution

This function returns an array of K random variates from a Dirichlet distribution of order K-1. The distribution function is

$$p(\theta_1, \dots, \theta_K) d\theta_1 \cdots d\theta_K = \frac{1}{Z} \prod_{i=1}^K \theta_i^{\alpha_i - 1} \delta(1 - \sum_{i=1}^K \theta_i) d\theta_1 \cdots d\theta_K$$

for $\theta_i \geq 0$ and $\alpha_i > 0$. The delta function ensures that $\sum \theta_i = 1$. The normalization factor Z is

$$Z = \frac{\prod_{i=1}^{K} \Gamma(\alpha_i)}{\Gamma(\sum_{i=1}^{K} \alpha_i)}$$

The random variates are generated by sampling K values from gamma distributions with parameters $a = \alpha_i$, b = 1, and renormalizing. See A.M. Law, W.D. Kelton, Simulation Modeling and Analysis (1991).

This function computes the probability density $p(\theta_1, \ldots, \theta_K)$ at theta[K] for a Dirichlet distribution with parameters alpha[K], using the formula given above.

This function computes the logarithm of the probability density $p(\theta_1, \ldots, \theta_K)$ for a Dirichlet distribution with parameters alpha[K].

20.28 General Discrete Distributions

Given K discrete events with different probabilities P[k], produce a random value k consistent with its probability.

The obvious way to do this is to preprocess the probability list by generating a cumulative probability array with K+1 elements:

$$C[0] = 0$$

$$C[k+1] = C[k] + P[k].$$

Note that this construction produces C[K] = 1. Now choose a uniform deviate u between 0 and 1, and find the value of k such that $C[k] \le u < C[k+1]$. Although this in principle requires of order $\log K$ steps per random number generation, they are fast steps, and if you use something like |uK| as a starting point, you can often do pretty well.

But faster methods have been devised. Again, the idea is to preprocess the probability list, and save the result in some form of lookup table; then the individual calls for a random discrete event can go rapidly. An approach invented by G. Marsaglia (Generating discrete random variables in a computer, Comm ACM 6, 37-38 (1963)) is very clever, and readers interested in examples of good algorithm design are directed to this short and well-written paper. Unfortunately, for large K, Marsaglia's lookup table can be quite large.

A much better approach is due to Alastair J. Walker (An efficient method for generating discrete random variables with general distributions, ACM Trans on Mathematical Software 3, 253–256 (1977); see also Knuth, v2, 3rd ed, p120–121,139). This requires two lookup tables, one floating point and one integer, but both only of size K. After preprocessing, the random numbers are generated in O(1) time, even for large K. The preprocessing suggested by Walker requires $O(K^2)$ effort, but that is not actually necessary, and the implementation provided here only takes O(K) effort. In general, more preprocessing leads to faster generation of the individual random numbers, but a diminishing return is reached pretty early. Knuth points out that the optimal preprocessing is combinatorially difficult for large K.

This method can be used to speed up some of the discrete random number generators below, such as the binomial distribution. To use it for something like the Poisson Distribution, a modification would have to be made, since it only takes a finite set of K outcomes.

This function returns a pointer to a structure that contains the lookup table for the discrete random number generator. The array P[] contains the probabilities of the discrete events; these array elements must all be positive, but they needn't add up to one (so you can think of them more generally as "weights")—the preprocessor will normalize appropriately. This return value is used as an argument for the $gsl_ran_discrete$ function below.

After the preprocessor, above, has been called, you use this function to get the discrete random numbers.

double gsl_ran_discrete_pdf (size_t k, const gsl_ran_discrete_t * g) [Function] Returns the probability P[k] of observing the variable k. Since P[k] is not stored as part of the lookup table, it must be recomputed; this computation takes O(K), so if K is large and you care about the original array P[k] used to create the lookup table, then you should just keep this original array P[k] around.

void $gsl_ran_discrete_free (gsl_ran_discrete_t * g)$ [Function] De-allocates the lookup table pointed to by g.

20.29 The Poisson Distribution

unsigned int gsl_ran_poisson (const gsl_rng * r, double mu) [Function]

This function returns a random integer from the Poisson distribution with mean mu.

The probability distribution for Poisson variates is,

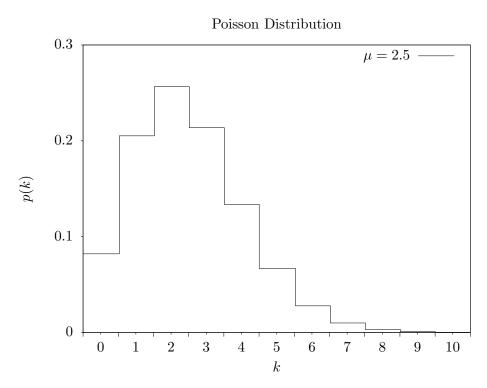
$$p(k) = \frac{\mu^k}{k!} \exp(-\mu)$$

for $k \geq 0$.

double gsl_ran_poisson_pdf (unsigned int k, double mu)

[Function]

This function computes the probability p(k) of obtaining k from a Poisson distribution with mean mu, using the formula given above.



double gsl_cdf_poisson_P (unsigned int k, double mu)
double gsl_cdf_poisson_Q (unsigned int k, double mu)
[Function]

These functions compute the cumulative distribution functions P(k), Q(k) for the Poisson distribution with parameter mu.

20.30 The Bernoulli Distribution

unsigned int gsl_ran_bernoulli (const gsl_rng * r, double p) [Function]
This function returns either 0 or 1, the result of a Bernoulli trial with probability p.
The probability distribution for a Bernoulli trial is,

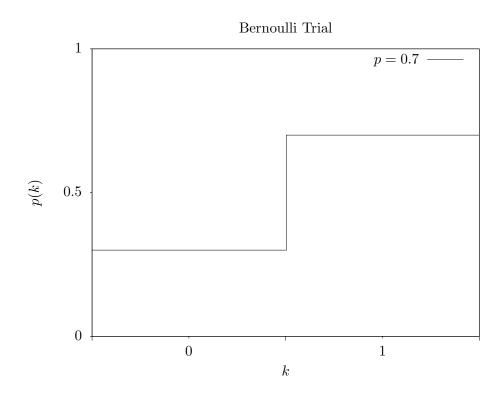
$$p(0) = 1 - p$$
$$p(1) = p$$

double gsl_ran_bernoulli_pdf (unsigned int k, double p)

This function computes the probability n(k) of obtaining k from a Bern

[Function]

This function computes the probability p(k) of obtaining k from a Bernoulli distribution with probability parameter p, using the formula given above.



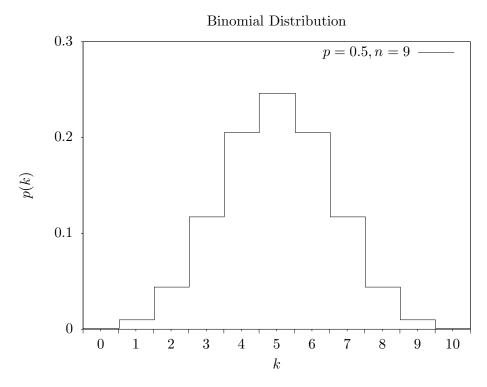
20.31 The Binomial Distribution

This function returns a random integer from the binomial distribution, the number of successes in n independent trials with probability p. The probability distribution for binomial variates is,

$$p(k) = \frac{n!}{k!(n-k)!}p^k(1-p)^{n-k}$$

for $0 \le k \le n$.

This function computes the probability p(k) of obtaining k from a binomial distribution with parameters p and n, using the formula given above.



double gsl_cdf_binomial_P (unsigned int k, double p, unsigned int n) [Function] double gsl_cdf_binomial_Q (unsigned int k, double p, unsigned int n) [Function] These functions compute the cumulative distribution functions P(k), Q(k) for the binomial distribution with parameters p and n.

20.32 The Multinomial Distribution

void gsl_ran_multinomial (const gsl_rng * r, size_t K, unsigned int N, [Function] const double p[], unsigned int n[])

This function computes a random sample n[] from the multinomial distribution formed by N trials from an underlying distribution p[K]. The distribution function for n[] is,

$$P(n_1, n_2, \cdots, n_K) = \frac{N!}{n_1! n_2! \cdots n_K!} p_1^{n_1} p_2^{n_2} \cdots p_K^{n_K}$$

where (n_1, n_2, \ldots, n_K) are nonnegative integers with $\sum_{k=1}^K n_k = N$, and (p_1, p_2, \ldots, p_K) is a probability distribution with $\sum p_i = 1$. If the array p[K] is not normalized then its entries will be treated as weights and normalized appropriately. The arrays n[] and p[] must both be of length K.

Random variates are generated using the conditional binomial method (see C.S. Davis, *The computer generation of multinomial random variates*, Comp. Stat. Data Anal. 16 (1993) 205–217 for details).

double gsl_ran_multinomial_pdf (size_t K, const double p[], const unsigned int n[]) [Function]

This function computes the probability $P(n_1, n_2, ..., n_K)$ of sampling n[K] from a multinomial distribution with parameters p[K], using the formula given above.

double gsl_ran_multinomial_lnpdf (size_t K, const double p[], const unsigned int n[]) [Function]

This function returns the logarithm of the probability for the multinomial distribution $P(n_1, n_2, ..., n_K)$ with parameters p[K].

20.33 The Negative Binomial Distribution

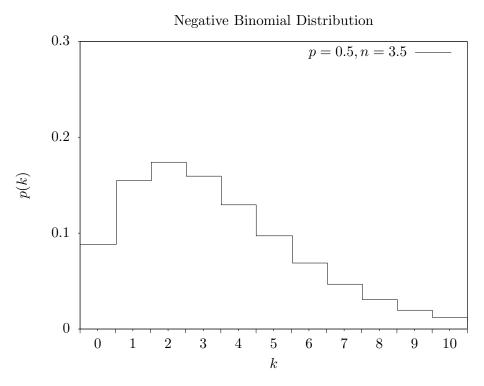
unsigned int gsl_ran_negative_binomial (const gsl_rng * r, double p, double n) [Function]

This function returns a random integer from the negative binomial distribution, the number of failures occurring before n successes in independent trials with probability p of success. The probability distribution for negative binomial variates is,

$$p(k) = \frac{\Gamma(n+k)}{\Gamma(k+1)\Gamma(n)} p^n (1-p)^k$$

Note that n is not required to be an integer.

This function computes the probability p(k) of obtaining k from a negative binomial distribution with parameters p and n, using the formula given above.



These functions compute the cumulative distribution functions P(k), Q(k) for the negative binomial distribution with parameters p and n.

20.34 The Pascal Distribution

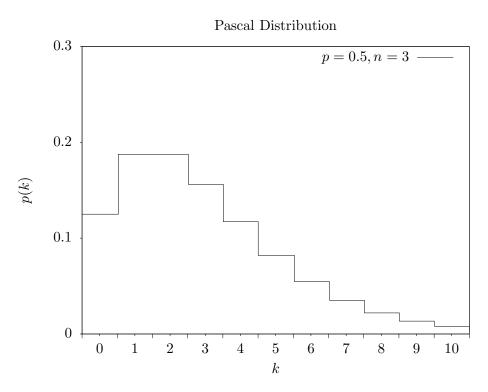
unsigned int gsl_ran_pascal (const gsl_rng * r, double p, unsigned int n)
[Function]

This function returns a random integer from the Pascal distribution. The Pascal distribution is simply a negative binomial distribution with an integer value of n.

$$p(k) = \frac{(n+k-1)!}{k!(n-1)!} p^n (1-p)^k$$

for $k \geq 0$

double gsl_ran_pascal_pdf (unsigned int k, double p, unsigned int n) [Function] This function computes the probability p(k) of obtaining k from a Pascal distribution with parameters p and n, using the formula given above.



double gsl_cdf_pascal_P (unsigned int k, double p, unsigned int n) [Function] double gsl_cdf_pascal_Q (unsigned int k, double p, unsigned int n) [Function] These functions compute the cumulative distribution functions P(k), Q(k) for the Pascal distribution with parameters p and n.

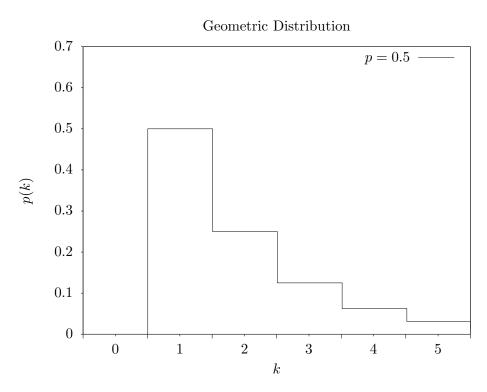
20.35 The Geometric Distribution

unsigned int gsl_ran_geometric ($const\ gsl_rng\ ^*r$, $double\ p$) [Function] This function returns a random integer from the geometric distribution, the number of independent trials with probability p until the first success. The probability distribution for geometric variates is,

$$p(k) = p(1-p)^{k-1}$$

for $k \ge 1$. Note that the distribution begins with k = 1 with this definition. There is another convention in which the exponent k - 1 is replaced by k.

double gsl_ran_geometric_pdf (unsigned int k, double p) [Function] This function computes the probability p(k) of obtaining k from a geometric distribution with probability parameter p, using the formula given above.



double gsl_cdf_geometric_P (unsigned int k, double p)
double gsl_cdf_geometric_Q (unsigned int k, double p)
[Function]

These functions compute the cumulative distribution functions P(k), Q(k) for the geometric distribution with parameter p.

20.36 The Hypergeometric Distribution

unsigned int gsl_ran_hypergeometric (const gsl_rng * r, unsigned int n1, unsigned int n2, unsigned int t) [Function]

This function returns a random integer from the hypergeometric distribution. The probability distribution for hypergeometric random variates is,

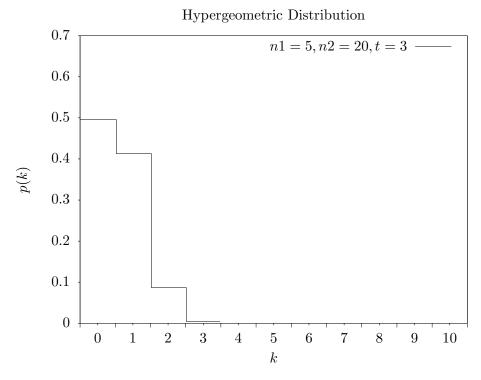
$$p(k) = C(n_1, k)C(n_2, t - k)/C(n_1 + n_2, t)$$

where C(a,b) = a!/(b!(a-b)!) and $t \leq n_1 + n_2$. The domain of k is $\max(0, t - n_2), \ldots, \min(t, n_1)$.

If a population contains n_1 elements of "type 1" and n_2 elements of "type 2" then the hypergeometric distribution gives the probability of obtaining k elements of "type 1" in t samples from the population without replacement.

double gsl_ran_hypergeometric_pdf (unsigned int k, unsigned int
n1, unsigned int n2, unsigned int t)
[Function]

This function computes the probability p(k) of obtaining k from a hypergeometric distribution with parameters n1, n2, t, using the formula given above.



double gsl_cdf_hypergeometric_Q (unsigned int k, unsigned int n1, unsigned int n2, unsigned int t) [Function]

These functions compute the cumulative distribution functions P(k), Q(k) for the hypergeometric distribution with parameters n1, n2 and t.

20.37 The Logarithmic Distribution

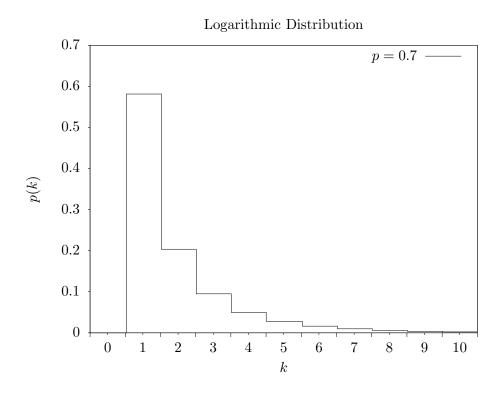
unsigned int gsl_ran_logarithmic (const gsl_rng * r, double p) [Function]

This function returns a random integer from the logarithmic distribution. The probability distribution for logarithmic random variates is,

$$p(k) = \frac{-1}{\log(1-p)} \left(\frac{p^k}{k}\right)$$

for $k \geq 1$.

double gsl_ran_logarithmic_pdf (unsigned int k, double p) [Function] This function computes the probability p(k) of obtaining k from a logarithmic distribution with probability parameter p, using the formula given above.



20.38 Shuffling and Sampling

The following functions allow the shuffling and sampling of a set of objects. The algorithms rely on a random number generator as a source of randomness and a poor quality generator can lead to correlations in the output. In particular it is important to avoid generators with a short period. For more information see Knuth, v2, 3rd ed, Section 3.4.2, "Random Sampling and Shuffling".

This function randomly shuffles the order of n objects, each of size size, stored in the array base[0..n-1]. The output of the random number generator r is used to produce the permutation. The algorithm generates all possible n! permutations with equal probability, assuming a perfect source of random numbers.

The following code shows how to shuffle the numbers from 0 to 51,

```
int a[52];
for (i = 0; i < 52; i++)
    {
        a[i] = i;
    }
gsl_ran_shuffle (r, a, 52, sizeof (int));</pre>
```

This function fills the array dest[k] with k objects taken randomly from the n elements of the array src[0..n-1]. The objects are each of size size. The output of the random number generator r is used to make the selection. The algorithm ensures all possible samples are equally likely, assuming a perfect source of randomness.

The objects are sampled without replacement, thus each object can only appear once in dest[k]. It is required that k be less than or equal to n. The objects in dest will be in the same relative order as those in sc. You will need to call $gsl_ran_shuffle(r, dest, n, size)$ if you want to randomize the order.

The following code shows how to select a random sample of three unique numbers from the set 0 to 99,

```
double a[3], b[100];

for (i = 0; i < 100; i++)
    {
      b[i] = (double) i;
    }

gsl_ran_choose (r, a, 3, b, 100, sizeof (double));</pre>
```

This function is like gsl_ran_choose but samples k items from the original array of n items src with replacement, so the same object can appear more than once in the output sequence dest. There is no requirement that k be less than n in this case.

20.39 Examples

The following program demonstrates the use of a random number generator to produce variates from a distribution. It prints 10 samples from the Poisson distribution with a mean of 3.

```
#include <stdio.h>
#include <gsl/gsl_rng.h>
#include <gsl/gsl_randist.h>
int
main (void)
{
  const gsl_rng_type * T;
  gsl_rng * r;
  int i, n = 10;
  double mu = 3.0;
  /* create a generator chosen by the
     environment variable GSL_RNG_TYPE */
  gsl_rng_env_setup();
  T = gsl_rng_default;
  r = gsl_rng_alloc (T);
  /* print n random variates chosen from
     the poisson distribution with mean
     parameter mu */
  for (i = 0; i < n; i++)
    {
      unsigned int k = gsl_ran_poisson (r, mu);
      printf (" %u", k);
    }
 printf ("\n");
  gsl_rng_free (r);
  return 0;
}
```

If the library and header files are installed under /usr/local (the default location) then the program can be compiled with these options,

```
$ gcc -Wall demo.c -lgsl -lgslcblas -lm
```

Here is the output of the program,

```
$ ./a.out
2 5 5 2 1 0 3 4 1 1
```

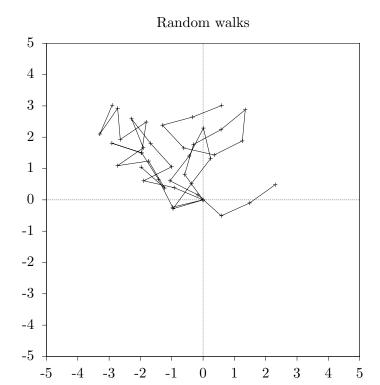
The variates depend on the seed used by the generator. The seed for the default generator type gsl_rng_default can be changed with the GSL_RNG_SEED environment variable to produce a different stream of variates,

```
$ GSL_RNG_SEED=123 ./a.out
GSL_RNG_SEED=123
4 5 6 3 3 1 4 2 5 5
```

The following program generates a random walk in two dimensions.

```
#include <stdio.h>
#include <gsl/gsl_rng.h>
#include <gsl/gsl_randist.h>
int
main (void)
{
  int i;
  double x = 0, y = 0, dx, dy;
  const gsl_rng_type * T;
  gsl_rng * r;
  gsl_rng_env_setup();
  T = gsl_rng_default;
  r = gsl_rng_alloc (T);
 printf ("%g %g\n", x, y);
  for (i = 0; i < 10; i++)
    {
      gsl_ran_dir_2d (r, &dx, &dy);
      x += dx; y += dy;
      printf ("%g %g\n", x, y);
    }
  gsl_rng_free (r);
  return 0;
```

Here is some output from the program, four 10-step random walks from the origin,



The following program computes the upper and lower cumulative distribution functions for the standard normal distribution at x=2.

```
#include <stdio.h>
#include <gsl/gsl_cdf.h>
int
main (void)
{
  double P, Q;
  double x = 2.0;
 P = gsl_cdf_ugaussian_P (x);
 printf ("prob(x < %f) = %f\n", x, P);
 Q = gsl_cdf_ugaussian_Q (x);
 printf ("prob(x > %f) = %f\n", x, Q);
 x = gsl_cdf_ugaussian_Pinv (P);
 printf ("Pinv(%f) = %f\n", P, x);
  x = gsl_cdf_ugaussian_Qinv (Q);
 printf ("Qinv(%f) = %f\n", Q, x);
 return 0;
}
```

Here is the output of the program,

```
prob(x < 2.000000) = 0.977250

prob(x > 2.000000) = 0.022750

Pinv(0.977250) = 2.000000

Qinv(0.022750) = 2.000000
```

20.40 References and Further Reading

For an encyclopaedic coverage of the subject readers are advised to consult the book *Non-Uniform Random Variate Generation* by Luc Devroye. It covers every imaginable distribution and provides hundreds of algorithms.

Luc Devroye, Non-Uniform Random Variate Generation, Springer-Verlag, ISBN 0-387-96305-7. Available online at http://cg.scs.carleton.ca/~luc/rnbookindex.html.

The subject of random variate generation is also reviewed by Knuth, who describes algorithms for all the major distributions.

Donald E. Knuth, The Art of Computer Programming: Seminumerical Algorithms (Vol 2, 3rd Ed, 1997), Addison-Wesley, ISBN 0201896842.

The Particle Data Group provides a short review of techniques for generating distributions of random numbers in the "Monte Carlo" section of its Annual Review of Particle Physics.

Review of Particle Properties R.M. Barnett et al., Physical Review D54, 1 (1996) http://pdg.lbl.gov/.

The Review of Particle Physics is available online in postscript and pdf format.

An overview of methods used to compute cumulative distribution functions can be found in *Statistical Computing* by W.J. Kennedy and J.E. Gentle. Another general reference is *Elements of Statistical Computing* by R.A. Thisted.

William E. Kennedy and James E. Gentle, *Statistical Computing* (1980), Marcel Dekker, ISBN 0-8247-6898-1.

Ronald A. Thisted, *Elements of Statistical Computing* (1988), Chapman & Hall, ISBN 0-412-01371-1.

The cumulative distribution functions for the Gaussian distribution are based on the following papers,

Rational Chebyshev Approximations Using Linear Equations, W.J. Cody, W. Fraser, J.F. Hart. Numerische Mathematik 12, 242–251 (1968).

Rational Chebyshev Approximations for the Error Function, W.J. Cody. Mathematics of Computation 23, n107, 631–637 (July 1969).

21 Statistics

This chapter describes the statistical functions in the library. The basic statistical functions include routines to compute the mean, variance and standard deviation. More advanced functions allow you to calculate absolute deviations, skewness, and kurtosis as well as the median and arbitrary percentiles. The algorithms use recurrence relations to compute average quantities in a stable way, without large intermediate values that might overflow.

The functions are available in versions for datasets in the standard floating-point and integer types. The versions for double precision floating-point data have the prefix gsl_stats and are declared in the header file gsl_statistics_double.h. The versions for integer data have the prefix gsl_stats_int and are declared in the header file gsl_statistics_int.h. All the functions operate on C arrays with a *stride* parameter specifying the spacing between elements.

21.1 Mean, Standard Deviation and Variance

double gsl_stats_mean (const double data[], size_t stride, size_t n) [Function] This function returns the arithmetic mean of data, a dataset of length n with stride stride. The arithmetic mean, or sample mean, is denoted by $\hat{\mu}$ and defined as,

$$\hat{\mu} = \frac{1}{N} \sum x_i$$

where x_i are the elements of the dataset data. For samples drawn from a gaussian distribution the variance of $\hat{\mu}$ is σ^2/N .

This function returns the estimated, or *sample*, variance of *data*, a dataset of length n with stride *stride*. The estimated variance is denoted by $\hat{\sigma}^2$ and is defined by,

$$\hat{\sigma}^2 = \frac{1}{(N-1)} \sum (x_i - \hat{\mu})^2$$

where x_i are the elements of the dataset data. Note that the normalization factor of 1/(N-1) results from the derivation of $\hat{\sigma}^2$ as an unbiased estimator of the population variance σ^2 . For samples drawn from a Gaussian distribution the variance of $\hat{\sigma}^2$ itself is $2\sigma^4/N$.

This function computes the mean via a call to gsl_stats_mean. If you have already computed the mean then you can pass it directly to gsl_stats_variance_m.

double gsl_stats_variance_m (const double data[], size_t stride, [Function] size_t n, double mean)

This function returns the sample variance of data relative to the given value of mean. The function is computed with $\hat{\mu}$ replaced by the value of mean that you supply,

$$\hat{\sigma}^2 = \frac{1}{(N-1)} \sum (x_i - mean)^2$$

The standard deviation is defined as the square root of the variance. These functions return the square root of the corresponding variance functions above.

These functions return the total sum of squares (TSS) of data about the mean. For gsl_stats_tss_m the user-supplied value of mean is used, and for gsl_stats_tss it is computed using gsl_stats_mean.

$$TSS = \sum (x_i - mean)^2$$

This function computes an unbiased estimate of the variance of data when the population mean mean of the underlying distribution is known a priori. In this case the estimator for the variance uses the factor 1/N and the sample mean $\hat{\mu}$ is replaced by the known population mean μ ,

$$\hat{\sigma}^2 = \frac{1}{N} \sum (x_i - \mu)^2$$

double gsl_stats_sd_with_fixed_mean (const double data[], size_t [Function] stride, size_t n, double mean)

This function calculates the standard deviation of data for a fixed population mean mean. The result is the square root of the corresponding variance function.

21.2 Absolute deviation

This function computes the absolute deviation from the mean of data, a dataset of length n with stride stride. The absolute deviation from the mean is defined as,

$$absdev = \frac{1}{N} \sum |x_i - \hat{\mu}|$$

where x_i are the elements of the dataset data. The absolute deviation from the mean provides a more robust measure of the width of a distribution than the variance. This function computes the mean of data via a call to gsl_stats_mean.

This function computes the absolute deviation of the dataset data relative to the given value of mean,

$$absdev = \frac{1}{N} \sum |x_i - mean|$$

This function is useful if you have already computed the mean of data (and want to avoid recomputing it), or wish to calculate the absolute deviation relative to another value (such as zero, or the median).

21.3 Higher moments (skewness and kurtosis)

double gsl_stats_skew (const double data[], size_t stride, size_t n) [Function]
This function computes the skewness of data, a dataset of length n with stride stride.
The skewness is defined as,

$$skew = \frac{1}{N} \sum \left(\frac{x_i - \hat{\mu}}{\hat{\sigma}} \right)^3$$

where x_i are the elements of the dataset data. The skewness measures the asymmetry of the tails of a distribution.

The function computes the mean and estimated standard deviation of data via calls to gsl_stats_mean and gsl_stats_sd.

double gsl_stats_skew_m_sd (const double data[], size_t stride, [Function] size_t n, double mean, double sd)

This function computes the skewness of the dataset data using the given values of the mean mean and standard deviation sd,

$$skew = \frac{1}{N} \sum \left(\frac{x_i - mean}{sd} \right)^3$$

These functions are useful if you have already computed the mean and standard deviation of *data* and want to avoid recomputing them.

This function computes the kurtosis of data, a dataset of length n with stride stride. The kurtosis is defined as,

$$kurtosis = \left(\frac{1}{N}\sum_{i}\left(\frac{x_i - \hat{\mu}}{\hat{\sigma}}\right)^4\right) - 3$$

The kurtosis measures how sharply peaked a distribution is, relative to its width. The kurtosis is normalized to zero for a Gaussian distribution.

This function computes the kurtosis of the dataset data using the given values of the mean mean and standard deviation sd,

$$kurtosis = \frac{1}{N} \left(\sum \left(\frac{x_i - mean}{sd} \right)^4 \right) - 3$$

This function is useful if you have already computed the mean and standard deviation of data and want to avoid recomputing them.

21.4 Autocorrelation

double gsl_stats_lag1_autocorrelation (const double data[], [Function] const size_t stride, const size_t n)

This function computes the lag-1 autocorrelation of the dataset data.

$$a_1 = \frac{\sum_{i=1}^{n} (x_i - \hat{\mu})(x_{i-1} - \hat{\mu})}{\sum_{i=1}^{n} (x_i - \hat{\mu})(x_i - \hat{\mu})}$$

double gsl_stats_lag1_autocorrelation_m (const double data[], const size_t stride, const size_t n, const double mean) [Function]

This function computes the lag-1 autocorrelation of the dataset data using the given value of the mean mean.

21.5 Covariance

double gsl_stats_covariance (const double data1[], const size_t [Function] stride1, const double data2[], const size_t stride2, const size_t n)

This function computes the covariance of the datasets data1 and data2 which must both be of the same length n.

$$covar = \frac{1}{(n-1)} \sum_{i=1}^{n} (x_i - \hat{x})(y_i - \hat{y})$$

double gsl_stats_covariance_m (const double data1[], const size_t [Function] stride1, const double data2[], const size_t stride2, const size_t n, const double mean1, const double mean2)

This function computes the covariance of the datasets data1 and data2 using the given values of the means, mean1 and mean2. This is useful if you have already computed the means of data1 and data2 and want to avoid recomputing them.

21.6 Correlation

double gsl_stats_correlation (const double data1[], const size_t [Function] stride1, const double data2[], const size_t stride2, const size_t n)

This function efficiently computes the Pearson correlation coefficient between the datasets data1 and data2 which must both be of the same length n.

$$r = \frac{cov(x,y)}{\hat{\sigma}_x \hat{\sigma}_y} = \frac{\frac{1}{n-1} \sum (x_i - \hat{x})(y_i - \hat{y})}{\sqrt{\frac{1}{n-1} \sum (x_i - \hat{x})^2} \sqrt{\frac{1}{n-1} \sum (y_i - \hat{y})^2}}$$

This function computes the Spearman rank correlation coefficient between the datasets data1 and data2 which must both be of the same length n. Additional workspace of size 2*n is required in work. The Spearman rank correlation between vectors x and y is equivalent to the Pearson correlation between the ranked vectors

 x_R and y_R , where ranks are defined to be the average of the positions of an element in the ascending order of the values.

21.7 Weighted Samples

The functions described in this section allow the computation of statistics for weighted samples. The functions accept an array of samples, x_i , with associated weights, w_i . Each sample x_i is considered as having been drawn from a Gaussian distribution with variance σ_i^2 . The sample weight w_i is defined as the reciprocal of this variance, $w_i = 1/\sigma_i^2$. Setting a weight to zero corresponds to removing a sample from a dataset.

This function returns the weighted mean of the dataset data with stride stride and length n, using the set of weights w with stride wstride and length n. The weighted mean is defined as,

$$\hat{\mu} = \frac{\sum w_i x_i}{\sum w_i}$$

This function returns the estimated variance of the dataset data with stride stride and length n, using the set of weights w with stride wstride and length n. The estimated variance of a weighted dataset is calculated as,

$$\hat{\sigma}^2 = \frac{\sum w_i}{(\sum w_i)^2 - \sum (w_i^2)} \sum w_i (x_i - \hat{\mu})^2$$

Note that this expression reduces to an unweighted variance with the familiar 1/(N-1) factor when there are N equal non-zero weights.

This function returns the estimated variance of the weighted dataset data using the given weighted mean wmean.

The standard deviation is defined as the square root of the variance. This function returns the square root of the corresponding variance function gsl_stats_wvariance above.

This function returns the square root of the corresponding variance function gsl_stats_wvariance_m above.

This function computes an unbiased estimate of the variance of the weighted dataset data when the population mean mean of the underlying distribution is known a priori. In this case the estimator for the variance replaces the sample mean $\hat{\mu}$ by the known population mean μ ,

$$\hat{\sigma}^2 = \frac{\sum w_i (x_i - \mu)^2}{\sum w_i}$$

double gsl_stats_wsd_with_fixed_mean (const double w[], size_t [Function] wstride, const double data[], size_t stride, size_t n, const double mean)

The standard deviation is defined as the square root of the variance. This function

returns the square root of the corresponding variance function above.

double gsl_stats_wtss_m (const double w[], const size_t wstride, const double data[], size_t stride, size_t n, double wmean) [Function]

These functions return the weighted total sum of squares (TSS) of data about the weighted mean. For gsl_stats_wtss_m the user-supplied value of wmean is used, and for gsl_stats_wtss it is computed using gsl_stats_wmean.

$$TSS = \sum w_i (x_i - wmean)^2$$

This function computes the weighted absolute deviation from the weighted mean of data. The absolute deviation from the mean is defined as,

$$absdev = \frac{\sum w_i |x_i - \hat{\mu}|}{\sum w_i}$$

This function computes the absolute deviation of the weighted dataset data about the given weighted mean wmean.

This function computes the weighted skewness of the dataset data.

$$skew = \frac{\sum w_i((x_i - \hat{x})/\hat{\sigma})^3}{\sum w_i}$$

double gsl_stats_wskew_m_sd (const double w[], size_t wstride, [Function] const double data[], size_t stride, size_t n, double wmean, double wsd)

This function computes the weighted skewness of the dataset data using the given values of the weighted mean and weighted standard deviation, wmean and wsd.

double gsl_stats_wkurtosis (const double w[], size_t wstride, const double data[], size_t stride, size_t n) [Function]

This function computes the weighted kurtosis of the dataset data.

$$kurtosis = rac{\sum w_i((x_i - \hat{x})/\hat{\sigma})^4}{\sum w_i} - 3$$

double gsl_stats_wkurtosis_m_sd (const double w[], size_t wstride, [Function] const double data[], size_t stride, size_t n, double wmean, double wsd)

This function computes the weighted kurtosis of the dataset *data* using the given values of the weighted mean and weighted standard deviation, *wmean* and *wsd*.

21.8 Maximum and Minimum values

The following functions find the maximum and minimum values of a dataset (or their indices). If the data contains NaNs then a NaN will be returned, since the maximum or minimum value is undefined. For functions which return an index, the location of the first NaN in the array is returned.

double gsl_stats_max (const double data[], size_t stride, size_t n) [Function] This function returns the maximum value in data, a dataset of length n with stride stride. The maximum value is defined as the value of the element x_i which satisfies $x_i \ge x_j$ for all j.

If you want instead to find the element with the largest absolute magnitude you will need to apply fabs or abs to your data before calling this function.

double gsl_stats_min (const double data[], size_t stride, size_t n) [Function] This function returns the minimum value in data, a dataset of length n with stride stride. The minimum value is defined as the value of the element x_i which satisfies $x_i \leq x_j$ for all j.

If you want instead to find the element with the smallest absolute magnitude you will need to apply fabs or abs to your data before calling this function.

This function finds both the minimum and maximum values min, max in data in a single pass.

size_t gsl_stats_max_index (const double data[], size_t stride, [Function] size_t n)

This function returns the index of the maximum value in data, a dataset of length n with stride stride. The maximum value is defined as the value of the element x_i which satisfies $x_i \geq x_j$ for all j. When there are several equal maximum elements then the first one is chosen.

This function returns the index of the minimum value in data, a dataset of length n with stride stride. The minimum value is defined as the value of the element x_i which

satisfies $x_i \geq x_j$ for all j. When there are several equal minimum elements then the first one is chosen.

This function returns the indexes min_index, max_index of the minimum and maximum values in data in a single pass.

21.9 Median and Percentiles

The median and percentile functions described in this section operate on sorted data. For convenience we use *quantiles*, measured on a scale of 0 to 1, instead of percentiles (which use a scale of 0 to 100).

This function returns the median value of $sorted_data$, a dataset of length n with stride stride. The elements of the array must be in ascending numerical order. There are no checks to see whether the data are sorted, so the function gsl_sort should always be used first.

When the dataset has an odd number of elements the median is the value of element (n-1)/2. When the dataset has an even number of elements the median is the mean of the two nearest middle values, elements (n-1)/2 and n/2. Since the algorithm for computing the median involves interpolation this function always returns a floating-point number, even for integer data types.

This function returns a quantile value of $sorted_data$, a double-precision array of length n with stride stride. The elements of the array must be in ascending numerical order. The quantile is determined by the f, a fraction between 0 and 1. For example, to compute the value of the 75th percentile f should have the value 0.75.

There are no checks to see whether the data are sorted, so the function gsl_sort should always be used first.

The quantile is found by interpolation, using the formula

quantile =
$$(1 - \delta)x_i + \delta x_{i+1}$$

where i is
$$floor((n-1)f)$$
 and δ is $(n-1)f - i$.

Thus the minimum value of the array (data[0*stride]) is given by f equal to zero, the maximum value (data[(n-1)*stride]) is given by f equal to one and the median value is given by f equal to 0.5. Since the algorithm for computing quantiles involves interpolation this function always returns a floating-point number, even for integer data types.

21.10 Examples

```
Here is a basic example of how to use the statistical functions:
```

```
#include <stdio.h>
  #include <gsl/gsl_statistics.h>
  int
  main(void)
    double data[5] = {17.2, 18.1, 16.5, 18.3, 12.6};
    double mean, variance, largest, smallest;
             = gsl_stats_mean(data, 1, 5);
    variance = gsl_stats_variance(data, 1, 5);
    largest = gsl_stats_max(data, 1, 5);
    smallest = gsl_stats_min(data, 1, 5);
    printf ("The dataset is %g, %g, %g, %g, %g\n",
           data[0], data[1], data[2], data[3], data[4]);
    printf ("The sample mean is %g\n", mean);
    printf ("The estimated variance is %g\n", variance);
    printf ("The largest value is %g\n", largest);
    printf ("The smallest value is %g\n", smallest);
    return 0;
  }
The program should produce the following output,
  The dataset is 17.2, 18.1, 16.5, 18.3, 12.6
  The sample mean is 16.54
  The estimated variance is 5.373
  The largest value is 18.3
  The smallest value is 12.6
Here is an example using sorted data,
  #include <stdio.h>
  #include <gsl/gsl_sort.h>
  #include <gsl/gsl_statistics.h>
  int
  main(void)
    double data[5] = {17.2, 18.1, 16.5, 18.3, 12.6};
    double median, upperq, lowerq;
    printf ("Original dataset:
                                   %g, %g, %g, %g\n",
           data[0], data[1], data[2], data[3], data[4]);
```

```
gsl_sort (data, 1, 5);
    printf ("Sorted dataset: %g, %g, %g, %g, %g\n",
           data[0], data[1], data[2], data[3], data[4]);
    median
      = gsl_stats_median_from_sorted_data (data,
                                            1, 5);
    upperq
      = gsl_stats_quantile_from_sorted_data (data,
                                              0.75);
    lowerq
      = gsl_stats_quantile_from_sorted_data (data,
                                              1, 5,
                                              0.25);
    printf ("The median is %g\n", median);
    printf ("The upper quartile is %g\n", upperq);
    printf ("The lower quartile is %g\n", lowerq);
    return 0;
This program should produce the following output,
  Original dataset: 17.2, 18.1, 16.5, 18.3, 12.6
  Sorted dataset: 12.6, 16.5, 17.2, 18.1, 18.3
  The median is 17.2
  The upper quartile is 18.1
  The lower quartile is 16.5
```

21.11 References and Further Reading

The standard reference for almost any topic in statistics is the multi-volume Advanced Theory of Statistics by Kendall and Stuart.

Maurice Kendall, Alan Stuart, and J. Keith Ord. The Advanced Theory of Statistics (multiple volumes) reprinted as Kendall's Advanced Theory of Statistics. Wiley, ISBN 047023380X.

Many statistical concepts can be more easily understood by a Bayesian approach. The following book by Gelman, Carlin, Stern and Rubin gives a comprehensive coverage of the subject.

Andrew Gelman, John B. Carlin, Hal S. Stern, Donald B. Rubin. *Bayesian Data Analysis*. Chapman & Hall, ISBN 0412039915.

For physicists the Particle Data Group provides useful reviews of Probability and Statistics in the "Mathematical Tools" section of its Annual Review of Particle Physics.

Review of Particle Properties R.M. Barnett et al., Physical Review D54, 1 (1996)

The Review of Particle Physics is available online at the website http://pdg.lbl.gov/.

22 Histograms

This chapter describes functions for creating histograms. Histograms provide a convenient way of summarizing the distribution of a set of data. A histogram consists of a set of bins which count the number of events falling into a given range of a continuous variable x. In GSL the bins of a histogram contain floating-point numbers, so they can be used to record both integer and non-integer distributions. The bins can use arbitrary sets of ranges (uniformly spaced bins are the default). Both one and two-dimensional histograms are supported.

Once a histogram has been created it can also be converted into a probability distribution function. The library provides efficient routines for selecting random samples from probability distributions. This can be useful for generating simulations based on real data.

The functions are declared in the header files gsl_histogram.h and gsl_histogram2d.h.

22.1 The histogram struct

A histogram is defined by the following struct,

gsl_histogram [Data Type]

size_t n This is the number of histogram bins

double * range

The ranges of the bins are stored in an array of n + 1 elements pointed to by range.

double * bin

The counts for each bin are stored in an array of n elements pointed to by bin. The bins are floating-point numbers, so you can increment them by non-integer values if necessary.

The range for bin[i] is given by range[i] to range[i+1]. For n bins there are n+1 entries in the array range. Each bin is inclusive at the lower end and exclusive at the upper end. Mathematically this means that the bins are defined by the following inequality,

bin[i] corresponds to range[i]
$$\leq x < \text{range}[i+1]$$

Here is a diagram of the correspondence between ranges and bins on the number-line for x,

In this picture the values of the range array are denoted by r. On the left-hand side of each bin the square bracket '[' denotes an inclusive lower bound $(r \leq x)$, and the round parentheses ')' on the right-hand side denote an exclusive upper bound (x < r). Thus any samples which fall on the upper end of the histogram are excluded. If you want to include this value for the last bin you will need to add an extra bin to your histogram.

The gsl_histogram struct and its associated functions are defined in the header file gsl_histogram.h.

22.2 Histogram allocation

The functions for allocating memory to a histogram follow the style of malloc and free. In addition they also perform their own error checking. If there is insufficient memory available to allocate a histogram then the functions call the error handler (with an error number of GSL_ENOMEM) in addition to returning a null pointer. Thus if you use the library error handler to abort your program then it isn't necessary to check every histogram alloc.

```
gsl_histogram * gsl_histogram_alloc (size_t n)
```

[Function]

This function allocates memory for a histogram with n bins, and returns a pointer to a newly created ${\tt gsl_histogram}$ struct. If insufficient memory is available a null pointer is returned and the error handler is invoked with an error code of ${\tt GSL_ENOMEM}$. The bins and ranges are not initialized, and should be prepared using one of the range-setting functions below in order to make the histogram ready for use.

```
int gsl_histogram_set_ranges (gsl_histogram * h, const double range[], size_t size) [Function]
```

This function sets the ranges of the existing histogram h using the array range of size size. The values of the histogram bins are reset to zero. The range array should contain the desired bin limits. The ranges can be arbitrary, subject to the restriction that they are monotonically increasing.

The following example shows how to create a histogram with logarithmic bins with ranges [1,10), [10,100) and [100,1000).

```
gsl_histogram * h = gsl_histogram_alloc (3);

/* bin[0] covers the range 1 <= x < 10 */
/* bin[1] covers the range 10 <= x < 100 */
/* bin[2] covers the range 100 <= x < 1000 */
double range[4] = { 1.0, 10.0, 100.0, 1000.0 };

gsl_histogram_set_ranges (h, range, 4);</pre>
```

Note that the size of the range array should be defined to be one element bigger than the number of bins. The additional element is required for the upper value of the final bin.

This function sets the ranges of the existing histogram h to cover the range xmin to xmax uniformly. The values of the histogram bins are reset to zero. The bin ranges are shown in the table below,

```
\begin{array}{lll} & \text{bin}[0] & \text{corresponds to} & xmin \leq x < xmin + d \\ & \text{bin}[1] & \text{corresponds to} & xmin + d \leq x < xmin + 2d \\ & \dots & & \dots \\ & \text{bin}[\text{n-1}] & \text{corresponds to} & xmin + (n-1)d \leq x < xmax \end{array}
```

where d is the bin spacing, d = (xmax - xmin)/n.

void gsl_histogram_free (gsl_histogram * h) [Function]
This function frees the histogram h and all of the memory associated with it.

22.3 Copying Histograms

This function copies the histogram src into the pre-existing histogram dest, making dest into an exact copy of src. The two histograms must be of the same size.

gsl_histogram * gsl_histogram_clone (const gsl_histogram * src) [Function]
This function returns a pointer to a newly created histogram which is an exact copy of the histogram src.

22.4 Updating and accessing histogram elements

There are two ways to access histogram bins, either by specifying an x coordinate or by using the bin-index directly. The functions for accessing the histogram through x coordinates use a binary search to identify the bin which covers the appropriate range.

int gsl_histogram_increment ($gsl_histogram * h$, double x) [Function] This function updates the histogram h by adding one (1.0) to the bin whose range contains the coordinate x.

If x lies in the valid range of the histogram then the function returns zero to indicate success. If x is less than the lower limit of the histogram then the function returns GSL_EDOM , and none of bins are modified. Similarly, if the value of x is greater than or equal to the upper limit of the histogram then the function returns GSL_EDOM , and none of the bins are modified. The error handler is not called, however, since it is often necessary to compute histograms for a small range of a larger dataset, ignoring the values outside the range of interest.

This function is similar to gsl_histogram_increment but increases the value of the appropriate bin in the histogram h by the floating-point number weight.

double gsl_histogram_get (const gsl_histogram * h, size_t i) [Function] This function returns the contents of the *i*-th bin of the histogram h. If i lies outside the valid range of indices for the histogram then the error handler is called with an error code of GSL_EDOM and the function returns 0.

This function finds the upper and lower range limits of the i-th bin of the histogram h. If the index i is valid then the corresponding range limits are stored in lower and upper. The lower limit is inclusive (i.e. events with this coordinate are included in the bin) and the upper limit is exclusive (i.e. events with the coordinate of the upper limit are excluded and fall in the neighboring higher bin, if it exists). The function returns 0 to indicate success. If i lies outside the valid range of indices for the histogram then the error handler is called and the function returns an error code of GSL_EDOM.

double gsl_histogram_max (const gsl_histogram * h)[Function]double gsl_histogram_min (const gsl_histogram * h)[Function]size_t gsl_histogram_bins (const gsl_histogram * h)[Function]

These functions return the maximum upper and minimum lower range limits and the number of bins of the histogram h. They provide a way of determining these values without accessing the $gsl_histogram$ struct directly.

void $gsl_histogram_reset$ ($gsl_histogram * h$) [Function]
This function resets all the bins in the histogram h to zero.

22.5 Searching histogram ranges

The following functions are used by the access and update routines to locate the bin which corresponds to a given x coordinate.

int gsl_histogram_find (const gsl_histogram * h, double x, size_t * i) [Function]

This function finds and sets the index i to the bin number which covers the coordinate x in the histogram h. The bin is located using a binary search. The search includes an optimization for histograms with uniform range, and will return the correct bin immediately in this case. If x is found in the range of the histogram then the function sets the index i and returns GSL_SUCCESS. If x lies outside the valid range of the histogram then the function returns GSL_EDOM and the error handler is invoked.

22.6 Histogram Statistics

- double gsl_histogram_max_val (const gsl_histogram * h) [Function]
 This function returns the maximum value contained in the histogram bins.
- size_t gsl_histogram_max_bin (const gsl_histogram * h) [Function]

 This function returns the index of the bin containing the maximum value. In the case where several bins contain the same maximum value the smallest index is returned.
- double gsl_histogram_min_val (const gsl_histogram * h) [Function] This function returns the minimum value contained in the histogram bins.
- size_t gsl_histogram_min_bin (const gsl_histogram * h) [Function]

 This function returns the index of the bin containing the minimum value. In the case where several bins contain the same maximum value the smallest index is returned.
- double gsl_histogram_mean (const gsl_histogram * h) [Function]

 This function returns the mean of the histogrammed variable, where the histogram is regarded as a probability distribution. Negative bin values are ignored for the purposes of this calculation. The accuracy of the result is limited by the bin width.
- double gsl_histogram_sigma (const gsl_histogram * h) [Function]

 This function returns the standard deviation of the histogrammed variable, where the histogram is regarded as a probability distribution. Negative bin values are ignored for the purposes of this calculation. The accuracy of the result is limited by the bin width.

double gsl_histogram_sum (const gsl_histogram * h) [Function]

This function returns the sum of all bin values. Negative bin values are included in the sum.

