

# C interfaces to GALAHAD LPA

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# **Chapter 1**

# GALAHAD C package Ipa

# 1.1 Introduction

# 1.1.1 Purpose

This package uses the simplex method to solve the linear programming problem

$$minimize q(x) = g^T x + f$$

subject to the general linear constraints

$$c_i^l \le a_i^T x \le c_i^u, \quad i = 1, \dots, m,$$

and the simple bound constraints

$$x_j^l \le x_j \le x_j^u, \quad j = 1, \dots, n,$$

where the vectors  $g, w, x^0, a_i, c^l, c^u, x^l, x^u$  and the scalar f are given. Any of the constraint bounds  $c_i^l, c_i^u, x_j^l$  and  $x_j^u$  may be infinite. Full advantage is taken of any zero coefficients in the matrix A whose rows are the transposes of the vectors  $a_i$ .

**N.B.** The package is simply a sophisticated interface to the HSL package LA04, and requires that a user has obtained the latter. **LA04 is not included in GALAHAD** but is available without charge to recognised academics, see <a href="http://www.hsl.rl.ac.uk/catalogue/la04.html">http://www.hsl.rl.ac.uk/catalogue/la04.html</a>. If LA04 is unavailable, the GALAHAD interiorpoint linear programming package LPB is an alternative.

#### 1.1.2 Authors

N. I. M. Gould and J. K. Reid, STFC-Rutherford Appleton Laboratory, England.

C interface, additionally J. Fowkes, STFC-Rutherford Appleton Laboratory.

Julia interface, additionally A. Montoison and D. Orban, Polytechnique Montréal.

### 1.1.3 Originally released

October 2018, C interface September 2021.

### 1.1.4 Terminology

The required solution x necessarily satisfies the primal optimality conditions

$$(1a) Ax = c$$

and

$$(1b) c^l \le c \le c^u, \quad x^l \le x \le x^u,$$

the dual optimality conditions

(2a) 
$$g = A^T y + z$$

where

(2b) 
$$y = y^l + y^u, z = z^l + z^u, y^l > 0, y^u < 0, z^l > 0 \text{ and } z^u < 0,$$

and the complementary slackness conditions

(3) 
$$(Ax - c^l)^T y^l = 0$$
,  $(Ax - c^u)^T y^u = 0$ ,  $(x - x^l)^T z^l = 0$  and  $(x - x^u)^T z^u = 0$ ,

where the vectors y and z are known as the Lagrange multipliers for the general linear constraints, and the dual variables for the bounds, respectively, and where the vector inequalities hold component-wise.

The so-called dual to this problem is another linear program

-minimize 
$$c^{lT}y^l + c^{uT}y^u + x^{lT}z^l + x^{uT}z^u + f$$
 subject to the constraints (2a) and (2b)

that uses the same data. The solution to the two problems, it is exists, is the same, but if one is infeasible, the other is unbounded. It can be more efficient to solve the dual, particularly if m is much larger than n.

#### 1.1.5 Method

The bulk of the work is performed by the HSL package LA04. The main subbroutine from this package requires that the input problem be transformed into the 'standard form'

$$\begin{array}{ll} \text{minimize} & g'^Tx'\\ (4) & \text{subject to} & A'x'=b\\ & l_i \leq x_i' \leq u_i, \ (i \leq k)\\ & \text{and} & x_l' \geq 0, \ (i \geq l) \end{array}$$

by introducing slack an surpulus variables, reordering and removing fixed variables and free constraints. The resulting problem involves n' unknowns and m' general constraints. In order to deal with the possibility that the general constraints are inconsistent or not of full rank, LA04 introduces additional 'artifical" variables v and replaces the constraints of (4) by

(5) 
$$A'x' + v = b$$

and gradually encourages v to zero as a first solution phase.

Once a selection of m' independent **non-basic** variables is made, the constraints (5) determine the remaining m' dependent **basic** variables. The **simplex method** is a scheme for systematically adjusting the choice of basic and non-basic variables until a set which defines an optimal solution of (4) is obtained. Each iteration of the simplex method requires the solution of a number of sets of linear equations whose coefficient matrix is the **basis** matrix B, made up of the columns of A' I corresponding to the basic variables, or its transpose  $B^T$ . As the basis matrices for consecutive iterations are closely related, it is normally advantageous to update (rather than recompute) their factorizations as the computation proceeds. If an initial basis is not provided by the user, a set of basic variables which provide a (permuted) triangular basis matrix is found by the simple crash algorithm of Gould and Reid (1989), and initial steepest-edge weights are calculated.