22.7 Histogram Operations

- - This function returns 1 if the all of the individual bin ranges of the two histograms are identical, and 0 otherwise.
- int gsl_histogram_add (gsl_histogram * h1, const gsl_histogram * h2) [Function] This function adds the contents of the bins in histogram h2 to the corresponding bins of histogram h1, i.e. $h'_1(i) = h_1(i) + h_2(i)$. The two histograms must have identical bin ranges.
- int gsl_histogram_sub (gsl_histogram * h1, const gsl_histogram * h2) [Function] This function subtracts the contents of the bins in histogram h2 from the corresponding bins of histogram h1, i.e. $h'_1(i) = h_1(i) h_2(i)$. The two histograms must have identical bin ranges.
- int gsl_histogram_mul (gsl_histogram * h1, const gsl_histogram * h2) [Function] This function multiplies the contents of the bins of histogram h1 by the contents of the corresponding bins in histogram h2, i.e. $h'_1(i) = h_1(i) * h_2(i)$. The two histograms must have identical bin ranges.
- int gsl_histogram_div (gsl_histogram * h1, const gsl_histogram * h2) [Function] This function divides the contents of the bins of histogram h1 by the contents of the corresponding bins in histogram h2, i.e. $h'_1(i) = h_1(i)/h_2(i)$. The two histograms must have identical bin ranges.
- int gsl_histogram_scale (gsl_histogram * h, double scale) [Function] This function multiplies the contents of the bins of histogram h by the constant scale, i.e. $h'_1(i) = h_1(i) * scale$.
- int gsl_histogram_shift (gsl_histogram * h, double offset) [Function] This function shifts the contents of the bins of histogram h by the constant offset, i.e. $h'_1(i) = h_1(i) + offset$.

22.8 Reading and writing histograms

The library provides functions for reading and writing histograms to a file as binary data or formatted text.

int gsl_histogram_fwrite (FILE * stream, const gsl_histogram * h) [Function] This function writes the ranges and bins of the histogram h to the stream stream in binary format. The return value is 0 for success and GSL_EFAILED if there was a problem writing to the file. Since the data is written in the native binary format it may not be portable between different architectures.

int gsl_histogram_fread (FILE * stream, gsl_histogram * h) [Function]

This function reads into the histogram h from the open stream stream in binary format. The histogram h must be preallocated with the correct size since the function uses the number of bins in h to determine how many bytes to read. The return value is 0 for success and GSL_EFAILED if there was a problem reading from the file. The data is assumed to have been written in the native binary format on the same architecture.

int gsl_histogram_fprintf (FILE * stream, const gsl_histogram * h, const char * range_format, const char * bin_format) [Function]

This function writes the ranges and bins of the histogram h line-by-line to the stream stream using the format specifiers $range_format$ and bin_format . These should be one of the %g, %e or %f formats for floating point numbers. The function returns 0 for success and $GSL_EFAILED$ if there was a problem writing to the file. The histogram output is formatted in three columns, and the columns are separated by spaces, like this,

```
range[0] range[1] bin[0]
range[1] range[2] bin[1]
range[2] range[3] bin[2]
....
range[n-1] range[n] bin[n-1]
```

The values of the ranges are formatted using range_format and the value of the bins are formatted using bin_format. Each line contains the lower and upper limit of the range of the bins and the value of the bin itself. Since the upper limit of one bin is the lower limit of the next there is duplication of these values between lines but this allows the histogram to be manipulated with line-oriented tools.

int gsl_histogram_fscanf (FILE * stream, gsl_histogram * h) [Function]

This function reads formatted data from the stream stream into the histogram h. The data is assumed to be in the three-column format used by gsl_histogram_fprintf. The histogram h must be preallocated with the correct length since the function uses the size of h to determine how many numbers to read. The function returns 0 for success and GSL_EFAILED if there was a problem reading from the file.

22.9 Resampling from histograms

A histogram made by counting events can be regarded as a measurement of a probability distribution. Allowing for statistical error, the height of each bin represents the probability of an event where the value of x falls in the range of that bin. The probability distribution function has the one-dimensional form p(x)dx where,

$$p(x) = n_i/(Nw_i)$$

In this equation n_i is the number of events in the bin which contains x, w_i is the width of the bin and N is the total number of events. The distribution of events within each bin is assumed to be uniform.

22.10 The histogram probability distribution struct

The probability distribution function for a histogram consists of a set of bins which measure the probability of an event falling into a given range of a continuous variable x. A probability

distribution function is defined by the following struct, which actually stores the cumulative probability distribution function. This is the natural quantity for generating samples via the inverse transform method, because there is a one-to-one mapping between the cumulative probability distribution and the range [0,1]. It can be shown that by taking a uniform random number in this range and finding its corresponding coordinate in the cumulative probability distribution we obtain samples with the desired probability distribution.

gsl_histogram_pdf

[Data Type]

size_t n This is the number of bins used to approximate the probability distribution function.

double * range

The ranges of the bins are stored in an array of n + 1 elements pointed to by range.

double * sum

The cumulative probability for the bins is stored in an array of n elements pointed to by sum.

The following functions allow you to create a gsl_histogram_pdf struct which represents this probability distribution and generate random samples from it.

gsl_histogram_pdf * gsl_histogram_pdf_alloc (size_t n) [Function]

This function allocates memory for a probability distribution with *n* bins and returns a pointer to a newly initialized gsl_histogram_pdf struct. If insufficient memory is available a null pointer is returned and the error handler is invoked with an error code of GSL_ENOMEM.

This function initializes the probability distribution p with the contents of the histogram h. If any of the bins of h are negative then the error handler is invoked with an error code of GSL_EDOM because a probability distribution cannot contain negative values.

$\verb|void gsl_histogram_pdf_free (gsl_histogram_pdf * p)| [Function]|$

This function frees the probability distribution function p and all of the memory associated with it.

double gsl_histogram_pdf_sample (
$$const gsl_histogram_pdf * p$$
, [Function] $double r$)

This function uses r, a uniform random number between zero and one, to compute a single random sample from the probability distribution p. The algorithm used to compute the sample s is given by the following formula,

$$s = \text{range}[i] + \delta * (\text{range}[i+1] - \text{range}[i])$$

where i is the index which satisfies $sum[i] \le r < sum[i+1]$ and delta is (r - sum[i])/(sum[i+1] - sum[i]).

22.11 Example programs for histograms

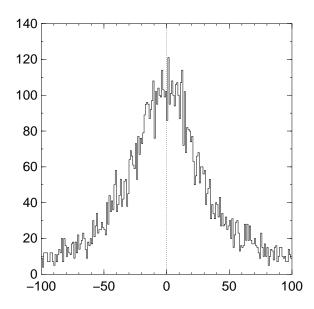
The following program shows how to make a simple histogram of a column of numerical data supplied on stdin. The program takes three arguments, specifying the upper and lower bounds of the histogram and the number of bins. It then reads numbers from stdin, one line at a time, and adds them to the histogram. When there is no more data to read it prints out the accumulated histogram using gsl_histogram_fprintf.

```
#include <stdio.h>
#include <stdlib.h>
#include <gsl/gsl_histogram.h>
int
main (int argc, char **argv)
 double a, b;
  size_t n;
  if (argc != 4)
    {
      printf ("Usage: gsl-histogram xmin xmax n\n"
              "Computes a histogram of the data "
              "on stdin using n bins from xmin "
              "to xmax\n");
      exit(0);
    }
  a = atof (argv[1]);
  b = atof (argv[2]);
  n = atoi (argv[3]);
  {
    double x;
    gsl_histogram * h = gsl_histogram_alloc (n);
    gsl_histogram_set_ranges_uniform (h, a, b);
    while (fscanf (stdin, "%lg", &x) == 1)
      {
        gsl_histogram_increment (h, x);
    gsl_histogram_fprintf (stdout, h, "%g", "%g");
    gsl_histogram_free (h);
  }
  exit (0);
}
```

Here is an example of the program in use. We generate 10000 random samples from a Cauchy distribution with a width of 30 and histogram them over the range -100 to 100, using 200 bins.

[Data Type]

A plot of the resulting histogram shows the familiar shape of the Cauchy distribution and the fluctuations caused by the finite sample size.



22.12 Two dimensional histograms

A two dimensional histogram consists of a set of bins which count the number of events falling in a given area of the (x,y) plane. The simplest way to use a two dimensional histogram is to record two-dimensional position information, n(x,y). Another possibility is to form a joint distribution by recording related variables. For example a detector might record both the position of an event (x) and the amount of energy it deposited E. These could be histogrammed as the joint distribution n(x, E).

22.13 The 2D histogram struct

Two dimensional histograms are defined by the following struct,

```
gsl_histogram2d
```

size_t nx, ny

This is the number of histogram bins in the x and y directions.

double * xrange

The ranges of the bins in the x-direction are stored in an array of nx + 1 elements pointed to by xrange.

double * yrange

The ranges of the bins in the y-direction are stored in an array of ny + 1 elements pointed to by yrange.

double * bin

The counts for each bin are stored in an array pointed to by bin. The bins are floating-point numbers, so you can increment them by non-integer values if necessary. The array bin stores the two dimensional array of bins in a single block of memory according to the mapping bin(i,j) = bin[i * ny + j].

The range for bin(i,j) is given by xrange[i] to xrange[i+1] in the x-direction and yrange[j] to yrange[j+1] in the y-direction. Each bin is inclusive at the lower end and exclusive at the upper end. Mathematically this means that the bins are defined by the following inequality,

```
bin(i,j) corresponds to xrange[i] \le x < xrange[i+1]
and yrange[j] \le y < yrange[j+1]
```

Note that any samples which fall on the upper sides of the histogram are excluded. If you want to include these values for the side bins you will need to add an extra row or column to your histogram.

The gsl_histogram2d struct and its associated functions are defined in the header file gsl_histogram2d.h.

22.14 2D Histogram allocation

zero.

The functions for allocating memory to a 2D histogram follow the style of malloc and free. In addition they also perform their own error checking. If there is insufficient memory available to allocate a histogram then the functions call the error handler (with an error number of GSL_ENOMEM) in addition to returning a null pointer. Thus if you use the library error handler to abort your program then it isn't necessary to check every 2D histogram alloc.

- gsl_histogram2d * gsl_histogram2d_alloc (size_t nx, size_t ny) [Function]

 This function allocates memory for a two-dimensional histogram with nx bins in the x direction and ny bins in the y direction. The function returns a pointer to a newly created gsl_histogram2d struct. If insufficient memory is available a null pointer is returned and the error handler is invoked with an error code of GSL_ENOMEM. The bins and ranges must be initialized with one of the functions below before the histogram is ready for use.
- int gsl_histogram2d_set_ranges (gsl_histogram2d * h, const double [Function] xrange[], size_t xsize, const double yrange[], size_t ysize)

 This function sets the ranges of the existing histogram h using the arrays xrange and yrange of size xsize and ysize respectively. The values of the histogram bins are reset to zero.
- int gsl_histogram2d_set_ranges_uniform (gsl_histogram2d * h, double xmin, double xmax, double ymin, double ymax)

 This function sets the ranges of the existing histogram h to cover the ranges xmin to xmax and ymin to ymax uniformly. The values of the histogram bins are reset to

void gsl_histogram2d_free (gsl_histogram2d * h)

[Function]

This function frees the 2D histogram h and all of the memory associated with it.

22.15 Copying 2D Histograms

[Function]

This function copies the histogram src into the pre-existing histogram dest, making dest into an exact copy of src. The two histograms must be of the same size.

[Function]

This function returns a pointer to a newly created histogram which is an exact copy of the histogram src.

22.16 Updating and accessing 2D histogram elements

You can access the bins of a two-dimensional histogram either by specifying a pair of (x,y) coordinates or by using the bin indices (i,j) directly. The functions for accessing the histogram through (x,y) coordinates use binary searches in the x and y directions to identify the bin which covers the appropriate range.

int gsl_histogram2d_increment (gsl_histogram2d * h, double x, double x) [Function]

This function updates the histogram h by adding one (1.0) to the bin whose x and y ranges contain the coordinates (x,y).

If the point (x,y) lies inside the valid ranges of the histogram then the function returns zero to indicate success. If (x,y) lies outside the limits of the histogram then the function returns GSL_EDOM, and none of the bins are modified. The error handler is not called, since it is often necessary to compute histograms for a small range of a larger dataset, ignoring any coordinates outside the range of interest.

int gsl_histogram2d_accumulate (gsl_histogram2d * h, double x, double x, double y, double weight) [Function]

This function is similar to $gsl_histogram2d_increment$ but increases the value of the appropriate bin in the histogram h by the floating-point number weight.

This function returns the contents of the (i,j)-th bin of the histogram h. If (i,j) lies outside the valid range of indices for the histogram then the error handler is called with an error code of GSL_EDOM and the function returns 0.

These functions find the upper and lower range limits of the i-th and j-th bins in the x and y directions of the histogram h. The range limits are stored in xlower and

xupper or ylower and yupper. The lower limits are inclusive (i.e. events with these coordinates are included in the bin) and the upper limits are exclusive (i.e. events with the value of the upper limit are not included and fall in the neighboring higher bin, if it exists). The functions return 0 to indicate success. If i or j lies outside the valid range of indices for the histogram then the error handler is called with an error code of GSL_EDOM .

```
double gsl_histogram2d_xmax (const gsl_histogram2d * h) [Function]
double gsl_histogram2d_xmin (const gsl_histogram2d * h) [Function]
size_t gsl_histogram2d_nx (const gsl_histogram2d * h) [Function]
double gsl_histogram2d_ymax (const gsl_histogram2d * h) [Function]
double gsl_histogram2d_ymin (const gsl_histogram2d * h) [Function]
size_t gsl_histogram2d_ny (const gsl_histogram2d * h) [Function]
```

These functions return the maximum upper and minimum lower range limits and the number of bins for the x and y directions of the histogram h. They provide a way of determining these values without accessing the gsl_histogram2d struct directly.

```
void gsl_histogram2d_reset (gsl_histogram2d * h) [Function] This function resets all the bins of the histogram h to zero.
```

22.17 Searching 2D histogram ranges

The following functions are used by the access and update routines to locate the bin which corresponds to a given (x, y) coordinate.

This function finds and sets the indices i and j to the bin which covers the coordinates (x,y). The bin is located using a binary search. The search includes an optimization for histograms with uniform ranges, and will return the correct bin immediately in this case. If (x,y) is found then the function sets the indices (i,j) and returns GSL_SUCCESS. If (x,y) lies outside the valid range of the histogram then the function returns GSL_EDOM and the error handler is invoked.

22.18 2D Histogram Statistics

```
double gsl_histogram2d_max_val (const gsl_histogram2d * h) [Function]
This function returns the maximum value contained in the histogram bins.
```

```
void gsl_histogram2d_max_bin (const \ gsl_histogram2d * h, size_t * i, [Function] size_t * j)
```

This function finds the indices of the bin containing the maximum value in the histogram h and stores the result in (i,j). In the case where several bins contain the same maximum value the first bin found is returned.

```
double gsl_histogram2d_min_val (const gsl_histogram2d * h) [Function]
This function returns the minimum value contained in the histogram bins.
```

This function finds the indices of the bin containing the minimum value in the histogram h and stores the result in (i,j). In the case where several bins contain the same maximum value the first bin found is returned.

double gsl_histogram2d_xmean (const gsl_histogram2d * h) [Function]

This function returns the mean of the histogrammed x variable, where the histogram is regarded as a probability distribution. Negative bin values are ignored for the purposes of this calculation.

double gsl_histogram2d_ymean (const gsl_histogram2d * h) [Function]

This function returns the mean of the histogrammed y variable, where the histogram is regarded as a probability distribution. Negative his values are ignored for the

is regarded as a probability distribution. Negative bin values are ignored for the purposes of this calculation.

- double gsl_histogram2d_xsigma (const gsl_histogram2d * h) [Function]

 This function returns the standard deviation of the histogrammed x variable, where the histogram is regarded as a probability distribution. Negative bin values are ignored for the purposes of this calculation.
- double gsl_histogram2d_ysigma (const gsl_histogram2d * h) [Function]

 This function returns the standard deviation of the histogrammed y variable, where the histogram is regarded as a probability distribution. Negative bin values are ignored for the purposes of this calculation.
- double gsl_histogram2d_cov (const gsl_histogram2d * h) [Function]
 This function returns the covariance of the histogrammed x and y variables, where the histogram is regarded as a probability distribution. Negative bin values are ignored for the purposes of this calculation.
- double gsl_histogram2d_sum ($const\ gsl_histogram2d\ ^*\ h$) [Function] This function returns the sum of all bin values. Negative bin values are included in the sum.

22.19 2D Histogram Operations

This function returns 1 if all the individual bin ranges of the two histograms are identical, and 0 otherwise.

int gsl_histogram2d_add (gsl_histogram2d * h1, const gsl_histogram2d * h2) [Function]

This function adds the contents of the bins in histogram h2 to the corresponding bins of histogram h1, i.e. $h'_1(i,j) = h_1(i,j) + h_2(i,j)$. The two histograms must have identical bin ranges.

int gsl_histogram2d_sub (gsl_histogram2d * h1, const gsl_histogram2d * h2) [Function]

This function subtracts the contents of the bins in histogram h2 from the corresponding bins of histogram h1, i.e. $h'_1(i,j) = h_1(i,j) - h_2(i,j)$. The two histograms must have identical bin ranges.

int gsl_histogram2d_mul (gsl_histogram2d * h1, const gsl_histogram2d * h2) [Function]

This function multiplies the contents of the bins of histogram h1 by the contents of the corresponding bins in histogram h2, i.e. $h'_1(i,j) = h_1(i,j) * h_2(i,j)$. The two histograms must have identical bin ranges.

This function divides the contents of the bins of histogram h1 by the contents of the corresponding bins in histogram h2, i.e. $h'_1(i,j) = h_1(i,j)/h_2(i,j)$. The two histograms must have identical bin ranges.

- int gsl_histogram2d_scale (gsl_histogram2d * h, double scale) [Function] This function multiplies the contents of the bins of histogram h by the constant scale, i.e. $h'_1(i,j) = h_1(i,j) * scale$.
- int gsl_histogram2d_shift (gsl_histogram2d * h, double offset) [Function] This function shifts the contents of the bins of histogram h by the constant offset, i.e. $h'_1(i,j) = h_1(i,j) + offset$.

22.20 Reading and writing 2D histograms

The library provides functions for reading and writing two dimensional histograms to a file as binary data or formatted text.

This function writes the ranges and bins of the histogram h to the stream stream in binary format. The return value is 0 for success and GSL_EFAILED if there was a problem writing to the file. Since the data is written in the native binary format it may not be portable between different architectures.

- int gsl_histogram2d_fread (FILE * stream, $gsl_histogram2d * h$) [Function] This function reads into the histogram h from the stream h in binary format. The histogram h must be preallocated with the correct size since the function uses the number of x and y bins in h to determine how many bytes to read. The return value is 0 for success and GSL_EFAILED if there was a problem reading from the file. The data is assumed to have been written in the native binary format on the same architecture.

This function writes the ranges and bins of the histogram h line-by-line to the stream stream using the format specifiers range_format and bin_format. These should be one

of the %g, %e or %f formats for floating point numbers. The function returns 0 for success and GSL_EFAILED if there was a problem writing to the file. The histogram output is formatted in five columns, and the columns are separated by spaces, like this,

```
xrange[0] xrange[1] yrange[0] yrange[1] bin(0,0)
xrange[0] xrange[1] yrange[1] yrange[2] bin(0,1)
xrange[0] xrange[1] yrange[2] yrange[3] bin(0,2)
....
xrange[0] xrange[1] yrange[ny-1] yrange[ny] bin(0,ny-1)

xrange[1] xrange[2] yrange[0] yrange[ny] bin(1,0)
xrange[1] xrange[2] yrange[1] yrange[2] bin(1,1)
xrange[1] xrange[2] yrange[1] yrange[2] bin(1,2)
....
xrange[1] xrange[2] yrange[ny-1] yrange[ny] bin(1,ny-1)
....

xrange[nx-1] xrange[nx] yrange[0] yrange[1] bin(nx-1,0)
xrange[nx-1] xrange[nx] yrange[1] yrange[2] bin(nx-1,1)
xrange[nx-1] xrange[nx] yrange[1] yrange[2] bin(nx-1,2)
....
xrange[nx-1] xrange[nx] yrange[ny-1] yrange[ny] bin(nx-1,ny-1)
```

Each line contains the lower and upper limits of the bin and the contents of the bin. Since the upper limits of the each bin are the lower limits of the neighboring bins there is duplication of these values but this allows the histogram to be manipulated with line-oriented tools.

int gsl_histogram2d_fscanf (FILE * stream, gsl_histogram2d * h) [Function]

This function reads formatted data from the stream stream into the histogram h. The data is assumed to be in the five-column format used by gsl_histogram2d_fprintf. The histogram h must be preallocated with the correct lengths since the function uses the sizes of h to determine how many numbers to read. The function returns 0 for success and GSL_EFAILED if there was a problem reading from the file.

22.21 Resampling from 2D histograms

As in the one-dimensional case, a two-dimensional histogram made by counting events can be regarded as a measurement of a probability distribution. Allowing for statistical error, the height of each bin represents the probability of an event where (x,y) falls in the range of that bin. For a two-dimensional histogram the probability distribution takes the form p(x,y)dxdy where,

$$p(x,y) = n_{ij}/(NA_{ij})$$

In this equation n_{ij} is the number of events in the bin which contains (x, y), A_{ij} is the area of the bin and N is the total number of events. The distribution of events within each bin is assumed to be uniform.

[Data Type]

size_t nx, ny

This is the number of histogram bins used to approximate the probability distribution function in the x and y directions.

double * xrange

The ranges of the bins in the x-direction are stored in an array of nx + 1 elements pointed to by xrange.

double * yrange

The ranges of the bins in the y-direction are stored in an array of ny + 1 pointed to by yrange.

double * sum

The cumulative probability for the bins is stored in an array of nx^*ny elements pointed to by sum.

The following functions allow you to create a gsl_histogram2d_pdf struct which represents a two dimensional probability distribution and generate random samples from it.

gsl_histogram2d_pdf * gsl_histogram2d_pdf_alloc (size_t nx, size_t ny) [Function]

This function allocates memory for a two-dimensional probability distribution of size nx-by-ny and returns a pointer to a newly initialized $gsl_histogram2d_pdf$ struct. If insufficient memory is available a null pointer is returned and the error handler is invoked with an error code of GSL_ENOMEM .

This function initializes the two-dimensional probability distribution calculated p from the histogram h. If any of the bins of h are negative then the error handler is invoked with an error code of GSL_EDOM because a probability distribution cannot contain negative values.

void $gsl_histogram2d_pdf_free (gsl_histogram2d_pdf * p)$ [Function] This function frees the two-dimensional probability distribution function p and all of the memory associated with it.

This function uses two uniform random numbers between zero and one, r1 and r2, to compute a single random sample from the two-dimensional probability distribution p.

22.22 Example programs for 2D histograms

This program demonstrates two features of two-dimensional histograms. First a 10-by-10 two-dimensional histogram is created with x and y running from 0 to 1. Then a few sample points are added to the histogram, at (0.3,0.3) with a height of 1, at (0.8,0.1) with a height of 5 and at (0.7,0.9) with a height of 0.5. This histogram with three events is used to generate a random sample of 1000 simulated events, which are printed out.

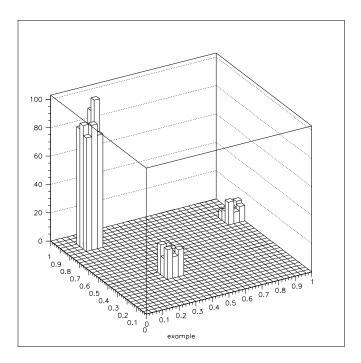
```
#include <stdio.h>
#include <gsl/gsl_rng.h>
#include <gsl/gsl_histogram2d.h>
int
main (void)
  const gsl_rng_type * T;
  gsl_rng * r;
  gsl_histogram2d * h = gsl_histogram2d_alloc (10, 10);
  gsl_histogram2d_set_ranges_uniform (h,
                                       0.0, 1.0,
                                       0.0, 1.0);
  gsl_histogram2d_accumulate (h, 0.3, 0.3, 1);
  gsl_histogram2d_accumulate (h, 0.8, 0.1, 5);
  gsl_histogram2d_accumulate (h, 0.7, 0.9, 0.5);
  gsl_rng_env_setup ();
  T = gsl_rng_default;
  r = gsl_rng_alloc (T);
  {
    int i;
    gsl_histogram2d_pdf * p
      = gsl_histogram2d_pdf_alloc (h->nx, h->ny);
    gsl_histogram2d_pdf_init (p, h);
    for (i = 0; i < 1000; i++) {
      double x, y;
      double u = gsl_rng_uniform (r);
      double v = gsl_rng_uniform (r);
      gsl_histogram2d_pdf_sample (p, u, v, &x, &y);
      printf ("%g %g\n", x, y);
```

```
gsl_histogram2d_pdf_free (p);
}

gsl_histogram2d_free (h);
gsl_rng_free (r);

return 0;
}
```

The following plot shows the distribution of the simulated events. Using a higher resolution grid we can see the original underlying histogram and also the statistical fluctuations caused by the events being uniformly distributed over the area of the original bins.



23 N-tuples

This chapter describes functions for creating and manipulating *ntuples*, sets of values associated with events. The ntuples are stored in files. Their values can be extracted in any combination and *booked* in a histogram using a selection function.

The values to be stored are held in a user-defined data structure, and an ntuple is created associating this data structure with a file. The values are then written to the file (normally inside a loop) using the ntuple functions described below.

A histogram can be created from ntuple data by providing a selection function and a value function. The selection function specifies whether an event should be included in the subset to be analyzed or not. The value function computes the entry to be added to the histogram for each event.

All the ntuple functions are defined in the header file gsl_ntuple.h

23.1 The ntuple struct

Ntuples are manipulated using the gsl_ntuple struct. This struct contains information on the file where the ntuple data is stored, a pointer to the current ntuple data row and the size of the user-defined ntuple data struct.

```
typedef struct {
    FILE * file;
    void * ntuple_data;
    size_t size;
} gsl_ntuple;
```

23.2 Creating ntuples

This function creates a new write-only ntuple file filename for ntuples of size size and returns a pointer to the newly created ntuple struct. Any existing file with the same name is truncated to zero length and overwritten. A pointer to memory for the current ntuple row ntuple_data must be supplied—this is used to copy ntuples in and out of the file.

23.3 Opening an existing ntuple file

```
gsl_ntuple * gsl_ntuple_open (char * filename, void * [Function] ntuple_data, size_t size)
```

This function opens an existing ntuple file *filename* for reading and returns a pointer to a corresponding ntuple struct. The ntuples in the file must have size *size*. A pointer to memory for the current ntuple row *ntuple_data* must be supplied—this is used to copy ntuples in and out of the file.

23.4 Writing ntuples

```
int gsl_ntuple_write (gsl_ntuple * ntuple)
```

[Function]

This function writes the current ntuple *ntuple->ntuple_data* of size *ntuple->size* to the corresponding file.

```
int gsl_ntuple_bookdata (gsl_ntuple * ntuple)
```

[Function]

This function is a synonym for gsl_ntuple_write.

23.5 Reading ntuples

```
int gsl_ntuple_read (gsl_ntuple * ntuple)
```

[Function]

This function reads the current row of the ntuple file for ntuple and stores the values in ntuple > data.

23.6 Closing an ntuple file

```
int gsl_ntuple_close (gsl_ntuple * ntuple)
```

[Function]

This function closes the ntuple file *ntuple* and frees its associated allocated memory.

23.7 Histogramming ntuple values

Once an ntuple has been created its contents can be histogrammed in various ways using the function gsl_ntuple_project. Two user-defined functions must be provided, a function to select events and a function to compute scalar values. The selection function and the value function both accept the ntuple row as a first argument and other parameters as a second argument.

The selection function determines which ntuple rows are selected for histogramming. It is defined by the following struct,

```
typedef struct {
  int (* function) (void * ntuple_data, void * params);
  void * params;
} gsl_ntuple_select_fn;
```

The struct component function should return a non-zero value for each ntuple row that is to be included in the histogram.

The value function computes scalar values for those ntuple rows selected by the selection function,

```
typedef struct {
  double (* function) (void * ntuple_data, void * params);
  void * params;
} gsl_ntuple_value_fn;
```

In this case the struct component function should return the value to be added to the histogram for the ntuple row.

This function updates the histogram h from the ntuple using the functions value_func and select_func. For each ntuple row where the selection function select_func is non-zero the corresponding value of that row is computed using the function value_func and added to the histogram. Those ntuple rows where select_func

returns zero are ignored. New entries are added to the histogram, so subsequent calls can be used to accumulate further data in the same histogram.

23.8 Examples

The following example programs demonstrate the use of ntuples in managing a large dataset. The first program creates a set of 10,000 simulated "events", each with 3 associated values (x, y, z). These are generated from a Gaussian distribution with unit variance, for demonstration purposes, and written to the ntuple file test.dat.

```
#include <gsl/gsl_ntuple.h>
#include <gsl/gsl_rng.h>
#include <gsl/gsl_randist.h>
struct data
{
  double x;
 double y;
 double z;
};
int
main (void)
  const gsl_rng_type * T;
  gsl_rng * r;
  struct data ntuple_row;
  int i;
  gsl_ntuple *ntuple
    = gsl_ntuple_create ("test.dat", &ntuple_row,
                         sizeof (ntuple_row));
  gsl_rng_env_setup ();
 T = gsl_rng_default;
  r = gsl_rng_alloc (T);
  for (i = 0; i < 10000; i++)
    {
      ntuple_row.x = gsl_ran_ugaussian (r);
      ntuple_row.y = gsl_ran_ugaussian (r);
      ntuple_row.z = gsl_ran_ugaussian (r);
      gsl_ntuple_write (ntuple);
```

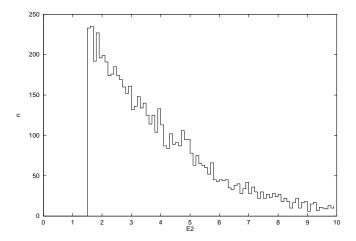
```
gsl_ntuple_close (ntuple);
gsl_rng_free (r);
return 0;
}
```

The next program analyses the ntuple data in the file test.dat. The analysis procedure is to compute the squared-magnitude of each event, $E^2 = x^2 + y^2 + z^2$, and select only those which exceed a lower limit of 1.5. The selected events are then histogrammed using their E^2 values.

```
#include <math.h>
#include <gsl/gsl_ntuple.h>
#include <gsl/gsl_histogram.h>
struct data
{
  double x;
 double y;
  double z;
};
int sel_func (void *ntuple_data, void *params);
double val_func (void *ntuple_data, void *params);
int
main (void)
{
  struct data ntuple_row;
  gsl_ntuple *ntuple
    = gsl_ntuple_open ("test.dat", &ntuple_row,
                       sizeof (ntuple_row));
  double lower = 1.5;
  gsl_ntuple_select_fn S;
  gsl_ntuple_value_fn V;
  gsl_histogram *h = gsl_histogram_alloc (100);
  gsl_histogram_set_ranges_uniform(h, 0.0, 10.0);
  S.function = &sel_func;
  S.params = &lower;
  V.function = &val_func;
  V.params = 0;
  gsl_ntuple_project (h, ntuple, &V, &S);
```

```
gsl_histogram_fprintf (stdout, h, "%f", "%f");
  gsl_histogram_free (h);
 gsl_ntuple_close (ntuple);
 return 0;
}
int
sel_func (void *ntuple_data, void *params)
 struct data * data = (struct data *) ntuple_data;
 double x, y, z, E2, scale;
 scale = *(double *) params;
 x = data -> x;
 y = data -> y;
 z = data -> z;
 E2 = x * x + y * y + z * z;
 return E2 > scale;
}
double
val_func (void *ntuple_data, void *params)
 struct data * data = (struct data *) ntuple_data;
 double x, y, z;
 x = data -> x;
 y = data -> y;
 z = data -> z;
 return x * x + y * y + z * z;
```

The following plot shows the distribution of the selected events. Note the cut-off at the lower bound.



23.9 References and Further Reading

Further information on the use of ntuples can be found in the documentation for the CERN packages PAW and HBOOK (available online).

24 Monte Carlo Integration

This chapter describes routines for multidimensional Monte Carlo integration. These include the traditional Monte Carlo method and adaptive algorithms such as VEGAS and MISER which use importance sampling and stratified sampling techniques. Each algorithm computes an estimate of a multidimensional definite integral of the form,

$$I = \int_{x_l}^{x_u} dx \, \int_{y_l}^{y_u} dy \, ... f(x, y, ...)$$

over a hypercubic region $((x_l, x_u), (y_l, y_u), ...)$ using a fixed number of function calls. The routines also provide a statistical estimate of the error on the result. This error estimate should be taken as a guide rather than as a strict error bound—random sampling of the region may not uncover all the important features of the function, resulting in an underestimate of the error.

The functions are defined in separate header files for each routine, gsl_monte_plain.h, gsl_monte_miser.h and gsl_monte_vegas.h.

24.1 Interface

All of the Monte Carlo integration routines use the same general form of interface. There is an allocator to allocate memory for control variables and workspace, a routine to initialize those control variables, the integrator itself, and a function to free the space when done.

Each integration function requires a random number generator to be supplied, and returns an estimate of the integral and its standard deviation. The accuracy of the result is determined by the number of function calls specified by the user. If a known level of accuracy is required this can be achieved by calling the integrator several times and averaging the individual results until the desired accuracy is obtained.

Random sample points used within the Monte Carlo routines are always chosen strictly within the integration region, so that endpoint singularities are automatically avoided.

The function to be integrated has its own datatype, defined in the header file gsl_monte.h.

gsl_monte_function

[Data Type]

This data type defines a general function with parameters for Monte Carlo integration.

double (* f) (double * x, size_t dim, void * params)

this function should return the value f(x, params) for the argument x and parameters params, where x is an array of size dim giving the coordinates of the point where the function is to be evaluated.

size_t dim

the number of dimensions for x.

void * params

a pointer to the parameters of the function.

Here is an example for a quadratic function in two dimensions,

$$f(x,y) = ax^2 + bxy + cy^2$$

with $a=3,\,b=2,\,c=1$. The following code defines a gsl_monte_function F which you could pass to an integrator:

```
struct my_f_params { double a; double b; double c; };
     double
     my_f (double x[], size_t dim, void * p) {
        struct my_f_params * fp = (struct my_f_params *)p;
        if (dim != 2)
           {
             fprintf (stderr, "error: dim != 2");
             abort ();
           }
        return fp->a * x[0] * x[0]
                  + fp -> b * x[0] * x[1]
                    + fp->c * x[1] * x[1];
     }
     gsl_monte_function F;
     struct my_f_params params = { 3.0, 2.0, 1.0 };
     F.f = &my_f;
     F.dim = 2;
     F.params = &params;
The function f(x) can be evaluated using the following macro,
     #define GSL_MONTE_FN_EVAL(F,x)
         (*((F)->f))(x,(F)->dim,(F)->params)
```

24.2 PLAIN Monte Carlo

The plain Monte Carlo algorithm samples points randomly from the integration region to estimate the integral and its error. Using this algorithm the estimate of the integral E(f; N) for N randomly distributed points x_i is given by,

$$E(f; N) = V\langle f \rangle = \frac{V}{N} \sum_{i}^{N} f(x_i)$$

where V is the volume of the integration region. The error on this estimate $\sigma(E; N)$ is calculated from the estimated variance of the mean,

$$\sigma^{2}(E; N) = \frac{V^{2}}{N^{2}} \sum_{i}^{N} (f(x_{i}) - \langle f \rangle)^{2}.$$

For large N this variance decreases asymptotically as Var(f)/N, where Var(f) is the true variance of the function over the integration region. The error estimate itself should decrease as $\sigma(f)/\sqrt{N}$. The familiar law of errors decreasing as $1/\sqrt{N}$ applies—to reduce the error by a factor of 10 requires a 100-fold increase in the number of sample points.

The functions described in this section are declared in the header file gsl_monte_plain.h.

- gsl_monte_plain_state * gsl_monte_plain_alloc (size_t dim) [Function]

 This function allocates and initializes a workspace for Monte Carlo integration in dim dimensions.
- int gsl_monte_plain_init (gsl_monte_plain_state* s) [Function]

 This function initializes a previously allocated integration state. This allows an existing workspace to be reused for different integrations.
- int gsl_monte_plain_integrate (gsl_monte_function * f, const double x1[], const double xu[], size_t dim, size_t calls, gsl_rng * r, gsl_monte_plain_state * s, double * result, double * abserr) [Function]

This routines uses the plain Monte Carlo algorithm to integrate the function f over the dim-dimensional hypercubic region defined by the lower and upper limits in the arrays xl and xu, each of size dim. The integration uses a fixed number of function calls calls, and obtains random sampling points using the random number generator r. A previously allocated workspace s must be supplied. The result of the integration is returned in result, with an estimated absolute error abserr.

void gsl_monte_plain_free (gsl_monte_plain_state * s) [Function]
This function frees the memory associated with the integrator state s.

24.3 MISER

The MISER algorithm of Press and Farrar is based on recursive stratified sampling. This technique aims to reduce the overall integration error by concentrating integration points in the regions of highest variance.

The idea of stratified sampling begins with the observation that for two disjoint regions a and b with Monte Carlo estimates of the integral $E_a(f)$ and $E_b(f)$ and variances $\sigma_a^2(f)$ and $\sigma_b^2(f)$, the variance $\operatorname{Var}(f)$ of the combined estimate $E(f) = \frac{1}{2}(E_a(f) + E_b(f))$ is given by,

$$\operatorname{Var}(f) = \frac{\sigma_a^2(f)}{4N_a} + \frac{\sigma_b^2(f)}{4N_b}.$$

It can be shown that this variance is minimized by distributing the points such that,

$$\frac{N_a}{N_a + N_b} = \frac{\sigma_a}{\sigma_a + \sigma_b}.$$

Hence the smallest error estimate is obtained by allocating sample points in proportion to the standard deviation of the function in each sub-region.

The MISER algorithm proceeds by bisecting the integration region along one coordinate axis to give two sub-regions at each step. The direction is chosen by examining all d possible bisections and selecting the one which will minimize the combined variance of the two sub-regions. The variance in the sub-regions is estimated by sampling with a fraction of the total number of points available to the current step. The same procedure is then repeated recursively for each of the two half-spaces from the best bisection. The remaining sample points are allocated to the sub-regions using the formula for N_a and N_b . This recursive

allocation of integration points continues down to a user-specified depth where each subregion is integrated using a plain Monte Carlo estimate. These individual values and their error estimates are then combined upwards to give an overall result and an estimate of its error.

The functions described in this section are declared in the header file gsl_monte_miser.h.

- gsl_monte_miser_state * gsl_monte_miser_alloc (size_t dim) [Function]

 This function allocates and initializes a workspace for Monte Carlo integration in dim dimensions. The workspace is used to maintain the state of the integration.
- int gsl_monte_miser_init (gsl_monte_miser_state* s) [Function]

 This function initializes a previously allocated integration state. This allows an existing workspace to be reused for different integrations.

This routines uses the MISER Monte Carlo algorithm to integrate the function f over the dim-dimensional hypercubic region defined by the lower and upper limits in the arrays xl and xu, each of size dim. The integration uses a fixed number of function calls calls, and obtains random sampling points using the random number generator r. A previously allocated workspace s must be supplied. The result of the integration is returned in result, with an estimated absolute error abserr.

void gsl_monte_miser_free (gsl_monte_miser_state * s) [Function]
This function frees the memory associated with the integrator state s.

The MISER algorithm has several configurable parameters which can be changed using the following two functions.¹

This function copies the parameters of the integrator state into the user-supplied params structure.

This function sets the integrator parameters based on values provided in the *params* structure.

Typically the values of the parameters are first read using gsl_monte_miser_params_get, the necessary changes are made to the fields of the params structure, and the values are copied back into the integrator state using gsl_monte_miser_params_set. The functions use the gsl_monte_miser_params structure which contains the following fields:

The previous method of accessing these fields directly through the gsl_monte_miser_state struct is now deprecated.

double estimate_frac

[Variable]

This parameter specifies the fraction of the currently available number of function calls which are allocated to estimating the variance at each recursive step. The default value is 0.1.

size_t min_calls [Variable]

This parameter specifies the minimum number of function calls required for each estimate of the variance. If the number of function calls allocated to the estimate using estimate_frac falls below min_calls then min_calls are used instead. This ensures that each estimate maintains a reasonable level of accuracy. The default value of min_calls is 16 * dim.

size_t min_calls_per_bisection

[Variable]

This parameter specifies the minimum number of function calls required to proceed with a bisection step. When a recursive step has fewer calls available than $min_calls_per_bisection$ it performs a plain Monte Carlo estimate of the current subregion and terminates its branch of the recursion. The default value of this parameter is $32 * min_calls$.

double alpha [Variable]

This parameter controls how the estimated variances for the two sub-regions of a bisection are combined when allocating points. With recursive sampling the overall variance should scale better than 1/N, since the values from the sub-regions will be obtained using a procedure which explicitly minimizes their variance. To accommodate this behavior the MISER algorithm allows the total variance to depend on a scaling parameter α ,

$$Var(f) = \frac{\sigma_a}{N_a^{\alpha}} + \frac{\sigma_b}{N_b^{\alpha}}.$$

The authors of the original paper describing MISER recommend the value $\alpha=2$ as a good choice, obtained from numerical experiments, and this is used as the default value in this implementation.

double dither [Variable]

This parameter introduces a random fractional variation of size *dither* into each bisection, which can be used to break the symmetry of integrands which are concentrated near the exact center of the hypercubic integration region. The default value of dither is zero, so no variation is introduced. If needed, a typical value of *dither* is 0.1.

24.4 **VEGAS**

The VEGAS algorithm of Lepage is based on importance sampling. It samples points from the probability distribution described by the function |f|, so that the points are concentrated in the regions that make the largest contribution to the integral.

In general, if the Monte Carlo integral of f is sampled with points distributed according to a probability distribution described by the function g, we obtain an estimate $E_q(f; N)$,

$$E_g(f;N) = E(f/g;N)$$

with a corresponding variance,

$$\operatorname{Var}_q(f; N) = \operatorname{Var}(f/g; N).$$

If the probability distribution is chosen as g = |f|/I(|f|) then it can be shown that the variance $V_g(f;N)$ vanishes, and the error in the estimate will be zero. In practice it is not possible to sample from the exact distribution g for an arbitrary function, so importance sampling algorithms aim to produce efficient approximations to the desired distribution.

The VEGAS algorithm approximates the exact distribution by making a number of passes over the integration region while histogramming the function f. Each histogram is used to define a sampling distribution for the next pass. Asymptotically this procedure converges to the desired distribution. In order to avoid the number of histogram bins growing like K^d the probability distribution is approximated by a separable function: $g(x_1, x_2, \ldots) = g_1(x_1)g_2(x_2)\ldots$ so that the number of bins required is only Kd. This is equivalent to locating the peaks of the function from the projections of the integrand onto the coordinate axes. The efficiency of VEGAS depends on the validity of this assumption. It is most efficient when the peaks of the integrand are well-localized. If an integrand can be rewritten in a form which is approximately separable this will increase the efficiency of integration with VEGAS.

VEGAS incorporates a number of additional features, and combines both stratified sampling and importance sampling. The integration region is divided into a number of "boxes", with each box getting a fixed number of points (the goal is 2). Each box can then have a fractional number of bins, but if the ratio of bins-per-box is less than two, Vegas switches to a kind variance reduction (rather than importance sampling).

- gsl_monte_vegas_state * gsl_monte_vegas_alloc (size_t dim) [Function]

 This function allocates and initializes a workspace for Monte Carlo integration in dim dimensions. The workspace is used to maintain the state of the integration.
- int gsl_monte_vegas_init (gsl_monte_vegas_state* s) [Function]

 This function initializes a previously allocated integration state. This allows an existing workspace to be reused for different integrations.
- int gsl_monte_vegas_integrate (gsl_monte_function * f, double x1[], [Function] double xu[], size_t dim, size_t calls, gsl_rng * r, gsl_monte_vegas_state * s, double * result, double * abserr)

This routines uses the VEGAS Monte Carlo algorithm to integrate the function f over the dim-dimensional hypercubic region defined by the lower and upper limits in the arrays xl and xu, each of size dim. The integration uses a fixed number of function calls calls, and obtains random sampling points using the random number generator r. A previously allocated workspace s must be supplied. The result of the integration is returned in result, with an estimated absolute error abserr. The result and its error estimate are based on a weighted average of independent samples. The chi-squared per degree of freedom for the weighted average is returned via the state struct component, s->chisq, and must be consistent with 1 for the weighted average to be reliable.

void gsl_monte_vegas_free (gsl_monte_vegas_state * s) [Function]
This function frees the memory associated with the integrator state s.

The VEGAS algorithm computes a number of independent estimates of the integral internally, according to the iterations parameter described below, and returns their weighted average. Random sampling of the integrand can occasionally produce an estimate where the error is zero, particularly if the function is constant in some regions. An estimate with zero error causes the weighted average to break down and must be handled separately. In the original Fortran implementations of VEGAS the error estimate is made non-zero by substituting a small value (typically 1e-30). The implementation in GSL differs from this and avoids the use of an arbitrary constant—it either assigns the value a weight which is the average weight of the preceding estimates or discards it according to the following procedure,

current estimate has zero error, weighted average has finite error

The current estimate is assigned a weight which is the average weight of the preceding estimates.

current estimate has finite error, previous estimates had zero error

The previous estimates are discarded and the weighted averaging procedure begins with the current estimate.

current estimate has zero error, previous estimates had zero error

The estimates are averaged using the arithmetic mean, but no error is computed.

The convergence of the algorithm can be tested using the overall chi-squared value of the results, which is available from the following function:

double gsl_monte_vegas_chisq (const gsl_monte_vegas_state * s) [Function] This function returns the chi-squared per degree of freedom for the weighted estimate of the integral. The returned value should be close to 1. A value which differs significantly from 1 indicates that the values from different iterations are inconsistent. In this case the weighted error will be under-estimated, and further iterations of the algorithm are needed to obtain reliable results.

This function returns the raw (unaveraged) values of the integral result and its error sigma from the most recent iteration of the algorithm.

The VEGAS algorithm is highly configurable. Several parameters can be changed using the following two functions.

This function copies the parameters of the integrator state into the user-supplied params structure.

This function sets the integrator parameters based on values provided in the *params* structure.

Typically the values of the parameters are first read using gsl_monte_vegas_params_get, the necessary changes are made to the fields of the params structure, and the values are copied back into the integrator state using gsl_monte_vegas_params_set. The functions use the gsl_monte_vegas_params structure which contains the following fields:

double alpha [Variable]

The parameter alpha controls the stiffness of the rebinning algorithm. It is typically set between one and two. A value of zero prevents rebinning of the grid. The default value is 1.5.

size_t iterations [Variable]

The number of iterations to perform for each call to the routine. The default value is 5 iterations.

int stage [Variable]

Setting this determines the stage of the calculation. Normally, stage = 0 which begins with a new uniform grid and empty weighted average. Calling VEGAS with stage = 1 retains the grid from the previous run but discards the weighted average, so that one can "tune" the grid using a relatively small number of points and then do a large run with stage = 1 on the optimized grid. Setting stage = 2 keeps the grid and the weighted average from the previous run, but may increase (or decrease) the number of histogram bins in the grid depending on the number of calls available. Choosing stage = 3 enters at the main loop, so that nothing is changed, and is equivalent to performing additional iterations in a previous call.

int mode [Variable]

The possible choices are GSL_VEGAS_MODE_IMPORTANCE, GSL_VEGAS_MODE_STRATIFIED, GSL_VEGAS_MODE_IMPORTANCE_ONLY. This determines whether VEGAS will use importance sampling or stratified sampling, or whether it can pick on its own. In low dimensions VEGAS uses strict stratified sampling (more precisely, stratified sampling is chosen if there are fewer than 2 bins per box).

int verbose [Variable]
FILE * ostream [Variable]

These parameters set the level of information printed by VEGAS. All information is written to the stream ostream. The default setting of verbose is -1, which turns off all output. A verbose value of 0 prints summary information about the weighted average and final result, while a value of 1 also displays the grid coordinates. A value of 2 prints information from the rebinning procedure for each iteration.

The above fields and the *chisq* value can also be accessed directly in the gsl_monte_vegas_state but such use is deprecated.

24.5 Examples

The example program below uses the Monte Carlo routines to estimate the value of the following 3-dimensional integral from the theory of random walks,

$$I = \int_{-\pi}^{+\pi} \frac{dk_x}{2\pi} \int_{-\pi}^{+\pi} \frac{dk_y}{2\pi} \int_{-\pi}^{+\pi} \frac{dk_z}{2\pi} \frac{1}{(1 - \cos(k_x)\cos(k_y)\cos(k_z))}.$$

The analytic value of this integral can be shown to be $I = \Gamma(1/4)^4/(4\pi^3) = 1.393203929685676859...$ The integral gives the mean time spent at the origin by a random walk on a body-centered cubic lattice in three dimensions.

For simplicity we will compute the integral over the region (0,0,0) to (π,π,π) and multiply by 8 to obtain the full result. The integral is slowly varying in the middle of the region but has integrable singularities at the corners (0,0,0), $(0,\pi,\pi)$, $(\pi,0,\pi)$ and $(\pi,\pi,0)$. The Monte Carlo routines only select points which are strictly within the integration region and so no special measures are needed to avoid these singularities.

```
#include <stdlib.h>
#include <gsl/gsl_math.h>
#include <gsl/gsl_monte.h>
#include <gsl/gsl_monte_plain.h>
#include <gsl/gsl_monte_miser.h>
#include <gsl/gsl_monte_vegas.h>
/* Computation of the integral,
      I = int (dx dy dz)/(2pi)^3 1/(1-cos(x)cos(y)cos(z))
   over (-pi,-pi,-pi) to (+pi, +pi, +pi).
                                             The exact answer
   is Gamma(1/4)^4/(4 pi^3). This example is taken from
   C.Itzykson, J.M.Drouffe, "Statistical Field Theory -
   Volume 1", Section 1.1, p21, which cites the original
   paper M.L.Glasser, I.J.Zucker, Proc.Natl.Acad.Sci.USA 74
   1800 (1977) */
/* For simplicity we compute the integral over the region
   (0,0,0) \rightarrow (pi,pi,pi) and multiply by 8 */
double exact = 1.3932039296856768591842462603255;
double
g (double *k, size_t dim, void *params)
 double A = 1.0 / (M_PI * M_PI * M_PI);
 return A / (1.0 - \cos (k[0]) * \cos (k[1]) * \cos (k[2]));
}
void
display_results (char *title, double result, double error)
 printf ("%s =========\n", title);
 printf ("result = % .6f\n", result);
 printf ("sigma = % .6f\n", error);
 printf ("exact = % .6f\n", exact);
 printf ("error = \% .6f = \%.2g sigma\n", result - exact,
         fabs (result - exact) / error);
}
int.
main (void)
 double res, err;
 double x1[3] = \{ 0, 0, 0 \};
 double xu[3] = { M_PI, M_PI, M_PI };
```

```
const gsl_rng_type *T;
gsl_rng *r;
gsl_monte_function G = { &g, 3, 0 };
size_t calls = 500000;
gsl_rng_env_setup ();
T = gsl_rng_default;
r = gsl_rng_alloc (T);
  gsl_monte_plain_state *s = gsl_monte_plain_alloc (3);
  gsl_monte_plain_integrate (&G, xl, xu, 3, calls, r, s,
                             &res, &err);
  gsl_monte_plain_free (s);
  display_results ("plain", res, err);
  gsl_monte_miser_state *s = gsl_monte_miser_alloc (3);
  gsl_monte_miser_integrate (&G, x1, xu, 3, calls, r, s,
                             &res, &err);
  gsl_monte_miser_free (s);
  display_results ("miser", res, err);
}
  gsl_monte_vegas_state *s = gsl_monte_vegas_alloc (3);
  gsl_monte_vegas_integrate (&G, xl, xu, 3, 10000, r, s,
                             &res, &err);
  display_results ("vegas warm-up", res, err);
  printf ("converging...\n");
  do
      gsl_monte_vegas_integrate (&G, xl, xu, 3, calls/5, r, s,
                                 &res, &err);
      printf ("result = % .6f sigma = % .6f "
              "chisq/dof = %.1f\n", res, err, gsl_monte_vegas_chisq (s));
    }
  while (fabs (gsl_monte_vegas_chisq (s) - 1.0) > 0.5);
  display_results ("vegas final", res, err);
  gsl_monte_vegas_free (s);
gsl_rng_free (r);
return 0;
```

With 500,000 function calls the plain Monte Carlo algorithm achieves a fractional error of 1%. The estimated error sigma is roughly consistent with the actual error—the computed result differs from the true result by about 1.4 standard deviations,

The MISER algorithm reduces the error by a factor of four, and also correctly estimates the error.

In the case of the VEGAS algorithm the program uses an initial warm-up run of 10,000 function calls to prepare, or "warm up", the grid. This is followed by a main run with five iterations of 100,000 function calls. The chi-squared per degree of freedom for the five iterations are checked for consistency with 1, and the run is repeated if the results have not converged. In this case the estimates are consistent on the first pass.

If the value of chisq had differed significantly from 1 it would indicate inconsistent results, with a correspondingly underestimated error. The final estimate from VEGAS (using a similar number of function calls) is significantly more accurate than the other two algorithms.

24.6 References and Further Reading

The MISER algorithm is described in the following article by Press and Farrar,

W.H. Press, G.R. Farrar, Recursive Stratified Sampling for Multidimensional Monte Carlo Integration, Computers in Physics, v4 (1990), pp190–195.

The VEGAS algorithm is described in the following papers,

G.P. Lepage, A New Algorithm for Adaptive Multidimensional Integration, Journal of Computational Physics 27, 192–203, (1978)

G.P. Lepage, VEGAS: An Adaptive Multi-dimensional Integration Program, Cornell preprint CLNS 80-447, March 1980

25 Simulated Annealing

Stochastic search techniques are used when the structure of a space is not well understood or is not smooth, so that techniques like Newton's method (which requires calculating Jacobian derivative matrices) cannot be used. In particular, these techniques are frequently used to solve combinatorial optimization problems, such as the traveling salesman problem.

The goal is to find a point in the space at which a real valued energy function (or cost function) is minimized. Simulated annealing is a minimization technique which has given good results in avoiding local minima; it is based on the idea of taking a random walk through the space at successively lower temperatures, where the probability of taking a step is given by a Boltzmann distribution.

The functions described in this chapter are declared in the header file gsl_siman.h.

25.1 Simulated Annealing algorithm

The simulated annealing algorithm takes random walks through the problem space, looking for points with low energies; in these random walks, the probability of taking a step is determined by the Boltzmann distribution,

$$p = e^{-(E_{i+1} - E_i)/(kT)}$$

if $E_{i+1} > E_i$, and p = 1 when $E_{i+1} \le E_i$.

In other words, a step will occur if the new energy is lower. If the new energy is higher, the transition can still occur, and its likelihood is proportional to the temperature T and inversely proportional to the energy difference $E_{i+1} - E_i$.

The temperature T is initially set to a high value, and a random walk is carried out at that temperature. Then the temperature is lowered very slightly according to a *cooling* schedule, for example: $T \to T/\mu_T$ where μ_T is slightly greater than 1.

The slight probability of taking a step that gives higher energy is what allows simulated annealing to frequently get out of local minima.

25.2 Simulated Annealing functions

This function performs a simulated annealing search through a given space. The space is specified by providing the functions Ef and distance. The simulated annealing steps are generated using the random number generator r and the function $take_step$.

The starting configuration of the system should be given by x0-p. The routine offers two modes for updating configurations, a fixed-size mode and a variable-size mode. In the fixed-size mode the configuration is stored as a single block of memory of size $element_size$. Copies of this configuration are created, copied and destroyed internally using the standard library functions malloc, memcpy and free. The function pointers copyfunc, $copy_constructor$ and destructor should be null pointers in fixed-size mode.

In the variable-size mode the functions *copyfunc*, *copy_constructor* and *destructor* are used to create, copy and destroy configurations internally. The variable *element_size* should be zero in the variable-size mode.

The params structure (described below) controls the run by providing the temperature schedule and other tunable parameters to the algorithm.

On exit the best result achieved during the search is placed in $*x0_p$. If the annealing process has been successful this should be a good approximation to the optimal point in the space.

If the function pointer *print_position* is not null, a debugging log will be printed to **stdout** with the following columns:

#-iter #-evals temperature position energy best_energy and the output of the function *print_position* itself. If *print_position* is null then no information is printed.

The simulated annealing routines require several user-specified functions to define the configuration space and energy function. The prototypes for these functions are given below.

gsl_siman_Efunc_t

[Data Type]

This function type should return the energy of a configuration xp.

double (*gsl_siman_Efunc_t) (void *xp)

gsl_siman_step_t

[Data Type]

This function type should modify the configuration xp using a random step taken from the generator r, up to a maximum distance of $step_size$.

gsl_siman_metric_t

[Data Type]

This function type should return the distance between two configurations xp and yp. double (*gsl_siman_metric_t) (void *xp, void *yp)

gsl_siman_print_t

[Data Type]

This function type should print the contents of the configuration xp.

void (*gsl_siman_print_t) (void *xp)

gsl_siman_copy_t

[Data Type]

This function type should copy the configuration source into dest.

void (*gsl_siman_copy_t) (void *source, void *dest)

gsl_siman_copy_construct_t

[Data Type]

This function type should create a new copy of the configuration xp.

void * (*gsl_siman_copy_construct_t) (void *xp)

gsl_siman_destroy_t

[Data Type]

This function type should destroy the configuration xp, freeing its memory.

void (*gsl_siman_destroy_t) (void *xp)

```
gsl_siman_params_t
```

[Data Type]

These are the parameters that control a run of gsl_siman_solve. This structure contains all the information needed to control the search, beyond the energy function, the step function and the initial guess.

double step_size

The maximum step size in the random walk.

double k, t_initial, mu_t, t_min

The parameters of the Boltzmann distribution and cooling schedule.

25.3 Examples

The simulated annealing package is clumsy, and it has to be because it is written in C, for C callers, and tries to be polymorphic at the same time. But here we provide some examples which can be pasted into your application with little change and should make things easier.

25.3.1 Trivial example

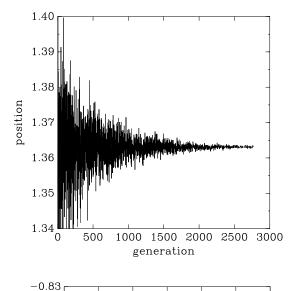
The first example, in one dimensional Cartesian space, sets up an energy function which is a damped sine wave; this has many local minima, but only one global minimum, somewhere between 1.0 and 1.5. The initial guess given is 15.5, which is several local minima away from the global minimum.

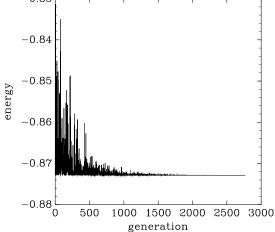
```
#include <math.h>
#include <stdlib.h>
#include <string.h>
#include <gsl/gsl_siman.h>
/* set up parameters for this simulated annealing run */
/* how many points do we try before stepping */
#define N_TRIES 200
/* how many iterations for each T? */
#define ITERS_FIXED_T 1000
/* max step size in random walk */
#define STEP_SIZE 1.0
/* Boltzmann constant */
#define K 1.0
/* initial temperature */
#define T_INITIAL 0.008
/* damping factor for temperature */
#define MU_T 1.003
#define T_MIN 2.0e-6
gsl_siman_params_t params
```

```
= {N_TRIES, ITERS_FIXED_T, STEP_SIZE,
    K, T_INITIAL, MU_T, T_MIN};
/* now some functions to test in one dimension */
double E1(void *xp)
 double x = * ((double *) xp);
 return exp(-pow((x-1.0),2.0))*sin(8*x);
double M1(void *xp, void *yp)
 double x = *((double *) xp);
 double y = *((double *) yp);
 return fabs(x - y);
void S1(const gsl_rng * r, void *xp, double step_size)
 double old_x = *((double *) xp);
 double new_x;
 double u = gsl_rng_uniform(r);
 new_x = u * 2 * step_size - step_size + old_x;
 memcpy(xp, &new_x, sizeof(new_x));
void P1(void *xp)
 printf ("%12g", *((double *) xp));
int
main(int argc, char *argv[])
 const gsl_rng_type * T;
 gsl_rng * r;
 double x_initial = 15.5;
 gsl_rng_env_setup();
 T = gsl_rng_default;
 r = gsl_rng_alloc(T);
 gsl_siman_solve(r, &x_initial, E1, S1, M1, P1,
                  NULL, NULL, NULL,
                  sizeof(double), params);
 gsl_rng_free (r);
 return 0;
```

Here are a couple of plots that are generated by running siman_test in the following way:

```
$ ./siman_test | awk '!/^#/ {print $1, $4}'
| graph -y 1.34 1.4 -WO -X generation -Y position
| plot -Tps > siman-test.eps
$ ./siman_test | awk '!/^#/ {print $1, $5}'
| graph -y -0.88 -0.83 -WO -X generation -Y energy
| plot -Tps > siman-energy.eps
```





Example of a simulated annealing run: at higher temperatures (early in the plot) you see that the solution can fluctuate, but at lower temperatures it converges.

25.3.2 Traveling Salesman Problem

The TSP (Traveling Salesman Problem) is the classic combinatorial optimization problem. I have provided a very simple version of it, based on the coordinates of twelve cities in the southwestern United States. This should maybe be called the Flying Salesman Problem, since I am using the great-circle distance between cities, rather than the driving distance. Also: I assume the earth is a sphere, so I don't use geoid distances.

The gsl_siman_solve routine finds a route which is 3490.62 Kilometers long; this is confirmed by an exhaustive search of all possible routes with the same initial city.

The full code can be found in siman/siman_tsp.c, but I include here some plots generated in the following way:

```
$ ./siman_tsp > tsp.output
$ grep -v "^#" tsp.output
| awk '{print $1, $NF}'
| graph -y 3300 6500 -WO -X generation -Y distance
    -L "TSP - 12 southwest cities"
| plot -Tps > 12-cities.eps
$ grep initial_city_coord tsp.output
  | awk '{print $2, $3}'
  | graph -X "longitude (- means west)" -Y "latitude"
     -L "TSP - initial-order" -f 0.03 -S 1 0.1
  | plot -Tps > initial-route.eps
$ grep final_city_coord tsp.output
  | awk '{print $2, $3}'
  | graph -X "longitude (- means west)" -Y "latitude"
     -L "TSP - final-order" -f 0.03 -S 1 0.1
  | plot -Tps > final-route.eps
```

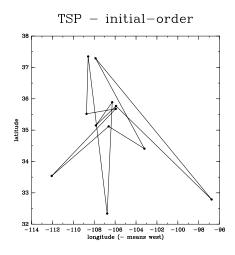
This is the output showing the initial order of the cities; longitude is negative, since it is west and I want the plot to look like a map.

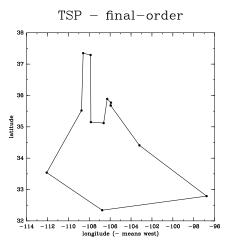
```
# initial coordinates of cities (longitude and latitude)
###initial_city_coord: -105.95 35.68 Santa Fe
###initial_city_coord: -112.07 33.54 Phoenix
###initial_city_coord: -106.62 35.12 Albuquerque
###initial_city_coord: -103.2 34.41 Clovis
###initial_city_coord: -107.87 37.29 Durango
###initial_city_coord: -96.77 32.79 Dallas
###initial_city_coord: -105.92 35.77 Tesuque
###initial_city_coord: -107.84 35.15 Grants
###initial_city_coord: -106.28 35.89 Los Alamos
###initial_city_coord: -106.76 32.34 Las Cruces
###initial_city_coord: -108.58 37.35 Cortez
###initial_city_coord: -108.74 35.52 Gallup
###initial_city_coord: -105.95 35.68 Santa Fe
```

The optimal route turns out to be:

```
# final coordinates of cities (longitude and latitude)
###final_city_coord: -105.95 35.68 Santa Fe
###final_city_coord: -103.2 34.41 Clovis
###final_city_coord: -96.77 32.79 Dallas
###final_city_coord: -106.76 32.34 Las Cruces
###final_city_coord: -112.07 33.54 Phoenix
###final_city_coord: -108.74 35.52 Gallup
###final_city_coord: -108.58 37.35 Cortez
###final_city_coord: -107.87 37.29 Durango
###final_city_coord: -107.84 35.15 Grants
```

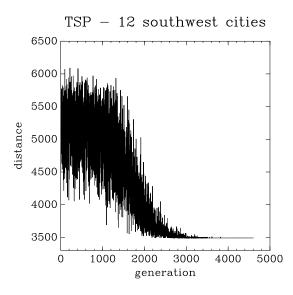
```
###final_city_coord: -106.62 35.12 Albuquerque
###final_city_coord: -106.28 35.89 Los Alamos
###final_city_coord: -105.92 35.77 Tesuque
###final_city_coord: -105.95 35.68 Santa Fe
```





Initial and final (optimal) route for the 12 southwestern cities Flying Salesman Problem.

Here's a plot of the cost function (energy) versus generation (point in the calculation at which a new temperature is set) for this problem:



Example of a simulated annealing run for the 12 southwestern cities Flying Salesman Problem.

25.4 References and Further Reading

Further information is available in the following book,

 $Modern\ Heuristic\ Techniques\ for\ Combinatorial\ Problems,\ Colin\ R.$ Reeves (ed.), McGraw-Hill, 1995 (ISBN 0-07-709239-2).

26 Ordinary Differential Equations

This chapter describes functions for solving ordinary differential equation (ODE) initial value problems. The library provides a variety of low-level methods, such as Runge-Kutta and Bulirsch-Stoer routines, and higher-level components for adaptive step-size control. The components can be combined by the user to achieve the desired solution, with full access to any intermediate steps. A driver object can be used as a high level wrapper for easy use of low level functions.

These functions are declared in the header file gsl_odeiv2.h. This is a new interface in version 1.15 and uses the prefix gsl_odeiv2 for all functions. It is recommended over the previous gsl_odeiv implementation defined in gsl_odeiv.h The old interface has been retained under the original name for backwards compatibility.

26.1 Defining the ODE System

The routines solve the general n-dimensional first-order system,

$$\frac{dy_i(t)}{dt} = f_i(t, y_1(t), \dots y_n(t))$$

for $i=1,\ldots,n$. The stepping functions rely on the vector of derivatives f_i and the Jacobian matrix, $J_{ij}=\partial f_i(t,y(t))/\partial y_j$. A system of equations is defined using the gsl_odeiv2_system datatype.

gsl_odeiv2_system

[Data Type]

This data type defines a general ODE system with arbitrary parameters.

int (* function) (double t, const double y[], double dydt[], void *
params)

This function should store the vector elements $f_i(t, y, params)$ in the array dydt, for arguments (t,y) and parameters params.

The function should return GSL_SUCCESS if the calculation was completed successfully. Any other return value indicates an error. A special return value GSL_EBADFUNC causes gsl_odeiv2 routines to immediately stop and return. The user must call an appropriate reset function (e.g. gsl_odeiv2_driver_reset or gsl_odeiv2_step_reset) before continuing. Use return values distinct from standard GSL error codes to distinguish your function as the source of the error.

int (* jacobian) (double t, const double y[], double * dfdy, double
dfdt[], void * params);

This function should store the vector of derivative elements $\partial f_i(t,y,params)/\partial t$ in the array dfdt and the Jacobian matrix J_{ij} in the array dfdy, regarded as a row-ordered matrix J(i,j) = dfdy[i*dimension + j] where dimension is the dimension of the system.

Not all of the stepper algorithms of gsl_odeiv2 make use of the Jacobian matrix, so it may not be necessary to provide this function (the jacobian element of the struct can be replaced by a null pointer for those algorithms).

The function should return GSL_SUCCESS if the calculation was completed successfully. Any other return value indicates an error. A special return value GSL_EBADFUNC causes gsl_odeiv2 routines to immediately stop and return. The user must call an appropriate reset function (e.g. gsl_odeiv2_driver_reset or gsl_odeiv2_step_reset) before continuing. Use return values distinct from standard GSL error codes to distinguish your function as the source of the error.

size_t dimension;

This is the dimension of the system of equations.

void * params

This is a pointer to the arbitrary parameters of the system.

26.2 Stepping Functions

The lowest level components are the stepping functions which advance a solution from time t to t + h for a fixed step-size h and estimate the resulting local error.

```
gsl_odeiv2_step * gsl_odeiv2_step_alloc (const gsl_odeiv2_step_type * T, size_t dim) [Function]
```

This function returns a pointer to a newly allocated instance of a stepping function of type T for a system of \dim dimensions. Please note that if you use a stepper method that requires access to a driver object, it is advisable to use a driver allocation method, which automatically allocates a stepper, too.

```
int gsl_odeiv2_step_reset (gsl_odeiv2_step * s) [Function]

This function resets the stepping function s. It should be used whenever the next use of s will not be a continuation of a previous step.
```

```
void gsl\_odeiv2\_step\_free (gsl\_odeiv2\_step * s) [Function] This function frees all the memory associated with the stepping function s.
```

```
const char * gsl_odeiv2_step_name (const gsl_odeiv2_step * s) [Function]

This function returns a pointer to the name of the stepping function. For example,
```

would print something like step method is 'rkf45'.

unsigned int gsl_odeiv2_step_order (const gsl_odeiv2_step * s) [Function]
This function returns the order of the stepping function on the previous step. The order can vary if the stepping function itself is adaptive.

This function sets a pointer of the driver object d for stepper s, to allow the stepper to access control (and evolve) object through the driver object. This is a requirement for some steppers, to get the desired error level for internal iteration of stepper. Allocation of a driver object calls this function automatically.

This function applies the stepping function s to the system of equations defined by sys, using the step-size h to advance the system from time t and state y to time t+h. The new state of the system is stored in y on output, with an estimate of the absolute error in each component stored in yerr. If the argument $dydt_-in$ is not null it should point an array containing the derivatives for the system at time t on input. This is optional as the derivatives will be computed internally if they are not provided, but allows the reuse of existing derivative information. On output the new derivatives of the system at time t+h will be stored in $dydt_-out$ if it is not null.

The stepping function returns GSL_FAILURE if it is unable to compute the requested step. Also, if the user-supplied functions defined in the system sys return a status other than GSL_SUCCESS the step will be aborted. In that case, the elements of y will be restored to their pre-step values and the error code from the user-supplied function will be returned. Failure may be due to a singularity in the system or too large step-size h. In that case the step should be attempted again with a smaller step-size, e.g. h/2.

If the driver object is not appropriately set via gsl_odeiv2_step_set_driver for those steppers that need it, the stepping function returns GSL_EFAULT. If the user-supplied functions defined in the system sys returns GSL_EBADFUNC, the function returns immediately with the same return code. In this case the user must call gsl_odeiv2_step_reset before calling this function again.

The following algorithms are available,

gsl_odeiv2_step_rk2

[Step Type]

Explicit embedded Runge-Kutta (2, 3) method.

gsl_odeiv2_step_rk4

[Step Type]

Explicit 4th order (classical) Runge-Kutta. Error estimation is carried out by the step doubling method. For more efficient estimate of the error, use the embedded methods described below.

gsl_odeiv2_step_rkf45

[Step Type]

Explicit embedded Runge-Kutta-Fehlberg (4, 5) method. This method is a good general-purpose integrator.

gsl_odeiv2_step_rkck

[Step Type]

Explicit embedded Runge-Kutta Cash-Karp (4, 5) method.

gsl_odeiv2_step_rk8pd

[Step Type]

Explicit embedded Runge-Kutta Prince-Dormand (8, 9) method.

gsl_odeiv2_step_rk1imp

[Step Type]

Implicit Gaussian first order Runge-Kutta. Also known as implicit Euler or backward Euler method. Error estimation is carried out by the step doubling method. This algorithm requires the Jacobian and access to the driver object via gsl_odeiv2_step_set_driver.

gsl_odeiv2_step_rk2imp

[Step Type]

Implicit Gaussian second order Runge-Kutta. Also known as implicit mid-point rule. Error estimation is carried out by the step doubling method. This stepper requires the Jacobian and access to the driver object via gsl_odeiv2_step_set_driver.

gsl_odeiv2_step_rk4imp

[Step Type]

Implicit Gaussian 4th order Runge-Kutta. Error estimation is carried out by the step doubling method. This algorithm requires the Jacobian and access to the driver object via gsl_odeiv2_step_set_driver.

gsl_odeiv2_step_bsimp

[Step Type]

Implicit Bulirsch-Stoer method of Bader and Deuflhard. The method is generally suitable for stiff problems. This stepper requires the Jacobian.

gsl_odeiv2_step_msadams

[Step Type]

A variable-coefficient linear multistep Adams method in Nordsieck form. This stepper uses explicit Adams-Bashforth (predictor) and implicit Adams-Moulton (corrector) methods in $P(EC)^m$ functional iteration mode. Method order varies dynamically between 1 and 12. This stepper requires the access to the driver object via gsl_odeiv2_step_set_driver.

gsl_odeiv2_step_msbdf

[Step Type]

A variable-coefficient linear multistep backward differentiation formula (BDF) method in Nordsieck form. This stepper uses the explicit BDF formula as predictor and implicit BDF formula as corrector. A modified Newton iteration method is used to solve the system of non-linear equations. Method order varies dynamically between 1 and 5. The method is generally suitable for stiff problems. This stepper requires the Jacobian and the access to the driver object via gsl_odeiv2_step_set_driver.

26.3 Adaptive Step-size Control

The control function examines the proposed change to the solution produced by a stepping function and attempts to determine the optimal step-size for a user-specified level of error.

The standard control object is a four parameter heuristic based on absolute and relative errors eps_abs and eps_rel , and scaling factors a_y and a_dydt for the system state y(t) and derivatives y'(t) respectively.

The step-size adjustment procedure for this method begins by computing the desired error level D_i for each component,

$$D_i = \epsilon_{abs} + \epsilon_{rel} * (a_y|y_i| + a_{dydt}h|y\prime_i|)$$

and comparing it with the observed error $E_i = |yerr_i|$. If the observed error E exceeds the desired error level D by more than 10% for any component then the method reduces the step-size by an appropriate factor,

$$h_{new} = h_{old} * S * (E/D)^{-1/q}$$

where q is the consistency order of the method (e.g. q = 4 for 4(5) embedded RK), and S is a safety factor of 0.9. The ratio E/D is taken to be the maximum of the ratios E_i/D_i .

If the observed error E is less than 50% of the desired error level D for the maximum ratio E_i/D_i then the algorithm takes the opportunity to increase the step-size to bring the error in line with the desired level,

$$h_{new} = h_{old} * S * (E/D)^{-1/(q+1)}$$

This encompasses all the standard error scaling methods. To avoid uncontrolled changes in the stepsize, the overall scaling factor is limited to the range 1/5 to 5.

gsl_odeiv2_control * gsl_odeiv2_control_y_new (double ps_abs, double eps_rel)
[Function]

This function creates a new control object which will keep the local error on each step within an absolute error of eps_abs and relative error of eps_rel with respect to the solution $y_i(t)$. This is equivalent to the standard control object with $a_y=1$ and $a_dydt=0$.

This function creates a new control object which will keep the local error on each step within an absolute error of eps_abs and relative error of eps_rel with respect to the derivatives of the solution $y'_i(t)$. This is equivalent to the standard control object with $a_y=0$ and $a_dydt=1$.

This function creates a new control object which uses the same algorithm as $gsl_odeiv2_control_standard_new$ but with an absolute error which is scaled for each component by the array $scale_abs$. The formula for D_i for this control object is,

$$D_i = \epsilon_{abs} s_i + \epsilon_{rel} * (a_y | y_i | + a_{dudt} h | y t_i |)$$

where s_i is the *i*-th component of the array $scale_abs$. The same error control heuristic is used by the Matlab ODE suite.

This function returns a pointer to a newly allocated instance of a control function of type T. This function is only needed for defining new types of control functions. For most purposes the standard control functions described above should be sufficient.

This function initializes the control function c with the parameters eps_abs (absolute error), eps_rel (relative error), a_y (scaling factor for y) and a_dydt (scaling factor for derivatives).

void $gsl_odeiv2_control_free\ (gsl_odeiv2_control * c)$ [Function] This function frees all the memory associated with the control function c.

This function adjusts the step-size h using the control function c, and the current values of y, yerr and dydt. The stepping function step is also needed to determine the order of the method. If the error in the y-values yerr is found to be too large then the step-size h is reduced and the function returns $GSL_ODEIV_HADJ_DEC$. If the error is sufficiently small then h may be increased and $GSL_ODEIV_HADJ_INC$ is returned. The function returns $GSL_ODEIV_HADJ_NIL$ if the step-size is unchanged. The goal of the function is to estimate the largest step-size which satisfies the user-specified accuracy requirements for the current point.

This function returns a pointer to the name of the control function. For example,

would print something like control method is 'standard'

This function calculates the desired error level of the *ind*-th component to *errlev*. It requires the value (y) and value of the derivative (dydt) of the component, and the current step size h.

This function sets a pointer of the driver object d for control object c.

26.4 Evolution

The evolution function combines the results of a stepping function and control function to reliably advance the solution forward one step using an acceptable step-size.

```
gsl_odeiv2_evolve * gsl_odeiv2_evolve_alloc (size_t dim) [Function]

This function returns a pointer to a newly allocated instance of an evolution function for a system of dim dimensions.
```

This function advances the system (e, sys) from time t and position y using the stepping function step. The new time and position are stored in t and y on output. The initial step-size is taken as h. The control function con is applied to check whether

the local error estimated by the stepping function step using step-size h exceeds the

required error tolerance. If the error is too high, the step is retried by calling *step* with a decreased step-size. This process is continued until an acceptable step-size is found. An estimate of the local error for the step can be obtained from the components of the array e->yerr[].

If the user-supplied functions defined in the system sys returns GSL_EBADFUNC, the function returns immediately with the same return code. In this case the user must call gsl_odeiv2_step_reset and gsl_odeiv2_evolve_reset before calling this function again.

Otherwise, if the user-supplied functions defined in the system sys or the stepping function step return a status other than $GSL_SUCCESS$, the step is retried with a decreased step-size. If the step-size decreases below machine precision, a status of $GSL_FAILURE$ is returned if the user functions returned $GSL_SUCCESS$. Otherwise the value returned by user function is returned. If no acceptable step can be made, t and y will be restored to their pre-step values and h contains the final attempted step-size.

If the step is successful the function returns a suggested step-size for the next step in h. The maximum time t1 is guaranteed not to be exceeded by the time-step. On the final time-step the value of t will be set to t1 exactly.

int gsl_odeiv2_evolve_apply_fixed_step (gsl_odeiv2_evolve * e, [Function] gsl_odeiv2_control * con, gsl_odeiv2_step * step, const gsl_odeiv2_system * sys, double * t, const double h, double y[])

This function advances the ODE-system (e, sys, con) from time t and position y using the stepping function step by a specified step size h. If the local error estimated by the stepping function exceeds the desired error level, the step is not taken and the function returns $GSL_FAILURE$. Otherwise the value returned by user function is returned.

int gsl_odeiv2_evolve_reset (gsl_odeiv2_evolve * e) [Function]

This function resets the evolution function e. It should be used whenever the next use of e will not be a continuation of a previous step.

void gsl_odeiv2_evolve_free (gsl_odeiv2_evolve * e) [Function]
This function frees all the memory associated with the evolution function e.

This function sets a pointer of the driver object d for evolve object e.

If a system has discontinuous changes in the derivatives at known points, it is advisable to evolve the system between each discontinuity in sequence. For example, if a step-change in an external driving force occurs at times t_a, t_b and t_c then evolution should be carried out over the ranges (t_0, t_a) , (t_a, t_b) , (t_b, t_c) , and (t_c, t_1) separately and not directly over the range (t_0, t_1) .

26.5 Driver

The driver object is a high level wrapper that combines the evolution, control and stepper objects for easy use.

- gsl_odeiv2_driver * gsl_odeiv2_driver_alloc_y_new (const gsl_odeiv2_system * sys, const gsl_odeiv2_step_type * T, const double hstart, const double epsabs, const double epsrel)
- gsl_odeiv2_driver * gsl_odeiv2_driver_alloc_yp_new (const [Function] gsl_odeiv2_system * sys, const gsl_odeiv2_step_type * T, const double hstart, const double epsabs, const double epsrel)
- gsl_odeiv2_driver * gsl_odeiv2_driver_alloc_standard_new [Function] (const gsl_odeiv2_system * sys, const gsl_odeiv2_step_type * T, const double hstart, const double epsabs, const double epsrel, const double a_y, const double a_dydt)

These functions return a pointer to a newly allocated instance of a driver object. The functions automatically allocate and initialise the evolve, control and stepper objects for ODE system sys using stepper type T. The initial step size is given in hstart. The rest of the arguments follow the syntax and semantics of the control functions with same name (gsl_odeiv2_control_*_new).

The function sets a minimum for allowed step size hmin for driver d. Default value is 0.

The function sets a maximum for allowed step size hmax for driver d. Default value is GSL_DBL_MAX .

int gsl_odeiv2_driver_set_nmax (gsl_odeiv2_driver * d, const unsigned long int nmax) [Function]

The function sets a maximum for allowed number of steps nmax for driver d. Default value of 0 sets no limit for steps.

This function evolves the driver system d from t to t1. Initially vector y should contain the values of dependent variables at point t. If the function is unable to complete the calculation, an error code from $gsl_odeiv2_evolve_apply$ is returned, and t and y contain the values from last successful step.

If maximum number of steps is reached, a value of GSL_EMAXITER is returned. If the step size drops below minimum value, the function returns with GSL_ENOPROG. If the user-supplied functions defined in the system sys returns GSL_EBADFUNC, the function returns immediately with the same return code. In this case the user must call gsl_odeiv2_driver_reset before calling this function again.

This function evolves the driver system d from t with n steps of size h. If the function is unable to complete the calculation, an error code from $gsl_odeiv2_evolve_apply_fixed_step$ is returned, and t and y contain the values from last successful step.

- int gsl_odeiv2_driver_reset (gsl_odeiv2_driver * d) [Function]
 This function resets the evolution and stepper objects.

The routine resets the evolution and stepper objects and sets new initial step size to *hstart*. This function can be used e.g. to change the direction of integration.

int gsl_odeiv2_driver_free (gsl_odeiv2_driver * d) [Function]

This function frees the driver object, and the related evolution, stepper and control objects.

26.6 Examples

The following program solves the second-order nonlinear Van der Pol oscillator equation,

$$u''(t) + \mu u'(t)(u(t)^2 - 1) + u(t) = 0$$

This can be converted into a first order system suitable for use with the routines described in this chapter by introducing a separate variable for the velocity, v = u'(t),

$$u' = v$$
$$v' = -u + \mu v(1 - u^2)$$

The program begins by defining functions for these derivatives and their Jacobian. The main function uses driver level functions to solve the problem. The program evolves the solution from (u, v) = (1, 0) at t = 0 to t = 100. The step-size h is automatically adjusted by the controller to maintain an absolute accuracy of 10^{-6} in the function values (u, v). The loop in the example prints the solution at the points $t_i = 1, 2, \ldots, 100$.

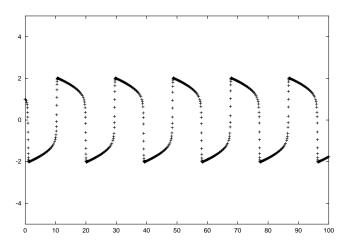
int

```
jac (double t, const double y[], double *dfdy,
     double dfdt[], void *params)
{
  double mu = *(double *)params;
  gsl_matrix_view dfdy_mat
    = gsl_matrix_view_array (dfdy, 2, 2);
  gsl_matrix * m = &dfdy_mat.matrix;
  gsl_matrix_set (m, 0, 0, 0.0);
  gsl_matrix_set (m, 0, 1, 1.0);
  gsl_matrix_set (m, 1, 0, -2.0*mu*y[0]*y[1] - 1.0);
  gsl_matrix_set (m, 1, 1, -mu*(y[0]*y[0] - 1.0));
  dfdt[0] = 0.0;
 dfdt[1] = 0.0;
 return GSL_SUCCESS;
}
int
main (void)
{
  double mu = 10;
  gsl_odeiv2_system sys = {func, jac, 2, &mu};
  gsl_odeiv2_driver * d =
    gsl_odeiv2_driver_alloc_y_new (&sys, gsl_odeiv2_step_rk8pd,
                                  1e-6, 1e-6, 0.0);
  int i;
  double t = 0.0, t1 = 100.0;
  double y[2] = \{ 1.0, 0.0 \};
  for (i = 1; i <= 100; i++)
    {
      double ti = i * t1 / 100.0;
      int status = gsl_odeiv2_driver_apply (d, &t, ti, y);
      if (status != GSL_SUCCESS)
        {
          printf ("error, return value=%d\n", status);
          break;
        }
      printf ("%.5e %.5e \n", t, y[0], y[1]);
  gsl_odeiv2_driver_free (d);
  return 0;
}
```

The user can work with the lower level functions directly, as in the following example. In this case an intermediate result is printed after each successful step instead of equidistant time points.

```
int
main (void)
  const gsl_odeiv2_step_type * T
    = gsl_odeiv2_step_rk8pd;
  gsl_odeiv2_step * s
    = gsl_odeiv2_step_alloc (T, 2);
  gsl_odeiv2_control * c
    = gsl_odeiv2_control_y_new (1e-6, 0.0);
  gsl_odeiv2_evolve * e
    = gsl_odeiv2_evolve_alloc (2);
  double mu = 10;
  gsl_odeiv2_system sys = {func, jac, 2, &mu};
  double t = 0.0, t1 = 100.0;
  double h = 1e-6;
  double y[2] = \{ 1.0, 0.0 \};
  while (t < t1)
    {
      int status = gsl_odeiv2_evolve_apply (e, c, s,
                                            &sys,
                                            &t, t1,
                                            &h, y);
      if (status != GSL_SUCCESS)
          break;
      printf ("%.5e %.5e %.5e\n", t, y[0], y[1]);
    }
  gsl_odeiv2_evolve_free (e);
  gsl_odeiv2_control_free (c);
  gsl_odeiv2_step_free (s);
 return 0;
}
```

For functions with multiple parameters, the appropriate information can be passed in through the *params* argument in gsl_odeiv2_system definition (*mu* in this example) by using a pointer to a struct.



Numerical solution of the Van der Pol oscillator equation using Prince-Dormand 8th order Runge-Kutta.

It is also possible to work with a non-adaptive integrator, using only the stepping function itself, gsl_odeiv2_driver_apply_fixed_step or gsl_odeiv2_evolve_apply_fixed_step. The following program uses the driver level function, with fourth-order Runge-Kutta stepping function with a fixed stepsize of 0.001.

```
int
main (void)
{
  double mu = 10;
  gsl_odeiv2_system sys = { func, jac, 2, &mu };
  gsl_odeiv2_driver *d =
    gsl_odeiv2_driver_alloc_y_new (&sys, gsl_odeiv2_step_rk4,
                                   1e-3, 1e-8, 1e-8);
  double t = 0.0;
  double y[2] = \{ 1.0, 0.0 \};
  int i, s;
  for (i = 0; i < 100; i++)
      s = gsl_odeiv2_driver_apply_fixed_step (d, &t, 1e-3, 1000, y);
      if (s != GSL_SUCCESS)
        {
          printf ("error: driver returned %d\n", s);
          break;
        }
      printf ("%.5e %.5e \n", t, y[0], y[1]);
    }
  gsl_odeiv2_driver_free (d);
```

```
return s;
}
```

26.7 References and Further Reading

Ascher, U.M., Petzold, L.R., Computer Methods for Ordinary Differential and Differential-Algebraic Equations, SIAM, Philadelphia, 1998.

Hairer, E., Norsett, S. P., Wanner, G., Solving Ordinary Differential Equations I: Nonstiff Problems, Springer, Berlin, 1993.

Hairer, E., Wanner, G., Solving Ordinary Differential Equations II: Stiff and Differential-Algebraic Problems, Springer, Berlin, 1996.

Many of the basic Runge-Kutta formulas can be found in the Handbook of Mathematical Functions,

Abramowitz & Stegun (eds.), Handbook of Mathematical Functions, Section 25.5.

The implicit Bulirsch-Stoer algorithm bsimp is described in the following paper,

G. Bader and P. Deuflhard, "A Semi-Implicit Mid-Point Rule for Stiff Systems of Ordinary Differential Equations.", Numer. Math. 41, 373–398, 1983.

The Adams and BDF multistep methods msadams and msbdf are based on the following articles,

- G. D. Byrne and A. C. Hindmarsh, "A Polyalgorithm for the Numerical Solution of Ordinary Differential Equations.", ACM Trans. Math. Software, 1, 71–96, 1975.
- P. N. Brown, G. D. Byrne and A. C. Hindmarsh, "VODE: A Variable-coefficient ODE Solver.", SIAM J. Sci. Stat. Comput. 10, 1038–1051, 1989.
- A. C. Hindmarsh, P. N. Brown, K. E. Grant, S. L. Lee, R. Serban, D. E. Shumaker and C. S. Woodward, "SUNDIALS: Suite of Nonlinear and Differential/Algebraic Equation Solvers.", ACM Trans. Math. Software 31, 363–396, 2005.

27 Interpolation

This chapter describes functions for performing interpolation. The library provides a variety of interpolation methods, including Cubic splines and Akima splines. The interpolation types are interchangeable, allowing different methods to be used without recompiling. Interpolations can be defined for both normal and periodic boundary conditions. Additional functions are available for computing derivatives and integrals of interpolating functions.

These interpolation methods produce curves that pass through each datapoint. To interpolate noisy data with a smoothing curve see Chapter 39 [Basis Splines], page 433.

The functions described in this section are declared in the header files gsl_interp.h and gsl_spline.h.

27.1 Introduction

Given a set of data points $(x_1, y_1) \dots (x_n, y_n)$ the routines described in this section compute a continuous interpolating function y(x) such that $y(x_i) = y_i$. The interpolation is piecewise smooth, and its behavior at the end-points is determined by the type of interpolation used.

27.2 Interpolation Functions

The interpolation function for a given dataset is stored in a gsl_interp object. These are created by the following functions.

This function returns a pointer to a newly allocated interpolation object of type T for size data-points.

int gsl_interp_init (gsl_interp * interp, const double xa[], const double ya[], size_t size) [Function]

This function initializes the interpolation object interp for the data (xa,ya) where xa and ya are arrays of size size. The interpolation object (gsl_interp) does not save the data arrays xa and ya and only stores the static state computed from the data. The xa data array is always assumed to be strictly ordered, with increasing x values; the behavior for other arrangements is not defined.

void gsl_interp_free (gsl_interp * interp) [Function]
This function frees the interpolation object interp.

27.3 Interpolation Types

The interpolation library provides six interpolation types:

gsl_interp_linear [Interpolation Type]
Linear interpolation. This interpolation method does not require any additional memory.

gsl_interp_polynomial [Interpolation Type]

Polynomial interpolation. This method should only be used for interpolating small numbers of points because polynomial interpolation introduces large oscillations, even

for well-behaved datasets. The number of terms in the interpolating polynomial is equal to the number of points.

gsl_interp_cspline

[Interpolation Type]

Cubic spline with natural boundary conditions. The resulting curve is piecewise cubic on each interval, with matching first and second derivatives at the supplied data-points. The second derivative is chosen to be zero at the first point and last point.

gsl_interp_cspline_periodic

[Interpolation Type]

Cubic spline with periodic boundary conditions. The resulting curve is piecewise cubic on each interval, with matching first and second derivatives at the supplied data-points. The derivatives at the first and last points are also matched. Note that the last point in the data must have the same y-value as the first point, otherwise the resulting periodic interpolation will have a discontinuity at the boundary.

gsl_interp_akima

[Interpolation Type]

Non-rounded Akima spline with natural boundary conditions. This method uses the non-rounded corner algorithm of Wodicka.

gsl_interp_akima_periodic

[Interpolation Type]

Non-rounded Akima spline with periodic boundary conditions. This method uses the non-rounded corner algorithm of Wodicka.

The following related functions are available:

```
const char * gsl_interp_name (const gsl_interp * interp)
```

[Function]

This function returns the name of the interpolation type used by interp. For example,

would print something like,

interp uses 'cspline' interpolation.

```
unsigned int gsl_interp_min_size (const gsl_interp * interp) [Function] unsigned int gsl_interp_type_min_size (const gsl_interp_type * T) [Function]
```

These functions return the minimum number of points required by the interpolation object interp or interpolation type T. For example, Akima spline interpolation requires a minimum of 5 points.

27.4 Index Look-up and Acceleration

The state of searches can be stored in a gsl_interp_accel object, which is a kind of iterator for interpolation lookups. It caches the previous value of an index lookup. When the subsequent interpolation point falls in the same interval its index value can be returned immediately.

This function returns the index i of the array x_array such that $x_array[i] \le x \le x_array[i+1]$. The index is searched for in the range $[index_lo,index_hi]$. An inline version of this function is used when HAVE_INLINE is defined.

gsl_interp_accel * gsl_interp_accel_alloc (void)

[Function]

This function returns a pointer to an accelerator object, which is a kind of iterator for interpolation lookups. It tracks the state of lookups, thus allowing for application of various acceleration strategies.

This function performs a lookup action on the data array x_array of size size, using the given accelerator a. This is how lookups are performed during evaluation of an interpolation. The function returns an index i such that $x_array[i] \le x \le x_array[i+1]$. An inline version of this function is used when HAVE_INLINE is defined.

int gsl_interp_accel_reset (gsl_interp_accel * acc); [Function]
This function reinitializes the accelerator object acc. It should be used when the cached information is no longer applicable—for example, when switching to a new dataset.

void gsl_interp_accel_free (gsl_interp_accel* acc)
This function frees the accelerator object acc.

[Function]

27.5 Evaluation of Interpolating Functions

double gsl_interp_eval (const gsl_interp * interp, const double xa[], [Function] const double ya[], double x, gsl_interp_accel * acc)

These functions return the interpolated value of y for a given point x, using the interpolation object *interp*, data arrays xa and ya and the accelerator acc. When x is outside the range of xa, the error code GSL_EDOM is returned with a value of GSL_NAN for y.

int gsl_interp_eval_deriv_e (const gsl_interp * interp, const double xa[], const double ya[], double x, gsl_interp_accel * acc, double * d)

These functions return the derivative d of an interpolated function for a given point

These functions return the derivative d of an interpolated function for a given point x, using the interpolation object interp, data arrays xa and ya and the accelerator acc.

double gsl_interp_eval_deriv2 (const gsl_interp * interp, const double xa[], const double ya[], double x, gsl_interp_accel * acc) [Function]

int gsl_interp_eval_deriv2_e (const gsl_interp * interp, const [Function] double xa[], const double ya[], double x, gsl_interp_accel * acc, double * d2)

These functions return the second derivative d2 of an interpolated function for a given point x, using the interpolation object interp, data arrays xa and ya and the accelerator acc.

double gsl_interp_eval_integ (const gsl_interp * interp, const double xa[], const double ya[], double a, double b, gsl_interp_accel * acc)

These functions return the numerical integral result of an interpolated function over the range [a, b], using the interpolation object interp, data arrays xa and ya and the accelerator acc.

27.6 Higher-level Interface

The functions described in the previous sections required the user to supply pointers to the x and y arrays on each call. The following functions are equivalent to the corresponding gsl_interp functions but maintain a copy of this data in the gsl_spline object. This removes the need to pass both xa and ya as arguments on each evaluation. These functions are defined in the header file $gsl_spline.h$.

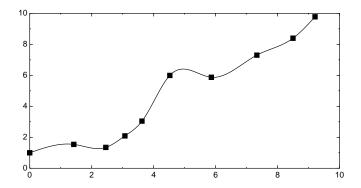
27.7 Examples

The following program demonstrates the use of the interpolation and spline functions. It computes a cubic spline interpolation of the 10-point dataset (x_i, y_i) where $x_i = i + \sin(i)/2$ and $y_i = i + \cos(i^2)$ for i = 0...9.

```
#include <stdlib.h>
#include <stdio.h>
#include <math.h>
#include <gsl/gsl_errno.h>
#include <gsl/gsl_spline.h>
int
main (void)
{
  int i;
 double xi, yi, x[10], y[10];
 printf ("#m=0,S=2\n");
  for (i = 0; i < 10; i++)
    {
      x[i] = i + 0.5 * sin (i);
      y[i] = i + cos (i * i);
      printf ("%g %g\n", x[i], y[i]);
    }
 printf ("#m=1,S=0\n");
    gsl_interp_accel *acc
      = gsl_interp_accel_alloc ();
    gsl_spline *spline
      = gsl_spline_alloc (gsl_interp_cspline, 10);
    gsl_spline_init (spline, x, y, 10);
    for (xi = x[0]; xi < x[9]; xi += 0.01)
        yi = gsl_spline_eval (spline, xi, acc);
        printf ("%g %g\n", xi, yi);
    gsl_spline_free (spline);
    gsl_interp_accel_free (acc);
 return 0;
}
```

The output is designed to be used with the GNU plotutils graph program,

```
$ ./a.out > interp.dat
$ graph -T ps < interp.dat > interp.ps
```



The result shows a smooth interpolation of the original points. The interpolation method can be changed simply by varying the first argument of gsl_spline_alloc.

The next program demonstrates a periodic cubic spline with 4 data points. Note that the first and last points must be supplied with the same y-value for a periodic spline.

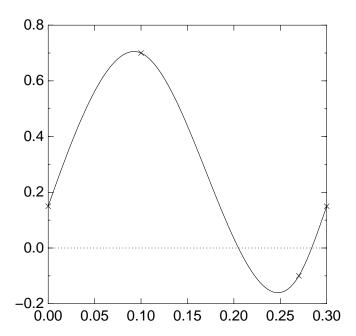
```
#include <stdlib.h>
#include <stdio.h>
#include <math.h>
#include <gsl/gsl_errno.h>
#include <gsl/gsl_spline.h>
int
main (void)
{
  int N = 4;
  double x[4] = \{0.00, 0.10, 0.27, 0.30\};
  double y[4] = \{0.15, 0.70, -0.10, 0.15\};
             /* Note: y[0] == y[3] for periodic data */
  gsl_interp_accel *acc = gsl_interp_accel_alloc ();
  const gsl_interp_type *t = gsl_interp_cspline_periodic;
  gsl_spline *spline = gsl_spline_alloc (t, N);
  int i; double xi, yi;
  printf ("#m=0,S=5\n");
  for (i = 0; i < N; i++)
      printf ("%g %g\n", x[i], y[i]);
    }
 printf ("#m=1,S=0\n");
  gsl_spline_init (spline, x, y, N);
  for (i = 0; i \le 100; i++)
    {
      xi = (1 - i / 100.0) * x[0] + (i / 100.0) * x[N-1];
      yi = gsl_spline_eval (spline, xi, acc);
```

```
printf ("%g %g\n", xi, yi);
}

gsl_spline_free (spline);
gsl_interp_accel_free (acc);
return 0;
}
```

The output can be plotted with GNU graph.

```
$ ./a.out > interp.dat
$ graph -T ps < interp.dat > interp.ps
```



The result shows a periodic interpolation of the original points. The slope of the fitted curve is the same at the beginning and end of the data, and the second derivative is also.