Phases one (finding a feasible solution) and two (solving (4) of the simplex method are applied, as appropriate, with the choice of entering variable as described by Goldfarb and Reid (1977) and the choice of leaving variable as proposed by Harris (1973). Refactorizations of the basis matrix are performed whenever doing so will reduce the average iteration time or there is insufficient memory for its factors. The reduced cost for the entering variable is computed afresh. If it is found to be of a different sign from the recurred value or more than 10% different in magnitude, a fresh computation of all the reduced costs is performed. Details of the factorization and updating procedures are given by Reid (1982). Iterative refinement is encouraged for the basic solution and for the reduced costs after each factorization of the basis matrix and when they are recomputed at the end of phase 1.

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#### 1.1.6 References

D. Goldfarb and J. K. Reid (1977). A practicable steepest-edge simplex algorithm. Mathematical Programming 12 361-371.

N. I. M. Gould and J. K. Reid (1989) New crash procedures for large systems of linear constraints. Mathematical Programming **45** 475-501.

P. M. J. Harris (1973). Pivot selection methods of the Devex LP code. Mathematical Programming 5 1-28.

J. K. Reid (1982) A sparsity-exploiting variant of the Bartels-Golub decomposition for linear-programming bases. Mathematical Programming **24** 55-69.

#### 1.1.7 Call order

To solve a given problem, functions from the lpa package must be called in the following order:

- lpa initialize provide default control parameters and set up initial data structures
- lpa\_read\_specfile (optional) override control values by reading replacement values from a file
- lpa\_import set up problem data structures and fixed values
- · lpa\_reset\_control (optional) possibly change control parameters if a sequence of problems are being solved
- lpa\_solve\_lp solve the linear program
- lpa information (optional) recover information about the solution and solution process
- lpa terminate deallocate data structures

See Section ?? for examples of use.

#### 1.1.8 Unsymmetric matrix storage formats

The unsymmetric m by n constraint matrix A may be presented and stored in a variety of convenient input formats.

Both C-style (0 based) and fortran-style (1-based) indexing is allowed. Choose control.f\_indexing as false for C style and true for fortran style; the discussion below presumes C style, but add 1 to indices for the corresponding fortran version.

Wrappers will automatically convert between 0-based (C) and 1-based (fortran) array indexing, so may be used transparently from C. This conversion involves both time and memory overheads that may be avoided by supplying data that is already stored using 1-based indexing.

#### 1.1.8.1 Dense storage format

The matrix A is stored as a compact dense matrix by rows, that is, the values of the entries of each row in turn are stored in order within an appropriate real one-dimensional array. In this case, component n\*i+j of the storage array A\_val will hold the value  $A_{ij}$  for  $0 \le i \le m-1$ ,  $0 \le j \le n-1$ .

#### 1.1.8.2 Sparse co-ordinate storage format

Only the nonzero entries of the matrices are stored. For the l-th entry,  $0 \le l \le ne-1$ , of A, its row index i, column index j and value  $A_{ij}$ ,  $0 \le i \le m-1$ ,  $0 \le j \le n-1$ , are stored as the l-th components of the integer arrays A\_row and A\_col and real array A\_val, respectively, while the number of nonzeros is recorded as A\_ne = ne.

### 1.1.8.3 Sparse row-wise storage format

Again only the nonzero entries are stored, but this time they are ordered so that those in row i appear directly before those in row i+1. For the i-th row of A the i-th component of the integer array A\_ptr holds the position of the first entry in this row, while A\_ptr(m) holds the total number of entries. The column indices j,  $0 \le j \le n-1$ , and values  $A_{ij}$  of the nonzero entries in the i-th row are stored in components I = A\_ptr(i), . . . , A\_ptr(i+1)-1,  $0 \le i \le m-1$ , of the integer array A\_col, and real array A\_val, respectively. For sparse matrices, this scheme almost always requires less storage than its predecessor.

# **Chapter 2**

# File Index

# 2.1 File List

Here is a list of all files with brief descriptions:			
galahad_lpa.h			??