27.8 References and Further Reading

Descriptions of the interpolation algorithms and further references can be found in the following books:

C.W. Ueberhuber, Numerical Computation (Volume 1), Chapter 9 "Interpolation", Springer (1997), ISBN 3-540-62058-3.

D.M. Young, R.T. Gregory A Survey of Numerical Mathematics (Volume 1), Chapter 6.8, Dover (1988), ISBN 0-486-65691-8.

28 Numerical Differentiation

The functions described in this chapter compute numerical derivatives by finite differencing. An adaptive algorithm is used to find the best choice of finite difference and to estimate the error in the derivative. These functions are declared in the header file gsl_deriv.h.

28.1 Functions

This function computes the numerical derivative of the function f at the point x using an adaptive central difference algorithm with a step-size of h. The derivative is returned in result and an estimate of its absolute error is returned in abserr.

The initial value of h is used to estimate an optimal step-size, based on the scaling of the truncation error and round-off error in the derivative calculation. The derivative is computed using a 5-point rule for equally spaced abscissae at x - h, x - h/2, x, x + h/2, x + h, with an error estimate taken from the difference between the 5-point rule and the corresponding 3-point rule x - h, x, x + h. Note that the value of the function at x does not contribute to the derivative calculation, so only 4-points are actually used.

This function computes the numerical derivative of the function f at the point x using an adaptive forward difference algorithm with a step-size of h. The function is evaluated only at points greater than x, and never at x itself. The derivative is returned in result and an estimate of its absolute error is returned in absert. This function should be used if f(x) has a discontinuity at x, or is undefined for values less than x.

The initial value of h is used to estimate an optimal step-size, based on the scaling of the truncation error and round-off error in the derivative calculation. The derivative at x is computed using an "open" 4-point rule for equally spaced abscissae at x+h/4, x+h/2, x+3h/4, x+h, with an error estimate taken from the difference between the 4-point rule and the corresponding 2-point rule x+h/2, x+h.

This function computes the numerical derivative of the function f at the point x using an adaptive backward difference algorithm with a step-size of h. The function is evaluated only at points less than x, and never at x itself. The derivative is returned in result and an estimate of its absolute error is returned in abserr. This function should be used if f(x) has a discontinuity at x, or is undefined for values greater than x

This function is equivalent to calling gsl_deriv_forward with a negative step-size.

28.2 Examples

The following code estimates the derivative of the function $f(x) = x^{3/2}$ at x = 2 and at x = 0. The function f(x) is undefined for x < 0 so the derivative at x = 0 is computed using gsl_deriv_forward.

```
#include <stdio.h>
     #include <gsl/gsl_math.h>
     #include <gsl/gsl_deriv.h>
     double f (double x, void * params)
       return pow (x, 1.5);
     int
     main (void)
     {
       gsl_function F;
       double result, abserr;
       F.function = &f;
       F.params = 0;
       printf ("f(x) = x^(3/2)\n");
       gsl_deriv_central (&F, 2.0, 1e-8, &result, &abserr);
       printf ("x = 2.0\n");
       printf ("f'(x) = %.10f +/- %.10f \n", result, abserr);
       printf ("exact = \%.10f\n', 1.5 * sqrt(2.0));
       gsl_deriv_forward (&F, 0.0, 1e-8, &result, &abserr);
       printf ("x = 0.0\n");
       printf ("f'(x) = %.10f +/- %.10f \n", result, abserr);
       printf ("exact = %.10f\n", 0.0);
       return 0;
Here is the output of the program,
     $ ./a.out
     f(x) = x^3(3/2)
     x = 2.0
     f'(x) = 2.1213203120 +/- 0.0000004064
     exact = 2.1213203436
     x = 0.0
     f'(x) = 0.0000000160 +/- 0.0000000339
     exact = 0.0000000000
```

28.3 References and Further Reading

The algorithms used by these functions are described in the following sources:

Abramowitz and Stegun, *Handbook of Mathematical Functions*, Section 25.3.4, and Table 25.5 (Coefficients for Differentiation).

S.D. Conte and Carl de Boor, Elementary Numerical Analysis: An Algorithmic Approach, McGraw-Hill, 1972.

29 Chebyshev Approximations

This chapter describes routines for computing Chebyshev approximations to univariate functions. A Chebyshev approximation is a truncation of the series $f(x) = \sum c_n T_n(x)$, where the Chebyshev polynomials $T_n(x) = \cos(n\arccos x)$ provide an orthogonal basis of polynomials on the interval [-1,1] with the weight function $1/\sqrt{1-x^2}$. The first few Chebyshev polynomials are, $T_0(x) = 1$, $T_1(x) = x$, $T_2(x) = 2x^2 - 1$. For further information see Abramowitz & Stegun, Chapter 22.

The functions described in this chapter are declared in the header file gsl_chebyshev.h.

29.1 Definitions

A Chebyshev series is stored using the following structure,

The approximation is made over the range [a, b] using order+1 terms, including the coefficient c[0]. The series is computed using the following convention,

$$f(x) = \frac{c_0}{2} + \sum_{n=1}^{\infty} c_n T_n(x)$$

which is needed when accessing the coefficients directly.

29.2 Creation and Calculation of Chebyshev Series

```
gsl_cheb_series * gsl_cheb_alloc (const size_t n) [Function]

This function allocates space for a Chebyshev series of order n and returns a pointer to a new gsl_cheb_series struct.
```

```
void gsl_cheb_free (gsl_cheb_series * cs) [Function]
This function frees a previously allocated Chebyshev series cs.
```

This function computes the Chebyshev approximation cs for the function f over the range (a,b) to the previously specified order. The computation of the Chebyshev approximation is an $O(n^2)$ process, and requires n function evaluations.

29.3 Auxiliary Functions

The following functions provide information about an existing Chebyshev series.

```
size_t gsl_cheb_order (const gsl_cheb_series * cs) [Function]
This function returns the order of Chebyshev series cs.
```

size_t gsl_cheb_size (const gsl_cheb_series * cs) [Function]
double * gsl_cheb_coeffs (const gsl_cheb_series * cs) [Function]
These functions return the size of the Chebyshev coefficient array c[] and a pointer

29.4 Chebyshev Series Evaluation

to its location in memory for the Chebyshev series cs.

double gsl_cheb_eval (const gsl_cheb_series * cs, double x) [Function]
This function evaluates the Chebyshev series cs at a given point x.

This function computes the Chebyshev series cs at a given point x, estimating both the series result and its absolute error abserr. The error estimate is made from the first neglected term in the series.

This function evaluates the Chebyshev series cs at a given point x, to (at most) the given order order.

This function evaluates a Chebyshev series cs at a given point x, estimating both the series result and its absolute error abserr, to (at most) the given order order. The error estimate is made from the first neglected term in the series.

29.5 Derivatives and Integrals

The following functions allow a Chebyshev series to be differentiated or integrated, producing a new Chebyshev series. Note that the error estimate produced by evaluating the derivative series will be underestimated due to the contribution of higher order terms being neglected.

int gsl_cheb_calc_deriv (gsl_cheb_series * deriv, const gsl_cheb_series * cs) [Function]

This function computes the derivative of the series cs, storing the derivative coefficients in the previously allocated deriv. The two series cs and deriv must have been allocated with the same order.

int gsl_cheb_calc_integ (gsl_cheb_series * integ, const gsl_cheb_series * cs) [Function]

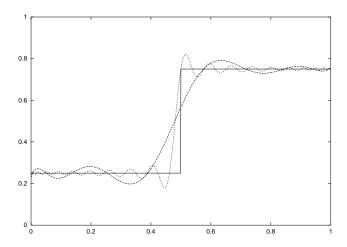
This function computes the integral of the series cs, storing the integral coefficients in the previously allocated *integ*. The two series cs and *integ* must have been allocated with the same order. The lower limit of the integration is taken to be the left hand end of the range a.

29.6 Examples

The following example program computes Chebyshev approximations to a step function. This is an extremely difficult approximation to make, due to the discontinuity, and was chosen as an example where approximation error is visible. For smooth functions the Chebyshev approximation converges extremely rapidly and errors would not be visible.

```
#include <stdio.h>
#include <gsl/gsl_math.h>
#include <gsl/gsl_chebyshev.h>
double
f (double x, void *p)
  if (x < 0.5)
    return 0.25;
  else
    return 0.75;
}
int
main (void)
{
  int i, n = 10000;
  gsl_cheb_series *cs = gsl_cheb_alloc (40);
  gsl_function F;
  F.function = f;
 F.params = 0;
  gsl_cheb_init (cs, &F, 0.0, 1.0);
  for (i = 0; i < n; i++)
    {
      double x = i / (double)n;
      double r10 = gsl_cheb_eval_n (cs, 10, x);
      double r40 = gsl_cheb_eval (cs, x);
      printf ("%g %g %g %g\n",
              x, GSL_FN_EVAL (&F, x), r10, r40);
    }
 gsl_cheb_free (cs);
  return 0;
}
```

The output from the program gives the original function, 10-th order approximation and 40-th order approximation, all sampled at intervals of 0.001 in x.



29.7 References and Further Reading

The following paper describes the use of Chebyshev series,

R. Broucke, "Ten Subroutines for the Manipulation of Chebyshev Series [C1] (Algorithm 446)". Communications of the ACM 16(4), 254-256 (1973)

30 Series Acceleration

The functions described in this chapter accelerate the convergence of a series using the Levin *u*-transform. This method takes a small number of terms from the start of a series and uses a systematic approximation to compute an extrapolated value and an estimate of its error. The *u*-transform works for both convergent and divergent series, including asymptotic series.

These functions are declared in the header file gsl_sum.h.

30.1 Acceleration functions

The following functions compute the full Levin u-transform of a series with its error estimate. The error estimate is computed by propagating rounding errors from each term through to the final extrapolation.

These functions are intended for summing analytic series where each term is known to high accuracy, and the rounding errors are assumed to originate from finite precision. They are taken to be relative errors of order GSL_DBL_EPSILON for each term.

The calculation of the error in the extrapolated value is an $O(N^2)$ process, which is expensive in time and memory. A faster but less reliable method which estimates the error from the convergence of the extrapolated value is described in the next section. For the method described here a full table of intermediate values and derivatives through to O(N) must be computed and stored, but this does give a reliable error estimate.

gsl_sum_levin_u_workspace * gsl_sum_levin_u_alloc (size_t n) [Function] This function allocates a workspace for a Levin u-transform of n terms. The size of the workspace is $O(2n^2 + 3n)$.

void gsl_sum_levin_u_free (gsl_sum_levin_u_workspace * w) [Function]
This function frees the memory associated with the workspace w.

This function takes the terms of a series in array of size array_size and computes the extrapolated limit of the series using a Levin u-transform. Additional working space must be provided in w. The extrapolated sum is stored in sum_accel, with an estimate of the absolute error stored in abserr. The actual term-by-term sum is returned in w->sum_plain. The algorithm calculates the truncation error (the difference between two successive extrapolations) and round-off error (propagated from the individual terms) to choose an optimal number of terms for the extrapolation. All the terms of the series passed in through array should be non-zero.

30.2 Acceleration functions without error estimation

The functions described in this section compute the Levin *u*-transform of series and attempt to estimate the error from the "truncation error" in the extrapolation, the difference between the final two approximations. Using this method avoids the need to compute an intermediate table of derivatives because the error is estimated from the behavior of the extrapolated

value itself. Consequently this algorithm is an O(N) process and only requires O(N) terms of storage. If the series converges sufficiently fast then this procedure can be acceptable. It is appropriate to use this method when there is a need to compute many extrapolations of series with similar convergence properties at high-speed. For example, when numerically integrating a function defined by a parameterized series where the parameter varies only slightly. A reliable error estimate should be computed first using the full algorithm described above in order to verify the consistency of the results.

This function allocates a workspace for a Levin u-transform of n terms, without error estimation. The size of the workspace is O(3n).

```
void gsl_sum_levin_utrunc_free (gsl_sum_levin_utrunc_workspace * [Function]
w)
```

This function frees the memory associated with the workspace w.

This function takes the terms of a series in array of size array_size and computes the extrapolated limit of the series using a Levin u-transform. Additional working space must be provided in w. The extrapolated sum is stored in sum_accel. The actual term-by-term sum is returned in w->sum_plain. The algorithm terminates when the difference between two successive extrapolations reaches a minimum or is sufficiently small. The difference between these two values is used as estimate of the error and is stored in abserr_trunc. To improve the reliability of the algorithm the extrapolated values are replaced by moving averages when calculating the truncation error, smoothing out any fluctuations.

30.3 Examples

The following code calculates an estimate of $\zeta(2) = \pi^2/6$ using the series,

$$\zeta(2) = 1 + 1/2^2 + 1/3^2 + 1/4^2 + \dots$$

After N terms the error in the sum is O(1/N), making direct summation of the series converge slowly.

```
#include <stdio.h>
#include <gsl/gsl_math.h>
#include <gsl/gsl_sum.h>
#define N 20
int
main (void)
{
   double t[N];
   double sum_accel, err;
```

```
double sum = 0;
 int n;
 gsl_sum_levin_u_workspace * w
   = gsl_sum_levin_u_alloc (N);
 const double zeta_2 = M_PI * M_PI / 6.0;
 /* terms for zeta(2) = \sum_{n=1}^{\sin y} 1/n^2 */
 for (n = 0; n < N; n++)
   {
     double np1 = n + 1.0;
     t[n] = 1.0 / (np1 * np1);
     sum += t[n];
   }
 gsl_sum_levin_u_accel (t, N, w, &sum_accel, &err);
 printf ("term-by-term sum = % .16f using %d terms\n",
         sum, N);
 printf ("term-by-term sum = % .16f using %d terms\n",
         w->sum_plain, w->terms_used);
 printf ("exact value = % .16f\n", zeta_2);
 printf ("accelerated sum = % .16f using %d terms\n",
          sum_accel, w->terms_used);
 printf ("estimated error = % .16f\n", err);
 printf ("actual error = \% .16f\n",
         sum_accel - zeta_2);
 gsl_sum_levin_u_free (w);
 return 0;
}
```

The output below shows that the Levin u-transform is able to obtain an estimate of the sum to 1 part in 10^{10} using the first eleven terms of the series. The error estimate returned by the function is also accurate, giving the correct number of significant digits.

```
$ ./a.out
term-by-term sum = 1.5961632439130233 using 20 terms
term-by-term sum = 1.5759958390005426 using 13 terms
exact value = 1.6449340668482264
accelerated sum = 1.6449340668166479 using 13 terms
estimated error = 0.0000000000508580
actual error = -0.0000000000315785
```

Note that a direct summation of this series would require 10^{10} terms to achieve the same precision as the accelerated sum does in 13 terms.

30.4 References and Further Reading

The algorithms used by these functions are described in the following papers,

T. Fessler, W.F. Ford, D.A. Smith, Hurry: An acceleration algorithm for scalar sequences and series *ACM Transactions on Mathematical Software*, 9(3):346–354, 1983. and Algorithm 602 9(3):355–357, 1983.

The theory of the *u*-transform was presented by Levin,

D. Levin, Development of Non-Linear Transformations for Improving Convergence of Sequences, *Intern. J. Computer Math.* B3:371–388, 1973.

A review paper on the Levin Transform is available online,

Herbert H. H. Homeier, Scalar Levin-Type Sequence Transformations, http://arxiv.org/abs/math/0005209.

31 Wavelet Transforms

This chapter describes functions for performing Discrete Wavelet Transforms (DWTs). The library includes wavelets for real data in both one and two dimensions. The wavelet functions are declared in the header files gsl_wavelet.h and gsl_wavelet2d.h.

31.1 Definitions

The continuous wavelet transform and its inverse are defined by the relations,

$$w(s,\tau) = \int_{-\infty}^{\infty} f(t) * \psi_{s,\tau}^{*}(t) dt$$

and,

$$f(t) = \int_0^\infty ds \int_{-\infty}^\infty w(s,\tau) * \psi_{s,\tau}(t) d\tau$$

where the basis functions $\psi_{s,\tau}$ are obtained by scaling and translation from a single function, referred to as the *mother wavelet*.

The discrete version of the wavelet transform acts on equally-spaced samples, with fixed scaling and translation steps (s, τ) . The frequency and time axes are sampled dyadically on scales of 2^j through a level parameter j. The resulting family of functions $\{\psi_{j,n}\}$ constitutes an orthonormal basis for square-integrable signals.

The discrete wavelet transform is an O(N) algorithm, and is also referred to as the fast wavelet transform.

31.2 Initialization

The gsl_wavelet structure contains the filter coefficients defining the wavelet and any associated offset parameters.

This function allocates and initializes a wavelet object of type T. The parameter k selects the specific member of the wavelet family. A null pointer is returned if insufficient memory is available or if a unsupported member is selected.

The following wavelet types are implemented:

gsl_wavelet_daubechies

[Wavelet]

gsl_wavelet_daubechies_centered

[Wavelet]

This is the Daubechies wavelet family of maximum phase with k/2 vanishing moments. The implemented wavelets are $k = 4, 6, \ldots, 20$, with k even.

gsl_wavelet_haar

[Wavelet]

gsl_wavelet_haar_centered

[Wavelet]

This is the Haar wavelet. The only valid choice of k for the Haar wavelet is k=2.

gsl_wavelet_bspline

[Wavelet]

gsl_wavelet_bspline_centered

[Wavelet]

This is the biorthogonal B-spline wavelet family of order (i, j). The implemented values of k = 100 * i + j are 103, 105, 202, 204, 206, 208, 301, 303, 305 307, 309.

The centered forms of the wavelets align the coefficients of the various sub-bands on edges. Thus the resulting visualization of the coefficients of the wavelet transform in the phase plane is easier to understand.

const char * gsl_wavelet_name (const gsl_wavelet * w) [Function]

This function returns a pointer to the name of the wavelet family for w.

void gsl_wavelet_free (gsl_wavelet * w)
This function frees the wavelet object w.
[Function]

The gsl_wavelet_workspace structure contains scratch space of the same size as the input data and is used to hold intermediate results during the transform.

This function allocates a workspace for the discrete wavelet transform. To perform a one-dimensional transform on n elements, a workspace of size n must be provided. For two-dimensional transforms of n-by-n matrices it is sufficient to allocate a workspace of size n, since the transform operates on individual rows and columns. A null pointer is returned if insufficient memory is available.

void gsl_wavelet_workspace_free (gsl_wavelet_workspace * work) [Function]
This function frees the allocated workspace work.

31.3 Transform Functions

This sections describes the actual functions performing the discrete wavelet transform. Note that the transforms use periodic boundary conditions. If the signal is not periodic in the sample length then spurious coefficients will appear at the beginning and end of each level of the transform.

31.3.1 Wavelet transforms in one dimension

int gsl_wavelet_transform_forward (const gsl_wavelet * w, double * [Function] data, size_t stride, size_t n, gsl_wavelet_workspace * work)

These functions compute in-place forward and inverse discrete wavelet transforms of length n with stride stride on the array data. The length of the transform n is restricted to powers of two. For the transform version of the function the argument dir can be either forward (+1) or backward (-1). A workspace work of length n must be provided.

For the forward transform, the elements of the original array are replaced by the discrete wavelet transform $f_i \to w_{j,k}$ in a packed triangular storage layout, where j is the index of the level $j = 0 \dots J - 1$ and k is the index of the coefficient within each

level, $k = 0 \dots 2^j - 1$. The total number of levels is $J = \log_2(n)$. The output data has the following form,

$$(s_{-1,0}, d_{0,0}, d_{1,0}, d_{1,1}, d_{2,0}, \cdots, d_{i,k}, \cdots, d_{J-1,2^{J-1}-1})$$

where the first element is the smoothing coefficient $s_{-1,0}$, followed by the detail coefficients $d_{j,k}$ for each level j. The backward transform inverts these coefficients to obtain the original data.

These functions return a status of $GSL_SUCCESS$ upon successful completion. GSL_EINVAL is returned if n is not an integer power of 2 or if insufficient workspace is provided.

31.3.2 Wavelet transforms in two dimension

The library provides functions to perform two-dimensional discrete wavelet transforms on square matrices. The matrix dimensions must be an integer power of two. There are two possible orderings of the rows and columns in the two-dimensional wavelet transform, referred to as the "standard" and "non-standard" forms.

The "standard" transform performs a complete discrete wavelet transform on the rows of the matrix, followed by a separate complete discrete wavelet transform on the columns of the resulting row-transformed matrix. This procedure uses the same ordering as a two-dimensional Fourier transform.

The "non-standard" transform is performed in interleaved passes on the rows and columns of the matrix for each level of the transform. The first level of the transform is applied to the matrix rows, and then to the matrix columns. This procedure is then repeated across the rows and columns of the data for the subsequent levels of the transform, until the full discrete wavelet transform is complete. The non-standard form of the discrete wavelet transform is typically used in image analysis.

The functions described in this section are declared in the header file gsl_wavelet2d.h.

These functions compute two-dimensional in-place forward and inverse discrete wavelet transforms in standard form on the array data stored in row-major form with dimensions size1 and size2 and physical row length tda. The dimensions must be equal (square matrix) and are restricted to powers of two. For the transform version of the function the argument dir can be either forward (+1) or backward (-1). A workspace work of the appropriate size must be provided. On exit, the appropriate elements of the array data are replaced by their two-dimensional wavelet transform.

The functions return a status of GSL_SUCCESS upon successful completion. GSL_EINVAL is returned if size1 and size2 are not equal and integer powers of 2, or if insufficient workspace is provided.

- int gsl_wavelet2d_transform_matrix (const gsl_wavelet * w, gsl_wavelet_direction dir, gsl_wavelet_workspace * work) [Function]

These functions compute the two-dimensional in-place wavelet transform on a matrix a.

These functions compute the two-dimensional wavelet transform in non-standard form.

- int gsl_wavelet2d_nstransform_matrix (const gsl_wavelet * w, gsl_wavelet_direction dir, gsl_wavelet_workspace * work) [Function]

These functions compute the non-standard form of the two-dimensional in-place wavelet transform on a matrix a.

31.4 Examples

The following program demonstrates the use of the one-dimensional wavelet transform functions. It computes an approximation to an input signal (of length 256) using the 20 largest components of the wavelet transform, while setting the others to zero.

```
#include <stdio.h>
#include <math.h>
#include <gsl/gsl_sort.h>
#include <gsl/gsl_wavelet.h>

int
main (int argc, char **argv)
{
  int i, n = 256, nc = 20;
```

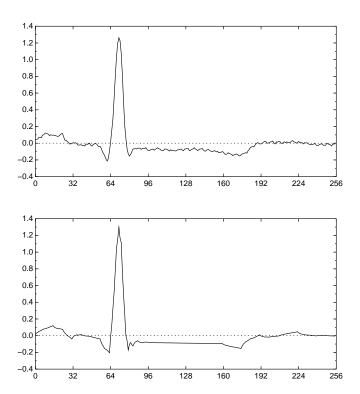
```
double *data = malloc (n * sizeof (double));
double *abscoeff = malloc (n * sizeof (double));
size_t *p = malloc (n * sizeof (size_t));
FILE * f;
gsl_wavelet *w;
gsl_wavelet_workspace *work;
w = gsl_wavelet_alloc (gsl_wavelet_daubechies, 4);
work = gsl_wavelet_workspace_alloc (n);
f = fopen (argv[1], "r");
for (i = 0; i < n; i++)
    fscanf (f, "%lg", &data[i]);
  }
fclose (f);
gsl_wavelet_transform_forward (w, data, 1, n, work);
for (i = 0; i < n; i++)
  {
    abscoeff[i] = fabs (data[i]);
  }
gsl_sort_index (p, abscoeff, 1, n);
for (i = 0; (i + nc) < n; i++)
  data[p[i]] = 0;
gsl_wavelet_transform_inverse (w, data, 1, n, work);
for (i = 0; i < n; i++)
  {
    printf ("%g\n", data[i]);
gsl_wavelet_free (w);
gsl_wavelet_workspace_free (work);
free (data);
free (abscoeff);
free (p);
return 0;
```

The output can be used with the GNU plotutils graph program,

}

```
$ ./a.out ecg.dat > dwt.dat
$ graph -T ps -x 0 256 32 -h 0.3 -a dwt.dat > dwt.ps
```

The graphs below show an original and compressed version of a sample ECG recording from the MIT-BIH Arrhythmia Database, part of the PhysioNet archive of public-domain of medical datasets.



Original (upper) and wavelet-compressed (lower) ECG signals, using the 20 largest components of the Daubechies(4) discrete wavelet transform.

31.5 References and Further Reading

The mathematical background to wavelet transforms is covered in the original lectures by Daubechies,

Ingrid Daubechies. Ten Lectures on Wavelets. CBMS-NSF Regional Conference Series in Applied Mathematics (1992), SIAM, ISBN 0898712742.

An easy to read introduction to the subject with an emphasis on the application of the wavelet transform in various branches of science is,

Paul S. Addison. The Illustrated Wavelet Transform Handbook. Institute of Physics Publishing (2002), ISBN 0750306920.

For extensive coverage of signal analysis by wavelets, wavelet packets and local cosine bases see,

S. G. Mallat. A wavelet tour of signal processing (Second edition). Academic Press (1999), ISBN 012466606X.

The concept of multiresolution analysis underlying the wavelet transform is described in,

- S. G. Mallat. Multiresolution Approximations and Wavelet Orthonormal Bases of $L^2(R)$. Transactions of the American Mathematical Society, 315(1), 1989, 69–87.
- S. G. Mallat. A Theory for Multiresolution Signal Decomposition—The Wavelet Representation. *IEEE Transactions on Pattern Analysis and Machine Intelligence*, 11, 1989, 674–693.

The coefficients for the individual wavelet families implemented by the library can be found in the following papers,

- I. Daubechies. Orthonormal Bases of Compactly Supported Wavelets. Communications on Pure and Applied Mathematics, 41 (1988) 909–996.
- A. Cohen, I. Daubechies, and J.-C. Feauveau. Biorthogonal Bases of Compactly Supported Wavelets. *Communications on Pure and Applied Mathematics*, 45 (1992) 485–560.

The PhysioNet archive of physiological datasets can be found online at http://www.physionet.org/ and is described in the following paper,

Goldberger et al. PhysioBank, PhysioToolkit, and PhysioNet: Components of a New Research Resource for Complex Physiologic Signals. *Circulation* 101(23):e215-e220 2000.

32 Discrete Hankel Transforms

This chapter describes functions for performing Discrete Hankel Transforms (DHTs). The functions are declared in the header file gsl_dht.h.

32.1 Definitions

The discrete Hankel transform acts on a vector of sampled data, where the samples are assumed to have been taken at points related to the zeroes of a Bessel function of fixed order; compare this to the case of the discrete Fourier transform, where samples are taken at points related to the zeroes of the sine or cosine function.

Specifically, let f(t) be a function on the unit interval and $j_{\nu,m}$ the m-th zero of the Bessel function $J_{\nu}(x)$. Then the finite ν -Hankel transform of f(t) is defined to be the set of numbers g_m given by,

$$g_m = \int_0^1 t dt \, J_{\nu}(j_{\nu,m}t) f(t),$$

so that,

$$f(t) = \sum_{m=1}^{\infty} \frac{2J_{\nu}(j_{\nu,m}t)}{J_{\nu+1}(j_{\nu,m})^2} g_m.$$

Suppose that f is band-limited in the sense that $g_m = 0$ for m > M. Then we have the following fundamental sampling theorem.

$$g_m = \frac{2}{j_{\nu,M}^2} \sum_{k=1}^{M-1} f\left(\frac{j_{\nu,k}}{j_{\nu,M}}\right) \frac{J_{\nu}(j_{\nu,m}j_{\nu,k}/j_{\nu,M})}{J_{\nu+1}(j_{\nu,k})^2}.$$

It is this discrete expression which defines the discrete Hankel transform. The kernel in the summation above defines the matrix of the ν -Hankel transform of size M-1. The coefficients of this matrix, being dependent on ν and M, must be precomputed and stored; the gsl_dht object encapsulates this data. The allocation function gsl_dht_alloc returns a gsl_dht object which must be properly initialized with gsl_dht_init before it can be used to perform transforms on data sample vectors, for fixed ν and M, using the gsl_dht_apply function. The implementation allows a scaling of the fundamental interval, for convenience, so that one can assume the function is defined on the interval [0,X], rather than the unit interval.

Notice that by assumption f(t) vanishes at the endpoints of the interval, consistent with the inversion formula and the sampling formula given above. Therefore, this transform corresponds to an orthogonal expansion in eigenfunctions of the Dirichlet problem for the Bessel differential equation.

32.2 Functions

gsl_dht * gsl_dht_alloc (size_t size)

[Function]

This function allocates a Discrete Hankel transform object of size size.

int gsl_dht_init (gsl_dht * t, double nu, double xmax)

[Function]

This function initializes the transform t for the given values of nu and xmax.

- gsl_dht * gsl_dht_new (size_t size, double nu, double xmax) [Function]
 This function allocates a Discrete Hankel transform object of size size and initializes it for the given values of nu and xmax.
- void gsl_dht_free ($gsl_dht * t$) [Function] This function frees the transform t.
- int gsl_dht_apply (const gsl_dht * t, double * f_in, double * f_out) [Function]
 This function applies the transform t to the array f_in whose size is equal to the size of the transform. The result is stored in the array f_out which must be of the same length.

Applying this function to its output gives the original data multiplied by $(1/j_{\nu,M})^2$, up to numerical errors.

- double gsl_dht_x_sample (const gsl_dht * t, int n) [Function] This function returns the value of the n-th sample point in the unit interval, $(j_{\nu,n+1}/j_{\nu,M})X$. These are the points where the function f(t) is assumed to be sampled.
- double gsl_dht_k_sample (const gsl_dht * t, int n) [Function]

 This function returns the value of the n-th sample point in "k-space", $j_{\nu,n+1}/X$.

32.3 References and Further Reading

The algorithms used by these functions are described in the following papers,

- H. Fisk Johnson, Comp. Phys. Comm. 43, 181 (1987).
- D. Lemoine, J. Chem. Phys. 101, 3936 (1994).

33 One dimensional Root-Finding

This chapter describes routines for finding roots of arbitrary one-dimensional functions. The library provides low level components for a variety of iterative solvers and convergence tests. These can be combined by the user to achieve the desired solution, with full access to the intermediate steps of the iteration. Each class of methods uses the same framework, so that you can switch between solvers at runtime without needing to recompile your program. Each instance of a solver keeps track of its own state, allowing the solvers to be used in multi-threaded programs.

The header file gsl_roots.h contains prototypes for the root finding functions and related declarations.

33.1 Overview

One-dimensional root finding algorithms can be divided into two classes, root bracketing and root polishing. Algorithms which proceed by bracketing a root are guaranteed to converge. Bracketing algorithms begin with a bounded region known to contain a root. The size of this bounded region is reduced, iteratively, until it encloses the root to a desired tolerance. This provides a rigorous error estimate for the location of the root.

The technique of root polishing attempts to improve an initial guess to the root. These algorithms converge only if started "close enough" to a root, and sacrifice a rigorous error bound for speed. By approximating the behavior of a function in the vicinity of a root they attempt to find a higher order improvement of an initial guess. When the behavior of the function is compatible with the algorithm and a good initial guess is available a polishing algorithm can provide rapid convergence.

In GSL both types of algorithm are available in similar frameworks. The user provides a high-level driver for the algorithms, and the library provides the individual functions necessary for each of the steps. There are three main phases of the iteration. The steps are,

- initialize solver state, s, for algorithm T
- update s using the iteration T
- test s for convergence, and repeat iteration if necessary

The state for bracketing solvers is held in a gsl_root_fsolver struct. The updating procedure uses only function evaluations (not derivatives). The state for root polishing solvers is held in a gsl_root_fdfsolver struct. The updates require both the function and its derivative (hence the name fdf) to be supplied by the user.

33.2 Caveats

Note that root finding functions can only search for one root at a time. When there are several roots in the search area, the first root to be found will be returned; however it is difficult to predict which of the roots this will be. In most cases, no error will be reported if you try to find a root in an area where there is more than one.

Care must be taken when a function may have a multiple root (such as $f(x) = (x-x_0)^2$ or $f(x) = (x-x_0)^3$). It is not possible to use root-bracketing algorithms on even-multiplicity roots. For these algorithms the initial interval must contain a zero-crossing, where the function is negative at one end of the interval and positive at the other end. Roots with

even-multiplicity do not cross zero, but only touch it instantaneously. Algorithms based on root bracketing will still work for odd-multiplicity roots (e.g. cubic, quintic, ...). Root polishing algorithms generally work with higher multiplicity roots, but at a reduced rate of convergence. In these cases the *Steffenson algorithm* can be used to accelerate the convergence of multiple roots.

While it is not absolutely required that f have a root within the search region, numerical root finding functions should not be used haphazardly to check for the *existence* of roots. There are better ways to do this. Because it is easy to create situations where numerical root finders can fail, it is a bad idea to throw a root finder at a function you do not know much about. In general it is best to examine the function visually by plotting before searching for a root.

33.3 Initializing the Solver

This function returns a pointer to a newly allocated instance of a solver of type T. For example, the following code creates an instance of a bisection solver,

If there is insufficient memory to create the solver then the function returns a null pointer and the error handler is invoked with an error code of GSL_ENOMEM.

This function returns a pointer to a newly allocated instance of a derivative-based solver of type T. For example, the following code creates an instance of a Newton-Raphson solver,

If there is insufficient memory to create the solver then the function returns a null pointer and the error handler is invoked with an error code of GSL_ENOMEM.

This function initializes, or reinitializes, an existing solver s to use the function f and the initial search interval $[x_lower, x_upper]$.

This function initializes, or reinitializes, an existing solver s to use the function and derivative fdf and the initial guess root.

33.4 Providing the function to solve

You must provide a continuous function of one variable for the root finders to operate on, and, sometimes, its first derivative. In order to allow for general parameters the functions are defined by the following data types:

```
gsl_function [Data Type]
```

This data type defines a general function with parameters.

```
double (* function) (double x, void * params) this function should return the value f(x, params) for argument x and parameters params
```

void * params

a pointer to the parameters of the function

Here is an example for the general quadratic function,

$$f(x) = ax^2 + bx + c$$

with a = 3, b = 2, c = 1. The following code defines a gsl_function F which you could pass to a root finder as a function pointer:

```
F.function = &my_f;
F.params = &params;
```

The function f(x) can be evaluated using the macro GSL_FN_EVAL(&F,x) defined in gsl_math.h

gsl_function_fdf

[Data Type]

This data type defines a general function with parameters and its first derivative.

```
double (* f) (double x, void * params)
```

this function should return the value of f(x, params) for argument x and parameters params

```
double (* df) (double x, void * params)
```

this function should return the value of the derivative of f with respect to x, f'(x, params), for argument x and parameters params

void (* fdf) (double x, void * params, double * f, double * df)

this function should set the values of the function f to f(x, params) and its derivative df to f'(x, params) for argument x and parameters params. This function provides an optimization of the separate functions for f(x) and f'(x)—it is always faster to compute the function and its derivative at the same time.

void * params

a pointer to the parameters of the function

```
Here is an example where f(x) = \exp(2x):
  double
  my_f (double x, void * params)
     return exp (2 * x);
  }
  my_df (double x, void * params)
  {
     return 2 * exp (2 * x);
  }
  my_fdf (double x, void * params,
          double * f, double * df)
  {
     double t = \exp(2 * x);
     *f = t;
     *df = 2 * t; /* uses existing value */
  }
```

```
gsl_function_fdf FDF;
FDF.f = &my_f;
FDF.df = &my_df;
FDF.fdf = &my_fdf;
FDF.params = 0;
```

The function f(x) can be evaluated using the macro GSL_FN_FDF_EVAL_F(&FDF,x) and the derivative f'(x) can be evaluated using the macro GSL_FN_FDF_EVAL_DF(&FDF,x). Both the function y = f(x) and its derivative dy = f'(x) can be evaluated at the same time using the macro GSL_FN_FDF_EVAL_F_DF(&FDF,x,y,dy). The macro stores f(x) in its y argument and f'(x) in its dy argument—both of these should be pointers to double.

33.5 Search Bounds and Guesses

You provide either search bounds or an initial guess; this section explains how search bounds and guesses work and how function arguments control them.

A guess is simply an x value which is iterated until it is within the desired precision of a root. It takes the form of a double.

Search bounds are the endpoints of an interval which is iterated until the length of the interval is smaller than the requested precision. The interval is defined by two values, the lower limit and the upper limit. Whether the endpoints are intended to be included in the interval or not depends on the context in which the interval is used.

33.6 Iteration

The following functions drive the iteration of each algorithm. Each function performs one iteration to update the state of any solver of the corresponding type. The same functions work for all solvers so that different methods can be substituted at runtime without modifications to the code.

These functions perform a single iteration of the solver s. If the iteration encounters an unexpected problem then an error code will be returned,

```
GSL_EBADFUNC
```

the iteration encountered a singular point where the function or its derivative evaluated to Inf or NaN.

```
GSL_EZERODIV
```

the derivative of the function vanished at the iteration point, preventing the algorithm from continuing without a division by zero.

The solver maintains a current best estimate of the root at all times. The bracketing solvers also keep track of the current best interval bounding the root. This information can be accessed with the following auxiliary functions,

```
double gsl_root_fsolver_root (const gsl_root_fsolver * s) [Function]
double gsl_root_fdfsolver_root (const gsl_root_fdfsolver * s) [Function]
```

These functions return the current estimate of the root for the solver s.

double gsl_root_fsolver_x_lower (const gsl_root_fsolver * s) [Function]
double gsl_root_fsolver_x_upper (const gsl_root_fsolver * s) [Function]
These functions return the current bracketing interval for the solver s.

33.7 Search Stopping Parameters

A root finding procedure should stop when one of the following conditions is true:

- A root has been found to within the user-specified precision.
- A user-specified maximum number of iterations has been reached.
- An error has occurred.

The handling of these conditions is under user control. The functions below allow the user to test the precision of the current result in several standard ways.

This function tests for the convergence of the interval $[x_lower, x_upper]$ with absolute error epsabs and relative error epsable. The test returns GSL_SUCCESS if the following condition is achieved,

$$|a-b| < epsabs + epsrel \min(|a|, |b|)$$

when the interval x = [a, b] does not include the origin. If the interval includes the origin then $\min(|a|, |b|)$ is replaced by zero (which is the minimum value of |x| over the interval). This ensures that the relative error is accurately estimated for roots close to the origin.

This condition on the interval also implies that any estimate of the root r in the interval satisfies the same condition with respect to the true root r^* ,

$$|r - r^*| < epsabs + epsrel r^*$$

assuming that the true root r^* is contained within the interval.

This function tests for the convergence of the sequence ..., x0, x1 with absolute error *epsabs* and relative error *epsrel*. The test returns GSL_SUCCESS if the following condition is achieved.

$$|x_1 - x_0| < epsabs + epsrel |x_1|$$

and returns GSL_CONTINUE otherwise.

int gsl_root_test_residual (double f, double epsabs) [Function]

This function tests the residual value f against the absolute error bound epsabs. The test returns $GSL_SUCCESS$ if the following condition is achieved,

and returns GSL_CONTINUE otherwise. This criterion is suitable for situations where the precise location of the root, x, is unimportant provided a value can be found where the residual, |f(x)|, is small enough.

33.8 Root Bracketing Algorithms

The root bracketing algorithms described in this section require an initial interval which is guaranteed to contain a root—if a and b are the endpoints of the interval then f(a) must differ in sign from f(b). This ensures that the function crosses zero at least once in the interval. If a valid initial interval is used then these algorithm cannot fail, provided the function is well-behaved.

Note that a bracketing algorithm cannot find roots of even degree, since these do not cross the x-axis.

gsl_root_fsolver_bisection

[Solver]

The bisection algorithm is the simplest method of bracketing the roots of a function. It is the slowest algorithm provided by the library, with linear convergence.

On each iteration, the interval is bisected and the value of the function at the midpoint is calculated. The sign of this value is used to determine which half of the interval does not contain a root. That half is discarded to give a new, smaller interval containing the root. This procedure can be continued indefinitely until the interval is sufficiently small.

At any time the current estimate of the root is taken as the midpoint of the interval.

gsl_root_fsolver_falsepos

[Solver]

The false position algorithm is a method of finding roots based on linear interpolation. Its convergence is linear, but it is usually faster than bisection.

On each iteration a line is drawn between the endpoints (a, f(a)) and (b, f(b)) and the point where this line crosses the x-axis taken as a "midpoint". The value of the function at this point is calculated and its sign is used to determine which side of the interval does not contain a root. That side is discarded to give a new, smaller interval containing the root. This procedure can be continued indefinitely until the interval is sufficiently small.

The best estimate of the root is taken from the linear interpolation of the interval on the current iteration.

gsl_root_fsolver_brent

[Solver]

The Brent-Dekker method (referred to here as Brent's method) combines an interpolation strategy with the bisection algorithm. This produces a fast algorithm which is still robust.

On each iteration Brent's method approximates the function using an interpolating curve. On the first iteration this is a linear interpolation of the two endpoints. For subsequent iterations the algorithm uses an inverse quadratic fit to the last three points, for higher accuracy. The intercept of the interpolating curve with the x-axis is taken as a guess for the root. If it lies within the bounds of the current interval then the interpolating point is accepted, and used to generate a smaller interval. If the interpolating point is not accepted then the algorithm falls back to an ordinary bisection step.

The best estimate of the root is taken from the most recent interpolation or bisection.

33.9 Root Finding Algorithms using Derivatives

The root polishing algorithms described in this section require an initial guess for the location of the root. There is no absolute guarantee of convergence—the function must be suitable for this technique and the initial guess must be sufficiently close to the root for it to work. When these conditions are satisfied then convergence is quadratic.

These algorithms make use of both the function and its derivative.

gsl_root_fdfsolver_newton

[Derivative Solver]

Newton's Method is the standard root-polishing algorithm. The algorithm begins with an initial guess for the location of the root. On each iteration, a line tangent to the function f is drawn at that position. The point where this line crosses the x-axis becomes the new guess. The iteration is defined by the following sequence,

$$x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$$

Newton's method converges quadratically for single roots, and linearly for multiple roots.

gsl_root_fdfsolver_secant

[Derivative Solver]

The secant method is a simplified version of Newton's method which does not require the computation of the derivative on every step.

On its first iteration the algorithm begins with Newton's method, using the derivative to compute a first step,

$$x_1 = x_0 - \frac{f(x_0)}{f'(x_0)}$$

Subsequent iterations avoid the evaluation of the derivative by replacing it with a numerical estimate, the slope of the line through the previous two points,

$$x_{i+1} = x_i - \frac{f(x_i)}{f'_{est}}$$
 where $f'_{est} = \frac{f(x_i) - f(x_{i-1})}{x_i - x_{i-1}}$

When the derivative does not change significantly in the vicinity of the root the secant method gives a useful saving. Asymptotically the secant method is faster than Newton's method whenever the cost of evaluating the derivative is more than 0.44 times the cost of evaluating the function itself. As with all methods of computing a numerical derivative the estimate can suffer from cancellation errors if the separation of the points becomes too small.

On single roots, the method has a convergence of order $(1 + \sqrt{5})/2$ (approximately 1.62). It converges linearly for multiple roots.

gsl_root_fdfsolver_steffenson

[Derivative Solver]

The Steffenson Method¹ provides the fastest convergence of all the routines. It combines the basic Newton algorithm with an Aitken "delta-squared" acceleration. If the

J.F. Steffensen (1873–1961). The spelling used in the name of the function is slightly incorrect, but has been preserved to avoid incompatibility.

Newton iterates are x_i then the acceleration procedure generates a new sequence R_i ,

$$R_i = x_i - \frac{(x_{i+1} - x_i)^2}{(x_{i+2} - 2x_{i+1} + x_i)}$$

which converges faster than the original sequence under reasonable conditions. The new sequence requires three terms before it can produce its first value so the method returns accelerated values on the second and subsequent iterations. On the first iteration it returns the ordinary Newton estimate. The Newton iterate is also returned if the denominator of the acceleration term ever becomes zero.

As with all acceleration procedures this method can become unstable if the function is not well-behaved.

33.10 Examples

For any root finding algorithm we need to prepare the function to be solved. For this example we will use the general quadratic equation described earlier. We first need a header file (demo_fn.h) to define the function parameters,

```
struct quadratic_params
       {
         double a, b, c;
       };
     double quadratic (double x, void *params);
     double quadratic_deriv (double x, void *params);
     void quadratic_fdf (double x, void *params,
                          double *y, double *dy);
We place the function definitions in a separate file (demo_fn.c),
     double
     quadratic (double x, void *params)
       struct quadratic_params *p
         = (struct quadratic_params *) params;
       double a = p->a;
       double b = p->b;
       double c = p->c;
       return (a * x + b) * x + c;
     }
     double
     quadratic_deriv (double x, void *params)
       struct quadratic_params *p
         = (struct quadratic_params *) params;
```

```
double a = p->a;
  double b = p->b;
  double c = p -> c;
 return 2.0 * a * x + b;
}
void
quadratic_fdf (double x, void *params,
               double *y, double *dy)
  struct quadratic_params *p
    = (struct quadratic_params *) params;
  double a = p->a;
  double b = p->b;
  double c = p -> c;
  *y = (a * x + b) * x + c;
  *dy = 2.0 * a * x + b;
}
```

The first program uses the function solver gsl_root_fsolver_brent for Brent's method and the general quadratic defined above to solve the following equation,

$$x^2 - 5 = 0$$

```
with solution x = \sqrt{5} = 2.236068...
     #include <stdio.h>
     #include <gsl/gsl_errno.h>
     #include <gsl/gsl_math.h>
     #include <gsl/gsl_roots.h>
     #include "demo_fn.h"
     #include "demo_fn.c"
     int
     main (void)
       int status;
       int iter = 0, max_iter = 100;
       const gsl_root_fsolver_type *T;
       gsl_root_fsolver *s;
       double r = 0, r_{expected} = sqrt (5.0);
       double x_{lo} = 0.0, x_{hi} = 5.0;
       gsl_function F;
       struct quadratic_params params = {1.0, 0.0, -5.0};
```

```
F.function = &quadratic;
       F.params = &params;
       T = gsl_root_fsolver_brent;
       s = gsl_root_fsolver_alloc (T);
       gsl_root_fsolver_set (s, &F, x_lo, x_hi);
       printf ("using %s method\n",
                gsl_root_fsolver_name (s));
       printf ("%5s [%9s, %9s] %9s %10s %9s\n",
                "iter", "lower", "upper", "root",
                "err", "err(est)");
       do
         {
            iter++;
            status = gsl_root_fsolver_iterate (s);
           r = gsl_root_fsolver_root (s);
           x_lo = gsl_root_fsolver_x_lower (s);
           x_hi = gsl_root_fsolver_x_upper (s);
            status = gsl_root_test_interval (x_lo, x_hi,
                                               0, 0.001);
            if (status == GSL_SUCCESS)
              printf ("Converged:\n");
           printf ("%5d [%.7f, %.7f] %.7f %+.7f %.7f\n",
                    iter, x_lo, x_hi,
                    r, r - r_expected,
                    x_hi - x_lo);
         }
       while (status == GSL_CONTINUE && iter < max_iter);</pre>
       gsl_root_fsolver_free (s);
       return status;
     }
Here are the results of the iterations,
     $ ./a.out
     using brent method
      iter [ lower,
                        upper]
                                   root.
                                              err err(est)
         1 [1.0000000, 5.0000000] 1.0000000 -1.2360680 4.0000000
         2 [1.0000000, 3.0000000] 3.0000000 +0.7639320 2.0000000
         3 [2.0000000, 3.0000000] 2.0000000 -0.2360680 1.0000000
         4 [2.2000000, 3.0000000] 2.2000000 -0.0360680 0.8000000
         5 [2.2000000, 2.2366300] 2.2366300 +0.0005621 0.0366300
     Converged:
```

```
6 [2.2360634, 2.2366300] 2.2360634 -0.0000046 0.0005666
```

If the program is modified to use the bisection solver instead of Brent's method, by changing gsl_root_fsolver_brent to gsl_root_fsolver_bisection the slower convergence of the Bisection method can be observed,

```
$ ./a.out
using bisection method
iter [
          lower,
                     upper]
                                 root
                                              err err(est)
    1 [0.0000000, 2.5000000] 1.2500000 -0.9860680 2.5000000
    2 [1.2500000, 2.5000000] 1.8750000 -0.3610680 1.2500000
    3 [1.8750000, 2.5000000] 2.1875000 -0.0485680 0.6250000
    4 [2.1875000, 2.5000000] 2.3437500 +0.1076820 0.3125000
   5 [2.1875000, 2.3437500] 2.2656250 +0.0295570 0.1562500
    6 [2.1875000, 2.2656250] 2.2265625 -0.0095055 0.0781250
   7 [2.2265625, 2.2656250] 2.2460938 +0.0100258 0.0390625
   8 [2.2265625, 2.2460938] 2.2363281 +0.0002601 0.0195312
   9 [2.2265625, 2.2363281] 2.2314453 -0.0046227 0.0097656
   10 [2.2314453, 2.2363281] 2.2338867 -0.0021813 0.0048828
   11 [2.2338867, 2.2363281] 2.2351074 -0.0009606 0.0024414
Converged:
   12 [2.2351074, 2.2363281] 2.2357178 -0.0003502 0.0012207
```

The next program solves the same function using a derivative solver instead.

```
#include <stdio.h>
#include <gsl/gsl_errno.h>
#include <gsl/gsl_math.h>
#include <gsl/gsl_roots.h>
#include "demo_fn.h"
#include "demo_fn.c"
int
main (void)
{
  int status;
  int iter = 0, max_iter = 100;
  const gsl_root_fdfsolver_type *T;
  gsl_root_fdfsolver *s;
  double x0, x = 5.0, r_{expected} = sqrt (5.0);
  gsl_function_fdf FDF;
  struct quadratic_params params = {1.0, 0.0, -5.0};
  FDF.f = &quadratic;
  FDF.df = &quadratic_deriv;
  FDF.fdf = &quadratic_fdf;
  FDF.params = &params;
  T = gsl_root_fdfsolver_newton;
  s = gsl_root_fdfsolver_alloc (T);
  gsl_root_fdfsolver_set (s, &FDF, x);
```

printf ("using %s method\n",

```
gsl_root_fdfsolver_name (s));
       printf ("%-5s %10s %10s %10s\n",
               "iter", "root", "err", "err(est)");
       do
         {
           iter++;
           status = gsl_root_fdfsolver_iterate (s);
           x0 = x;
           x = gsl_root_fdfsolver_root (s);
           status = gsl_root_test_delta (x, x0, 0, 1e-3);
           if (status == GSL_SUCCESS)
             printf ("Converged:\n");
           printf ("%5d %10.7f %+10.7f %10.7f\n",
                   iter, x, x - r_{expected}, x - x0);
         }
       while (status == GSL_CONTINUE && iter < max_iter);</pre>
       gsl_root_fdfsolver_free (s);
       return status;
Here are the results for Newton's method,
     $ ./a.out
     using newton method
     iter
                 root
                              err
                                    err(est)
         1 3.0000000 +0.7639320 -2.0000000
         2 2.3333333 +0.0972654 -0.6666667
         3 2.2380952 +0.0020273 -0.0952381
     Converged:
         4 2.2360689 +0.0000009 -0.0020263
```

Note that the error can be estimated more accurately by taking the difference between the current iterate and next iterate rather than the previous iterate. The other derivative solvers can be investigated by changing gsl_root_fdfsolver_newton to gsl_root_fdfsolver_secant or gsl_root_fdfsolver_steffenson.

33.11 References and Further Reading

For information on the Brent-Dekker algorithm see the following two papers,

- R. P. Brent, "An algorithm with guaranteed convergence for finding a zero of a function", Computer Journal, 14 (1971) 422–425
- J. C. P. Bus and T. J. Dekker, "Two Efficient Algorithms with Guaranteed Convergence for Finding a Zero of a Function", *ACM Transactions of Mathematical Software*, Vol. 1 No. 4 (1975) 330–345

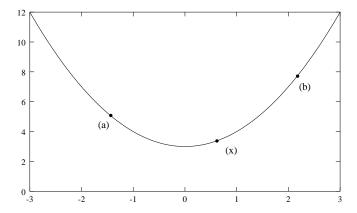
34 One dimensional Minimization

This chapter describes routines for finding minima of arbitrary one-dimensional functions. The library provides low level components for a variety of iterative minimizers and convergence tests. These can be combined by the user to achieve the desired solution, with full access to the intermediate steps of the algorithms. Each class of methods uses the same framework, so that you can switch between minimizers at runtime without needing to recompile your program. Each instance of a minimizer keeps track of its own state, allowing the minimizers to be used in multi-threaded programs.

The header file gsl_min.h contains prototypes for the minimization functions and related declarations. To use the minimization algorithms to find the maximum of a function simply invert its sign.

34.1 Overview

The minimization algorithms begin with a bounded region known to contain a minimum. The region is described by a lower bound a and an upper bound b, with an estimate of the location of the minimum x.



The value of the function at x must be less than the value of the function at the ends of the interval,

This condition guarantees that a minimum is contained somewhere within the interval. On each iteration a new point x' is selected using one of the available algorithms. If the new point is a better estimate of the minimum, i.e. where f(x') < f(x), then the current estimate of the minimum x is updated. The new point also allows the size of the bounded interval to be reduced, by choosing the most compact set of points which satisfies the constraint f(a) > f(x) < f(b). The interval is reduced until it encloses the true minimum to a desired tolerance. This provides a best estimate of the location of the minimum and a rigorous error estimate.

Several bracketing algorithms are available within a single framework. The user provides a high-level driver for the algorithm, and the library provides the individual functions necessary for each of the steps. There are three main phases of the iteration. The steps are,

- initialize minimizer state, s, for algorithm T
- update s using the iteration T
- test s for convergence, and repeat iteration if necessary

The state for the minimizers is held in a gsl_min_fminimizer struct. The updating procedure uses only function evaluations (not derivatives).

34.2 Caveats

Note that minimization functions can only search for one minimum at a time. When there are several minima in the search area, the first minimum to be found will be returned; however it is difficult to predict which of the minima this will be. In most cases, no error will be reported if you try to find a minimum in an area where there is more than one.

With all minimization algorithms it can be difficult to determine the location of the minimum to full numerical precision. The behavior of the function in the region of the minimum x^* can be approximated by a Taylor expansion,

$$y = f(x^*) + \frac{1}{2}f''(x^*)(x - x^*)^2$$

and the second term of this expansion can be lost when added to the first term at finite precision. This magnifies the error in locating x^* , making it proportional to $\sqrt{\epsilon}$ (where ϵ is the relative accuracy of the floating point numbers). For functions with higher order minima, such as x^4 , the magnification of the error is correspondingly worse. The best that can be achieved is to converge to the limit of numerical accuracy in the function values, rather than the location of the minimum itself.

34.3 Initializing the Minimizer

```
gsl_min_fminimizer * gsl_min_fminimizer_alloc (const gsl_min_fminimizer_type * T) [Function]
```

This function returns a pointer to a newly allocated instance of a minimizer of type T. For example, the following code creates an instance of a golden section minimizer,

If there is insufficient memory to create the minimizer then the function returns a null pointer and the error handler is invoked with an error code of GSL_ENOMEM.

```
int gsl_min_fminimizer_set (gsl_min_fminimizer * s, gsl_function * [Function]
f, double x_minimum, double x_lower, double x_upper)
```

This function sets, or resets, an existing minimizer s to use the function f and the initial search interval $[x_lower, x_upper]$, with a guess for the location of the minimum $x_minimum$.

If the interval given does not contain a minimum, then the function returns an error code of GSL_EINVAL.

This function is equivalent to $gsl_min_fminimizer_set$ but uses the values $f_minimum$, f_min

void gsl_min_fminimizer_free (gsl_min_fminimizer * s) [Function]
This function frees all the memory associated with the minimizer s.

const char * gsl_min_fminimizer_name (const gsl_min_fminimizer * [Function]
s)

This function returns a pointer to the name of the minimizer. For example,

would print something like s is a 'brent' minimizer.

34.4 Providing the function to minimize

You must provide a continuous function of one variable for the minimizers to operate on. In order to allow for general parameters the functions are defined by a gsl_function data type (see Section 33.4 [Providing the function to solve], page 359).

34.5 Iteration

The following functions drive the iteration of each algorithm. Each function performs one iteration to update the state of any minimizer of the corresponding type. The same functions work for all minimizers so that different methods can be substituted at runtime without modifications to the code.

int gsl_min_fminimizer_iterate (gsl_min_fminimizer * s) [Function]
This function performs a single iteration of the minimizer s. If the iteration encounters

an unexpected problem then an error code will be returned,

GSL_EBADFUNC

the iteration encountered a singular point where the function evaluated to Inf or NaN.

GSL_FAILURE

the algorithm could not improve the current best approximation or bounding interval.

The minimizer maintains a current best estimate of the position of the minimum at all times, and the current interval bounding the minimum. This information can be accessed with the following auxiliary functions,

double gsl_min_fminimizer_x_minimum (const gsl_min_fminimizer * [Function]
s)

This function returns the current estimate of the position of the minimum for the minimizer s.

double gsl_min_fminimizer_x_upper (const gsl_min_fminimizer * s) [Function] double gsl_min_fminimizer_x_lower (const gsl_min_fminimizer * s) [Function] These functions return the current upper and lower bound of the interval for the minimizer s.

double gsl_min_fminimizer_f_minimum (const gsl_min_fminimizer * [Function]
s)

double gsl_min_fminimizer_f_upper (const gsl_min_fminimizer * s) [Function] double gsl_min_fminimizer_f_lower (const gsl_min_fminimizer * s) [Function]

These functions return the value of the function at the current estimate of the minimum and at the upper and lower bounds of the interval for the minimizer s.

34.6 Stopping Parameters

A minimization procedure should stop when one of the following conditions is true:

- A minimum has been found to within the user-specified precision.
- A user-specified maximum number of iterations has been reached.
- An error has occurred.

The handling of these conditions is under user control. The function below allows the user to test the precision of the current result.

This function tests for the convergence of the interval $[x_lower, x_upper]$ with absolute error epsabs and relative error epsrel. The test returns GSL_SUCCESS if the following condition is achieved,

$$|a-b| < epsabs + epsrel \min(|a|, |b|)$$

when the interval x = [a, b] does not include the origin. If the interval includes the origin then $\min(|a|, |b|)$ is replaced by zero (which is the minimum value of |x| over the interval). This ensures that the relative error is accurately estimated for minima close to the origin.

This condition on the interval also implies that any estimate of the minimum x_m in the interval satisfies the same condition with respect to the true minimum x_m^* ,

$$|x_m - x_m^*| < epsabs + epsrel x_m^*$$

assuming that the true minimum x_m^* is contained within the interval.

34.7 Minimization Algorithms

The minimization algorithms described in this section require an initial interval which is guaranteed to contain a minimum—if a and b are the endpoints of the interval and x is an estimate of the minimum then f(a) > f(x) < f(b). This ensures that the function has at least one minimum somewhere in the interval. If a valid initial interval is used then these algorithm cannot fail, provided the function is well-behaved.

gsl_min_fminimizer_goldensection

[Minimizer]

The golden section algorithm is the simplest method of bracketing the minimum of a function. It is the slowest algorithm provided by the library, with linear convergence.

On each iteration, the algorithm first compares the subintervals from the endpoints to the current minimum. The larger subinterval is divided in a golden section (using the famous ratio $(3-\sqrt{5})/2=0.3189660...$) and the value of the function at this new point is calculated. The new value is used with the constraint f(a') > f(x') < f(b') to a select new interval containing the minimum, by discarding the least useful point. This procedure can be continued indefinitely until the interval is sufficiently small. Choosing the golden section as the bisection ratio can be shown to provide the fastest convergence for this type of algorithm.

gsl_min_fminimizer_brent

[Minimizer]

The Brent minimization algorithm combines a parabolic interpolation with the golden section algorithm. This produces a fast algorithm which is still robust.

The outline of the algorithm can be summarized as follows: on each iteration Brent's method approximates the function using an interpolating parabola through three existing points. The minimum of the parabola is taken as a guess for the minimum. If it lies within the bounds of the current interval then the interpolating point is accepted, and used to generate a smaller interval. If the interpolating point is not accepted then the algorithm falls back to an ordinary golden section step. The full details of Brent's method include some additional checks to improve convergence.

gsl_min_fminimizer_quad_golden

[Minimizer]

This is a variant of Brent's algorithm which uses the safeguarded step-length algorithm of Gill and Murray.

34.8 Examples

The following program uses the Brent algorithm to find the minimum of the function $f(x) = \cos(x) + 1$, which occurs at $x = \pi$. The starting interval is (0,6), with an initial guess for the minimum of 2.

```
#include <stdio.h>
#include <gsl/gsl_errno.h>
#include <gsl/gsl_math.h>
#include <gsl/gsl_min.h>

double fn1 (double x, void * params)
{
   return cos(x) + 1.0;
}

int
main (void)
{
   int status;
   int iter = 0, max_iter = 100;
```

```
const gsl_min_fminimizer_type *T;
gsl_min_fminimizer *s;
double m = 2.0, m_expected = M_PI;
double a = 0.0, b = 6.0;
gsl_function F;
F.function = &fn1;
F.params = 0;
T = gsl_min_fminimizer_brent;
s = gsl_min_fminimizer_alloc (T);
gsl_min_fminimizer_set (s, &F, m, a, b);
printf ("using %s method\n",
        gsl_min_fminimizer_name (s));
printf ("%5s [%9s, %9s] %9s %10s %9s\n",
        "iter", "lower", "upper", "min",
        "err", "err(est)");
printf ("%5d [%.7f, %.7f] %.7f %+.7f %.7f\n",
        iter, a, b,
        m, m - m_{expected}, b - a);
do
  {
    iter++;
    status = gsl_min_fminimizer_iterate (s);
    m = gsl_min_fminimizer_x_minimum (s);
    a = gsl_min_fminimizer_x_lower (s);
    b = gsl_min_fminimizer_x_upper (s);
    status
      = gsl_min_test_interval (a, b, 0.001, 0.0);
    if (status == GSL_SUCCESS)
      printf ("Converged:\n");
    printf ("%5d [%.7f, %.7f] "
            "%.7f %+.7f %.7f\n",
            iter, a, b,
            m, m - m_expected, b - a);
  }
while (status == GSL_CONTINUE && iter < max_iter);</pre>
gsl_min_fminimizer_free (s);
```

```
return status;
Here are the results of the minimization procedure.
      $ ./a.out
          0 [0.0000000, 6.0000000] 2.0000000 -1.1415927 6.0000000
          1 [2.0000000, 6.0000000] 3.2758640 +0.1342713 4.0000000
          2 [2.0000000, 3.2831929] 3.2758640 +0.1342713 1.2831929
          3 [2.8689068, 3.2831929] 3.2758640 +0.1342713 0.4142862
          4 [2.8689068, 3.2831929] 3.2758640 +0.1342713 0.4142862
          5 [2.8689068, 3.2758640] 3.1460585 +0.0044658 0.4069572
          6 [3.1346075, 3.2758640] 3.1460585 +0.0044658 0.1412565
          7 [3.1346075, 3.1874620] 3.1460585 +0.0044658 0.0528545
         8 [3.1346075, 3.1460585] 3.1460585 +0.0044658 0.0114510
         9 [3.1346075, 3.1460585] 3.1424060 +0.0008133 0.0114510
         10 [3.1346075, 3.1424060] 3.1415885 -0.0000041 0.0077985
      Converged:
         11 [3.1415885, 3.1424060] 3.1415927 -0.0000000 0.0008175
```

34.9 References and Further Reading

Further information on Brent's algorithm is available in the following book,

Richard Brent, Algorithms for minimization without derivatives, Prentice-Hall (1973), republished by Dover in paperback (2002), ISBN 0-486-41998-3.

35 Multidimensional Root-Finding

This chapter describes functions for multidimensional root-finding (solving nonlinear systems with n equations in n unknowns). The library provides low level components for a variety of iterative solvers and convergence tests. These can be combined by the user to achieve the desired solution, with full access to the intermediate steps of the iteration. Each class of methods uses the same framework, so that you can switch between solvers at runtime without needing to recompile your program. Each instance of a solver keeps track of its own state, allowing the solvers to be used in multi-threaded programs. The solvers are based on the original Fortran library MINPACK.

The header file gsl_multiroots.h contains prototypes for the multidimensional root finding functions and related declarations.

35.1 Overview

The problem of multidimensional root finding requires the simultaneous solution of n equations, f_i , in n variables, x_i ,

$$f_i(x_1, ..., x_n) = 0$$
 for $i = 1...n$.

In general there are no bracketing methods available for n dimensional systems, and no way of knowing whether any solutions exist. All algorithms proceed from an initial guess using a variant of the Newton iteration,

$$x \to x' = x - J^{-1}f(x)$$

where x, f are vector quantities and J is the Jacobian matrix $J_{ij} = \partial f_i/\partial x_j$. Additional strategies can be used to enlarge the region of convergence. These include requiring a decrease in the norm |f| on each step proposed by Newton's method, or taking steepest-descent steps in the direction of the negative gradient of |f|.

Several root-finding algorithms are available within a single framework. The user provides a high-level driver for the algorithms, and the library provides the individual functions necessary for each of the steps. There are three main phases of the iteration. The steps are,

- initialize solver state, s, for algorithm T
- \bullet update s using the iteration T
- test s for convergence, and repeat iteration if necessary

The evaluation of the Jacobian matrix can be problematic, either because programming the derivatives is intractable or because computation of the n^2 terms of the matrix becomes too expensive. For these reasons the algorithms provided by the library are divided into two classes according to whether the derivatives are available or not.

The state for solvers with an analytic Jacobian matrix is held in a gsl_multiroot_fdfsolver struct. The updating procedure requires both the function and its derivatives to be supplied by the user.

The state for solvers which do not use an analytic Jacobian matrix is held in a gsl_multiroot_fsolver struct. The updating procedure uses only function evaluations (not derivatives). The algorithms estimate the matrix J or J^{-1} by approximate methods.

35.2 Initializing the Solver

The following functions initialize a multidimensional solver, either with or without derivatives. The solver itself depends only on the dimension of the problem and the algorithm and can be reused for different problems.

This function returns a pointer to a newly allocated instance of a solver of type T for a system of n dimensions. For example, the following code creates an instance of a hybrid solver, to solve a 3-dimensional system of equations.

If there is insufficient memory to create the solver then the function returns a null pointer and the error handler is invoked with an error code of GSL_ENOMEM.

This function returns a pointer to a newly allocated instance of a derivative solver of type T for a system of n dimensions. For example, the following code creates an instance of a Newton-Raphson solver, for a 2-dimensional system of equations.

If there is insufficient memory to create the solver then the function returns a null pointer and the error handler is invoked with an error code of GSL_ENOMEM.

```
int gsl_multiroot_fsolver_set (gsl_multiroot_fsolver * s,

gsl_multiroot_function * f, const gsl_vector * x)

int gsl_multiroot_fdfsolver_set (gsl_multiroot_fdfsolver * g

[Function]
```

These functions set, or reset, an existing solver s to use the function f or function and derivative fdf, and the initial guess x. Note that the initial position is copied from x, this argument is not modified by subsequent iterations.

```
void gsl_multiroot_fsolver_free (gsl_multiroot_fsolver * s) [Function] void gsl_multiroot_fdfsolver_free (gsl_multiroot_fdfsolver * s) [Function] These functions free all the memory associated with the solver s.
```

These functions return a pointer to the name of the solver. For example,

35.3 Providing the function to solve

You must provide n functions of n variables for the root finders to operate on. In order to allow for general parameters the functions are defined by the following data types:

gsl_multiroot_function

[Data Type]

This data type defines a general system of functions with parameters.

```
int (* f) (const gsl_vector * x, void * params, gsl_vector * f) this function should store the vector result f(x, params) in f for argument x and parameters params, returning an appropriate error code if the function cannot be computed.
```

size_t n the dimension of the system, i.e. the number of components of the vectors x and f.

void * params

a pointer to the parameters of the function.

Here is an example using Powell's test function,

$$f_1(x) = Ax_0x_1 - 1, f_2(x) = \exp(-x_0) + \exp(-x_1) - (1 + 1/A)$$

with $A = 10^4$. The following code defines a gsl_multiroot_function system F which you could pass to a solver:

```
struct powell_params { double A; };
int
powell (gsl_vector * x, void * p, gsl_vector * f) {
  struct powell_params * params
     = (struct powell_params *)p;
  const double A = (params->A);
  const double x0 = gsl_vector_get(x,0);
  const double x1 = gsl_vector_get(x,1);
  gsl\_vector\_set (f, 0, A * x0 * x1 - 1);
  gsl\_vector\_set (f, 1, (exp(-x0) + exp(-x1)
                          -(1.0 + 1.0/A));
  return GSL_SUCCESS
}
gsl_multiroot_function F;
struct powell_params params = { 10000.0 };
F.f = \&powell;
F.n = 2;
F.params = &params;
```

```
gsl_multiroot_function_fdf
```

[Data Type]

This data type defines a general system of functions with parameters and the corresponding Jacobian matrix of derivatives,

- int (* f) (const gsl_vector * x, void * params, gsl_vector * f) this function should store the vector result f(x, params) in f for argument x and parameters params, returning an appropriate error code if the function cannot be computed.
- int (* df) (const gsl_vector * x, void * params, gsl_matrix * J) this function should store the n-by-n matrix result $J_{ij} = \partial f_i(x, params)/\partial x_j$ in J for argument x and parameters params, returning an appropriate error code if the function cannot be computed.

```
int (* fdf) (const gsl_vector * x, void * params, gsl_vector * f,
gsl_matrix * J)
```

This function should set the values of the f and J as above, for arguments x and parameters params. This function provides an optimization of the separate functions for f(x) and J(x)—it is always faster to compute the function and its derivative at the same time.

size_t n the dimension of the system, i.e. the number of components of the vectors x and f.

void * params

a pointer to the parameters of the function.

The example of Powell's test function defined above can be extended to include analytic derivatives using the following code,

```
powell_df (gsl_vector * x, void * p, gsl_matrix * J)
  struct powell_params * params
     = (struct powell_params *)p;
  const double A = (params->A);
   const double x0 = gsl_vector_get(x,0);
  const double x1 = gsl_vector_get(x,1);
  gsl_matrix_set (J, 0, 0, A * x1);
  gsl_matrix_set (J, 0, 1, A * x0);
  gsl_matrix_set (J, 1, 0, -exp(-x0));
  gsl_matrix_set (J, 1, 1, -exp(-x1));
  return GSL_SUCCESS
}
powell_fdf (gsl_vector * x, void * p,
           gsl_matrix * f, gsl_matrix * J) {
  struct powell_params * params
     = (struct powell_params *)p;
  const double A = (params->A);
```

```
const double x0 = gsl_vector_get(x,0);
   const double x1 = gsl_vector_get(x,1);
   const double u0 = exp(-x0);
   const double u1 = exp(-x1);
   gsl\_vector\_set (f, 0, A * x0 * x1 - 1);
   gsl\_vector\_set (f, 1, u0 + u1 - (1 + 1/A));
   gsl_matrix_set (J, 0, 0, A * x1);
   gsl_matrix_set (J, 0, 1, A * x0);
  gsl_matrix_set (J, 1, 0, -u0);
   gsl_matrix_set (J, 1, 1, -u1);
  return GSL_SUCCESS
}
gsl_multiroot_function_fdf FDF;
FDF.f = &powell_f;
FDF.df = &powell_df;
FDF.fdf = &powell_fdf;
FDF.n = 2;
FDF.params = 0;
```

Note that the function powell_fdf is able to reuse existing terms from the function when calculating the Jacobian, thus saving time.

35.4 Iteration

The following functions drive the iteration of each algorithm. Each function performs one iteration to update the state of any solver of the corresponding type. The same functions work for all solvers so that different methods can be substituted at runtime without modifications to the code.

```
int gsl_multiroot_fsolver_iterate (gsl_multiroot_fsolver * s) [Function] int gsl_multiroot_fdfsolver_iterate (gsl_multiroot_fdfsolver * s) [Function] These functions perform a single iteration of the solver s. If the iteration encounters an unexpected problem then an error code will be returned,
```

GSL_EBADFUNC

the iteration encountered a singular point where the function or its derivative evaluated to Inf or NaN.

GSL_ENOPROG

the iteration is not making any progress, preventing the algorithm from continuing.

The solver maintains a current best estimate of the root s->x and its function value s->f at all times. This information can be accessed with the following auxiliary functions,

These functions return the current estimate of the root for the solver s, given by s->x.

These functions return the function value f(x) at the current estimate of the root for the solver s, given by s->f.

These functions return the last step dx taken by the solver s, given by s->dx.

35.5 Search Stopping Parameters

A root finding procedure should stop when one of the following conditions is true:

- A multidimensional root has been found to within the user-specified precision.
- A user-specified maximum number of iterations has been reached.
- An error has occurred.

The handling of these conditions is under user control. The functions below allow the user to test the precision of the current result in several standard ways.

This function tests for the convergence of the sequence by comparing the last step dx with the absolute error epsabs and relative error epsrel to the current position x. The test returns $GSL_SUCCESS$ if the following condition is achieved,

$$|dx_i| < epsabs + epsrel |x_i|$$

for each component of x and returns GSL_CONTINUE otherwise.

This function tests the residual value f against the absolute error bound epsabs. The test returns $GSL_SUCCESS$ if the following condition is achieved,

$$\sum_{i} |f_i| < epsabs$$

and returns $GSL_CONTINUE$ otherwise. This criterion is suitable for situations where the precise location of the root, x, is unimportant provided a value can be found where the residual is small enough.

35.6 Algorithms using Derivatives

The root finding algorithms described in this section make use of both the function and its derivative. They require an initial guess for the location of the root, but there is no absolute guarantee of convergence—the function must be suitable for this technique and the initial guess must be sufficiently close to the root for it to work. When the conditions are satisfied then convergence is quadratic.

gsl_multiroot_fdfsolver_hybridsj

[Derivative Solver]

This is a modified version of Powell's Hybrid method as implemented in the HYBRJ algorithm in MINPACK. Minpack was written by Jorge J. Moré, Burton S. Garbow and Kenneth E. Hillstrom. The Hybrid algorithm retains the fast convergence of Newton's method but will also reduce the residual when Newton's method is unreliable.

The algorithm uses a generalized trust region to keep each step under control. In order to be accepted a proposed new position x' must satisfy the condition $|D(x'-x)| < \delta$, where D is a diagonal scaling matrix and δ is the size of the trust region. The components of D are computed internally, using the column norms of the Jacobian to estimate the sensitivity of the residual to each component of x. This improves the behavior of the algorithm for badly scaled functions.

On each iteration the algorithm first determines the standard Newton step by solving the system Jdx = -f. If this step falls inside the trust region it is used as a trial step in the next stage. If not, the algorithm uses the linear combination of the Newton and gradient directions which is predicted to minimize the norm of the function while staying inside the trust region,

$$dx = -\alpha J^{-1}f(x) - \beta \nabla |f(x)|^2.$$

This combination of Newton and gradient directions is referred to as a dogleg step.

The proposed step is now tested by evaluating the function at the resulting point, x'. If the step reduces the norm of the function sufficiently then it is accepted and size of the trust region is increased. If the proposed step fails to improve the solution then the size of the trust region is decreased and another trial step is computed.

The speed of the algorithm is increased by computing the changes to the Jacobian approximately, using a rank-1 update. If two successive attempts fail to reduce the residual then the full Jacobian is recomputed. The algorithm also monitors the progress of the solution and returns an error if several steps fail to make any improvement,

GSL_ENOPROG

the iteration is not making any progress, preventing the algorithm from continuing.

GSL_ENOPROGJ

re-evaluations of the Jacobian indicate that the iteration is not making any progress, preventing the algorithm from continuing.

gsl_multiroot_fdfsolver_hybridj

[Derivative Solver]

This algorithm is an unscaled version of hybridsj. The steps are controlled by a spherical trust region $|x'-x| < \delta$, instead of a generalized region. This can be useful if the generalized region estimated by hybridsj is inappropriate.

gsl_multiroot_fdfsolver_newton

[Derivative Solver]

Newton's Method is the standard root-polishing algorithm. The algorithm begins with an initial guess for the location of the solution. On each iteration a linear approximation to the function F is used to estimate the step which will zero all the components of the residual. The iteration is defined by the following sequence,

$$x \to x' = x - J^{-1} f(x)$$

where the Jacobian matrix J is computed from the derivative functions provided by f. The step dx is obtained by solving the linear system,

$$J dx = -f(x)$$

using LU decomposition. If the Jacobian matrix is singular, an error code of ${\tt GSL_EDOM}$ is returned.

gsl_multiroot_fdfsolver_gnewton

[Derivative Solver]

This is a modified version of Newton's method which attempts to improve global convergence by requiring every step to reduce the Euclidean norm of the residual, |f(x)|. If the Newton step leads to an increase in the norm then a reduced step of relative size,

$$t = (\sqrt{1+6r} - 1)/(3r)$$

is proposed, with r being the ratio of norms $|f(x')|^2/|f(x)|^2$. This procedure is repeated until a suitable step size is found.

35.7 Algorithms without Derivatives

The algorithms described in this section do not require any derivative information to be supplied by the user. Any derivatives needed are approximated by finite differences. Note that if the finite-differencing step size chosen by these routines is inappropriate, an explicit user-supplied numerical derivative can always be used with the algorithms described in the previous section.

gsl_multiroot_fsolver_hybrids

[Solver]

This is a version of the Hybrid algorithm which replaces calls to the Jacobian function by its finite difference approximation. The finite difference approximation is computed using gsl_multiroots_fdjac with a relative step size of GSL_SQRT_DBL_EPSILON. Note that this step size will not be suitable for all problems.

gsl_multiroot_fsolver_hybrid

[Solver]

This is a finite difference version of the Hybrid algorithm without internal scaling.

gsl_multiroot_fsolver_dnewton

[Solver]

The discrete Newton algorithm is the simplest method of solving a multidimensional system. It uses the Newton iteration

$$x \to x - J^{-1}f(x)$$

where the Jacobian matrix J is approximated by taking finite differences of the function f. The approximation scheme used by this implementation is,

$$J_{ij} = (f_i(x + \delta_j) - f_i(x))/\delta_j$$

where δ_j is a step of size $\sqrt{\epsilon}|x_j|$ with ϵ being the machine precision ($\epsilon \approx 2.22 \times 10^{-16}$). The order of convergence of Newton's algorithm is quadratic, but the finite differences require n^2 function evaluations on each iteration. The algorithm may become unstable if the finite differences are not a good approximation to the true derivatives.

gsl_multiroot_fsolver_broyden

[Solver]

The Broyden algorithm is a version of the discrete Newton algorithm which attempts to avoids the expensive update of the Jacobian matrix on each iteration. The changes to the Jacobian are also approximated, using a rank-1 update,

$$J^{-1} \to J^{-1} - (J^{-1}df - dx)dx^T J^{-1}/dx^T J^{-1}df$$

where the vectors dx and df are the changes in x and f. On the first iteration the inverse Jacobian is estimated using finite differences, as in the discrete Newton algorithm.

This approximation gives a fast update but is unreliable if the changes are not small, and the estimate of the inverse Jacobian becomes worse as time passes. The algorithm has a tendency to become unstable unless it starts close to the root. The Jacobian is refreshed if this instability is detected (consult the source for details).

This algorithm is included only for demonstration purposes, and is not recommended for serious use.

35.8 Examples

The multidimensional solvers are used in a similar way to the one-dimensional root finding algorithms. This first example demonstrates the hybrids scaled-hybrid algorithm, which does not require derivatives. The program solves the Rosenbrock system of equations,

$$f_1(x,y) = a(1-x), \ f_2(x,y) = b(y-x^2)$$

with a = 1, b = 10. The solution of this system lies at (x, y) = (1, 1) in a narrow valley.

The first stage of the program is to define the system of equations,

```
const double x0 = gsl_vector_get (x, 0);
const double x1 = gsl_vector_get (x, 1);

const double y0 = a * (1 - x0);
const double y1 = b * (x1 - x0 * x0);

gsl_vector_set (f, 0, y0);
gsl_vector_set (f, 1, y1);

return GSL_SUCCESS;
}
```

The main program begins by creating the function object f, with the arguments (x,y) and parameters (a,b). The solver s is initialized to use this function, with the hybrids method.

```
int
main (void)
{
  const gsl_multiroot_fsolver_type *T;
  gsl_multiroot_fsolver *s;
  int status;
  size_t i, iter = 0;
  const size_t n = 2;
  struct rparams p = \{1.0, 10.0\};
  gsl_multiroot_function f = {&rosenbrock_f, n, &p};
  double x_{init}[2] = \{-10.0, -5.0\};
  gsl_vector *x = gsl_vector_alloc (n);
  gsl_vector_set (x, 0, x_init[0]);
  gsl_vector_set (x, 1, x_init[1]);
 T = gsl_multiroot_fsolver_hybrids;
  s = gsl_multiroot_fsolver_alloc (T, 2);
  gsl_multiroot_fsolver_set (s, &f, x);
 print_state (iter, s);
  do
    {
      iter++;
      status = gsl_multiroot_fsolver_iterate (s);
      print_state (iter, s);
```

```
if (status) /* check if solver is stuck */
    break;

status =
    gsl_multiroot_test_residual (s->f, 1e-7);
}
while (status == GSL_CONTINUE && iter < 1000);
printf ("status = %s\n", gsl_strerror (status));
gsl_multiroot_fsolver_free (s);
gsl_vector_free (x);
return 0;
}</pre>
```

Note that it is important to check the return status of each solver step, in case the algorithm becomes stuck. If an error condition is detected, indicating that the algorithm cannot proceed, then the error can be reported to the user, a new starting point chosen or a different algorithm used.

The intermediate state of the solution is displayed by the following function. The solver state contains the vector s->x which is the current position, and the vector s->f with corresponding function values.

Here are the results of running the program. The algorithm is started at (-10, -5) far from the solution. Since the solution is hidden in a narrow valley the earliest steps follow the gradient of the function downhill, in an attempt to reduce the large value of the residual. Once the root has been approximately located, on iteration 8, the Newton behavior takes over and convergence is very rapid.

```
iter = 10 \text{ x} = 1.000   0.989 \text{ f(x)} = 1.124 \text{e} - 11 - 1.080 \text{e} - 01
iter = 11 \text{ x} = 1.000   1.000 \text{ f(x)} = 0.000 \text{e} + 00   0.000 \text{e} + 00
status = success
```

Note that the algorithm does not update the location on every iteration. Some iterations are used to adjust the trust-region parameter, after trying a step which was found to be divergent, or to recompute the Jacobian, when poor convergence behavior is detected.

The next example program adds derivative information, in order to accelerate the solution. There are two derivative functions rosenbrock_df and rosenbrock_fdf. The latter computes both the function and its derivative simultaneously. This allows the optimization of any common terms. For simplicity we substitute calls to the separate f and df functions at this point in the code below.

```
int
rosenbrock_df (const gsl_vector * x, void *params,
               gsl_matrix * J)
{
  const double a = ((struct rparams *) params)->a;
  const double b = ((struct rparams *) params)->b;
  const double x0 = gsl_vector_get (x, 0);
  const double df00 = -a;
  const double df01 = 0;
  const double df10 = -2 * b * x0;
  const double df11 = b;
 gsl_matrix_set (J, 0, 0, df00);
  gsl_matrix_set (J, 0, 1, df01);
 gsl_matrix_set (J, 1, 0, df10);
  gsl_matrix_set (J, 1, 1, df11);
 return GSL_SUCCESS;
}
int
rosenbrock_fdf (const gsl_vector * x, void *params,
                gsl_vector * f, gsl_matrix * J)
{
 rosenbrock_f (x, params, f);
 rosenbrock_df (x, params, J);
 return GSL_SUCCESS;
}
```

The main program now makes calls to the corresponding fdfsolver versions of the functions,

```
int
main (void)
```

```
{
  const gsl_multiroot_fdfsolver_type *T;
  gsl_multiroot_fdfsolver *s;
  int status;
  size_t i, iter = 0;
  const size_t n = 2;
  struct rparams p = \{1.0, 10.0\};
  gsl_multiroot_function_fdf f = {&rosenbrock_f,
                                   &rosenbrock_df,
                                   &rosenbrock_fdf,
                                   n, &p};
  double x_{init}[2] = \{-10.0, -5.0\};
  gsl_vector *x = gsl_vector_alloc (n);
  gsl_vector_set (x, 0, x_init[0]);
  gsl_vector_set (x, 1, x_init[1]);
  T = gsl_multiroot_fdfsolver_gnewton;
  s = gsl_multiroot_fdfsolver_alloc (T, n);
  gsl_multiroot_fdfsolver_set (s, &f, x);
 print_state (iter, s);
  do
    {
      iter++;
      status = gsl_multiroot_fdfsolver_iterate (s);
      print_state (iter, s);
      if (status)
        break;
      status = gsl_multiroot_test_residual (s->f, 1e-7);
 while (status == GSL_CONTINUE && iter < 1000);</pre>
 printf ("status = %s\n", gsl_strerror (status));
 gsl_multiroot_fdfsolver_free (s);
  gsl_vector_free (x);
  return 0;
}
```

The addition of derivative information to the hybrids solver does not make any significant difference to its behavior, since it able to approximate the Jacobian numerically with sufficient accuracy. To illustrate the behavior of a different derivative solver we switch to gnewton. This is a traditional Newton solver with the constraint that it scales back its step if the full step would lead "uphill". Here is the output for the gnewton algorithm,

```
iter = 0 x = -10.000 -5.000 f(x) = 1.100e+01 -1.050e+03 iter = 1 x = -4.231 -65.317 f(x) = 5.231e+00 -8.321e+02 iter = 2 x = 1.000 -26.358 f(x) = -8.882e-16 -2.736e+02 iter = 3 x = 1.000 1.000 f(x) = -2.220e-16 -4.441e-15 status = success
```

The convergence is much more rapid, but takes a wide excursion out to the point (-4.23, -65.3). This could cause the algorithm to go astray in a realistic application. The hybrid algorithm follows the downhill path to the solution more reliably.

35.9 References and Further Reading

The original version of the Hybrid method is described in the following articles by Powell, M.J.D. Powell, "A Hybrid Method for Nonlinear Equations" (Chap 6, p 87–114) and "A Fortran Subroutine for Solving systems of Nonlinear Algebraic Equations" (Chap 7, p 115–161), in *Numerical Methods for Nonlinear Algebraic Equations*, P. Rabinowitz, editor. Gordon and Breach, 1970.

The following papers are also relevant to the algorithms described in this section,

- J.J. Moré, M.Y. Cosnard, "Numerical Solution of Nonlinear Equations", ACM Transactions on Mathematical Software, Vol 5, No 1, (1979), p 64–85
- C.G. Broyden, "A Class of Methods for Solving Nonlinear Simultaneous Equations", Mathematics of Computation, Vol 19 (1965), p 577–593
- J.J. Moré, B.S. Garbow, K.E. Hillstrom, "Testing Unconstrained Optimization Software", ACM Transactions on Mathematical Software, Vol 7, No 1 (1981), p 17–41

36 Multidimensional Minimization

This chapter describes routines for finding minima of arbitrary multidimensional functions. The library provides low level components for a variety of iterative minimizers and convergence tests. These can be combined by the user to achieve the desired solution, while providing full access to the intermediate steps of the algorithms. Each class of methods uses the same framework, so that you can switch between minimizers at runtime without needing to recompile your program. Each instance of a minimizer keeps track of its own state, allowing the minimizers to be used in multi-threaded programs. The minimization algorithms can be used to maximize a function by inverting its sign.

The header file gsl_multimin.h contains prototypes for the minimization functions and related declarations.

36.1 Overview

The problem of multidimensional minimization requires finding a point x such that the scalar function,

$$f(x_1,\ldots,x_n)$$

takes a value which is lower than at any neighboring point. For smooth functions the gradient $g = \nabla f$ vanishes at the minimum. In general there are no bracketing methods available for the minimization of n-dimensional functions. The algorithms proceed from an initial guess using a search algorithm which attempts to move in a downhill direction.

Algorithms making use of the gradient of the function perform a one-dimensional line minimisation along this direction until the lowest point is found to a suitable tolerance. The search direction is then updated with local information from the function and its derivatives, and the whole process repeated until the true n-dimensional minimum is found.

Algorithms which do not require the gradient of the function use different strategies. For example, the Nelder-Mead Simplex algorithm maintains n+1 trial parameter vectors as the vertices of a n-dimensional simplex. On each iteration it tries to improve the worst vertex of the simplex by geometrical transformations. The iterations are continued until the overall size of the simplex has decreased sufficiently.

Both types of algorithms use a standard framework. The user provides a high-level driver for the algorithms, and the library provides the individual functions necessary for each of the steps. There are three main phases of the iteration. The steps are,

- initialize minimizer state, s, for algorithm T
- update s using the iteration T
- test s for convergence, and repeat iteration if necessary

Each iteration step consists either of an improvement to the line-minimisation in the current direction or an update to the search direction itself. The state for the minimizers is held in a gsl_multimin_fdfminimizer struct or a gsl_multimin_fminimizer struct.

36.2 Caveats

Note that the minimization algorithms can only search for one local minimum at a time. When there are several local minima in the search area, the first minimum to be found will be returned; however it is difficult to predict which of the minima this will be. In most cases, no error will be reported if you try to find a local minimum in an area where there is more than one.

It is also important to note that the minimization algorithms find local minima; there is no way to determine whether a minimum is a global minimum of the function in question.

36.3 Initializing the Multidimensional Minimizer

The following function initializes a multidimensional minimizer. The minimizer itself depends only on the dimension of the problem and the algorithm and can be reused for different problems.

```
gsl_multimin_fdfminimizer * [Function]
    gsl_multimin_fdfminimizer_alloc (const
        gsl_multimin_fdfminimizer_type * T, size_t n)
```

gsl_multimin_fminimizer * gsl_multimin_fminimizer_alloc [Function] (const gsl_multimin_fminimizer_type * T, size_t n)

This function returns a pointer to a newly allocated instance of a minimizer of type T for an n-dimension function. If there is insufficient memory to create the minimizer then the function returns a null pointer and the error handler is invoked with an error code of GSL_ENOMEM .

int gsl_multimin_fminimizer_set (gsl_multimin_fminimizer * s, gsl_vector * x, const gsl_vector * step_size) The function gsl_multimin_fdfminimizer_set initializes the minimizer s to minimize the function fdf starting from the initial point x. The size of the first trial step is given by step_size. The accuracy of the line minimization is specified by tol. The precise meaning of this parameter depends on the method used. Typically the line minimization is considered successful if the gradient of the function g is orthogonal to the current search direction p to a relative accuracy of tol, where $p \cdot g < tol|p||g|$. A tol value of 0.1 is suitable for most purposes, since line minimization only needs to be carried out approximately. Note that setting tol to zero will force the use of "exact" line-searches, which are extremely expensive.

The function gsl_multimin_fminimizer_set initializes the minimizer s to minimize the function f, starting from the initial point x. The size of the initial trial steps is given in vector step_size. The precise meaning of this parameter depends on the method used.

```
void gsl_multimin_fdfminimizer_free (gsl_multimin_fdfminimizer * [Function]
s)
```

void gsl_multimin_fminimizer_free (gsl_multimin_fminimizer * s) [Function]
This function frees all the memory associated with the minimizer s.

36.4 Providing a function to minimize

would print something like s is a 'conjugate_pr' minimizer.

You must provide a parametric function of n variables for the minimizers to operate on. You may also need to provide a routine which calculates the gradient of the function and a third routine which calculates both the function value and the gradient together. In order to allow for general parameters the functions are defined by the following data types:

gsl_multimin_function_fdf

[Data Type]

This data type defines a general function of n variables with parameters and the corresponding gradient vector of derivatives,

double (* f) (const gsl_vector * x, void * params) this function should return the result f(x, params) for argument x and parameters params. If the function cannot be computed, an error value of GSL_NAN should be returned.

void (* df) (const gsl_vector * x, void * params, gsl_vector * g) this function should store the n-dimensional gradient $g_i = \partial f(x, params)/\partial x_i$ in the vector g for argument x and parameters params, returning an appropriate error code if the function cannot be computed.

void (* fdf) (const gsl_vector * x, void * params, double * f, gsl_vector *
g)

This function should set the values of the f and g as above, for arguments x and parameters params. This function provides an optimization of the separate functions for f(x) and g(x)—it is always faster to compute the function and its derivative at the same time.

size_t n the dimension of the system, i.e. the number of components of the vectors x.

void * params

a pointer to the parameters of the function.

gsl_multimin_function

[Data Type]

This data type defines a general function of n variables with parameters,

```
double (* f) (const gsl_vector * x, void * params)
```

this function should return the result f(x, params) for argument x and parameters params. If the function cannot be computed, an error value of GSL_NAN should be returned.

The following example function defines a simple two-dimensional paraboloid with five parameters,

```
/* Paraboloid centered on (p[0],p[1]), with
   scale factors (p[2],p[3]) and minimum p[4] */
double
my_f (const gsl_vector *v, void *params)
 double x, y;
 double *p = (double *)params;
 x = gsl_vector_get(v, 0);
 y = gsl_vector_get(v, 1);
 return p[2] * (x - p[0]) * (x - p[0]) +
           p[3] * (y - p[1]) * (y - p[1]) + p[4];
}
/* The gradient of f, df = (df/dx, df/dy). */
void
my_df (const gsl_vector *v, void *params,
      gsl_vector *df)
{
  double x, y;
  double *p = (double *)params;
 x = gsl_vector_get(v, 0);
 y = gsl_vector_get(v, 1);
 gsl_vector_set(df, 0, 2.0 * p[2] * (x - p[0]));
  gsl_vector_set(df, 1, 2.0 * p[3] * (y - p[1]));
}
/* Compute both f and df together. */
my_fdf (const gsl_vector *x, void *params,
        double *f, gsl_vector *df)
{
  *f = my_f(x, params);
 my_df(x, params, df);
}
```

The function can be initialized using the following code,

```
gsl_multimin_function_fdf my_func;

/* Paraboloid center at (1,2), scale factors (10, 20),
    minimum value 30 */
double p[5] = { 1.0, 2.0, 10.0, 20.0, 30.0 };

my_func.n = 2;  /* number of function components */
my_func.f = &my_f;
my_func.df = &my_df;
my_func.fdf = &my_fdf;
my_func.params = (void *)p;
```

36.5 Iteration

The following function drives the iteration of each algorithm. The function performs one iteration to update the state of the minimizer. The same function works for all minimizers so that different methods can be substituted at runtime without modifications to the code.

These functions perform a single iteration of the minimizer s. If the iteration encounters an unexpected problem then an error code will be returned. The error code GSL_ENOPROG signifies that the minimizer is unable to improve on its current estimate, either due to numerical difficulty or because a genuine local minimum has been reached.

The minimizer maintains a current best estimate of the minimum at all times. This information can be accessed with the following auxiliary functions,

```
gsl_vector * gsl_multimin_fdfminimizer_x (const
                                                                      [Function]
         gsl_multimin_fdfminimizer * s)
gsl_vector * gsl_multimin_fminimizer_x (const
                                                                      [Function]
         gsl_multimin_fminimizer * s)
double gsl_multimin_fdfminimizer_minimum (const
                                                                      [Function]
         gsl_multimin_fdfminimizer * s)
double gsl_multimin_fminimizer_minimum (const
                                                                      [Function]
         gsl_multimin_fminimizer * s)
gsl_vector * gsl_multimin_fdfminimizer_gradient (const
                                                                      [Function]
         gsl_multimin_fdfminimizer * s)
double gsl_multimin_fminimizer_size (const
                                                                      [Function]
         gsl_multimin_fminimizer * s)
```

These functions return the current best estimate of the location of the minimum, the value of the function at that point, its gradient, and minimizer specific characteristic size for the minimizer s.

int gsl_multimin_fdfminimizer_restart

[Function]

 $(gsl_multimin_fdfminimizer * s)$

This function resets the minimizer s to use the current point as a new starting point.

36.6 Stopping Criteria

A minimization procedure should stop when one of the following conditions is true:

- A minimum has been found to within the user-specified precision.
- A user-specified maximum number of iterations has been reached.
- An error has occurred.

The handling of these conditions is under user control. The functions below allow the user to test the precision of the current result.

This function tests the norm of the gradient g against the absolute tolerance epsabs. The gradient of a multidimensional function goes to zero at a minimum. The test returns GSL_SUCCESS if the following condition is achieved,

and returns GSL_CONTINUE otherwise. A suitable choice of epsabs can be made from the desired accuracy in the function for small variations in x. The relationship between these quantities is given by $\delta f = g \, \delta x$.

int gsl_multimin_test_size (const double size, double epsabs) [Function]

This function tests the minimizer specific characteristic size (if applicable to the used minimizer) against absolute tolerance epsabs. The test returns GSL_SUCCESS if the size is smaller than tolerance, otherwise GSL_CONTINUE is returned.

36.7 Algorithms with Derivatives

There are several minimization methods available. The best choice of algorithm depends on the problem. The algorithms described in this section use the value of the function and its gradient at each evaluation point.

gsl_multimin_fdfminimizer_conjugate_fr

[Minimizer]

This is the Fletcher-Reeves conjugate gradient algorithm. The conjugate gradient algorithm proceeds as a succession of line minimizations. The sequence of search directions is used to build up an approximation to the curvature of the function in the neighborhood of the minimum.

An initial search direction p is chosen using the gradient, and line minimization is carried out in that direction. The accuracy of the line minimization is specified by the parameter tol. The minimum along this line occurs when the function gradient g and the search direction p are orthogonal. The line minimization terminates when $p \cdot g < tol|p||g|$. The search direction is updated using the Fletcher-Reeves formula $p' = g' - \beta g$ where $\beta = -|g'|^2/|g|^2$, and the line minimization is then repeated for the new search direction.

gsl_multimin_fdfminimizer_conjugate_pr

[Minimizer]

This is the Polak-Ribiere conjugate gradient algorithm. It is similar to the Fletcher-Reeves method, differing only in the choice of the coefficient β . Both methods work well when the evaluation point is close enough to the minimum of the objective function that it is well approximated by a quadratic hypersurface.

[Minimizer]

[Minimizer]

These methods use the vector Broyden-Fletcher-Goldfarb-Shanno (BFGS) algorithm. This is a quasi-Newton method which builds up an approximation to the second derivatives of the function f using the difference between successive gradient vectors. By combining the first and second derivatives the algorithm is able to take Newton-type steps towards the function minimum, assuming quadratic behavior in that region. The bfgs2 version of this minimizer is the most efficient version available, and is a faithful implementation of the line minimization scheme described in Fletcher's Practical Methods of Optimization, Algorithms 2.6.2 and 2.6.4. It supersedes the original bfgs routine and requires substantially fewer function and gradient evaluations. The user-supplied tolerance tol corresponds to the parameter σ used by Fletcher. A value of 0.1 is recommended for typical use (larger values correspond to less accurate line

gsl_multimin_fdfminimizer_steepest_descent

searches).