6 File Index

# **Chapter 3**

# **File Documentation**

# 3.1 galahad lpa.h File Reference

```
#include <stdbool.h>
#include <stdint.h>
#include "galahad_precision.h"
#include "galahad_cfunctions.h"
#include "galahad_rpd.h"
```

#### **Data Structures**

- struct lpa\_control\_type
- struct lpa\_time\_type
- struct lpa\_inform\_type

#### **Functions**

- void lpa\_initialize (void \*\*data, struct lpa\_control\_type \*control, int \*status)
- void lpa\_read\_specfile (struct lpa\_control\_type \*control, const char specfile[])
- void lpa\_import (struct lpa\_control\_type \*control, void \*\*data, int \*status, int n, int m, const char A\_type[], int A\_ne, const int A\_row[], const int A\_col[], const int A\_ptr[])
- void lpa\_reset\_control (struct lpa\_control\_type \*control, void \*\*data, int \*status)
- void lpa\_solve\_lp (void \*\*data, int \*status, int n, int m, const real\_wp\_ g[], const real\_wp\_ f, int a\_ne, const real\_wp\_ A\_val[], const real\_wp\_ c\_l[], const real\_wp\_ c\_u[], const real\_wp\_ x\_l[], const real\_wp\_ x\_u[], real\_wp\_ x[], real\_wp\_ c[], real\_wp\_ z[], int x\_stat[], int c\_stat[])
- void lpa\_information (void \*\*data, struct lpa\_inform\_type \*inform, int \*status)
- void lpa terminate (void \*\*data, struct lpa control type \*control, struct lpa inform type \*inform)

#### 3.1.1 Data Structure Documentation

#### 3.1.1.1 struct lpa\_control\_type

control derived type as a C struct

#### **Examples**

lpat.c, and lpatf.c.

# **Data Fields**

bool	f_indexing	use C or Fortran sparse matrix indexing
int	error	error and warning diagnostics occur on stream error
int	out	general output occurs on stream out
int	print_level	the level of output required is specified by print_level (>= 2 turns on LA04 output)
int	start_print	any printing will start on this iteration
int	stop_print	any printing will stop on this iteration
int	maxit	at most maxit inner iterations are allowed
int	max_iterative_refinements	maximum number of iterative refinements allowed
int	min_real_factor_size	initial size for real array for the factors and other data
int	min_integer_factor_size	initial size for integer array for the factors and other data
int	random_number_seed	the initial seed used when generating random numbers
int	sif_file_device	specifies the unit number to write generated SIF file describing the current problem
int	qplib_file_device	specifies the unit number to write generated QPLIB file describing the current problem
real_wp_	infinity	any bound larger than infinity in modulus will be regarded as infinite
real_wp_	tol_data	the tolerable relative perturbation of the data (A,g,) defining the problem
real_wp_	feas_tol	any constraint violated by less than feas_tol will be considered to be satisfied
real_wp_	relative_pivot_tolerance	pivot threshold used to control the selection of pivot elements in the matrix factorization. Any potential pivot which is less than the largest entry in its row times the threshold is excluded as a candidate
real_wp_	growth_limit	limit to control growth in the upated basis factors. A refactorization occurs if the growth exceeds this limit
real_wp_	zero_tolerance	any entry in the basis smaller than this is considered zero
real_wp_	change_tolerance	any solution component whose change is smaller than a tolerence times the largest change may be considered to be zero
real_wp_	identical_bounds_tol	any pair of constraint bounds (c_l,c_u) or (x_l,x_u) that are closer than identical_bounds_tol will be reset to the average of their values
real_wp_	cpu_time_limit	the maximum CPU time allowed (-ve means infinite)
real_wp_	clock_time_limit	the maximum elapsed clock time allowed (-ve means infinite)
bool	scale	if .scale is true, the problem will be automatically scaled prior to solution. This may improve computation time and accuracy
bool	dual	should the dual problem be solved rather than the primal?
bool	warm_start	should a warm start using the data in C_stat and X_stat be attempted?
bool	steepest_edge	should steepest-edge weights be used to determine the variable leaving the basis?
bool	space_critical	if .space_critical is true, every effort will be made to use as little space as possible. This may result in longer computation time
bool	deallocate_error_fatal	if .deallocate_error_fatal is true, any array/pointer deallocation error will terminate execution. Otherwise, computation will continue
bool	generate_sif_file	if .generate_sif_file is .true. if a SIF file describing the current problem is to be generated
bool	generate_qplib_file	if .generate_qplib_file is .true. if a QPLIB file describing the current problem is to be generated
char	sif_file_name[31]	name of generated SIF file containing input problem
char	qplib_file_name[31]	name of generated QPLIB file containing input problem
	L	

#### **Data Fields**

char	prefix[31]	all output lines will be prefixed by .prefix(2:LEN(TRIM(.prefix))-1)
		where .prefix contains the required string enclosed in quotes, e.g.
		"string" or 'string'

# 3.1.1.2 struct lpa\_time\_type

time derived type as a C struct

# Data Fields

real_wp_	total	the total CPU time spent in the package
real_wp_	preprocess	the CPU time spent preprocessing the problem
real_wp_	clock_total	the total clock time spent in the package
real_wp_	clock_preprocess	the clock time spent preprocessing the problem

# 3.1.1.3 struct lpa\_inform\_type

inform derived type as a C struct

# Examples

lpat.c, and lpatf.c.

#### **Data Fields**

int	status	return status. See LPA_solve for details
int	alloc_status	the status of the last attempted allocation/deallocation
char	bad_alloc[81]	the name of the array for which an allocation/deallocation error ocurred
int	iter	the total number of iterations required
int	la04_job	the final value of la04's job argument
int	la04_job_info	any extra information from an unsuccesfull call to LA04 (LA04's RINFO(35)
real_wp_	obj	the value of the objective function at the best estimate of the solution determined by LPA_solve
real_wp_	primal_infeasibility	the value of the primal infeasibility
bool	feasible	is the returned "solution" feasible?
real_wp_	RINFO[40]	the information array from LA04
struct lpa_time_type	time	timings (see above)
struct rpd_inform_type	rpd_inform	inform parameters for RPD

# 3.1.2 Function Documentation

#### 3.1.2.1 | lpa\_initialize()

Set default control values and initialize private data

#### **Parameters**

in,out	data	holds private internal data
out	control	is a struct containing control information (see lpa_control_type)
out	status	is a scalar variable of type int, that gives the exit status from the package. Possible values are (currently):  • 0. The import was successful.

# **Examples**

lpat.c, and lpatf.c.

# 3.1.2.2 lpa\_read\_specfile()

Read the content of a specification file, and assign values associated with given keywords to the corresponding control parameters. By default, the spcification file will be named RUNLPA.SPC and lie in the current directory. Refer to Table 2.1 in the fortran documentation provided in \$GALAHAD/doc/lpa.pdf for a list of keywords that may be set.

#### Parameters

in,out	control	is a struct containing control information (see lpa_control_type)
in	specfile	is a character string containing the name of the specification file

#### 3.1.2.3 lpa\_import()

```
int m,
const char A_type[],
int A_ne,
const int A_row[],
const int A_col[],
const int A_ptr[])
```

Import problem data into internal storage prior to solution.

#### **Parameters**

in	control	is a struct whose members provide control paramters for the remaining proedures (see lpa_control_type)
in,out	data	holds private internal data
in, out	status	is a scalar variable of type int, that gives the exit status from the package. Possible values are:  • 0. The import was succesful  • -1. An allocation error occurred. A message indicating the offending array is written on unit control.error, and the returned allocation status and a string containing the name of the offending array are held in inform.alloc_status and inform.bad_alloc respectively.  • -2. A deallocation error occurred. A message indicating the offending array is written on unit control.error and the returned allocation status and a string
		<ul> <li>containing the name of the offending array are held in inform.alloc_status and inform.bad_alloc respectively.</li> <li>-3. The restrictions n &gt; 0 or m &gt; 0 or requirement that A_type contains its relevant string 'dense', 'coordinate' or 'sparse_by_rows' has been violated.</li> </ul>
in	n	is a scalar variable of type int, that holds the number of variables.
in	m	is a scalar variable of type int, that holds the number of general linear constraints.
in	A_type	is a one-dimensional array of type char that specifies the unsymmetric storage scheme used for the constraint Jacobian, $A$ . It should be one of 'coordinate', 'sparse_by_rows' or 'dense; lower or upper case variants are allowed.
in	A_ne	is a scalar variable of type int, that holds the number of entries in $A$ in the sparse co-ordinate storage scheme. It need not be set for any of the other schemes.
in	A_row	is a one-dimensional array of size A_ne and type int, that holds the row indices of $A$ in the sparse co-ordinate storage scheme. It need not be set for any of the other schemes, and in this case can be NULL.
in	A_col	is a one-dimensional array of size A_ne and type int, that holds the column indices of $A$ in either the sparse co-ordinate, or the sparse row-wise storage scheme. It need not be set when the dense or diagonal storage schemes are used, and in this case can be NULL.
in	A_ptr	is a one-dimensional array of size $n+1$ and type int, that holds the starting position of each row of $A$ , as well as the total number of entries, in the sparse row-wise storage scheme. It need not be set when the other schemes are used, and in this case can be NULL.

# **Examples**

lpat.c, and lpatf.c.

#### 3.1.2.4 | lpa\_reset\_control()

Reset control parameters after import if required.

#### **Parameters**

in	control	is a struct whose members provide control paramters for the remaining proedures (see lpa_control_type)
in,out	data	holds private internal data
in,out	status	is a scalar variable of type int, that gives the exit status from the package. Possible values are:  • 0. The import was succesful.