[Minimizer]

The steepest descent algorithm follows the downhill gradient of the function at each step. When a downhill step is successful the step-size is increased by a factor of two. If the downhill step leads to a higher function value then the algorithm backtracks and the step size is decreased using the parameter tol. A suitable value of tol for most applications is 0.1. The steepest descent method is inefficient and is included only for demonstration purposes.

36.8 Algorithms without Derivatives

The algorithms described in this section use only the value of the function at each evaluation point.

[Minimizer]

[Minimizer]

These methods use the Simplex algorithm of Nelder and Mead. Starting from the initial vector $x = p_0$, the algorithm constructs an additional n vectors p_i using the step size vector $s = step_size$ as follows:

$$p_0 = (x_0, x_1, \dots, x_n)$$

$$p_1 = (x_0 + s_0, x_1, \dots, x_n)$$

$$p_2 = (x_0, x_1 + s_1, \dots, x_n)$$

$$\dots = \dots$$

$$p_n = (x_0, x_1, \dots, x_n + s_n)$$

These vectors form the n+1 vertices of a simplex in n dimensions. On each iteration the algorithm uses simple geometrical transformations to update the vector corresponding to the highest function value. The geometric transformations are reflection,

reflection followed by expansion, contraction and multiple contraction. Using these transformations the simplex moves through the space towards the minimum, where it contracts itself.

After each iteration, the best vertex is returned. Note, that due to the nature of the algorithm not every step improves the current best parameter vector. Usually several iterations are required.

The minimizer-specific characteristic size is calculated as the average distance from the geometrical center of the simplex to all its vertices. This size can be used as a stopping criteria, as the simplex contracts itself near the minimum. The size is returned by the function gsl_multimin_fminimizer_size.

The nmsimplex2 version of this minimiser is a new O(N) operations implementation of the earlier $O(N^2)$ operations nmsimplex minimiser. It uses the same underlying algorithm, but the simplex updates are computed more efficiently for high-dimensional problems. In addition, the size of simplex is calculated as the RMS distance of each vertex from the center rather than the mean distance, allowing a linear update of this quantity on each step. The memory usage is $O(N^2)$ for both algorithms.

gsl_multimin_fminimizer_nmsimplex2rand

[Minimizer]

This method is a variant of nmsimplex2 which initialises the simplex around the starting point x using a randomly-oriented set of basis vectors instead of the fixed coordinate axes. The final dimensions of the simplex are scaled along the coordinate axes by the vector $step_size$. The randomisation uses a simple deterministic generator so that repeated calls to $gsl_multimin_fminimizer_set$ for a given solver object will vary the orientation in a well-defined way.

36.9 Examples

This example program finds the minimum of the paraboloid function defined earlier. The location of the minimum is offset from the origin in x and y, and the function value at the minimum is non-zero. The main program is given below, it requires the example function given earlier in this chapter.

```
int
main (void)
{
 size_t iter = 0;
  int status:
  const gsl_multimin_fdfminimizer_type *T;
 gsl_multimin_fdfminimizer *s;
  /* Position of the minimum (1,2), scale factors
     10,20, height 30. */
 double par[5] = { 1.0, 2.0, 10.0, 20.0, 30.0 };
 gsl_vector *x;
 gsl_multimin_function_fdf my_func;
 my_func.n = 2;
 my_func.f = my_f;
 my_func.df = my_df;
 my_func.fdf = my_fdf;
```

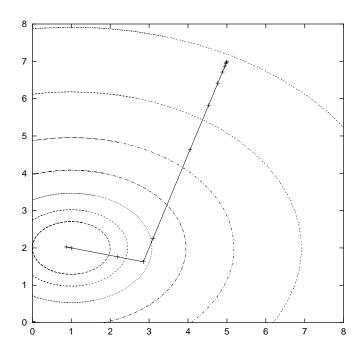
```
my_func.params = par;
/* Starting point, x = (5,7) */
x = gsl_vector_alloc (2);
gsl_vector_set (x, 0, 5.0);
gsl_vector_set (x, 1, 7.0);
T = gsl_multimin_fdfminimizer_conjugate_fr;
s = gsl_multimin_fdfminimizer_alloc (T, 2);
gsl_multimin_fdfminimizer_set (s, &my_func, x, 0.01, 1e-4);
do
    iter++;
    status = gsl_multimin_fdfminimizer_iterate (s);
    if (status)
      break;
    status = gsl_multimin_test_gradient (s->gradient, 1e-3);
    if (status == GSL_SUCCESS)
      printf ("Minimum found at:\n");
    printf ("%5d %.5f %.5f %10.5f\n", iter,
            gsl_vector_get (s->x, 0),
            gsl_vector_get (s->x, 1),
            s->f);
 }
while (status == GSL_CONTINUE && iter < 100);</pre>
gsl_multimin_fdfminimizer_free (s);
gsl_vector_free (x);
return 0;
```

The initial step-size is chosen as 0.01, a conservative estimate in this case, and the line minimization parameter is set at 0.0001. The program terminates when the norm of the gradient has been reduced below 0.001. The output of the program is shown below,

```
1 4.99629 6.99072 687.84780
2 4.98886 6.97215
                   683.55456
3 4.97400 6.93501
                   675.01278
4 4.94429 6.86073
                   658.10798
5 4.88487 6.71217
                   625.01340
6 4.76602 6.41506 561.68440
7 4.52833 5.82083 446.46694
8 4.05295 4.63238 261.79422
9 3.10219 2.25548
                   75.49762
10 2.85185 1.62963
                    67.03704
11 2.19088 1.76182
                    45.31640
12 0.86892 2.02622
                    30.18555
```

```
Minimum found at:
13 1.00000 2.00000 30.00000
```

Note that the algorithm gradually increases the step size as it successfully moves downhill, as can be seen by plotting the successive points.



The conjugate gradient algorithm finds the minimum on its second direction because the function is purely quadratic. Additional iterations would be needed for a more complicated function.

Here is another example using the Nelder-Mead Simplex algorithm to minimize the same example object function, as above.

```
int
main(void)
{
  double par[5] = \{1.0, 2.0, 10.0, 20.0, 30.0\};
  const gsl_multimin_fminimizer_type *T =
    gsl_multimin_fminimizer_nmsimplex2;
  gsl_multimin_fminimizer *s = NULL;
 gsl_vector *ss, *x;
 gsl_multimin_function minex_func;
  size_t iter = 0;
  int status;
 double size;
  /* Starting point */
 x = gsl_vector_alloc (2);
 gsl_vector_set (x, 0, 5.0);
 gsl_vector_set (x, 1, 7.0);
```

/* Set initial step sizes to 1 */

```
ss = gsl_vector_alloc (2);
       gsl_vector_set_all (ss, 1.0);
       /* Initialize method and iterate */
       minex_func.n = 2;
       minex_func.f = my_f;
       minex_func.params = par;
       s = gsl_multimin_fminimizer_alloc (T, 2);
       gsl_multimin_fminimizer_set (s, &minex_func, x, ss);
       do
         {
           iter++;
           status = gsl_multimin_fminimizer_iterate(s);
           if (status)
             break;
           size = gsl_multimin_fminimizer_size (s);
           status = gsl_multimin_test_size (size, 1e-2);
           if (status == GSL_SUCCESS)
             {
               printf ("converged to minimum at\n");
           printf ("%5d %10.3e %10.3e f() = %7.3f size = \%.3f\n",
                   gsl_vector_get (s->x, 0),
                   gsl_vector_get (s->x, 1),
                   s->fval, size);
         }
       while (status == GSL_CONTINUE && iter < 100);</pre>
       gsl_vector_free(x);
       gsl_vector_free(ss);
       gsl_multimin_fminimizer_free (s);
       return status;
The minimum search stops when the Simplex size drops to 0.01. The output is shown below.
          1 6.500e+00 5.000e+00 f() = 512.500 size = 1.130
          2 	 5.250e+00 	 4.000e+00 	 f() = 290.625 	 size = 1.409
          3 	 5.250e+00 	 4.000e+00 	 f() = 290.625 	 size = 1.409
          4 5.500e+00 1.000e+00 f() = 252.500 size = 1.409
          5 \quad 2.625e+00 \quad 3.500e+00 \text{ f()} = 101.406 \text{ size} = 1.847
          6 2.625e+00 3.500e+00 f() = 101.406 size = 1.847
          7 0.000e+00 3.000e+00 f() = 60.000 size = 1.847
          8 \quad 2.094e+00 \quad 1.875e+00 \text{ f()} = 42.275 \text{ size} = 1.321
          9 2.578e-01 1.906e+00 f() = 35.684 size = 1.069
         10 5.879e-01 2.445e+00 f() = 35.664 size = 0.841
         11 1.258e+00 2.025e+00 f() = 30.680 \text{ size} = 0.476
         12 \quad 1.258e+00 \quad 2.025e+00 \text{ f()} = 30.680 \text{ size} = 0.367
```

```
1.849e+00 f() =
        1.093e+00
                                           30.539 \text{ size} = 0.300
   13
        8.830e-01 2.004e+00 f() =
   14
                                           30.137 \text{ size} = 0.172
   15 8.830e-01 2.004e+00 f() =
                                           30.137 \text{ size} = 0.126
   16 9.582e-01 2.060e+00 f() =
                                           30.090 \text{ size} = 0.106
        1.022e+00 \quad 2.004e+00 f() =
   17
                                           30.005 \text{ size} = 0.063
        1.022e+00 \quad 2.004e+00 f() =
   18
                                           30.005 \text{ size} = 0.043
        1.022e+00 \quad 2.004e+00 f() =
                                            30.005 \text{ size} = 0.043
   19
   20 \quad 1.022e+00 \quad 2.004e+00 \text{ f()} =
                                           30.005 \text{ size} = 0.027
   21 \quad 1.022e+00 \quad 2.004e+00 f() =
                                            30.005 \text{ size} = 0.022
   9.920e-01 \quad 1.997e+00 \text{ f()} =
                                           30.001 \text{ size} = 0.016
   23 \quad 9.920e-01 \quad 1.997e+00 \text{ f()} =
                                           30.001 \text{ size} = 0.013
converged to minimum at
   24 	 9.920e-01 	 1.997e+00 	 f() = 30.001 	 size = 0.008
```

The simplex size first increases, while the simplex moves towards the minimum. After a while the size begins to decrease as the simplex contracts around the minimum.

36.10 References and Further Reading

The conjugate gradient and BFGS methods are described in detail in the following book,

R. Fletcher, $Practical\ Methods$ of $Optimization\ (Second\ Edition)$ Wiley (1987), ISBN 0471915475.

A brief description of multidimensional minimization algorithms and more recent references can be found in,

C.W. Ueberhuber, Numerical Computation (Volume 2), Chapter 14, Section 4.4 "Minimization Methods", p. 325–335, Springer (1997), ISBN 3-540-62057-5.

The simplex algorithm is described in the following paper,

J.A. Nelder and R. Mead, A simplex method for function minimization, Computer Journal vol. 7 (1965), 308–313.

37 Least-Squares Fitting

This chapter describes routines for performing least squares fits to experimental data using linear combinations of functions. The data may be weighted or unweighted, i.e. with known or unknown errors. For weighted data the functions compute the best fit parameters and their associated covariance matrix. For unweighted data the covariance matrix is estimated from the scatter of the points, giving a variance-covariance matrix.

The functions are divided into separate versions for simple one- or two-parameter regression and multiple-parameter fits. The functions are declared in the header file gsl_fit.h.

37.1 Overview

Least-squares fits are found by minimizing χ^2 (chi-squared), the weighted sum of squared residuals over n experimental datapoints (x_i, y_i) for the model Y(c, x),

$$\chi^{2} = \sum_{i} w_{i} (y_{i} - Y(c, x_{i}))^{2}$$

The p parameters of the model are $c = \{c_0, c_1, \ldots\}$. The weight factors w_i are given by $w_i = 1/\sigma_i^2$, where σ_i is the experimental error on the data-point y_i . The errors are assumed to be Gaussian and uncorrelated. For unweighted data the chi-squared sum is computed without any weight factors.

The fitting routines return the best-fit parameters c and their $p \times p$ covariance matrix. The covariance matrix measures the statistical errors on the best-fit parameters resulting from the errors on the data, σ_i , and is defined as $C_{ab} = \langle \delta c_a \delta c_b \rangle$ where $\langle \rangle$ denotes an average over the Gaussian error distributions of the underlying datapoints.

The covariance matrix is calculated by error propagation from the data errors σ_i . The change in a fitted parameter δc_a caused by a small change in the data δy_i is given by

$$\delta c_a = \sum_i \frac{\partial c_a}{\partial y_i} \delta y_i$$

allowing the covariance matrix to be written in terms of the errors on the data,

$$C_{ab} = \sum_{i,j} \frac{\partial c_a}{\partial y_i} \frac{\partial c_b}{\partial y_j} \langle \delta y_i \delta y_j \rangle$$

For uncorrelated data the fluctuations of the underlying datapoints satisfy $\langle \delta y_i \delta y_j \rangle = \sigma_i^2 \delta_{ij}$, giving a corresponding parameter covariance matrix of

$$C_{ab} = \sum_{i} \frac{1}{w_i} \frac{\partial c_a}{\partial y_i} \frac{\partial c_b}{\partial y_i}$$

When computing the covariance matrix for unweighted data, i.e. data with unknown errors, the weight factors w_i in this sum are replaced by the single estimate $w = 1/\sigma^2$, where σ^2 is the computed variance of the residuals about the best-fit model, $\sigma^2 = \sum (y_i - Y(c, x_i))^2/(n-p)$. This is referred to as the variance-covariance matrix.

The standard deviations of the best-fit parameters are given by the square root of the corresponding diagonal elements of the covariance matrix, $\sigma_{c_a} = \sqrt{C_{aa}}$. The correlation coefficient of the fit parameters c_a and c_b is given by $\rho_{ab} = C_{ab}/\sqrt{C_{aa}C_{bb}}$.

37.2 Linear regression

The functions described in this section can be used to perform least-squares fits to a straight line model, $Y(c, x) = c_0 + c_1 x$.

This function computes the best-fit linear regression coefficients (c0,c1) of the model $Y = c_0 + c_1 X$ for the dataset (x, y), two vectors of length n with strides xstride and ystride. The errors on y are assumed unknown so the variance-covariance matrix for the parameters (c0, c1) is estimated from the scatter of the points around the best-fit line and returned via the parameters (cov00, cov01, cov11). The sum of squares of the residuals from the best-fit line is returned in sumsq. Note: the correlation coefficient of the data can be computed using $gsl_stats_correlation$ (see Section 21.6 [Correlation], page 265), it does not depend on the fit.

This function computes the best-fit linear regression coefficients (c0,c1) of the model $Y = c_0 + c_1 X$ for the weighted dataset (x, y), two vectors of length n with strides xstride and ystride. The vector w, of length n and stride wstride, specifies the weight of each datapoint. The weight is the reciprocal of the variance for each datapoint in y.

The covariance matrix for the parameters (c0, c1) is computed using the weights and returned via the parameters (cov00, cov01, cov11). The weighted sum of squares of the residuals from the best-fit line, χ^2 , is returned in *chisq*.

int gsl_fit_linear_est (double x, double c0, double c1, double cov00, double cov01, double cov11, double * y, double * y_err)

This function uses the best-fit linear regression coefficients c0, c1 and their covariance cov00, cov01, cov11 to compute the fitted function y and its standard deviation y_err for the model $Y = c_0 + c_1 X$ at the point x.

37.3 Linear fitting without a constant term

The functions described in this section can be used to perform least-squares fits to a straight line model without a constant term, $Y = c_1 X$.

This function computes the best-fit linear regression coefficient c1 of the model $Y = c_1X$ for the datasets (x, y), two vectors of length n with strides xstride and ystride. The errors on y are assumed unknown so the variance of the parameter c1 is estimated from the scatter of the points around the best-fit line and returned via the parameter cov11. The sum of squares of the residuals from the best-fit line is returned in sumsq.

This function computes the best-fit linear regression coefficient c1 of the model $Y = c_1X$ for the weighted datasets (x, y), two vectors of length n with strides xstride and ystride. The vector w, of length n and stride wstride, specifies the weight of each datapoint. The weight is the reciprocal of the variance for each datapoint in y.

The variance of the parameter c1 is computed using the weights and returned via the parameter cov11. The weighted sum of squares of the residuals from the best-fit line, χ^2 , is returned in chisq.

This function uses the best-fit linear regression coefficient c1 and its covariance cov11 to compute the fitted function y and its standard deviation y_err for the model $Y = c_1 X$ at the point x.

37.4 Multi-parameter fitting

The functions described in this section perform least-squares fits to a general linear model, y = Xc where y is a vector of n observations, X is an n by p matrix of predictor variables, and the elements of the vector c are the p unknown best-fit parameters which are to be estimated. The chi-squared value is given by $\chi^2 = \sum_i w_i (y_i - \sum_j X_{ij} c_j)^2$.

This formulation can be used for fits to any number of functions and/or variables by preparing the n-by-p matrix X appropriately. For example, to fit to a p-th order polynomial in x, use the following matrix,

$$X_{ij} = x_i^j$$

where the index i runs over the observations and the index j runs from 0 to p-1.

To fit to a set of p sinusoidal functions with fixed frequencies $\omega_1, \omega_2, \ldots, \omega_n$, use,

$$X_{ij} = \sin(\omega_i x_i)$$

To fit to p independent variables x_1, x_2, \ldots, x_p , use,

$$X_{ij} = x_i(i)$$

where $x_i(i)$ is the *i*-th value of the predictor variable x_i .

The functions described in this section are declared in the header file gsl_multifit.h.

The solution of the general linear least-squares system requires an additional working space for intermediate results, such as the singular value decomposition of the matrix X.

This function allocates a workspace for fitting a model to n observations using p parameters.

This function frees the memory associated with the workspace w.

This function computes the best-fit parameters c of the model y = Xc for the observations y and the matrix of predictor variables X, using the preallocated workspace provided in work. The p-by-p variance-covariance matrix of the model parameters cov is set to $\sigma^2(X^TX)^{-1}$, where σ is the standard deviation of the fit residuals. The sum of squares of the residuals from the best-fit, χ^2 , is returned in chisq. If the coefficient of determination is desired, it can be computed from the expression $R^2 = 1 - \chi^2/TSS$, where the total sum of squares (TSS) of the observations y may be computed from gsl_stats_ts .

The best-fit is found by singular value decomposition of the matrix X using the modified Golub-Reinsch SVD algorithm, with column scaling to improve the accuracy of the singular values. Any components which have zero singular value (to machine precision) are discarded from the fit.

This function computes the best-fit parameters c of the weighted model y = Xc for the observations y with weights w and the matrix of predictor variables X, using the preallocated workspace provided in work. The p-by-p covariance matrix of the model parameters cov is computed as $(X^TWX)^{-1}$. The weighted sum of squares of the residuals from the best-fit, χ^2 , is returned in chisq. If the coefficient of determination is desired, it can be computed from the expression $R^2 = 1 - \chi^2/WTSS$, where the weighted total sum of squares (WTSS) of the observations y may be computed from gsl_stats_wtss .

In these functions components of the fit are discarded if the ratio of singular values s_i/s_0 falls below the user-specified tolerance tol, and the effective rank is returned in rank.

- int gsl_multifit_wlinear_usvd (const gsl_matrix * X, const [Function]
 gsl_vector * w, const gsl_vector * y, double tol, size_t * rank, gsl_vector * c,
 gsl_matrix * cov, double * chisq, gsl_multifit_linear_workspace * work)
 These functions compute the fit using an SVD without column scaling.

This function uses the best-fit multilinear regression coefficients c and their covariance matrix cov to compute the fitted function value y and its standard deviation y_err for the model y = x.c at the point x.

int gsl_multifit_linear_residuals (const gsl_matrix * X, const gsl_vector * y, const gsl_vector * c, gsl_vector * r) [Function]

This function computes the vector of residuals r = y - Xc for the observations y, coefficients c and matrix of predictor variables X.

37.5 Robust linear regression

Ordinary least squares (OLS) models are often heavily influenced by the presence of outliers. Outliers are data points which do not follow the general trend of the other observations, although there is strictly no precise definition of an outlier. Robust linear regression refers to regression algorithms which are robust to outliers. The most common type of robust regression is M-estimation. The general M-estimator minimizes the objective function

$$\sum_{i} \rho(e_i) = \sum_{i} \rho(y_i - Y(c, x_i))$$

where $e_i = y_i - Y(c, x_i)$ is the residual of the ith data point, and $\rho(e_i)$ is a function which should have the following properties:

$$\rho(e) \ge 0$$
 $\rho(0) = 0$
 $\rho(-e) = \rho(e)$
 $\rho(e_1) > \rho(e_2) \text{ for } |e_1| > |e_2|$

The special case of ordinary least squares is given by $\rho(e_i) = e_i^2$. Letting $\psi = \rho'$ be the derivative of ρ , differentiating the objective function with respect to the coefficients c and setting the partial derivatives to zero produces the system of equations

$$\sum_{i} \psi(e_i) X_i = 0$$

where X_i is a vector containing row i of the design matrix X. Next, we define a weight function $w(e) = \psi(e)/e$, and let $w_i = w(e_i)$:

$$\sum_{i} w_i e_i X_i = 0$$

This system of equations is equivalent to solving a weighted ordinary least squares problem, minimizing $\chi^2 = \sum_i w_i e_i^2$. The weights however, depend on the residuals e_i , which depend on the coefficients c, which depend on the weights. Therefore, an iterative solution is used, called Iteratively Reweighted Least Squares (IRLS).

- 1. Compute initial estimates of the coefficients $c^{(0)}$ using ordinary least squares
- 2. For iteration k, form the residuals $e_i^{(k)} = (y_i X_i c^{(k-1)})/(t\sigma^{(k)}\sqrt{1-h_i})$, where t is a tuning constant depending on the choice of ψ , and h_i are the statistical leverages (diagonal elements of the matrix $X(X^TX)^{-1}X^T$). Including t and h_i in the residual calculation has been shown to improve the convergence of the method. The residual

standard deviation is approximated as $\sigma^{(k)} = MAD/0.6745$, where MAD is the Median-Absolute-Deviation of the n-p largest residuals from the previous iteration.

- 3. Compute new weights $w_i^{(k)} = \psi(e_i^{(k)})/e_i^{(k)}$.
- 4. Compute new coefficients $c^{(k)}$ by solving the weighted least squares problem with weights $w_i^{(k)}$.
- 5. Steps 2 through 4 are iterated until the coefficients converge or until some maximum iteration limit is reached. Coefficients are tested for convergence using the critera:

$$|c_i^{(k)} - c_i^{(k-1)}| \le \epsilon \times \max(|c_i^{(k)}|, |c_i^{(k-1)}|)$$

for all $0 \le i < p$ where ϵ is a small tolerance factor.

The key to this method lies in selecting the function $\psi(e_i)$ to assign smaller weights to large residuals, and larger weights to smaller residuals. As the iteration proceeds, outliers are assigned smaller and smaller weights, eventually having very little or no effect on the fitted model.

This function allocates a workspace for fitting a model to n observations using p parameters. The type T specifies the function ψ and can be selected from the following choices.

gsl_multifit_robust_default

[Robust type]

This specifies the gsl_multifit_robust_bisquare type (see below) and is a good general purpose choice for robust regression.

gsl_multifit_robust_bisquare

[Robust type]

This is Tukey's biweight (bisquare) function and is a good general purpose choice for robust regression. The weight function is given by

$$w(e) = \begin{cases} (1 - e^2)^2, & |e| \le 1\\ 0, & |e| > 1 \end{cases}$$

and the default tuning constant is t = 4.685.

gsl_multifit_robust_cauchy

[Robust type]

This is Cauchy's function, also known as the Lorentzian function. This function does not guarantee a unique solution, meaning different choices of the coefficient vector c could minimize the objective function. Therefore this option should be used with care. The weight function is given by

$$w(e) = \frac{1}{1 + e^2}$$

and the default tuning constant is t = 2.385.

gsl_multifit_robust_fair

[Robust type]

This is the fair ρ function, which guarantees a unique solution and has continuous derivatives to three orders. The weight function is given by

$$w(e) = \frac{1}{1 + |e|}$$

and the default tuning constant is t = 1.400.

gsl_multifit_robust_huber

[Robust type]

This specifies Huber's ρ function, which is a parabola in the vicinity of zero and increases linearly for a given threshold |e| > t. This function is also considered an excellent general purpose robust estimator, however, occasional difficulties can be encountered due to the discontinuous first derivative of the ψ function. The weight function is given by

$$w(e) = \begin{cases} 1, & |e| \le 1\\ \frac{1}{|e|}, & |e| > 1 \end{cases}$$

and the default tuning constant is t = 1.345.

gsl_multifit_robust_ols

[Robust type]

This specifies the ordinary least squares solution, which can be useful for quickly checking the difference between the various robust and OLS solutions. The weight function is given by

$$w(e) = 1$$

and the default tuning constant is t = 1.

gsl_multifit_robust_welsch

[Robust type]

This specifies the Welsch function which can perform well in cases where the residuals have an exponential distribution. The weight function is given by

$$w(e) = \exp\left(-e^2\right)$$

and the default tuning constant is t = 2.985.

void gsl_multifit_robust_free (gsl_multifit_robust_workspace * w) [Function]
This function frees the memory associated with the workspace w.

const char * gsl_multifit_robust_name (const

[Function]

gsl_multifit_robust_workspace * w)

This function returns the name of the robust type T specified to $gsl_multifit_robust_alloc$.

int gsl_multifit_robust_tune (const double tune,

[Function]

gsl_multifit_robust_workspace * w)

This function sets the tuning constant t used to adjust the residuals at each iteration to tune. Decreasing the tuning constant increases the downweight assigned to large residuals, while increasing the tuning constant decreases the downweight assigned to large residuals.

int gsl_multifit_robust (const gsl_matrix * X, const gsl_vector * y, gsl_vector * c, gsl_matrix * cov, gsl_multifit_robust_workspace * w) [Function]

This function computes the best-fit parameters c of the model y = Xc for the observations y and the matrix of predictor variables X, attemping to reduce the influence of outliers using the algorithm outlined above. The p-by-p variance-covariance matrix of the model parameters cov is estimated as $\sigma^2(X^TX)^{-1}$, where σ is an approximation of the residual standard deviation using the theory of robust regression. Special care must be taken when estimating σ and other statistics such as R^2 , and so these are

computed internally and are available by calling the function gsl_multifit_robust_statistics.

This function uses the best-fit robust regression coefficients c and their covariance matrix cov to compute the fitted function value y and its standard deviation y_err for the model y = x.c at the point x.

gsl_multifit_robust_stats gsl_multifit_robust_statistics [Function] (const gsl_multifit_robust_workspace * w)

This function returns a structure containing relevant statistics from a robust regression. The function <code>gsl_multifit_robust</code> must be called first to perform the regression and calculate these statistics. The returned <code>gsl_multifit_robust_stats</code> structure contains the following fields.

double sigma_ols This contains the standard deviation of the residuals as computed from ordinary least squares (OLS).

double sigma_mad This contains an estimate of the standard deviation of the final residuals using the Median-Absolute-Deviation statistic

double sigma_rob This contains an estimate of the standard deviation of the final residuals from the theory of robust regression (see Street et al, 1988).

double sigma This contains an estimate of the standard deviation of the final residuals by attemping to reconcile sigma_rob and sigma_ols in a reasonable way.

double Rsq This contains the \mathbb{R}^2 coefficient of determination statistic using the estimate sigma.

double adj_Rsq This contains the adjusted R^2 coefficient of determination statistic using the estimate sigma.

double rmse This contains the root mean squared error of the final residuals double sse This contains the residual sum of squares taking into account the robust covariance matrix.

size_t dof This contains the number of degrees of freedom n-p

size_t numit Upon successful convergence, this contains the number of iterations performed

gsl_vector * weights This contains the final weight vector of length n gsl_vector * r This contains the final residual vector of length n, r = y - Xc

37.6 Troubleshooting

When using models based on polynomials, care should be taken when constructing the design matrix X. If the x values are large, then the matrix X could be ill-conditioned since its columns are powers of x, leading to unstable least-squares solutions. In this case it can often help to center and scale the x values using the mean and standard deviation:

$$x' = \frac{x - \mu(x)}{\sigma(x)}$$

and then construct the X matrix using the transformed values x'.

37.7 Examples

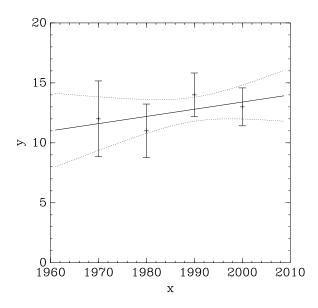
The following program computes a least squares straight-line fit to a simple dataset, and outputs the best-fit line and its associated one standard-deviation error bars.

```
#include <stdio.h>
#include <gsl/gsl_fit.h>
int
main (void)
{
  int i, n = 4;
 double x[4] = \{ 1970, 1980, 1990, 2000 \};
 double y[4] = \{ 12, 
                        11,
                              14,
                                      13 };
 double w[4] = \{ 0.1, 0.2, 0.3, 0.4 \};
 double c0, c1, cov00, cov01, cov11, chisq;
 gsl_fit_wlinear (x, 1, w, 1, y, 1, n,
                   &c0, &c1, &cov00, &cov01, &cov11,
                   &chisq);
 printf ("# best fit: Y = %g + %g X n", c0, c1);
 printf ("# covariance matrix:\n");
 printf ("# [ %g, %g\n# %g, %g]\n",
          cov00, cov01, cov01, cov11);
 printf ("# chisq = %g\n", chisq);
 for (i = 0; i < n; i++)
   printf ("data: %g %g %g\n",
                   x[i], y[i], 1/sqrt(w[i]));
 printf ("\n");
 for (i = -30; i < 130; i++)
    {
      double xf = x[0] + (i/100.0) * (x[n-1] - x[0]);
      double yf, yf_err;
      gsl_fit_linear_est (xf,
                          c0, c1,
                          cov00, cov01, cov11,
                          &yf, &yf_err);
     printf ("fit: %g %g\n", xf, yf);
     printf ("hi : %g %g\n", xf, yf + yf_err);
     printf ("lo : %g %g\n", xf, yf - yf_err);
    }
```

```
return 0;
}
```

The following commands extract the data from the output of the program and display it using the GNU plotutils graph utility,

```
$ ./demo > tmp
$ more tmp
# best fit: Y = -106.6 + 0.06 X
# covariance matrix:
# [ 39602, -19.9
# -19.9, 0.01]
# chisq = 0.8
$ for n in data fit hi lo;
    do
        grep "^$n" tmp | cut -d: -f2 > $n;
    done
$ graph -T X -X x -Y y -y 0 20 -m 0 -S 2 -Ie data
        -S 0 -I a -m 1 fit -m 2 hi -m 2 lo
```



The next program performs a quadratic fit $y = c_0 + c_1 x + c_2 x^2$ to a weighted dataset using the generalised linear fitting function gsl_multifit_wlinear. The model matrix X for a quadratic fit is given by,

$$X = \begin{pmatrix} 1 & x_0 & x_0^2 \\ 1 & x_1 & x_1^2 \\ 1 & x_2 & x_2^2 \\ \dots & \dots & \dots \end{pmatrix}$$

where the column of ones corresponds to the constant term c_0 . The two remaining columns corresponds to the terms c_1x and c_2x^2 .

The program reads n lines of data in the format (x, y, err) where err is the error (standard deviation) in the value y.

```
#include <stdio.h>
#include <gsl/gsl_multifit.h>
int
main (int argc, char **argv)
  int i, n;
 double xi, yi, ei, chisq;
  gsl_matrix *X, *cov;
  gsl_vector *y, *w, *c;
  if (argc != 2)
    {
      fprintf (stderr, "usage: fit n < data\n");</pre>
      exit (-1);
 n = atoi (argv[1]);
 X = gsl_matrix_alloc (n, 3);
  y = gsl_vector_alloc (n);
  w = gsl_vector_alloc (n);
  c = gsl_vector_alloc (3);
  cov = gsl_matrix_alloc (3, 3);
  for (i = 0; i < n; i++)
    {
      int count = fscanf (stdin, "%lg %lg %lg",
                          &xi, &yi, &ei);
      if (count != 3)
        {
          fprintf (stderr, "error reading file\n");
          exit (-1);
        }
      printf ("%g %g +/- %g\n", xi, yi, ei);
      gsl_matrix_set (X, i, 0, 1.0);
      gsl_matrix_set (X, i, 1, xi);
      gsl_matrix_set (X, i, 2, xi*xi);
      gsl_vector_set (y, i, yi);
```

```
gsl_vector_set (w, i, 1.0/(ei*ei));
   }
 {
   gsl_multifit_linear_workspace * work
     = gsl_multifit_linear_alloc (n, 3);
   gsl_multifit_wlinear (X, w, y, c, cov,
                          &chisq, work);
   gsl_multifit_linear_free (work);
 }
#define C(i) (gsl_vector_get(c,(i)))
#define COV(i,j) (gsl_matrix_get(cov,(i),(j)))
 {
   printf ("# best fit: Y = %g + %g X + %g X^2\n",
           C(0), C(1), C(2);
   printf ("# covariance matrix:\n");
   printf ("[ \%+.5e, \%+.5e, \%+.5e \n",
               COV(0,0), COV(0,1), COV(0,2);
   printf (" %+.5e, %+.5e, %+.5e \n",
               COV(1,0), COV(1,1), COV(1,2);
   printf (" %+.5e, %+.5e, %+.5e]\n",
               COV(2,0), COV(2,1), COV(2,2);
   printf ("# chisq = %g\n", chisq);
 gsl_matrix_free (X);
 gsl_vector_free (y);
 gsl_vector_free (w);
 gsl_vector_free (c);
 gsl_matrix_free (cov);
 return 0;
}
```

A suitable set of data for fitting can be generated using the following program. It outputs a set of points with gaussian errors from the curve $y = e^x$ in the region 0 < x < 2.

```
#include <stdio.h>
#include <math.h>
#include <gsl/gsl_randist.h>
int
main (void)
{
   double x;
```

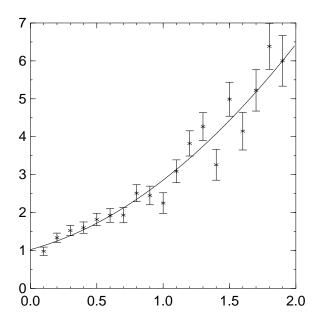
```
const gsl_rng_type * T;
       gsl_rng * r;
       gsl_rng_env_setup ();
       T = gsl_rng_default;
       r = gsl_rng_alloc (T);
       for (x = 0.1; x < 2; x+= 0.1)
         {
           double y0 = exp(x);
           double sigma = 0.1 * y0;
           double dy = gsl_ran_gaussian (r, sigma);
           printf ("%g %g %g\n", x, y0 + dy, sigma);
         }
       gsl_rng_free(r);
      return 0;
     }
The data can be prepared by running the resulting executable program,
     $ GSL_RNG_TYPE=mt19937_1999 ./generate > exp.dat
     $ more exp.dat
     0.1 0.97935 0.110517
     0.2 1.3359 0.12214
     0.3 1.52573 0.134986
     0.4 1.60318 0.149182
     0.5 1.81731 0.164872
     0.6 1.92475 0.182212
```

To fit the data use the previous program, with the number of data points given as the first argument. In this case there are 19 data points.

```
$ ./fit 19 < exp.dat
0.1 0.97935 +/- 0.110517
0.2 1.3359 +/- 0.12214
...
# best fit: Y = 1.02318 + 0.956201 X + 0.876796 X^2
# covariance matrix:
[ +1.25612e-02, -3.64387e-02, +1.94389e-02
    -3.64387e-02, +1.42339e-01, -8.48761e-02
    +1.94389e-02, -8.48761e-02, +5.60243e-02 ]
# chisq = 23.0987</pre>
```

The parameters of the quadratic fit match the coefficients of the expansion of e^x , taking into account the errors on the parameters and the $O(x^3)$ difference between the exponential and quadratic functions for the larger values of x. The errors on the parameters are given

by the square-root of the corresponding diagonal elements of the covariance matrix. The chi-squared per degree of freedom is 1.4, indicating a reasonable fit to the data.



The next program demonstrates the advantage of robust least squares on a dataset with outliers. The program generates linear (x, y) data pairs on the line y = 1.45x + 3.88, adds some random noise, and inserts 3 outliers into the dataset. Both the robust and ordinary least squares (OLS) coefficients are computed for comparison.

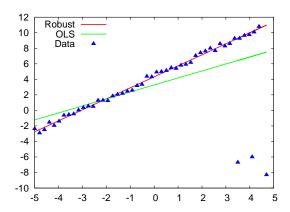
```
#include <stdio.h>
#include <gsl/gsl_multifit.h>
#include <gsl/gsl_randist.h>
int
dofit(const gsl_multifit_robust_type *T,
      const gsl_matrix *X, const gsl_vector *y,
      gsl_vector *c, gsl_matrix *cov)
{
  int s;
  gsl_multifit_robust_workspace * work
    = gsl_multifit_robust_alloc (T, X->size1, X->size2);
  s = gsl_multifit_robust (X, y, c, cov, work);
  gsl_multifit_robust_free (work);
  return s;
}
int
main (int argc, char **argv)
```

```
{
  int i;
  size_t n;
  const size_t p = 2; /* linear fit */
  gsl_matrix *X, *cov;
  gsl_vector *x, *y, *c, *c_ols;
  const double a = 1.45; /* slope */
  const double b = 3.88; /* intercept */
  gsl_rng *r;
  if (argc != 2)
    {
      fprintf (stderr, "usage: robfit n\n");
      exit (-1);
    }
 n = atoi (argv[1]);
 X = gsl_matrix_alloc (n, p);
  x = gsl_vector_alloc (n);
 y = gsl_vector_alloc (n);
  c = gsl_vector_alloc (p);
  c_ols = gsl_vector_alloc (p);
  cov = gsl_matrix_alloc (p, p);
 r = gsl_rng_alloc(gsl_rng_default);
  /* generate linear dataset */
  for (i = 0; i < n - 3; i++)
    {
      double dx = 10.0 / (n - 1.0);
      double ei = gsl_rng_uniform(r);
      double xi = -5.0 + i * dx;
      double yi = a * xi + b;
      gsl_vector_set (x, i, xi);
      gsl_vector_set (y, i, yi + ei);
    }
  /* add a few outliers */
  gsl\_vector\_set(x, n - 3, 4.7);
  gsl_vector_set(y, n - 3, -8.3);
  gsl_vector_set(x, n - 2, 3.5);
  gsl\_vector\_set(y, n - 2, -6.7);
```

```
gsl_vector_set(x, n - 1, 4.1);
  gsl_vector_set(y, n - 1, -6.0);
  /* construct design matrix X for linear fit */
  for (i = 0; i < n; ++i)
    {
      double xi = gsl_vector_get(x, i);
      gsl_matrix_set (X, i, 0, 1.0);
      gsl_matrix_set (X, i, 1, xi);
  /* perform robust and OLS fit */
 dofit(gsl_multifit_robust_ols, X, y, c_ols, cov);
  dofit(gsl_multifit_robust_bisquare, X, y, c, cov);
  /* output data and model */
  for (i = 0; i < n; ++i)
   {
      double xi = gsl_vector_get(x, i);
      double yi = gsl_vector_get(y, i);
      gsl_vector_view v = gsl_matrix_row(X, i);
      double y_ols, y_rob, y_err;
     gsl_multifit_robust_est(&v.vector, c, cov, &y_rob, &y_err);
     gsl_multifit_robust_est(&v.vector, c_ols, cov, &y_ols, &y_err);
     printf("%g %g %g %g\n", xi, yi, y_rob, y_ols);
    }
#define C(i) (gsl_vector_get(c,(i)))
#define COV(i,j) (gsl_matrix_get(cov,(i),(j)))
  {
    printf ("# best fit: Y = %g + %g X n",
           C(0), C(1));
    printf ("# covariance matrix:\n");
    printf ("# [ %+.5e, %+.5e\n",
              COV(0,0), COV(0,1));
   printf ("#
                %+.5e, %+.5e\n",
               COV(1,0), COV(1,1));
 }
  gsl_matrix_free (X);
  gsl_vector_free (x);
  gsl_vector_free (y);
```

```
gsl_vector_free (c);
gsl_vector_free (c_ols);
gsl_matrix_free (cov);
gsl_rng_free(r);
return 0;
}
```

The output from the program is shown in the following plot.



37.8 References and Further Reading

A summary of formulas and techniques for least squares fitting can be found in the "Statistics" chapter of the Annual Review of Particle Physics prepared by the Particle Data Group,

Review of Particle Properties, R.M. Barnett et al., Physical Review D54, 1 (1996) http://pdg.lbl.gov/

The Review of Particle Physics is available online at the website given above.

The tests used to prepare these routines are based on the NIST Statistical Reference Datasets. The datasets and their documentation are available from NIST at the following website,

```
http://www.nist.gov/itl/div898/strd/index.html.
```

The GSL implementation of robust linear regression closely follows the publications

DuMouchel, W. and F. O'Brien (1989), "Integrating a robust option into a multiple regression computing environment," Computer Science and Statistics: Proceedings of the 21st Symposium on the Interface, American Statistical Association

Street, J.O., R.J. Carroll, and D. Ruppert (1988), "A note on computing robust regression estimates via iteratively reweighted least squares," The American Statistician, v. 42, pp. 152-154.

38 Nonlinear Least-Squares Fitting

This chapter describes functions for multidimensional nonlinear least-squares fitting. The library provides low level components for a variety of iterative solvers and convergence tests. These can be combined by the user to achieve the desired solution, with full access to the intermediate steps of the iteration. Each class of methods uses the same framework, so that you can switch between solvers at runtime without needing to recompile your program. Each instance of a solver keeps track of its own state, allowing the solvers to be used in multi-threaded programs.

The header file gsl_multifit_nlin.h contains prototypes for the multidimensional non-linear fitting functions and related declarations.

38.1 Overview

The problem of multidimensional nonlinear least-squares fitting requires the minimization of the squared residuals of n functions, f_i , in p parameters, x_i ,

$$\Phi(x) = \frac{1}{2}||F(x)||^2 = \frac{1}{2}\sum_{i=1}^n f_i(x_1, \dots, x_p)^2$$

All algorithms proceed from an initial guess using the linearization,

$$\psi(p) = ||F(x+p)|| \approx ||F(x) + Jp||$$

where x is the initial point, p is the proposed step and J is the Jacobian matrix $J_{ij} = \partial f_i/\partial x_j$. Additional strategies are used to enlarge the region of convergence. These include requiring a decrease in the norm ||F|| on each step or using a trust region to avoid steps which fall outside the linear regime.

To perform a weighted least-squares fit of a nonlinear model Y(x,t) to data (t_i, y_i) with independent Gaussian errors σ_i , use function components of the following form,

$$f_i = \frac{(Y(x, t_i) - y_i)}{\sigma_i}$$

Note that the model parameters are denoted by x in this chapter since the non-linear least-squares algorithms are described geometrically (i.e. finding the minimum of a surface). The independent variable of any data to be fitted is denoted by t.

With the definition above the Jacobian is $J_{ij} = (1/\sigma_i)\partial Y_i/\partial x_j$, where $Y_i = Y(x, t_i)$.

38.2 Initializing the Solver

This function returns a pointer to a newly allocated instance of a solver of type T for n observations and p parameters. The number of observations n must be greater than or equal to parameters p.

If there is insufficient memory to create the solver then the function returns a null pointer and the error handler is invoked with an error code of GSL_ENOMEM.

This function returns a pointer to a newly allocated instance of a derivative solver of type T for n observations and p parameters. For example, the following code creates an instance of a Levenberg-Marquardt solver for 100 data points and 3 parameters,

The number of observations n must be greater than or equal to parameters p.

If there is insufficient memory to create the solver then the function returns a null pointer and the error handler is invoked with an error code of GSL_ENOMEM.

This function initializes, or reinitializes, an existing solver s to use the function f and the initial guess x.

This function initializes, or reinitializes, an existing solver s to use the function and derivative fdf and the initial guess x.

```
void gsl_multifit_fsolver_free (gsl_multifit_fsolver * s) [Function]
void gsl_multifit_fdfsolver_free (gsl_multifit_fdfsolver * s) [Function]
These functions free all the memory associated with the solver s.
```

These functions return a pointer to the name of the solver. For example,

would print something like s is a 'lmder' solver.

38.3 Providing the Function to be Minimized

You must provide n functions of p variables for the minimization algorithms to operate on. In order to allow for arbitrary parameters the functions are defined by the following data types:

```
gsl_multifit_function [Data Type]
```

This data type defines a general system of functions with arbitrary parameters.

```
int (* f) (const gsl_vector * x, void * params, gsl_vector * f) this function should store the vector result f(x, params) in f for argument x and arbitrary parameters params, returning an appropriate error code if the function cannot be computed.
```

- $size_t n$ the number of functions, i.e. the number of components of the vector f.
- size_t p the number of independent variables, i.e. the number of components of the vector x.

void * params

a pointer to the arbitrary parameters of the function.

gsl_multifit_function_fdf

[Data Type]

This data type defines a general system of functions with arbitrary parameters and the corresponding Jacobian matrix of derivatives,

- int (* f) (const gsl_vector * x, void * params, gsl_vector * f) this function should store the vector result f(x, params) in f for argument x and arbitrary parameters params, returning an appropriate error code if the function cannot be computed.
- int (* df) (const gsl_vector * x, void * params, gsl_matrix * J) this function should store the n-by-p matrix result $J_{ij} = \partial f_i(x, params)/\partial x_j$ in J for argument x and arbitrary parameters params, returning an appropriate error code if the function cannot be computed. If an analytic Jacobian is unavailable, or too expensive to compute, this function pointer may be set to NULL, in which case the Jacobian will be internally computed using finite difference approximations of the function f.
- int (* fdf) (const gsl_vector * x, void * params, gsl_vector * f,
 gsl_matrix * J)

This function should set the values of the f and J as above, for arguments x and arbitrary parameters params. This function provides an optimization of the separate functions for f(x) and J(x)—it is always faster to compute the function and its derivative at the same time. If an analytic Jacobian is unavailable, or too expensive to compute, this function pointer may be set to NULL, in which case the Jacobian will be internally computed using finite difference approximations of the function f.

- size_t n the number of functions, i.e. the number of components of the vector f.
- size_t p the number of independent variables, i.e. the number of components of the vector x.

void * params

a pointer to the arbitrary parameters of the function.

Note that when fitting a non-linear model against experimental data, the data is passed to the functions above using the params argument and the trial best-fit parameters through the x argument.

38.4 Finite Difference Jacobian

For the algorithms which require a Jacobian matrix of derivatives of the fit functions, there are times when an analytic Jacobian may be unavailable or too expensive to compute.

Therefore GSL supports approximating the Jacobian numerically using finite differences of the fit functions. This is typically done by setting the relevant function pointers of the gsl_multifit_function_fdf data type to NULL, however the following functions allow the user to access the approximate Jacobian directly if needed.

This function takes as input the current position x with the function values computed at the current position f, along with fdf which specifies the fit function and parameters and approximates the n-by-p Jacobian J using forward finite differences: $J_{ij} = \partial f_i(x, params)/\partial x_j \approx (f_i(x^*, params) - f_i(x, params))/\Delta x_j$. where x^* has the jth element perturbed by Δx_j and $\Delta x_j = \epsilon |x_j|$, where ϵ is the square root of the machine precision GSL_DBL_EPSILON.

This function computes the vector of function values f and the approximate Jacobian J at the position vector x using the system described in fdf. See gsl_multifit_fdfsolver_dif_df for a description of how the Jacobian is computed.

38.5 Iteration

The following functions drive the iteration of each algorithm. Each function performs one iteration to update the state of any solver of the corresponding type. The same functions work for all solvers so that different methods can be substituted at runtime without modifications to the code.

```
int gsl_multifit_fsolver_iterate (gsl_multifit_fsolver * s) [Function] int gsl_multifit_fdfsolver_iterate (gsl_multifit_fdfsolver * s) [Function]
```

These functions perform a single iteration of the solver s. If the iteration encounters an unexpected problem then an error code will be returned. The solver maintains a current estimate of the best-fit parameters at all times.

The solver struct s contains the following entries, which can be used to track the progress of the solution:

```
gsl_vector * x
```

The current position.

```
gsl_vector * f
```

The function value at the current position.

```
gsl_vector * dx
```

The difference between the current position and the previous position, i.e. the last step, taken as a vector.

```
gsl_matrix * J
```

The Jacobian matrix at the current position (for the gsl_multifit_fdfsolver struct only)

The best-fit information also can be accessed with the following auxiliary functions,

 $gsl_vector * gsl_multifit_fdfsolver_position (const gsl_multifit_fdfsolver * s)$ [F

These functions return the current position (i.e. best-fit parameters) $s\rightarrow x$ of the solver s.

38.6 Search Stopping Parameters

A minimization procedure should stop when one of the following conditions is true:

- A minimum has been found to within the user-specified precision.
- A user-specified maximum number of iterations has been reached.
- An error has occurred.

The handling of these conditions is under user control. The functions below allow the user to test the current estimate of the best-fit parameters in several standard ways.

This function tests for the convergence of the sequence by comparing the last step dx with the absolute error epsabs and relative error epsrel to the current position x. The test returns $GSL_SUCCESS$ if the following condition is achieved,

$$|dx_i| < epsabs + epsrel |x_i|$$

for each component of x and returns $GSL_CONTINUE$ otherwise.

This function tests the residual gradient g against the absolute error bound epsabs. Mathematically, the gradient should be exactly zero at the minimum. The test returns $GSL_SUCCESS$ if the following condition is achieved,

$$\sum_{i} |g_i| < epsabs$$

and returns ${\tt GSL_CONTINUE}$ otherwise. This criterion is suitable for situations where the precise location of the minimum, x, is unimportant provided a value can be found where the gradient is small enough.

This function computes the gradient g of $\Phi(x) = (1/2)||F(x)||^2$ from the Jacobian matrix J and the function values f, using the formula $g = J^T f$.

38.7 High Level Driver

These routines provide a high level wrapper that combine the iteration and convergence testing for easy use.

int gsl_multifit_fsolver_driver (gsl_multifit_fsolver * s, const size_t maxiter, const double epsabs, const double epsrel) [Function]

int gsl_multifit_fdfsolver_driver (gsl_multifit_fdfsolver * s, const size_t maxiter, const double epsabs, const double epsrel) [Function]

These functions iterate the solver s for a maximum of maxiter iterations. After each iteration, the system is tested for convergence using gsl_multifit_test_delta with the error tolerances epsabs and epsrel.

38.8 Minimization Algorithms using Derivatives

The minimization algorithms described in this section make use of both the function and its derivative. They require an initial guess for the location of the minimum. There is no absolute guarantee of convergence—the function must be suitable for this technique and the initial guess must be sufficiently close to the minimum for it to work.

gsl_multifit_fdfsolver_lmsder

[Derivative Solver]

This is a robust and efficient version of the Levenberg-Marquardt algorithm as implemented in the scaled LMDER routine in MINPACK. Minpack was written by Jorge J. Moré, Burton S. Garbow and Kenneth E. Hillstrom.

The algorithm uses a generalized trust region to keep each step under control. In order to be accepted a proposed new position x' must satisfy the condition $|D(x'-x)| < \delta$, where D is a diagonal scaling matrix and δ is the size of the trust region. The components of D are computed internally, using the column norms of the Jacobian to estimate the sensitivity of the residual to each component of x. This improves the behavior of the algorithm for badly scaled functions.

On each iteration the algorithm attempts to minimize the linear system |F + Jp| subject to the constraint $|Dp| < \Delta$. The solution to this constrained linear system is found using the Levenberg-Marquardt method.

The proposed step is now tested by evaluating the function at the resulting point, x'. If the step reduces the norm of the function sufficiently, and follows the predicted behavior of the function within the trust region, then it is accepted and the size of the trust region is increased. If the proposed step fails to improve the solution, or differs significantly from the expected behavior within the trust region, then the size of the trust region is decreased and another trial step is computed.

The algorithm also monitors the progress of the solution and returns an error if the changes in the solution are smaller than the machine precision. The possible error codes are,

GSL_ETOLF

the decrease in the function falls below machine precision

GSL_ETOLX

the change in the position vector falls below machine precision

GSL_ETOLG

the norm of the gradient, relative to the norm of the function, falls below machine precision GSL_ENOPROG

the routine has made 10 or more attempts to find a suitable trial step without success (but subsequent calls can be made to continue the search).¹

These error codes indicate that further iterations will be unlikely to change the solution from its current value.

gsl_multifit_fdfsolver_lmder

[Derivative Solver]

This is an unscaled version of the LMDER algorithm. The elements of the diagonal scaling matrix D are set to 1. This algorithm may be useful in circumstances where the scaled version of LMDER converges too slowly, or the function is already scaled appropriately.

38.9 Minimization Algorithms without Derivatives

There are no algorithms implemented in this section at the moment.

38.10 Computing the covariance matrix of best fit parameters

This function uses the Jacobian matrix J to compute the covariance matrix of the best-fit parameters, covar. The parameter epsrel is used to remove linear-dependent columns when J is rank deficient.

The covariance matrix is given by,

$$C = (J^T J)^{-1}$$

and is computed by QR decomposition of J with column-pivoting. Any columns of R which satisfy

$$|R_{kk}| \leq epsrel|R_{11}|$$

are considered linearly-dependent and are excluded from the covariance matrix (the corresponding rows and columns of the covariance matrix are set to zero).

If the minimisation uses the weighted least-squares function $f_i = (Y(x, t_i) - y_i)/\sigma_i$ then the covariance matrix above gives the statistical error on the best-fit parameters resulting from the Gaussian errors σ_i on the underlying data y_i . This can be verified from the relation $\delta f = J\delta c$ and the fact that the fluctuations in f from the data y_i are normalised by σ_i and so satisfy $\langle \delta f \delta f^T \rangle = I$.

For an unweighted least-squares function $f_i = (Y(x, t_i) - y_i)$ the covariance matrix above should be multiplied by the variance of the residuals about the best-fit $\sigma^2 = \sum (y_i - Y(x, t_i))^2/(n-p)$ to give the variance-covariance matrix $\sigma^2 C$. This estimates the statistical error on the best-fit parameters from the scatter of the underlying data. For more information about covariance matrices see Section 37.1 [Fitting Overview], page 403.

 $^{^{1}}$ The return code <code>GSL_CONTINUE</code> was used for this case in versions prior to 1.14.

38.11 Examples

The following example program fits a weighted exponential model with background to experimental data, $Y = A \exp(-\lambda t) + b$. The first part of the program sets up the functions $expb_f$ and $expb_d$ to calculate the model and its Jacobian. The appropriate fitting function is given by,

$$f_i = ((A \exp(-\lambda t_i) + b) - y_i)/\sigma_i$$

where we have chosen $t_i = i$. The Jacobian matrix J is the derivative of these functions with respect to the three parameters (A, λ, b) . It is given by,

$$J_{ij} = \frac{\partial f_i}{\partial x_j}$$

```
where x_0 = A, x_1 = \lambda and x_2 = b.
     /* expfit.c -- model functions for exponential + background */
     struct data {
       size_t n;
       double * y;
       double * sigma;
     };
     int
     expb_f (const gsl_vector * x, void *data,
             gsl_vector * f)
     {
       size_t n = ((struct data *)data)->n;
       double *y = ((struct data *)data)->y;
       double *sigma = ((struct data *) data)->sigma;
       double A = gsl_vector_get (x, 0);
       double lambda = gsl_vector_get (x, 1);
       double b = gsl_vector_get (x, 2);
       size_t i;
       for (i = 0; i < n; i++)
           /* Model Yi = A * exp(-lambda * i) + b */
           double t = i;
           double Yi = A * exp (-lambda * t) + b;
           gsl_vector_set (f, i, (Yi - y[i])/sigma[i]);
         }
       return GSL_SUCCESS;
     }
```

```
int
expb_df (const gsl_vector * x, void *data,
         gsl_matrix * J)
{
  size_t n = ((struct data *)data)->n;
  double *sigma = ((struct data *) data)->sigma;
  double A = gsl_vector_get (x, 0);
  double lambda = gsl_vector_get (x, 1);
  size_t i;
  for (i = 0; i < n; i++)
      /* Jacobian matrix J(i,j) = dfi / dxj, */
      /* where fi = (Yi - yi)/sigma[i],
              Yi = A * exp(-lambda * i) + b */
      /* and the xj are the parameters (A,lambda,b) */
      double t = i;
      double s = sigma[i];
      double e = \exp(-lambda * t);
      gsl_matrix_set (J, i, 0, e/s);
      gsl_matrix_set (J, i, 1, -t * A * e/s);
      gsl_matrix_set (J, i, 2, 1/s);
  return GSL_SUCCESS;
}
int
expb_fdf (const gsl_vector * x, void *data,
          gsl_vector * f, gsl_matrix * J)
  expb_f (x, data, f);
  expb_df (x, data, J);
 return GSL_SUCCESS;
}
```

The main part of the program sets up a Levenberg-Marquardt solver and some simulated random data. The data uses the known parameters (1.0,5.0,0.1) combined with Gaussian noise (standard deviation = 0.1) over a range of 40 timesteps. The initial guess for the parameters is chosen as (0.0, 1.0, 0.0).

```
#include <stdlib.h>
#include <stdio.h>
#include <gsl/gsl_rng.h>
#include <gsl/gsl_randist.h>
#include <gsl/gsl_vector.h>
```

```
#include <gsl/gsl_blas.h>
#include <gsl/gsl_multifit_nlin.h>
#include "expfit.c"
#define N 40
void print_state (size_t iter, gsl_multifit_fdfsolver * s);
int
main (void)
{
  const gsl_multifit_fdfsolver_type *T;
  gsl_multifit_fdfsolver *s;
  int status;
 unsigned int i, iter = 0;
  const size_t n = N;
  const size_t p = 3;
  gsl_matrix *covar = gsl_matrix_alloc (p, p);
  double y[N], sigma[N];
  struct data d = { n, y, sigma};
  gsl_multifit_function_fdf f;
  double x_{init}[3] = \{ 1.0, 0.0, 0.0 \};
  gsl_vector_view x = gsl_vector_view_array (x_init, p);
  const gsl_rng_type * type;
  gsl_rng * r;
 gsl_rng_env_setup();
 type = gsl_rng_default;
 r = gsl_rng_alloc (type);
 f.f = \&expb_f;
 f.df = &expb_df;
 f.fdf = &expb_fdf;
  f.n = n;
 f.p = p;
  f.params = &d;
  /* This is the data to be fitted */
  for (i = 0; i < n; i++)
    {
      double t = i;
      y[i] = 1.0 + 5 * exp (-0.1 * t)
                 + gsl_ran_gaussian (r, 0.1);
```

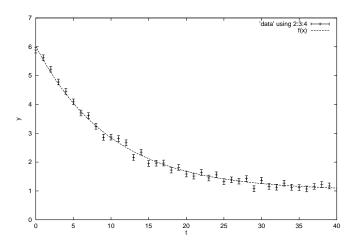
```
sigma[i] = 0.1;
     printf ("data: %u %g %g\n", i, y[i], sigma[i]);
    };
 T = gsl_multifit_fdfsolver_lmsder;
  s = gsl_multifit_fdfsolver_alloc (T, n, p);
 gsl_multifit_fdfsolver_set (s, &f, &x.vector);
 print_state (iter, s);
 do
    {
      iter++;
      status = gsl_multifit_fdfsolver_iterate (s);
     printf ("status = %s\n", gsl_strerror (status));
     print_state (iter, s);
      if (status)
       break;
      status = gsl_multifit_test_delta (s->dx, s->x,
                                         1e-4, 1e-4);
    }
 while (status == GSL_CONTINUE && iter < 500);</pre>
 gsl_multifit_covar (s->J, 0.0, covar);
#define FIT(i) gsl_vector_get(s->x, i)
#define ERR(i) sqrt(gsl_matrix_get(covar,i,i))
    double chi = gsl_blas_dnrm2(s->f);
    double dof = n - p;
    double c = GSL_MAX_DBL(1, chi / sqrt(dof));
   printf("chisq/dof = %g\n", pow(chi, 2.0) / dof);
                    = \%.5f +/- \%.5f n'', FIT(0), c*ERR(0));
   printf ("lambda = \%.5f +/- \%.5f n", FIT(1), c*ERR(1));
                   = \%.5f +/- \%.5f n'', FIT(2), c*ERR(2));
    printf ("b
  }
 printf ("status = %s\n", gsl_strerror (status));
  gsl_multifit_fdfsolver_free (s);
```

The iteration terminates when the change in x is smaller than 0.0001, as both an absolute and relative change. Here are the results of running the program:

```
iter: 0 x=1.00000000 0.00000000 0.00000000 |f(x)|=117.349
status=success
iter: 1 x=1.64659312 0.01814772 0.64659312 |f(x)|=76.4578
status=success
iter: 2 \times 2.85876037 \cdot 0.08092095 \cdot 1.44796363 \cdot |f(x)| = 37.6838
status=success
iter: 3 \times 4.94899512 \cdot 0.11942928 \cdot 1.09457665 \mid f(x) \mid = 9.58079
status=success
iter: 4 \times = 5.02175572 \ 0.10287787 \ 1.03388354 \ |f(x)| = 5.63049
status=success
iter: 5 \times 5.04520433 \cdot 0.10405523 \cdot 1.01941607 \cdot |f(x)| = 5.44398
status=success
iter: 6 x=5.04535782 0.10404906 1.01924871 |f(x)|=5.44397
chisq/dof = 0.800996
       = 5.04536 +/- 0.06028
lambda = 0.10405 +/- 0.00316
       = 1.01925 +/- 0.03782
status = success
```

The approximate values of the parameters are found correctly, and the chi-squared value indicates a good fit (the chi-squared per degree of freedom is approximately 1). In this case the errors on the parameters can be estimated from the square roots of the diagonal elements of the covariance matrix.

If the chi-squared value shows a poor fit (i.e. $\chi^2/(n-p)\gg 1$) then the error estimates obtained from the covariance matrix will be too small. In the example program the error estimates are multiplied by $\sqrt{\chi^2/(n-p)}$ in this case, a common way of increasing the errors for a poor fit. Note that a poor fit will result from the use an inappropriate model, and the scaled error estimates may then be outside the range of validity for Gaussian errors.



38.12 References and Further Reading

The MINPACK algorithm is described in the following article,

J.J. Moré, The Levenberg-Marquardt Algorithm: Implementation and Theory, Lecture Notes in Mathematics, v630 (1978), ed G. Watson.

The following paper is also relevant to the algorithms described in this section,

J.J. Moré, B.S. Garbow, K.E. Hillstrom, "Testing Unconstrained Optimization Software", ACM Transactions on Mathematical Software, Vol 7, No 1 (1981), p 17–41.

39 Basis Splines

This chapter describes functions for the computation of smoothing basis splines (B-splines). A smoothing spline differs from an interpolating spline in that the resulting curve is not required to pass through each datapoint. See Chapter 27 [Interpolation], page 330, for information about interpolating splines.

The header file gsl_bspline.h contains the prototypes for the bspline functions and related declarations.

39.1 Overview

B-splines are commonly used as basis functions to fit smoothing curves to large data sets. To do this, the abscissa axis is broken up into some number of intervals, where the endpoints of each interval are called *breakpoints*. These breakpoints are then converted to *knots* by imposing various continuity and smoothness conditions at each interface. Given a nondecreasing knot vector $t = \{t_0, t_1, \ldots, t_{n+k-1}\}$, the *n* basis splines of order *k* are defined by

$$B_{i,1}(x) = \begin{cases} 1, & t_i \le x < t_{i+1} \\ 0, & else \end{cases}$$

$$B_{i,k}(x) = \frac{(x-t_i)}{(t_{i+k-1}-t_i)} B_{i,k-1}(x) + \frac{(t_{i+k}-x)}{(t_{i+k}-t_{i+1})} B_{i+1,k-1}(x)$$

for i = 0, ..., n - 1. The common case of cubic B-splines is given by k = 4. The above recurrence relation can be evaluated in a numerically stable way by the de Boor algorithm.

If we define appropriate knots on an interval [a, b] then the B-spline basis functions form a complete set on that interval. Therefore we can expand a smoothing function as

$$f(x) = \sum_{i=0}^{n-1} c_i B_{i,k}(x)$$

given enough $(x_j, f(x_j))$ data pairs. The coefficients c_i can be readily obtained from a least-squares fit.

39.2 Initializing the B-splines solver

The computation of B-spline functions requires a preallocated workspace of type gsl_bspline_workspace. If B-spline derivatives are also required, an additional gsl_bspline_deriv_workspace is needed.

This function allocates a workspace for computing B-splines of order k. The number of breakpoints is given by nbreak. This leads to n = nbreak + k - 2 basis functions. Cubic B-splines are specified by k = 4. The size of the workspace is O(5k + nbreak).

void gsl_bspline_free (gsl_bspline_workspace * w) [Function]
This function frees the memory associated with the workspace w.

This function allocates a workspace for computing the derivatives of a B-spline basis function of order k. The size of the workspace is $O(2k^2)$.

void gsl_bspline_deriv_free (gsl_bspline_deriv_workspace * w) [Function]
This function frees the memory associated with the derivative workspace w.

39.3 Constructing the knots vector

This function computes the knots associated with the given breakpoints and stores them internally in w->knots.

This function assumes uniformly spaced breakpoints on [a, b] and constructs the corresponding knot vector using the previously specified *nbreak* parameter. The knots are stored in w->knots.

39.4 Evaluation of B-splines

This function evaluates all B-spline basis functions at the position x and stores them in the vector B, so that the i-th element is $B_i(x)$. The vector B must be of length n = nbreak + k - 2. This value may also be obtained by calling gsl_bspline_ncoeffs. Computing all the basis functions at once is more efficient than computing them individually, due to the nature of the defining recurrence relation.

This function evaluates all potentially nonzero B-spline basis functions at the position x and stores them in the vector Bk, so that the i-th element is $B_{(istart+i)}(x)$. The last element of Bk is $B_{iend}(x)$. The vector Bk must be of length k. By returning only the nonzero basis functions, this function allows quantities involving linear combinations of the $B_i(x)$ to be computed without unnecessary terms (such linear combinations occur, for example, when evaluating an interpolated function).

size_t gsl_bspline_ncoeffs (gsl_bspline_workspace * w) [Function] This function returns the number of B-spline coefficients given by n = nbreak + k - 2.

39.5 Evaluation of B-spline derivatives

int gsl_bspline_deriv_eval (const double x, const size_t nderiv, [Function]

gsl_matrix * dB, gsl_bspline_workspace * w, gsl_bspline_deriv_workspace * dw)

This function evaluates all B-spline basis function derivatives of orders 0 through

nderiv (inclusive) at the position x and stores them in the matrix dB. The (i, j)-th

element of dB is $d^jB_i(x)/dx^j$. The matrix dB must be of size n=nbreak+k-2 by nderiv+1. The value n may also be obtained by calling $gsl_bspline_ncoeffs$. Note that function evaluations are included as the zeroth order derivatives in dB. Computing all the basis function derivatives at once is more efficient than computing them individually, due to the nature of the defining recurrence relation.

```
int gsl_bspline_deriv_eval_nonzero (const double x, const size_t [Function] nderiv, gsl_matrix * dB, size_t * istart, size_t * iend, gsl_bspline_workspace * w, gsl_bspline_deriv_workspace * dw)
```

This function evaluates all potentially nonzero B-spline basis function derivatives of orders 0 through nderiv (inclusive) at the position x and stores them in the matrix dB. The (i, j)-th element of dB is $d^j B_{(istart+i)}(x)/dx^j$. The last row of dB contains $d^j B_{iend}(x)/dx^j$. The matrix dB must be of size k by at least nderiv + 1. Note that function evaluations are included as the zeroth order derivatives in dB. By returning only the nonzero basis functions, this function allows quantities involving linear combinations of the $B_i(x)$ and their derivatives to be computed without unnecessary terms.

39.6 Working with the Greville abscissae

The Greville abscissae are defined to be the mean location of k-1 consecutive knots in the knot vector for each basis spline function of order k. With the first and last knots in the $gsl_bspline_workspace$ knot vector excluded, there are $gsl_bspline_ncoeffs$ Greville abscissae for any given B-spline basis. These values are often used in B-spline collocation applications and may also be called Marsden-Schoenberg points.

Returns the location of the *i*-th Greville abscissa for the given B-spline basis. For the ill-defined case when k=1, the implementation chooses to return breakpoint interval midpoints.

39.7 Examples

The following program computes a linear least squares fit to data using cubic B-spline basis functions with uniform breakpoints. The data is generated from the curve $y(x) = \cos(x) \exp(-x/10)$ on the interval [0, 15] with Gaussian noise added.

```
#include <stdio.h>
#include <stdib.h>
#include <math.h>
#include <gsl/gsl_bspline.h>
#include <gsl/gsl_multifit.h>
#include <gsl/gsl_rng.h>
#include <gsl/gsl_randist.h>
#include <gsl/gsl_statistics.h>

/* number of data points to fit */
#define N 200
```

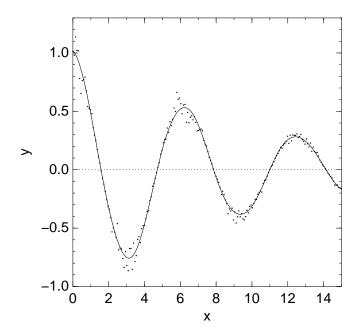
```
/* number of fit coefficients */
#define NCOEFFS 12
/* nbreak = ncoeffs + 2 - k = ncoeffs - 2 since k = 4 */
#define NBREAK (NCOEFFS - 2)
int
main (void)
{
  const size_t n = N;
  const size_t ncoeffs = NCOEFFS;
  const size_t nbreak = NBREAK;
  size_t i, j;
  gsl_bspline_workspace *bw;
  gsl_vector *B;
  double dy;
  gsl_rng *r;
  gsl_vector *c, *w;
  gsl_vector *x, *y;
  gsl_matrix *X, *cov;
  gsl_multifit_linear_workspace *mw;
  double chisq, Rsq, dof, tss;
  gsl_rng_env_setup();
  r = gsl_rng_alloc(gsl_rng_default);
  /* allocate a cubic bspline workspace (k = 4) */
  bw = gsl_bspline_alloc(4, nbreak);
  B = gsl_vector_alloc(ncoeffs);
  x = gsl_vector_alloc(n);
  y = gsl_vector_alloc(n);
  X = gsl_matrix_alloc(n, ncoeffs);
  c = gsl_vector_alloc(ncoeffs);
  w = gsl_vector_alloc(n);
  cov = gsl_matrix_alloc(ncoeffs, ncoeffs);
  mw = gsl_multifit_linear_alloc(n, ncoeffs);
  printf("#m=0,S=0\n");
  /* this is the data to be fitted */
  for (i = 0; i < n; ++i)
      double sigma;
      double xi = (15.0 / (N - 1)) * i;
      double yi = cos(xi) * exp(-0.1 * xi);
```

```
sigma = 0.1 * yi;
    dy = gsl_ran_gaussian(r, sigma);
    yi += dy;
    gsl_vector_set(x, i, xi);
    gsl_vector_set(y, i, yi);
    gsl_vector_set(w, i, 1.0 / (sigma * sigma));
    printf("%f %f\n", xi, yi);
  }
/* use uniform breakpoints on [0, 15] */
gsl_bspline_knots_uniform(0.0, 15.0, bw);
/* construct the fit matrix X */
for (i = 0; i < n; ++i)
  {
    double xi = gsl_vector_get(x, i);
    /* compute B_j(xi) for all j */
    gsl_bspline_eval(xi, B, bw);
    /* fill in row i of X */
    for (j = 0; j < ncoeffs; ++j)
      {
        double Bj = gsl_vector_get(B, j);
        gsl_matrix_set(X, i, j, Bj);
  }
/* do the fit */
gsl_multifit_wlinear(X, w, y, c, cov, &chisq, mw);
dof = n - ncoeffs;
tss = gsl_stats_wtss(w->data, 1, y->data, 1, y->size);
Rsq = 1.0 - chisq / tss;
fprintf(stderr, "chisq/dof = %e, Rsq = %f\n",
                 chisq / dof, Rsq);
/* output the smoothed curve */
  double xi, yi, yerr;
  printf("#m=1,S=0\n");
  for (xi = 0.0; xi < 15.0; xi += 0.1)
    {
```

```
gsl_bspline_eval(xi, B, bw);
        gsl_multifit_linear_est(B, c, cov, &yi, &yerr);
        printf("%f %f\n", xi, yi);
      }
  }
  gsl_rng_free(r);
  gsl_bspline_free(bw);
  gsl_vector_free(B);
  gsl_vector_free(x);
  gsl_vector_free(y);
  gsl_matrix_free(X);
  gsl_vector_free(c);
  gsl_vector_free(w);
  gsl_matrix_free(cov);
  gsl_multifit_linear_free(mw);
  return 0;
} /* main() */
```

The output can be plotted with GNU graph.

```
$ ./a.out > bspline.dat
chisq/dof = 1.118217e+00, Rsq = 0.989771
$ graph -T ps -X x -Y y -x 0 15 -y -1 1.3 < bspline.dat > bspline.ps
```



39.8 References and Further Reading

Further information on the algorithms described in this section can be found in the following book,

C. de Boor, A Practical Guide to Splines (1978), Springer-Verlag, ISBN 0-387-90356-9.

Further information of Greville abscissae and B-spline collocation can be found in the following paper,

Richard W. Johnson, Higher order B-spline collocation at the Greville abscissae. Applied Numerical Mathematics. vol. 52, 2005, 63–75.

A large collection of B-spline routines is available in the PPPACK library available at http://www.netlib.org/pppack, which is also part of SLATEC.

40 Physical Constants

This chapter describes macros for the values of physical constants, such as the speed of light, c, and gravitational constant, G. The values are available in different unit systems, including the standard MKSA system (meters, kilograms, seconds, amperes) and the CGSM system (centimeters, grams, seconds, gauss), which is commonly used in Astronomy.

The definitions of constants in the MKSA system are available in the file gsl_const_mksa.h. The constants in the CGSM system are defined in gsl_const_cgsm.h. Dimensionless constants, such as the fine structure constant, which are pure numbers are defined in gsl_const_num.h.

The full list of constants is described briefly below. Consult the header files themselves for the values of the constants used in the library.

40.1 Fundamental Constants

GSL_CONST_MKSA_SPEED_OF_LIGHT

The speed of light in vacuum, c.

GSL_CONST_MKSA_VACUUM_PERMEABILITY

The permeability of free space, μ_0 . This constant is defined in the MKSA system only.

GSL_CONST_MKSA_VACUUM_PERMITTIVITY

The permittivity of free space, ϵ_0 . This constant is defined in the MKSA system only.

GSL_CONST_MKSA_PLANCKS_CONSTANT_H Planck's constant, h.

GSL_CONST_MKSA_PLANCKS_CONSTANT_HBAR Planck's constant divided by 2π , \hbar .

GSL_CONST_NUM_AVOGADRO

Avogadro's number, N_a .

GSL_CONST_MKSA_FARADAY

The molar charge of 1 Faraday.

GSL_CONST_MKSA_BOLTZMANN

The Boltzmann constant, k.

GSL_CONST_MKSA_MOLAR_GAS

The molar gas constant, R_0 .

GSL_CONST_MKSA_STANDARD_GAS_VOLUME

The standard gas volume, V_0 .

GSL_CONST_MKSA_STEFAN_BOLTZMANN_CONSTANT

The Stefan-Boltzmann radiation constant, σ .

GSL_CONST_MKSA_GAUSS

The magnetic field of 1 Gauss.

40.2 Astronomy and Astrophysics

GSL_CONST_MKSA_ASTRONOMICAL_UNIT

The length of 1 astronomical unit (mean earth-sun distance), au.

GSL_CONST_MKSA_GRAVITATIONAL_CONSTANT

The gravitational constant, G.

GSL_CONST_MKSA_LIGHT_YEAR

The distance of 1 light-year, ly.

GSL_CONST_MKSA_PARSEC

The distance of 1 parsec, pc.

GSL_CONST_MKSA_GRAV_ACCEL

The standard gravitational acceleration on Earth, g.

GSL_CONST_MKSA_SOLAR_MASS

The mass of the Sun.

40.3 Atomic and Nuclear Physics

GSL_CONST_MKSA_ELECTRON_CHARGE

The charge of the electron, e.

GSL_CONST_MKSA_ELECTRON_VOLT

The energy of 1 electron volt, eV.

GSL_CONST_MKSA_UNIFIED_ATOMIC_MASS

The unified atomic mass, amu.

GSL_CONST_MKSA_MASS_ELECTRON

The mass of the electron, m_e .

GSL_CONST_MKSA_MASS_MUON

The mass of the muon, m_{μ} .

GSL_CONST_MKSA_MASS_PROTON

The mass of the proton, m_p .

GSL_CONST_MKSA_MASS_NEUTRON

The mass of the neutron, m_n .

GSL_CONST_NUM_FINE_STRUCTURE

The electromagnetic fine structure constant α .

GSL_CONST_MKSA_RYDBERG

The Rydberg constant, Ry, in units of energy. This is related to the Rydberg inverse wavelength R_{∞} by $Ry = hcR_{\infty}$.

GSL_CONST_MKSA_BOHR_RADIUS

The Bohr radius, a_0 .

GSL_CONST_MKSA_ANGSTROM

The length of 1 angstrom.

GSL_CONST_MKSA_BARN

The area of 1 barn.

GSL_CONST_MKSA_BOHR_MAGNETON

The Bohr Magneton, μ_B .

GSL_CONST_MKSA_NUCLEAR_MAGNETON

The Nuclear Magneton, μ_N .

GSL_CONST_MKSA_ELECTRON_MAGNETIC_MOMENT

The absolute value of the magnetic moment of the electron, μ_e . The physical magnetic moment of the electron is negative.

GSL_CONST_MKSA_PROTON_MAGNETIC_MOMENT

The magnetic moment of the proton, μ_p .

GSL_CONST_MKSA_THOMSON_CROSS_SECTION

The Thomson cross section, σ_T .

GSL_CONST_MKSA_DEBYE

The electric dipole moment of 1 Debye, D.

40.4 Measurement of Time

GSL_CONST_MKSA_MINUTE

The number of seconds in 1 minute.

GSL_CONST_MKSA_HOUR

The number of seconds in 1 hour.

GSL_CONST_MKSA_DAY

The number of seconds in 1 day.

GSL_CONST_MKSA_WEEK

The number of seconds in 1 week.

40.5 Imperial Units

GSL_CONST_MKSA_INCH

The length of 1 inch.

GSL_CONST_MKSA_FOOT

The length of 1 foot.

GSL_CONST_MKSA_YARD

The length of 1 yard.

GSL_CONST_MKSA_MILE

The length of 1 mile.

GSL_CONST_MKSA_MIL

The length of 1 mil (1/1000th of an inch).

40.6 Speed and Nautical Units

- $\begin{tabular}{ll} {\tt GSL_CONST_MKSA_KILOMETERS_PER_HOUR}\\ {\tt The speed of 1 kilometer per hour.} \end{tabular}$
- $\begin{tabular}{ll} {\tt GSL_CONST_MKSA_MILES_PER_HOUR} \\ {\tt The speed of 1 mile per hour.} \end{tabular}$
- GSL_CONST_MKSA_NAUTICAL_MILE

 The length of 1 nautical mile.
- GSL_CONST_MKSA_FATHOM

 The length of 1 fathom.
- $\begin{array}{c} \mathtt{GSL_CONST_MKSA_KNOT} \\ & \text{The speed of 1 knot.} \end{array}$

40.7 Printers Units

- GSL_CONST_MKSA_POINT

 The length of 1 printer's point (1/72 inch).
- GSL_CONST_MKSA_TEXPOINT

 The length of 1 TeX point (1/72.27 inch).

40.8 Volume, Area and Length

- $\begin{tabular}{ll} {\tt GSL_CONST_MKSA_MICRON} \\ {\tt The length of 1 micron.} \\ \end{tabular}$
- GSL_CONST_MKSA_HECTARE

 The area of 1 hectare.
- $\begin{array}{c} {\tt GSL_CONST_MKSA_ACRE} \\ {\tt The~area~of~1~acre.} \end{array}$
- $\begin{array}{c} {\tt GSL_CONST_MKSA_LITER} \\ {\tt The\ volume\ of\ 1\ liter}. \end{array}$
- GSL_CONST_MKSA_US_GALLON

 The volume of 1 US gallon.
- $\begin{array}{c} {\tt GSL_CONST_MKSA_CANADIAN_GALLON} \\ {\tt The\ volume\ of\ 1\ Canadian\ gallon.} \end{array}$
- $\begin{array}{c} {\tt GSL_CONST_MKSA_UK_GALLON} \\ {\tt The\ volume\ of\ 1\ UK\ gallon.} \end{array}$
- $\begin{tabular}{ll} {\tt GSL_CONST_MKSA_QUART}\\ & {\tt The\ volume\ of\ 1\ quart.} \end{tabular}$
- GSL_CONST_MKSA_PINT

 The volume of 1 pint.

40.9 Mass and Weight

 $\begin{array}{c} \mathtt{GSL_CONST_MKSA_POUND_MASS} \\ \\ \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \\ \phantom{\mathsf{MSS}} \phantom{\mathsf$

 $\begin{array}{c} \mathtt{GSL_CONST_MKSA_OUNCE_MASS} \\ \\ \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \\ \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \phantom{\mathsf{MSS}} \\ \phantom{\mathsf{MSS}} \phantom$

 $\begin{array}{c} \mathtt{GSL_CONST_MKSA_TON} \\ & \quad \text{The mass of 1 ton.} \end{array}$

 $\begin{tabular}{ll} {\tt GSL_CONST_MKSA_METRIC_TON} \\ {\tt The mass of 1 metric ton (1000 kg)}. \end{tabular}$

GSL_CONST_MKSA_UK_TON

The mass of 1 UK ton.

 $\begin{array}{c} {\tt GSL_CONST_MKSA_TROY_OUNCE} \\ {\tt The \ mass \ of \ 1 \ troy \ ounce.} \end{array}$

 $\begin{array}{c} \mathtt{GSL_CONST_MKSA_CARAT} \\ \text{The mass of 1 carat.} \end{array}$

 $\begin{tabular}{ll} {\tt GSL_CONST_MKSA_GRAM_FORCE} \\ {\tt The force of 1 gram weight.} \end{tabular}$

GSL_CONST_MKSA_POUND_FORCE

The force of 1 pound weight.

GSL_CONST_MKSA_KILOPOUND_FORCE

The force of 1 kilopound weight.

 $\begin{array}{c} {\tt GSL_CONST_MKSA_POUNDAL} \\ {\tt The force of 1 poundal.} \end{array}$

40.10 Thermal Energy and Power

 $\begin{tabular}{ll} {\tt GSL_CONST_MKSA_CALORIE} \\ {\tt The~energy~of~1~calorie.} \end{tabular}$

GSL_CONST_MKSA_BTU

The energy of 1 British Thermal Unit, btu.

GSL_CONST_MKSA_THERM

The energy of 1 Therm.

 $\begin{tabular}{ll} {\tt GSL_CONST_MKSA_HORSEPOWER} \\ {\tt The~power~of~1~horsepower}. \end{tabular}$

40.11 Pressure

GSL_CONST_MKSA_BAR

The pressure of 1 bar.

GSL_CONST_MKSA_STD_ATMOSPHERE

The pressure of 1 standard atmosphere.

GSL_CONST_MKSA_TORR

The pressure of 1 torr.

GSL_CONST_MKSA_METER_OF_MERCURY

The pressure of 1 meter of mercury.

GSL_CONST_MKSA_INCH_OF_MERCURY

The pressure of 1 inch of mercury.

GSL_CONST_MKSA_INCH_OF_WATER

The pressure of 1 inch of water.

GSL_CONST_MKSA_PSI

The pressure of 1 pound per square inch.

40.12 Viscosity

GSL_CONST_MKSA_POISE

The dynamic viscosity of 1 poise.

GSL_CONST_MKSA_STOKES

The kinematic viscosity of 1 stokes.

40.13 Light and Illumination

GSL_CONST_MKSA_STILB

The luminance of 1 stilb.

GSL_CONST_MKSA_LUMEN

The luminous flux of 1 lumen.

GSL_CONST_MKSA_LUX

The illuminance of 1 lux.

GSL_CONST_MKSA_PHOT

The illuminance of 1 phot.

GSL_CONST_MKSA_FOOTCANDLE

The illuminance of 1 footcandle.

GSL_CONST_MKSA_LAMBERT

The luminance of 1 lambert.

GSL_CONST_MKSA_FOOTLAMBERT

The luminance of 1 footlambert.

40.14 Radioactivity

GSL_CONST_MKSA_CURIE

The activity of 1 curie.

GSL_CONST_MKSA_ROENTGEN

The exposure of 1 roentgen.

GSL_CONST_MKSA_RAD

The absorbed dose of 1 rad.

40.15 Force and Energy

GSL_CONST_MKSA_NEWTON

The SI unit of force, 1 Newton.

GSL_CONST_MKSA_DYNE

The force of 1 Dyne = 10^{-5} Newton.

GSL_CONST_MKSA_JOULE

The SI unit of energy, 1 Joule.

GSL_CONST_MKSA_ERG

The energy 1 erg = 10^{-7} Joule.

40.16 Prefixes

These constants are dimensionless scaling factors.

GSL_CONST_NUM_YOTTA

 10^{24}

GSL_CONST_NUM_ZETTA

 10^{21}

GSL_CONST_NUM_EXA

 10^{18}

GSL_CONST_NUM_PETA

 10^{15}

GSL_CONST_NUM_TERA

 10^{12}

GSL_CONST_NUM_GIGA

 10^{9}

GSL_CONST_NUM_MEGA

 10^{6}

GSL_CONST_NUM_KILO

 10^{3}

GSL_CONST_NUM_MILLI

 10^{-3}

GSL_CONST_NUM_MICRO

 10^{-6}

GSL_CONST_NUM_NANO

 10^{-9}

GSL_CONST_NUM_PICO

 10^{-12}

GSL_CONST_NUM_FEMTO

 10^{-15}

```
\begin{array}{c} {\rm GSL\_CONST\_NUM\_ATTO} \\ 10^{-18} \\ \\ {\rm GSL\_CONST\_NUM\_ZEPTO} \\ 10^{-21} \\ \\ {\rm GSL\_CONST\_NUM\_YOCTO} \\ 10^{-24} \end{array}
```

40.17 Examples

#include <stdio.h>

The following program demonstrates the use of the physical constants in a calculation. In this case, the goal is to calculate the range of light-travel times from Earth to Mars.

The required data is the average distance of each planet from the Sun in astronomical units (the eccentricities and inclinations of the orbits will be neglected for the purposes of this calculation). The average radius of the orbit of Mars is 1.52 astronomical units, and for the orbit of Earth it is 1 astronomical unit (by definition). These values are combined with the MKSA values of the constants for the speed of light and the length of an astronomical unit to produce a result for the shortest and longest light-travel times in seconds. The figures are converted into minutes before being displayed.

```
#include <gsl/gsl_const_mksa.h>
     int
     main (void)
       double c = GSL_CONST_MKSA_SPEED_OF_LIGHT;
       double au = GSL_CONST_MKSA_ASTRONOMICAL_UNIT;
       double minutes = GSL_CONST_MKSA_MINUTE;
       /* distance stored in meters */
       double r_{earth} = 1.00 * au;
       double r_mars = 1.52 * au;
       double t_min, t_max;
       t_min = (r_mars - r_earth) / c;
       t_max = (r_mars + r_earth) / c;
       printf ("light travel time from Earth to Mars:\n");
       printf ("minimum = %.1f minutes\n", t_min / minutes);
       printf ("maximum = %.1f minutes\n", t_max / minutes);
       return 0;
     }
Here is the output from the program,
     light travel time from Earth to Mars:
     minimum = 4.3 minutes
```

maximum = 21.0 minutes

40.18 References and Further Reading

The authoritative sources for physical constants are the 2006 CODATA recommended values, published in the article below. Further information on the values of physical constants is also available from the NIST website.

P.J. Mohr, B.N. Taylor, D.B. Newell, "CODATA Recommended Values of the Fundamental Physical Constants: 2006", Reviews of Modern Physics, 80(2), pp. 633–730 (2008).

http://www.physics.nist.gov/cuu/Constants/index.html

http://physics.nist.gov/Pubs/SP811/appenB9.html

41 IEEE floating-point arithmetic

This chapter describes functions for examining the representation of floating point numbers and controlling the floating point environment of your program. The functions described in this chapter are declared in the header file gsl_ieee_utils.h.

41.1 Representation of floating point numbers

The IEEE Standard for Binary Floating-Point Arithmetic defines binary formats for single and double precision numbers. Each number is composed of three parts: a sign bit (s), an exponent (E) and a fraction (f). The numerical value of the combination (s, E, f) is given by the following formula,

$$(-1)^s (1 \cdot fffff \dots) 2^E$$

The sign bit is either zero or one. The exponent ranges from a minimum value E_{min} to a maximum value E_{max} depending on the precision. The exponent is converted to an unsigned number e, known as the biased exponent, for storage by adding a bias parameter, e = E + bias. The sequence fffff... represents the digits of the binary fraction f. The binary digits are stored in normalized form, by adjusting the exponent to give a leading digit of 1. Since the leading digit is always 1 for normalized numbers it is assumed implicitly and does not have to be stored. Numbers smaller than $2^{E_{min}}$ are be stored in denormalized form with a leading zero,

$$(-1)^s(0 \cdot fffff \dots)2^{E_{min}}$$

This allows gradual underflow down to $2^{E_{min}-p}$ for p bits of precision. A zero is encoded with the special exponent of $2^{E_{min}-1}$ and infinities with the exponent of $2^{E_{max}+1}$.

The format for single precision numbers uses 32 bits divided in the following way,

seeeeeeffffffffffffffffffffff

following forms,

```
s = sign bit, 1 bit
e = exponent, 8 bits (E_min=-126, E_max=127, bias=127)
f = fraction, 23 bits
```

The format for double precision numbers uses 64 bits divided in the following way,


```
s = sign bit, 1 bit
e = exponent, 11 bits (E_min=-1022, E_max=1023, bias=1023)
f = fraction, 52 bits
```

It is often useful to be able to investigate the behavior of a calculation at the bit-level and the library provides functions for printing the IEEE representations in a human-readable form.

```
void gsl_ieee_fprintf_float (FILE * stream, const float * x) [Function] void gsl_ieee_fprintf_double (FILE * stream, const double * x) [Function] These functions output a formatted version of the IEEE floating-point number pointed to by x to the stream stream. A pointer is used to pass the number indirectly, to avoid any undesired promotion from float to double. The output takes one of the
```

```
NaN the Not-a-Number symbol

Inf, -Inf positive or negative infinity

1.ffffff...*2^E, -1.ffffff...*2^E
a normalized floating point number

0.fffff...*2^E, -0.ffffff...*2^E
a denormalized floating point number

0, -0 positive or negative zero
```

The output can be used directly in GNU Emacs Calc mode by preceding it with 2# to indicate binary.

The following program demonstrates the use of the functions by printing the single and double precision representations of the fraction 1/3. For comparison the representation of the value promoted from single to double precision is also printed.

```
#include <stdio.h>
#include <gsl/gsl_ieee_utils.h>

int
main (void)
{
   float f = 1.0/3.0;
   double d = 1.0/3.0;

   double fd = f; /* promote from float to double */
   printf (" f="); gsl_ieee_printf_float(&f);
   printf ("\n");

   printf ("fd="); gsl_ieee_printf_double(&fd);
   printf ("\n");

   printf (" d="); gsl_ieee_printf_double(&d);
   printf ("\n");

   return 0;
}
```

The binary representation of 1/3 is 0.01010101... The output below shows that the IEEE format normalizes this fraction to give a leading digit of 1,

The output also shows that a single-precision number is promoted to double-precision by adding zeros in the binary representation.

41.2 Setting up your IEEE environment

The IEEE standard defines several *modes* for controlling the behavior of floating point operations. These modes specify the important properties of computer arithmetic: the direction used for rounding (e.g. whether numbers should be rounded up, down or to the nearest number), the rounding precision and how the program should handle arithmetic exceptions, such as division by zero.

Many of these features can now be controlled via standard functions such as fpsetround, which should be used whenever they are available. Unfortunately in the past there has been no universal API for controlling their behavior—each system has had its own low-level way of accessing them. To help you write portable programs GSL allows you to specify modes in a platform-independent way using the environment variable GSL_IEEE_MODE. The library then takes care of all the necessary machine-specific initializations for you when you call the function gsl_ieee_env_setup.

```
void gsl_ieee_env_setup ()
```

[Function]

This function reads the environment variable GSL_IEEE_MODE and attempts to set up the corresponding specified IEEE modes. The environment variable should be a list of keywords, separated by commas, like this,

```
GSL_IEEE_MODE = "keyword,keyword,..."
```

where keyword is one of the following mode-names,

```
single-precision
double-precision
extended-precision
round-to-nearest
round-down
round-up
round-to-zero
mask-all
mask-invalid
mask-denormalized
mask-division-by-zero
mask-overflow
mask-underflow
trap-inexact
trap-common
```

If GSL_IEEE_MODE is empty or undefined then the function returns immediately and no attempt is made to change the system's IEEE mode. When the modes from GSL_IEEE_MODE are turned on the function prints a short message showing the new settings to remind you that the results of the program will be affected.

If the requested modes are not supported by the platform being used then the function calls the error handler and returns an error code of GSL_EUNSUP.

When options are specified using this method, the resulting mode is based on a default setting of the highest available precision (double precision or extended precision, depending on the platform) in round-to-nearest mode, with all exceptions enabled apart from the INEXACT exception. The INEXACT exception is generated whenever rounding occurs, so it must generally be disabled in typical scientific calculations. All other floating-point exceptions are enabled by default, including underflows and the use of denormalized numbers, for safety. They can be disabled with the individual mask-settings or together using mask-all.

The following adjusted combination of modes is convenient for many purposes,

This choice ignores any errors relating to small numbers (either denormalized, or underflowing to zero) but traps overflows, division by zero and invalid operations.

Note that on the x86 series of processors this function sets both the original x87 mode and the newer MXCSR mode, which controls SSE floating-point operations. The SSE floating-point units do not have a precision-control bit, and always work in double-precision. The single-precision and extended-precision keywords have no effect in this case.

To demonstrate the effects of different rounding modes consider the following program which computes e, the base of natural logarithms, by summing a rapidly-decreasing series,

```
e = 1 + \frac{1}{2!} + \frac{1}{3!} + \frac{1}{4!} + \dots = 2.71828182846\dots
#include <stdio.h>
#include <gsl/gsl_math.h>
#include <gsl/gsl_ieee_utils.h>

int
main (void)
{
    double x = 1, oldsum = 0, sum = 0;
    int i = 0;

    gsl_ieee_env_setup (); /* read GSL_IEEE_MODE */

    do
     {
        i++;
        oldsum = sum;
        sum += x;
        x = x / i;
```

Here are the results of running the program in round-to-nearest mode. This is the IEEE default so it isn't really necessary to specify it here,

After nineteen terms the sum converges to within 4×10^{-16} of the correct value. If we now change the rounding mode to round-down the final result is less accurate,

The result is about 4×10^{-15} below the correct value, an order of magnitude worse than the result obtained in the round-to-nearest mode.

If we change to rounding mode to round-up then the final result is higher than the correct value (when we add each term to the sum the final result is always rounded up, which increases the sum by at least one tick until the added term underflows to zero). To avoid this problem we would need to use a safer converge criterion, such as while (fabs(sum - oldsum) > epsilon), with a suitably chosen value of epsilon.

Finally we can see the effect of computing the sum using single-precision rounding, in the default round-to-nearest mode. In this case the program thinks it is still using double precision numbers but the CPU rounds the result of each floating point operation to singleprecision accuracy. This simulates the effect of writing the program using single-precision float variables instead of double variables. The iteration stops after about half the number of iterations and the final result is much less accurate,

```
$ GSL_IEEE_MODE="single-precision" ./a.out
....
i=12 sum=2.718281984329223633 error=1.5587e-07
```

with an error of $O(10^{-7})$, which corresponds to single precision accuracy (about 1 part in 10^{7}). Continuing the iterations further does not decrease the error because all the subsequent results are rounded to the same value.

41.3 References and Further Reading

The reference for the IEEE standard is,

ANSI/IEEE Std 754-1985, IEEE Standard for Binary Floating-Point Arithmetic.

A more pedagogical introduction to the standard can be found in the following paper,

David Goldberg: What Every Computer Scientist Should Know About Floating-Point Arithmetic. ACM Computing Surveys, Vol. 23, No. 1 (March 1991), pages 5–48.

Corrigendum: ACM Computing Surveys, Vol. 23, No. 3 (September 1991), page 413. and see also the sections by B. A. Wichmann and Charles B. Dunham in Surveyor's Forum: "What Every Computer Scientist Should Know About Floating-Point Arithmetic". ACM Computing Surveys, Vol. 24, No. 3 (September 1992), page 319.

A detailed textbook on IEEE arithmetic and its practical use is available from SIAM Press,

Michael L. Overton, Numerical Computing with IEEE Floating Point Arithmetic, SIAM Press, ISBN 0898715717.

Appendix A Debugging Numerical Programs

This chapter describes some tips and tricks for debugging numerical programs which use GSL.

A.1 Using gdb

Any errors reported by the library are passed to the function gsl_error. By running your programs under gdb and setting a breakpoint in this function you can automatically catch any library errors. You can add a breakpoint for every session by putting

```
break gsl_error
```

into your .gdbinit file in the directory where your program is started.

If the breakpoint catches an error then you can use a backtrace (bt) to see the call-tree, and the arguments which possibly caused the error. By moving up into the calling function you can investigate the values of variables at that point. Here is an example from the program fft/test_trap, which contains the following line,

```
status = gsl_fft_complex_wavetable_alloc (0, &complex_wavetable);
```

The function gsl_fft_complex_wavetable_alloc takes the length of an FFT as its first argument. When this line is executed an error will be generated because the length of an FFT is not allowed to be zero.

To debug this problem we start gdb, using the file .gdbinit to define a breakpoint in gsl_error,

```
$ gdb test_trap
```

```
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```

```
Breakpoint 1 at 0x8050b1e: file error.c, line 14.
```

When we run the program this breakpoint catches the error and shows the reason for it.

```
(gdb) rum
Starting program: test_trap
Breakpoint 1, gsl_error (reason=0x8052b0d
    "length n must be positive integer",
    file=0x8052b04 "c_init.c", line=108, gsl_errno=1)
    at error.c:14
14    if (gsl_error_handler)
```

The first argument of gsl_error is always a string describing the error. Now we can look at the backtrace to see what caused the problem,

```
(gdb) bt
#0  gsl_error (reason=0x8052b0d
    "length n must be positive integer",
    file=0x8052b04 "c_init.c", line=108, gsl_errno=1)
    at error.c:14
#1  0x8049376 in gsl_fft_complex_wavetable_alloc (n=0,
    wavetable=0xbffff778) at c_init.c:108
#2  0x8048a00 in main (argc=1, argv=0xbffff9bc)
```

```
at test_trap.c:94
#3 0x80488be in ___crt_dummy__ ()
```

We can see that the error was generated in the function $gsl_fft_complex_wavetable_alloc when it was called with an argument of <math>n=0$. The original call came from line 94 in the file $test_tap.c$.

By moving up to the level of the original call we can find the line that caused the error,

Thus we have found the line that caused the problem. From this point we could also print out the values of other variables such as complex_wavetable.

A.2 Examining floating point registers

The contents of floating point registers can be examined using the command info float (on supported platforms).

```
(gdb) info float
   st0: 0xc4018b895aa17a945000 Valid Normal -7.838871e+308
    st1: 0x3ff9ea3f50e4d7275000 Valid Normal 0.0285946
    st2: 0x3fe790c64ce27dad4800 Valid Normal 6.7415931e-08
    st3: 0x3ffaa3ef0df6607d7800 Spec Normal 0.0400229
    st4: 0x3c02800000000000000 Valid Normal 4.4501477e-308
    st5: 0x3ffef5412c22219d9000 Zero Normal 0.9580257
    st6: 0x3fff80000000000000 Valid Normal 1
    fctrl: 0x0272 53 bit; NEAR; mask DENOR UNDER LOS;
  fstat: Oxb9ba flags 0001; top 7; excep DENOR OVERF UNDER LOS
   ftag: 0x3fff
    fip: 0x08048b5c
    fcs: 0x051a0023
 fopoff: 0x08086820
 fopsel: 0x002b
```

Individual registers can be examined using the variables \$reg, where reg is the register name.

```
(gdb) p $st1
$1 = 0.02859464454261210347719
```

A.3 Handling floating point exceptions

It is possible to stop the program whenever a SIGFPE floating point exception occurs. This can be useful for finding the cause of an unexpected infinity or NaN. The current handler settings can be shown with the command info signal SIGFPE.

```
(gdb) info signal SIGFPE
Signal Stop Print Pass to program Description
SIGFPE Yes Yes Yes Arithmetic exception
```

Unless the program uses a signal handler the default setting should be changed so that SIGFPE is not passed to the program, as this would cause it to exit. The command handle SIGFPE stop nopass prevents this.

```
(gdb) handle SIGFPE stop nopass
Signal Stop Print Pass to program Description
SIGFPE Yes Yes No Arithmetic exception
```

Depending on the platform it may be necessary to instruct the kernel to generate signals for floating point exceptions. For programs using GSL this can be achieved using the GSL_IEEE_MODE environment variable in conjunction with the function gsl_ieee_env_setup as described in see Chapter 41 [IEEE floating-point arithmetic], page 449.

(gdb) set env GSL_IEEE_MODE=double-precision

A.4 GCC warning options for numerical programs

Writing reliable numerical programs in C requires great care. The following GCC warning options are recommended when compiling numerical programs:

```
gcc -ansi -pedantic -Werror -Wall -W
  -Wmissing-prototypes -Wstrict-prototypes
  -Wconversion -Wshadow -Wpointer-arith
  -Wcast-qual -Wcast-align
  -Wwrite-strings -Wnested-externs
  -fshort-enums -fno-common -Dinline= -g -02
```

For details of each option consult the manual *Using and Porting GCC*. The following table gives a brief explanation of what types of errors these options catch.

-ansi -pedantic

Use ANSI C, and reject any non-ANSI extensions. These flags help in writing portable programs that will compile on other systems.

- -Werror Consider warnings to be errors, so that compilation stops. This prevents warnings from scrolling off the top of the screen and being lost. You won't be able to compile the program until it is completely warning-free.
- -Wall This turns on a set of warnings for common programming problems. You need -Wall, but it is not enough on its own.
- -02 Turn on optimization. The warnings for uninitialized variables in -Wall rely on the optimizer to analyze the code. If there is no optimization then these warnings aren't generated.
- -W This turns on some extra warnings not included in -Wall, such as missing return values and comparisons between signed and unsigned integers.

-Wmissing-prototypes -Wstrict-prototypes

Warn if there are any missing or inconsistent prototypes. Without prototypes it is harder to detect problems with incorrect arguments.

-Wconversion

The main use of this option is to warn about conversions from signed to unsigned integers. For example, unsigned int x = -1. If you need to perform such a conversion you can use an explicit cast.

-Wshadow This warns whenever a local variable shadows another local variable. If two variables have the same name then it is a potential source of confusion.

-Wpointer-arith -Wcast-qual -Wcast-align

These options warn if you try to do pointer arithmetic for types which don't have a size, such as void, if you remove a const cast from a pointer, or if you cast a pointer to a type which has a different size, causing an invalid alignment.

-Wwrite-strings

This option gives string constants a **const** qualifier so that it will be a compiletime error to attempt to overwrite them.

-fshort-enums

This option makes the type of enum as short as possible. Normally this makes an enum different from an int. Consequently any attempts to assign a pointer-to-int to a pointer-to-enum will generate a cast-alignment warning.

-fno-common

This option prevents global variables being simultaneously defined in different object files (you get an error at link time). Such a variable should be defined in one file and referred to in other files with an extern declaration.

-Wnested-externs

This warns if an extern declaration is encountered within a function.

-Dinline=

The inline keyword is not part of ANSI C. Thus if you want to use -ansi with a program which uses inline functions you can use this preprocessor definition to remove the inline keywords.

-g It always makes sense to put debugging symbols in the executable so that you can debug it using gdb. The only effect of debugging symbols is to increase the size of the file, and you can use the strip command to remove them later if necessary.

A.5 References and Further Reading

The following books are essential reading for anyone writing and debugging numerical programs with GCC and GDB.

R.M. Stallman, *Using and Porting GNU CC*, Free Software Foundation, ISBN 1882114388

R.M. Stallman, R.H. Pesch, Debugging with GDB: The GNU Source-Level Debugger, Free Software Foundation, ISBN 1882114779

For a tutorial introduction to the GNU C Compiler and related programs, see

B.J. Gough, An Introduction to GCC, Network Theory Ltd, ISBN 0954161793

Appendix B Contributors to GSL

(See the AUTHORS file in the distribution for up-to-date information.)

Mark Galassi

Conceived GSL (with James Theiler) and wrote the design document. Wrote the simulated annealing package and the relevant chapter in the manual.

James Theiler

Conceived GSL (with Mark Galassi). Wrote the random number generators and the relevant chapter in this manual.

Jim Davies

Wrote the statistical routines and the relevant chapter in this manual.

Brian Gough

FFTs, numerical integration, random number generators and distributions, root finding, minimization and fitting, polynomial solvers, complex numbers, physical constants, permutations, vector and matrix functions, histograms, statistics, ieee-utils, revised CBLAS Level 2 & 3, matrix decompositions, eigensystems, cumulative distribution functions, testing, documentation and releases.

Reid Priedhorsky

Wrote and documented the initial version of the root finding routines while at Los Alamos National Laboratory, Mathematical Modeling and Analysis Group.

Gerard Jungman

Special Functions, Series acceleration, ODEs, BLAS, Linear Algebra, Eigensystems, Hankel Transforms.

Mike Booth

Wrote the Monte Carlo library.

Jorma Olavi Tähtinen

Wrote the initial complex arithmetic functions.

Thomas Walter

Wrote the initial heapsort routines and Cholesky decomposition.

Fabrice Rossi

Multidimensional minimization.

Carlo Perassi

Implementation of the random number generators in Knuth's Seminumerical Algorithms, 3rd Ed.

Szymon Jaroszewicz

Wrote the routines for generating combinations.

Nicolas Darnis

Wrote the cyclic functions and the initial functions for canonical permutations.

Jason H. Stover

Wrote the major cumulative distribution functions.

Ivo Alxneit

Wrote the routines for wavelet transforms.

Tuomo Keskitalo

Improved the implementation of the ODE solvers and wrote the ode-initval2 routines.

Lowell Johnson

Implementation of the Mathieu functions.

Patrick Alken

Implementation of nonsymmetric and generalized eigensystems, B-splines, and robust linear regression.

Rhys Ulerich

Wrote the multiset routines.

Pavel Holoborodko

Wrote the fixed order Gauss-Legendre quadrature routines.

Pedro Gonnet

Wrote the CQUAD integration routines.

Thanks to Nigel Lowry for help in proofreading the manual.

The non-symmetric eigensystems routines contain code based on the LAPACK linear algebra library. LAPACK is distributed under the following license:

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Appendix C Autoconf Macros

For applications using autoconf the standard macro AC_CHECK_LIB can be used to link with GSL automatically from a configure script. The library itself depends on the presence of a CBLAS and math library as well, so these must also be located before linking with the main libgsl file. The following commands should be placed in the configure.ac file to perform these tests,

```
AC_CHECK_LIB([m],[cos])
AC_CHECK_LIB([gslcblas],[cblas_dgemm])
AC_CHECK_LIB([gsl],[gsl_blas_dgemm])
```

It is important to check for libm and libgslcblas before libgsl, otherwise the tests will fail. Assuming the libraries are found the output during the configure stage looks like this,

```
checking for cos in -lm... yes checking for cblas_dgemm in -lgslcblas... yes checking for gsl_blas_dgemm in -lgsl... yes
```

If the library is found then the tests will define the macros HAVE_LIBGSL, HAVE_LIBGSLCBLAS, HAVE_LIBM and add the options -lgsl-lgslcblas-lm to the variable LIBS.

The tests above will find any version of the library. They are suitable for general use, where the versions of the functions are not important. An alternative macro is available in the file gsl.m4 to test for a specific version of the library. To use this macro simply add the following line to your configure.in file instead of the tests above:

The argument GSL_VERSION should be the two or three digit MAJOR.MINOR or MAJOR.MINOR.MICRO version number of the release you require. A suitable choice for action-if-not-found is,

```
AC_MSG_ERROR(could not find required version of GSL)
```

Then you can add the variables GSL_LIBS and GSL_CFLAGS to your Makefile.am files to obtain the correct compiler flags. GSL_LIBS is equal to the output of the gsl-config -- libs command and GSL_CFLAGS is equal to gsl-config --cflags command. For example,

```
libfoo_la_LDFLAGS = -lfoo $(GSL_LIBS) -lgslcblas
```

Note that the macro AX_PATH_GSL needs to use the C compiler so it should appear in the configure.in file before the macro AC_LANG_CPLUSPLUS for programs that use C++.

To test for inline the following test should be placed in your configure.in file,

```
AC_C_INLINE

if test "$ac_cv_c_inline" != no; then
   AC_DEFINE(HAVE_INLINE,1)
   AC_SUBST(HAVE_INLINE)
fi
```

and the macro will then be defined in the compilation flags or by including the file config.h before any library headers.

```
The following autoconf test will check for extern inline,

dnl Check for "extern inline", using a modified version
dnl of the test for AC_C_INLINE from acspecific.mt
dnl

AC_CACHE_CHECK([for extern inline], ac_cv_c_extern_inline,
[ac_cv_c_extern_inline=no
AC_TRY_COMPILE([extern $ac_cv_c_inline double foo(double x);
extern $ac_cv_c_inline double foo(double x) { return x+1.0; };
double foo (double x) { return x + 1.0; };
[ foo(1.0) ],
[ac_cv_c_extern_inline="yes"])
])

if test "$ac_cv_c_extern_inline" != no; then
    AC_DEFINE(HAVE_INLINE,1)
    AC_SUBST(HAVE_INLINE)
fi
```

The substitution of portability functions can be made automatically if you use autoconf. For example, to test whether the BSD function hypot is available you can include the following line in the configure file configure.in for your application,

```
AC_CHECK_FUNCS(hypot)

and place the following macro definitions in the file config.h.in,

/* Substitute gsl_hypot for missing system hypot */

#ifndef HAVE_HYPOT

#define hypot gsl_hypot

#endif
```

The application source files can then use the include command #include <config.h> to substitute gsl_hypot for each occurrence of hypot when hypot is not available.

Appendix D GSL CBLAS Library

The prototypes for the low-level CBLAS functions are declared in the file gsl_cblas.h. For the definition of the functions consult the documentation available from Netlib (see Section 13.3 [BLAS References and Further Reading], page 132).

D.1 Level 1

<pre>float cblas_sdsdot (const int N, const float alpha, const float * x,</pre>	[Function]
<pre>double cblas_dsdot (const int N, const float * x, const int incx, const</pre>	[Function]
<pre>float cblas_sdot (const int N, const float * x, const int incx, const</pre>	[Function]
<pre>double cblas_ddot (const int N, const double * x, const int incx, const</pre>	[Function]
<pre>void cblas_cdotu_sub (const int N, const void * x, const int incx,</pre>	[Function]
<pre>void cblas_cdotc_sub (const int N, const void * x, const int incx,</pre>	[Function]
<pre>void cblas_zdotu_sub (const int N, const void * x, const int incx,</pre>	[Function]
<pre>void cblas_zdotc_sub (const int N, const void * x, const int incx,</pre>	[Function]
float cblas_snrm2 (const int N, const float * x, const int incx)	[Function]
float cblas_snrm2 (const int N, const float * x, const int incx) float cblas_sasum (const int N, const float * x, const int incx)	[Function]
•	. ,
float cblas_sasum (const int N, const float * x, const int incx)	[Function]
float cblas_sasum (const int N, const float * x, const int incx) double cblas_dnrm2 (const int N, const double * x, const int incx)	[Function]
float cblas_sasum (const int N, const float * x, const int incx) double cblas_dnrm2 (const int N, const double * x, const int incx) double cblas_dasum (const int N, const double * x, const int incx)	[Function] [Function]
float cblas_sasum (const int N, const float * x, const int incx) double cblas_dnrm2 (const int N, const double * x, const int incx) double cblas_dasum (const int N, const double * x, const int incx) float cblas_scnrm2 (const int N, const void * x, const int incx)	[Function] [Function] [Function]
float cblas_sasum (const int N, const float * x, const int incx) double cblas_dnrm2 (const int N, const double * x, const int incx) double cblas_dasum (const int N, const double * x, const int incx) float cblas_scnrm2 (const int N, const void * x, const int incx) float cblas_scasum (const int N, const void * x, const int incx)	[Function] [Function] [Function] [Function]
float cblas_sasum (const int N, const float * x, const int incx) double cblas_dnrm2 (const int N, const double * x, const int incx) double cblas_dasum (const int N, const double * x, const int incx) float cblas_scnrm2 (const int N, const void * x, const int incx) float cblas_scasum (const int N, const void * x, const int incx) double cblas_dznrm2 (const int N, const void * x, const int incx)	[Function] [Function] [Function] [Function] [Function]
float cblas_sasum (const int N, const float * x, const int incx) double cblas_dnrm2 (const int N, const double * x, const int incx) double cblas_dasum (const int N, const double * x, const int incx) float cblas_scnrm2 (const int N, const void * x, const int incx) float cblas_scasum (const int N, const void * x, const int incx) double cblas_dznrm2 (const int N, const void * x, const int incx) double cblas_dzasum (const int N, const void * x, const int incx)	[Function] [Function] [Function] [Function] [Function] [Function]
float cblas_sasum (const int N, const float * x, const int incx) double cblas_dnrm2 (const int N, const double * x, const int incx) double cblas_dasum (const int N, const double * x, const int incx) float cblas_scnrm2 (const int N, const void * x, const int incx) float cblas_scasum (const int N, const void * x, const int incx) double cblas_dznrm2 (const int N, const void * x, const int incx) double cblas_dzasum (const int N, const void * x, const int incx) CBLAS_INDEX cblas_isamax (const int N, const float * x, const int incx) CBLAS_INDEX cblas_idamax (const int N, const double * x, const int	[Function] [Function] [Function] [Function] [Function] [Function] [Function]

[Function]

[Function]

void	<pre>cblas_sswap (const int N, float * x, const int incx, float * y, const int incy)</pre>	[Function]
void	<pre>cblas_scopy (const int N, const float * x, const int incx, float * y,</pre>	[Function]
void	<pre>cblas_saxpy (const int N, const float alpha, const float * x, const int incx, float * y, const int incy)</pre>	[Function]
void	<pre>cblas_dswap (const int N, double * x, const int incx, double * y, const int incy)</pre>	[Function]
void	<pre>cblas_dcopy (const int N, const double * x, const int incx, double * y, const int incy)</pre>	[Function]
void	<pre>cblas_daxpy (const int N, const double alpha, const double * x, const int incx, double * y, const int incy)</pre>	[Function]
void	<pre>cblas_cswap (const int N, void * x, const int incx, void * y, const int incy)</pre>	[Function]
void	<pre>cblas_ccopy (const int N, const void * x, const int incx, void * y, const int incy)</pre>	[Function]
void	<pre>cblas_caxpy (const int N, const void * alpha, const void * x, const int incx, void * y, const int incy)</pre>	[Function]
void	<pre>cblas_zswap (const int N, void * x, const int incx, void * y, const int incy)</pre>	[Function]
void	<pre>cblas_zcopy (const int N, const void * x, const int incx, void * y, const int incy)</pre>	[Function]
void	<pre>cblas_zaxpy (const int N, const void * alpha, const void * x, const int incx, void * y, const int incy)</pre>	[Function]
void	cblas_srotg (float * a, float * b, float * c, float * s)	[Function]
void	<pre>cblas_srotmg (float * d1, float * d2, float * b1, const float b2, float * P)</pre>	[Function]
void	<pre>cblas_srot (const int N, float * x, const int incx, float * y, const int incy, const float c, const float s)</pre>	[Function]
void	<pre>cblas_srotm (const int N, float * x, const int incx, float * y, const int incy, const float * P)</pre>	[Function]
void	cblas_drotg (double * a, double * b, double * c, double * s)	[Function]
void	<pre>cblas_drotmg (double * d1, double * d2, double * b1, const double b2, double * P)</pre>	[Function]
void	<pre>cblas_drot (const int N, double * x, const int incx, double * y, const int incy, const double c, const double s)</pre>	[Function]

void cblas_drotm (const int N, double * x, const int incx, double * y,

 $\verb"void cblas_sscal" (const int N, const float \verb"alpha", float "x", const int")$

const int incy, const double * P)

incx)

- void cblas_dscal (const int N, const double alpha, double * x, const
 int incx)
 [Function]

D.2 Level 2

- void cblas_sgbmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_TRANSPOSE TransA, const int M, const int N, const int KL, const int KU, const float alpha, const float * A, const int lda, const float * x, const int incx, const float beta, float * y, const int incy)
- void cblas_stbmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const int K, const float * A, const int lda,

 float * x, const int incx)
- void cblas_stpmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const float * Ap, float * x, const int incx)
- void cblas_strsv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const float * A, const int lda, float * x, const int incx)
- void cblas_stpsv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum CBLAS_DIAG Diag, const int N, const float * Ap, float * x, const int incx)

- void cblas_dgemv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_TRANSPOSE TransA, const int M, const int N, const double alpha,

 const double * A, const int lda, const double * x, const int incx, const double

 beta, double * y, const int incy)
- void cblas_dgbmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_TRANSPOSE TransA, const int M, const int N, const int KL, const int KU, const double alpha, const double * A, const int lda, const double * x,

 const int incx, const double beta, double * y, const int incy)
- void cblas_dtbmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const int K, const double * A, const int lda,

 double * x, const int incx)
- void cblas_dtpmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const double * Ap, double * x, const int incx)
- void cblas_dtrsv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const double * A, const int lda, double * x,

 const int incx)
- void cblas_dtpsv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum CBLAS_DIAG Diag, const int N, const double * Ap, double * x, const int incx)
- void cblas_cgbmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_TRANSPOSE TransA, const int M, const int N, const int KL, const int KU, const void * alpha, const void * A, const int lda, const void * x, const int incx, const void * beta, void * y, const int incy)

 [Function]

- void cblas_ctbmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const int K, const void * A, const int lda,

 void * x, const int incx)
- void cblas_ctpmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const void * Ap, void * x, const int incx)
- void cblas_ctbsv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const int K, const void * A, const int lda,

 void * x, const int incx)
- void cblas_ctpsv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum CBLAS_DIAG Diag, const int N, const void * Ap, void * x, const int incx)

- void cblas_ztrmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const void * A, const int lda, void * x, const int incx)
- void cblas_ztbmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const int K, const void * A, const int lda,

 void * x, const int incx)
- void cblas_ztpmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const void * Ap, void * x, const int incx)

- void cblas_ztrsv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum CBLAS_DIAG Diag, const int N, const void * A, const int lda, void * x, const int incx)
- void cblas_ztbsv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const int K, const void * A, const int lda,

 void * x, const int incx)
- void cblas_ztpsv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE TransA, const enum

 CBLAS_DIAG Diag, const int N, const void * Ap, void * x, const int incx)
- void cblas_ssymv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const int N, const float alpha, const float * A, const int lda, const float * x, const int incx, const float beta, float * y, const int incy)

- void cblas_ssyr2 (const enum CBLAS_ORDER order, const enum CBLAS_UPLO Uplo, const int N, const float alpha, const float * x, const int incx, const float * y, const int incy, float * A, const int lda) [Function]
- void cblas_sspr2 (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const int N, const float alpha, const float * x, const int incx, const float * y, const int incy, float * A)

- void cblas_dspmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const int N, const double alpha, const double * Ap, const double * x, const int incx, const double beta, double * y, const int incy)

- void cblas_dsyr2 (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const int N, const double alpha, const double * x, const int incx, const double * y, const int incy, double * A, const int lda)
- void cblas_dspr2 (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const int N, const double alpha, const double * x, const int incx, const double * y, const int incy, double * A)

- void cblas_chpmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const int N, const void * alpha, const void * Ap, const void * x, const int incx, const void * beta, void * y, const int incy)

- void cblas_cher2 (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const int N, const void * alpha, const void * x, const int incx, const void * y, const int incy, void * A, const int lda)

- void cblas_zhpmv (const enum CBLAS_ORDER order, const enum [Function]

 CBLAS_UPLO Uplo, const int N, const void * alpha, const void * Ap, const void * x, const int incx, const void * beta, void * y, const int incy)

- void cblas_zher2 (const enum CBLAS_ORDER order, const enum [Function] CBLAS_UPLO Uplo, const int N, const void * alpha, const void * x, const int incx, const void * y, const int incy, void * A, const int lda)

D.3 Level 3

- void cblas_sgemm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_TRANSPOSE TransA, const enum CBLAS_TRANSPOSE TransB,

 const int M, const int N, const int K, const float alpha, const float * A, const int

 lda, const float * B, const int ldb, const float beta, float * C, const int ldc)
- void cblas_ssymm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const int M, const int N,

 const float alpha, const float * A, const int lda, const float * B, const int ldb,

 const float beta, float * C, const int ldc)
- void cblas_ssyrk (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE Trans, const int N,

 const int K, const float alpha, const float * A, const int lda, const float beta,

 float * C, const int ldc)
- void cblas_ssyr2k (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE Trans, const int N,

 const int K, const float alpha, const float * A, const int lda, const float * B,

 const int ldb, const float beta, float * C, const int ldc)
- void cblas_strmm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const enum

 CBLAS_TRANSPOSE TransA, const enum CBLAS_DIAG Diag, const int M,

 const int N, const float alpha, const float * A, const int lda, float * B, const int ldb)
- void cblas_strsm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const enum

 CBLAS_TRANSPOSE TransA, const enum CBLAS_DIAG Diag, const int M,

 const int N, const float alpha, const float * A, const int lda, float * B, const int ldb)
- void cblas_dgemm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_TRANSPOSE TransA, const enum CBLAS_TRANSPOSE TransB,

 const int M, const int N, const int K, const double alpha, const double * A,

 const int lda, const double * B, const int ldb, const double beta, double * C,

 const int ldc)
- void cblas_dsymm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const int M, const int N,

 const double alpha, const double * A, const int lda, const double * B, const

 int ldb, const double beta, double * C, const int ldc)

- void cblas_dsyrk (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE Trans, const int N,

 const int K, const double alpha, const double * A, const int lda, const double beta, double * C, const int ldc)
- void cblas_dsyr2k (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE Trans, const int N,

 const int K, const double alpha, const double * A, const int lda, const double

 * B, const int ldb, const double beta, double * C, const int ldc)
- void cblas_dtrmm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const enum

 CBLAS_TRANSPOSE TransA, const enum CBLAS_DIAG Diag, const int M,

 const int N, const double alpha, const double * A, const int lda, double * B,

 const int ldb)
- void cblas_dtrsm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const enum

 CBLAS_TRANSPOSE TransA, const enum CBLAS_DIAG Diag, const int M,

 const int N, const double alpha, const double * A, const int lda, double * B,

 const int ldb)
- void cblas_cgemm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_TRANSPOSE TransA, const enum CBLAS_TRANSPOSE TransB,

 const int M, const int N, const int K, const void * alpha, const void * A, const int

 lda, const void * B, const int ldb, const void * beta, void * C, const int ldc)

- void cblas_ctrmm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const enum

 CBLAS_TRANSPOSE TransA, const enum CBLAS_DIAG Diag, const int M,

 const int N, const void * alpha, const void * A, const int lda, void * B, const int ldb)

- void cblas_ctrsm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const enum

 CBLAS_TRANSPOSE TransA, const enum CBLAS_DIAG Diag, const int M,

 const int N, const void * alpha, const void * A, const int lda, void * B, const int ldb)
- void cblas_zgemm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_TRANSPOSE TransA, const enum CBLAS_TRANSPOSE TransB,

 const int M, const int N, const int K, const void * alpha, const void * A, const int lda, const void * B, const int ldb, const void * beta, void * C, const int ldc)
- void cblas_zsymm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const int M, const int N,

 const void * alpha, const void * A, const int lda, const void * B, const int ldb,

 const void * beta, void * C, const int ldc)
- void cblas_zsyrk (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE Trans, const int N,

 const int K, const void * alpha, const void * A, const int lda, const void *

 beta, void * C, const int ldc)
- void cblas_ztrmm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const enum

 CBLAS_TRANSPOSE TransA, const enum CBLAS_DIAG Diag, const int M,

 const int N, const void * alpha, const void * A, const int lda, void * B, const int ldb)
- void cblas_ztrsm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const enum

 CBLAS_TRANSPOSE TransA, const enum CBLAS_DIAG Diag, const int M,

 const int N, const void * alpha, const void * A, const int lda, void * B, const int ldb)
- void cblas_chemm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const int M, const int N,

 const void * alpha, const void * A, const int lda, const void * B, const int ldb,

 const void * beta, void * C, const int ldc)

- void cblas_cher2k (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE Trans, const int N,

 const int K, const void * alpha, const void * A, const int lda, const void * B,

 const int ldb, const float beta, void * C, const int ldc)
- void cblas_zhemm (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_SIDE Side, const enum CBLAS_UPLO Uplo, const int M, const int N,

 const void * alpha, const void * A, const int lda, const void * B, const int ldb,

 const void * beta, void * C, const int ldc)
- void cblas_zher2k (const enum CBLAS_ORDER Order, const enum [Function]

 CBLAS_UPLO Uplo, const enum CBLAS_TRANSPOSE Trans, const int N,

 const int K, const void * alpha, const void * A, const int lda, const void * B,

 const int ldb, const double beta, void * C, const int ldc)

void cblas_xerbla (int p, const char * rout, const char * form, ...)
[Function]

D.4 Examples

The following program computes the product of two matrices using the Level-3 BLAS function SGEMM,

$$\begin{pmatrix} 0.11 & 0.12 & 0.13 \\ 0.21 & 0.22 & 0.23 \end{pmatrix} \begin{pmatrix} 1011 & 1012 \\ 1021 & 1022 \\ 1031 & 1032 \end{pmatrix} = \begin{pmatrix} 367.76 & 368.12 \\ 674.06 & 674.72 \end{pmatrix}$$

The matrices are stored in row major order but could be stored in column major order if the first argument of the call to cblas_sgemm was changed to CblasColMajor.

To compile the program use the following command line,

```
$ gcc -Wall demo.c -lgslcblas
```

There is no need to link with the main library -lgsl in this case as the CBLAS library is an independent unit. Here is the output from the program,

```
$ ./a.out
[ 367.76, 368.12
674.06, 674.72 ]
```

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Version 3, 29 June 2007

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gsl_acosh. gsl_asinh. gsl_atanh. gsl_blas_caxpy. gsl_blas_ccopy. gsl_blas_cdotc. gsl_blas_cdotu. gsl_blas_cgemm. gsl_blas_cgemv. gsl_blas_cgerc. gsl_blas_cgerc. gsl_blas_cgeru. gsl_blas_chemm. gsl_blas_chemv. gsl_blas_cher. gsl_blas_cher2 gsl_blas_cher2k. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_cscau.	. 17 . 17 124 124 123 123 128 125 127 126 128 126 127 127 131 130 124 124 123	gsl_blas_sasum gsl_blas_saxpy gsl_blas_scasum gsl_blas_scnrm2 gsl_blas_scopy gsl_blas_sdot gsl_blas_sdedt gsl_blas_sgemm gsl_blas_sgemv gsl_blas_sger gsl_blas_srot gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssyr2 gsl_blas_ssyr2k	$\begin{array}{c} 124 \\ 123 \\ 124 \\ 122 \\ 122 \\ 128 \\ 125 \\ 126 \\ 123 \\ 124 \\ 125 \\ 124 \\ 123 \\ 128 \\ 126 \\ 127 \\ 130 \\ 130 \\ \end{array}$
gsl_acosh. gsl_asinh. gsl_atanh. gsl_blas_caxpy. gsl_blas_ccopy. gsl_blas_cdotc. gsl_blas_cdotu. gsl_blas_cgemm. gsl_blas_cgemv. gsl_blas_cgerc. gsl_blas_cgerc. gsl_blas_cgeru. gsl_blas_chemm. gsl_blas_chemv. gsl_blas_cher. gsl_blas_cher2 gsl_blas_cher2k. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_cscau. gsl_blas_cscau. gsl_blas_cscau. gsl_blas_cscau. gsl_blas_cscau. gsl_blas_cscau. gsl_blas_cscau.	. 17 . 17 124 124 123 123 123 125 127 126 127 127 127 131 130 124 124 123 128	gsl_blas_sasum gsl_blas_saxpy gsl_blas_scasum gsl_blas_scnrm2 gsl_blas_scopy gsl_blas_sdot gsl_blas_sdedt gsl_blas_sgemm gsl_blas_sgemv gsl_blas_sger gsl_blas_srot gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssyr2 gsl_blas_ssyrk	$\begin{array}{c} 124 \\ 123 \\ 124 \\ 122 \\ 122 \\ 128 \\ 125 \\ 126 \\ 123 \\ 124 \\ 125 \\ 124 \\ 123 \\ 128 \\ 126 \\ 127 \\ 130 \\ 130 \\ 129 \\ \end{array}$
gsl_acosh. gsl_asinh. gsl_atanh. gsl_blas_caxpy. gsl_blas_ccopy. gsl_blas_cdotc. gsl_blas_cdotu. gsl_blas_cgemm. gsl_blas_cgemv. gsl_blas_cgerc. gsl_blas_cgerc. gsl_blas_cgeru. gsl_blas_chemw. gsl_blas_chemv. gsl_blas_cher. gsl_blas_cher2 gsl_blas_cher2k. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_csymm. gsl_blas_csyy2k.	. 17 . 17 124 124 123 123 128 125 127 126 128 126 127 127 131 130 124 123 128 130	gsl_blas_sasum gsl_blas_saxpy gsl_blas_scasum gsl_blas_scnrm2 gsl_blas_scopy gsl_blas_sdot gsl_blas_sdedt gsl_blas_sgemm gsl_blas_sgemv gsl_blas_sger gsl_blas_srot gsl_blas_srot gsl_blas_srot gsl_blas_srot gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssyr2 gsl_blas_ssyrk gsl_blas_strmm gsl_blas_strmm	$\begin{array}{c} 124 \\ 123 \\ 124 \\ 122 \\ 122 \\ 128 \\ 125 \\ 126 \\ 123 \\ 124 \\ 125 \\ 124 \\ 123 \\ 128 \\ 126 \\ 127 \\ 130 \\ 130 \\ 129 \\ 125 \\ \end{array}$
gsl_acosh. gsl_asinh. gsl_atanh. gsl_blas_caxpy. gsl_blas_ccopy. gsl_blas_cdotc. gsl_blas_cdotu. gsl_blas_cgemm. gsl_blas_cgemv. gsl_blas_cgerc. gsl_blas_cgerc. gsl_blas_cgeru. gsl_blas_chemm. gsl_blas_chemv. gsl_blas_cher. gsl_blas_cher2 gsl_blas_cher2k. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_csymm. gsl_blas_csyynk.	. 17 . 17 124 124 123 123 128 125 127 126 128 126 127 127 131 130 124 123 128 130 130	gsl_blas_sasum gsl_blas_saxpy gsl_blas_scasum gsl_blas_scnrm2 gsl_blas_scopy gsl_blas_sdot gsl_blas_sdedt gsl_blas_sgemm gsl_blas_sgemv gsl_blas_sger gsl_blas_srot gsl_blas_ssyn gsl_blas_ssyn gsl_blas_ssyn gsl_blas_ssyn gsl_blas_ssyr gsl_blas_ssyr2 gsl_blas_ssyrk gsl_blas_strmm gsl_blas_strmm gsl_blas_strmm	$\begin{array}{c} 124 \\ 123 \\ 124 \\ 122 \\ 128 \\ 125 \\ 126 \\ 123 \\ 124 \\ 125 \\ 124 \\ 125 \\ 124 \\ 127 \\ 120 \\ 127 \\ 130 \\ 129 \\ 125 \\ 129 \\ \end{array}$
gsl_acosh. gsl_asinh. gsl_atanh. gsl_blas_caxpy. gsl_blas_ccopy. gsl_blas_cdotc. gsl_blas_cdotu. gsl_blas_cgemm. gsl_blas_cgemv. gsl_blas_cgerc. gsl_blas_cgerc. gsl_blas_cgeru. gsl_blas_chemm. gsl_blas_chemv. gsl_blas_cher. gsl_blas_cher2 gsl_blas_cher2k gsl_blas_cherk gsl_blas_cscal gsl_blas_cscal gsl_blas_cscal gsl_blas_csymm. gsl_blas_csyv2k gsl_blas_csyvk. gsl_blas_ctrmm.	. 17 . 17 124 124 123 123 128 125 127 126 128 126 127 127 131 130 124 123 128 130 130 129	gsl_blas_sasum gsl_blas_saxpy gsl_blas_scasum gsl_blas_scnrm2 gsl_blas_scopy gsl_blas_sdot gsl_blas_sdedt gsl_blas_sgemm gsl_blas_sger gsl_blas_srot gsl_blas_srot gsl_blas_srot gsl_blas_srot gsl_blas_srot gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssyr2 gsl_blas_ssyr2 gsl_blas_ssyrk gsl_blas_strmm gsl_blas_strsm gsl_blas_strsm gsl_blas_strsm	$\begin{array}{c} 124 \\ 123 \\ 124 \\ 122 \\ 128 \\ 125 \\ 126 \\ 123 \\ 124 \\ 125 \\ 124 \\ 125 \\ 122 \\ 123 \\ 128 \\ 126 \\ 127 \\ 130 \\ 129 \\ 125 \\ 129 \\ 125 \\ 125 \\ \end{array}$
gsl_acosh. gsl_asinh. gsl_atanh. gsl_blas_caxpy. gsl_blas_ccopy. gsl_blas_cdotc. gsl_blas_cdotu. gsl_blas_cgemm. gsl_blas_cgemv. gsl_blas_cgerc. gsl_blas_cgerc. gsl_blas_cderu. gsl_blas_chemm. gsl_blas_chemv. gsl_blas_cher. gsl_blas_cher2 gsl_blas_cher2k. gsl_blas_cherk. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_csymm. gsl_blas_csyvr2k. gsl_blas_csyrk. gsl_blas_ctrmw.	. 17 . 17 . 17 124 123 123 123 128 125 127 126 128 126 127 127 131 130 124 123 128 130 130 129 129 125	gsl_blas_sasum gsl_blas_saxpy gsl_blas_scasum gsl_blas_scnrm2 gsl_blas_scopy gsl_blas_sdot gsl_blas_sdedt gsl_blas_sgemm gsl_blas_sger gsl_blas_srot gsl_blas_srot gsl_blas_srot gsl_blas_srot gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssyr2 gsl_blas_ssyr2 gsl_blas_strmm gsl_blas_strmm gsl_blas_strmm gsl_blas_strsm gsl_blas_strsv gsl_blas_strsv gsl_blas_strsv	$\begin{array}{c} 124 \\ 123 \\ 124 \\ 122 \\ 128 \\ 125 \\ 126 \\ 123 \\ 124 \\ 125 \\ 124 \\ 125 \\ 126 \\ 127 \\ 130 \\ 129 \\ 125 \\ 129 \\ 125 \\ 124 \\ \end{array}$
gsl_acosh. gsl_asinh. gsl_atanh. gsl_blas_caxpy. gsl_blas_ccopy. gsl_blas_cdotc. gsl_blas_cdotu. gsl_blas_cgemm. gsl_blas_cgemv. gsl_blas_cgerc. gsl_blas_cgerc. gsl_blas_cgeru. gsl_blas_chemm. gsl_blas_chemv. gsl_blas_cher2 gsl_blas_cher2 gsl_blas_cher2k gsl_blas_cscal gsl_blas_cscal gsl_blas_cscal gsl_blas_cscypm. gsl_blas_csypnm. gsl_blas_csyprk gsl_blas_ctrmm. gsl_blas_ctrmv. gsl_blas_ctrsm.	. 17 . 17 . 17 124 123 123 128 125 127 126 128 126 127 127 131 130 124 123 128 130 130 129 129 125 129	gsl_blas_sasum gsl_blas_saxpy gsl_blas_scasum gsl_blas_scnrm2 gsl_blas_scopy gsl_blas_sdot gsl_blas_sdedt gsl_blas_sgemm gsl_blas_sger gsl_blas_srot gsl_blas_srot gsl_blas_srot gsl_blas_srot gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_ssvap gsl_blas_ssyv gsl_blas_ssyv gsl_blas_ssyv gsl_blas_ssyv gsl_blas_ssyv gsl_blas_ssyv gsl_blas_ssyv gsl_blas_ssyv gsl_blas_ssyv gsl_blas_strmm gsl_blas_strmw gsl_blas_strsm gsl_blas_strsw gsl_blas_strsv gsl_blas_strsv gsl_blas_zaxpy gsl_blas_zaxpy	$\begin{array}{c} 124 \\ 123 \\ 124 \\ 122 \\ 128 \\ 125 \\ 126 \\ 123 \\ 124 \\ 125 \\ 124 \\ 125 \\ 126 \\ 127 \\ 130 \\ 129 \\ 125 \\ 129 \\ 125 \\ 124 \\ 124 \\ 124 \\ 124 \\ 125 \\ 129 \\ 125 \\ 124 \\ 125 \\ 124 \\ 124 \\ 125 \\ 125 \\ 124 \\ 124 \\ 125 \\ 125 \\ 124 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 125 \\ 125 \\ 124 \\ 125 \\$
gsl_acosh. gsl_asinh. gsl_atanh. gsl_blas_caxpy. gsl_blas_ccopy. gsl_blas_cdotc. gsl_blas_cdotu. gsl_blas_cgemm. gsl_blas_cgemv. gsl_blas_cgerc. gsl_blas_cgerc. gsl_blas_cderu. gsl_blas_chemm. gsl_blas_chemv. gsl_blas_cher. gsl_blas_cher2 gsl_blas_cher2k. gsl_blas_cherk. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_cscal. gsl_blas_csymm. gsl_blas_csyvr2k. gsl_blas_csyrk. gsl_blas_ctrmw.	. 17 . 17 . 17 124 123 123 128 125 127 126 128 126 127 127 131 130 124 123 128 130 130 129 125 129 125	gsl_blas_sasum gsl_blas_saxpy gsl_blas_scasum gsl_blas_scnrm2 gsl_blas_scopy gsl_blas_sdot gsl_blas_sdedt gsl_blas_sgemm gsl_blas_sger gsl_blas_srot gsl_blas_srot gsl_blas_srot gsl_blas_srot gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_srotm gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssynt gsl_blas_ssyr2 gsl_blas_ssyr2 gsl_blas_strmm gsl_blas_strmm gsl_blas_strmm gsl_blas_strsm gsl_blas_strsv gsl_blas_strsv gsl_blas_strsv	$\begin{array}{c} 124 \\ 123 \\ 124 \\ 122 \\ 128 \\ 125 \\ 126 \\ 123 \\ 124 \\ 125 \\ 124 \\ 125 \\ 127 \\ 130 \\ 129 \\ 125 \\ 129 \\ 125 \\ 124 \\ 123 \\ 124 \\ 123 \\ 124 \\ 123 \\ 124 \\ 124 \\ 123 \\ 124 \\ 123 \\ 124 \\ 123 \\ 124 \\ 123 \\ 124 \\ 123 \\ 124 \\ 123 \\ 124 \\ 123 \\ 124 \\ 123 \\ 124 \\ 123 \\ 124 \\ 124 \\ 123 \\ 124 \\ 124 \\ 123 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 124 \\ 125 \\ 124 \\ 124 \\ 125 \\ 124 \\ 125 \\ 124 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 125 \\ 124 \\ 125 \\ 125 \\ 125 \\ 125 \\ 124 \\ 125 \\$

gsl_blas_zdscal	124	gsl_cdf_fdist_Q	234
gsl_blas_zgemm	128	gsl_cdf_fdist_Qinv	234
gsl_blas_zgemv	125	gsl_cdf_flat_P	231
gsl_blas_zgerc	127	gsl_cdf_flat_Pinv	231
gsl_blas_zgeru	126	gsl_cdf_flat_Q	231
gsl_blas_zhemm	128	gsl_cdf_flat_Qinv	231
gsl_blas_zhemv	126	gsl_cdf_gamma_P	229
gsl_blas_zher	127	gsl_cdf_gamma_Pinv	229
gsl_blas_zher2	127	gsl_cdf_gamma_Q	229
gsl_blas_zher2k	131	gsl_cdf_gamma_Qinv	
gsl_blas_zherk	130	gsl_cdf_gaussian_P	
gsl_blas_zscal	124	gsl_cdf_gaussian_Pinv	
gsl_blas_zswap		gsl_cdf_gaussian_Q	216
gsl_blas_zsymm	128	gsl_cdf_gaussian_Qinv	
gsl_blas_zsyr2k		gsl_cdf_geometric_P	254
gsl_blas_zsyrk		gsl_cdf_geometric_Q	
gsl_blas_ztrmm		gsl_cdf_gumbel1_P	
gsl_blas_ztrmv		gsl_cdf_gumbel1_Pinv	
gsl_blas_ztrsm		gsl_cdf_gumbel1_Q	
gsl_blas_ztrsv		gsl_cdf_gumbel1_Qinv	
gsl_block_alloc		gsl_cdf_gumbel2_P	
gsl_block_calloc		gsl_cdf_gumbel2_Pinv	
gsl_block_fprintf		gsl_cdf_gumbel2_Q	
gsl_block_fread		gsl_cdf_gumbel2_Qinv	
gsl_block_free		gsl_cdf_hypergeometric_P	
gsl_block_fscanf		gsl_cdf_hypergeometric_Q	
gsl_block_fwrite		gsl_cdf_laplace_P	
gsl_bspline_alloc		gsl_cdf_laplace_Pinv	
gsl_bspline_deriv_alloc		gsl_cdf_laplace_Q	
gsl_bspline_deriv_eval		gsl_cdf_laplace_Qinv	
gsl_bspline_deriv_eval_nonzero		gsl_cdf_logistic_P	
gsl_bspline_deriv_free		gsl_cdf_logistic_Pinv	
gsl_bspline_eval		gsl_cdf_logistic_Q	
gsl_bspline_eval_nonzero		gsl_cdf_logistic_Qinv	
gsl_bspline_free		gsl_cdf_lognormal_P	
gsl_bspline_greville_abscissa		gsl_cdf_lognormal_Pinv	
gsl_bspline_knots		gsl_cdf_lognormal_Q	
gsl_bspline_knots_uniform		gsl_cdf_lognormal_Qinv	
gsl_bspline_ncoeffs		gsl_cdf_negative_binomial_P	
gsl_cdf_beta_P		gsl_cdf_negative_binomial_Q	
gsl_cdf_beta_Pinv		gsl_cdf_pareto_P	
gsl_cdf_beta_Q		gsl_cdf_pareto_Pinv	
gsl_cdf_beta_Qinv	238	•	240
gsl_cdf_binomial_P		gsl_cdf_pareto_Qinv	-
gsl_cdf_binomial_Q		gsl_cdf_pascal_P	
gsl_cdf_cauchy_P		gsl_cdf_pascal_Q	
gsl_cdf_cauchy_Pinv		gsl_cdf_poisson_P	
gsl_cdf_cauchy_Q		gsl_cdf_poisson_Q	
gsl_cdf_cauchy_Qinv		gsl_cdf_rayleigh_P	
gsl_cdf_chisq_P		gsl_cdf_rayleigh_Pinv	
gsl_cdf_chisq_Pinv		gsl_cdf_rayleigh_Q	
gsl_cdf_chisq_Q		gsl_cdf_rayleigh_Qinv	
gsl_cdf_chisq_Qinv		gsl_cdf_tdist_P	
gsl_cdf_exponential_P		gsl_cdf_tdist_Pinv	
gsl_cdf_exponential_Pinv		gsl_cdf_tdist_Q	
gsl_cdf_exponential_Q		gsl_cdf_tdist_Qinv	
gsl_cdf_exponential_Qinv		gsl_cdf_ugaussian_P	
gsl_cdf_expow_P		gsl_cdf_ugaussian_Pinv	
gsl_cdf_exppow_Q		gsl_cdf_ugaussian_Q	
gsl_cdf_fdist_P		gsl_cdf_ugaussian_Qinv	
gsl_cdf_fdist_Pinv		gsl_cdf_weibull_P	
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gsl_cdf_weibull_Pinv			
		gsl_complex_csc	
gsl_cdf_weibull_Q		gsl_complex_csch	
gsl_cdf_weibull_Qinv		gsl_complex_div	
gsl_cheb_alloc		gsl_complex_div_imag	
gsl_cheb_calc_deriv		gsl_complex_div_real	
gsl_cheb_calc_integ		gsl_complex_exp	
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gsl_cheb_eval	341	gsl_complex_log	24
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gsl_cheb_free		gsl_complex_mul	
gsl_cheb_init		gsl_complex_mul_imag	
gsl_cheb_order		gsl_complex_mul_real	
gsl_cheb_size		gsl_complex_negative	
gsl_combination_alloc		gsl_complex_polar	
gsl_combination_calloc		gsl_complex_poly_complex_eval	
gsl_combination_data		gsl_complex_pow	
gsl_combination_fprintf		gsl_complex_pow_real	
gsl_combination_fread		gsl_complex_rect	
gsl_combination_free		gsl_complex_sec	
gsl_combination_fscanf		gsl_complex_sech	
gsl_combination_fwrite		gsl_complex_secngsl_complex_sin	
gsl_combination_get		gsl_complex_sinh	
-		gsl_complex_simigsl_complex_sqrt	
gsl_combination_init_first			
gsl_combination_init_last		gsl_complex_sqrt_real	
gsl_combination_k		gsl_complex_sub	
gsl_combination_memcpy		gsl_complex_sub_imag	
gsl_combination_n		gsl_complex_sub_real	
gsl_combination_next		gsl_complex_tan	
gsl_combination_prev		gsl_complex_tanh	
gsl_combination_valid		gsl_deriv_backward	
gsl_complex_abs		gsl_deriv_central	
gsl_complex_abs2		gsl_deriv_forward	
gsl_complex_add		gsl_dht_alloc	
gsl_complex_add_imag		gsl_dht_apply	
<pre>gsl_complex_add_real</pre>		gsl_dht_free	
gsl_complex_arccos		_	
		gsl_dht_init	
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gsl_complex_arccoshgsl_complex_arccosh_real	. 25 . 26 . 26	gsl_dht_k_samplegsl_dht_newgsl_dht_x_sample	356 356 356
gsl_complex_arccoshgsl_complex_arccosh_realgsl_complex_arccot	. 25 . 26 . 26 . 25	gsl_dht_k_sample gsl_dht_new gsl_dht_x_sample gsl_eigen_gen	356 356 356 155
gsl_complex_arccoshgsl_complex_arccosh_real	. 25 . 26 . 26 . 25	gsl_dht_k_sample gsl_dht_new gsl_dht_x_sample gsl_eigen_gen gsl_eigen_gen_alloc	356 356 356 155 155
gsl_complex_arccosh gsl_complex_arccosh_real gsl_complex_arccot gsl_complex_arccoth gsl_complex_arccoth gsl_complex_arccsc	. 25 . 26 . 26 . 25 . 26 . 25	gsl_dht_k_sample gsl_dht_new gsl_dht_x_sample gsl_eigen_gen	356 356 356 155 155
gsl_complex_arccosh	. 25 . 26 . 26 . 25 . 26 . 25	gsl_dht_k_sample gsl_dht_new gsl_dht_x_sample gsl_eigen_gen gsl_eigen_gen_alloc	356 356 356 155 155
gsl_complex_arccosh gsl_complex_arccosh_real gsl_complex_arccot gsl_complex_arccoth gsl_complex_arccoth gsl_complex_arccsc	. 25 . 26 . 26 . 25 . 26 . 25 . 25	gsl_dht_k_sample. gsl_dht_new gsl_dht_x_sample. gsl_eigen_gen. gsl_eigen_gen_alloc gsl_eigen_gen_free	356 356 356 155 155 155
gsl_complex_arccosh	. 25 . 26 . 26 . 25 . 26 . 25 . 25 . 25	gsl_dht_k_sample. gsl_dht_new gsl_dht_x_sample. gsl_eigen_gen. gsl_eigen_gen_alloc gsl_eigen_gen_free gsl_eigen_gen_params.	356 356 356 155 155 155 156
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