#### 3.1.2.5 lpa\_solve\_lp()

```
void lpa_solve_lp (
             void ** data,
             int * status,
             int n_{i}
             int m_{r}
             const real_wp_ g[],
             const real_wp_ f,
             int a_ne,
             const real_wp_ A_val[],
             const real_wp_ c_1[],
             const real_wp_ c_u[],
             const real_wp_ x_1[],
             const real_wp_ x_u[],
             real_wp_ x[],
             real_wp_ c[],
             real_wp_ y[],
             real_wp_ z[],
             int x_stat[],
             int c\_stat[] )
```

Solve the linear program.

#### **Parameters**

in,out	data	holds private internal data
, out	uaia	noids private internal data

# **Parameters**

in,out	status	is a scalar variable of type int, that gives the entry and exit status from the package.  Possible exit are:
		0. The run was succesful.
		<ul> <li>-1. An allocation error occurred. A message indicating the offending array is written on unit control.error, and the returned allocation status and a string containing the name of the offending array are held in inform.alloc_status and inform.bad_alloc respectively.</li> </ul>
		<ul> <li>-2. A deallocation error occurred. A message indicating the offending array is written on unit control.error and the returned allocation status and a string containing the name of the offending array are held in inform.alloc_status and inform.bad_alloc respectively.</li> </ul>
		<ul> <li>-3. The restrictions n &gt; 0 and m &gt; 0 or requirement that A_type contains its relevant string 'dense', 'coordinate' or 'sparse_by_rows' has been violated.</li> </ul>
		-5. The simple-bound constraints are inconsistent.
		-7. The constraints appear to have no feasible point.
		<ul> <li>-9. The analysis phase of the factorization failed; the return status from the factorization package is given in the component inform.factor_status</li> </ul>
		<ul> <li>-10. The factorization failed; the return status from the factorization package is given in the component inform.factor_status.</li> </ul>
		<ul> <li>-11. The solution of a set of linear equations using factors from the factorization package failed; the return status from the factorization package is given in the component inform.factor_status.</li> </ul>
		-16. The problem is so ill-conditioned that further progress is impossible.
		-17. The step is too small to make further impact.
		<ul> <li>-18. Too many iterations have been performed. This may happen if control.maxit is too small, but may also be symptomatic of a badly scaled problem.</li> </ul>
		<ul> <li>-19. The CPU time limit has been reached. This may happen if control.cpu_time_limit is too small, but may also be symptomatic of a badly scaled problem.</li> </ul>
in	n	is a scalar variable of type int, that holds the number of variables
in	m	is a scalar variable of type int, that holds the number of general linear constraints.
in	g	is a one-dimensional array of size n and type double, that holds the linear term $g$ of the objective function. The j-th component of g, j = 0,, n-1, contains $g_j$ .
in	f	is a scalar of type double, that holds the constant term $f$ of the objective function.
in	a_ne	is a scalar variable of type int, that holds the number of entries in the constraint Jacobian matrix ${\cal A}.$
in	A_val	is a one-dimensional array of size a_ne and type double, that holds the values of the entries of the constraint Jacobian matrix $A$ in any of the available storage schemes.
in	c_I	is a one-dimensional array of size m and type double, that holds the lower bounds $c^l$ on the constraints $Ax$ . The i-th component of c_l, i = 0,, m-1, contains $c^l_i$ .
in	c_u	is a one-dimensional array of size m and type double, that holds the upper bounds $c^l$ on the constraints $Ax$ . The i-th component of c_u, i = 0,, m-1, contains $c^u_i$ .
in	x_I	is a one-dimensional array of size n and type double, that holds the lower bounds $x^l$ on the variables $x$ . The j-th component of x_l, j = 0,, n-1, contains $x^l_j$ .

# **Parameters**

in	x_u	is a one-dimensional array of size n and type double, that holds the upper bounds $x^l$ on the variables $x$ . The j-th component of x_u, j = 0,, n-1, contains $x^l_j$ .
in,out	X	is a one-dimensional array of size n and type double, that holds the values $x$ of the optimization variables. The j-th component of x, j = 0,, n-1, contains $x_j$ .
out	С	is a one-dimensional array of size m and type double, that holds the residual $c(x)$ . The i-th component of c, i = 0,, m-1, contains $c_i(x)$ .
in,out	У	is a one-dimensional array of size n and type double, that holds the values $y$ of the Lagrange multipliers for the general linear constraints. The j-th component of y, i = 0,, m-1, contains $y_i$ .
in,out	Z	is a one-dimensional array of size n and type double, that holds the values $z$ of the dual variables. The j-th component of z, j = 0,, n-1, contains $z_j$ .
out	x_stat	is a one-dimensional array of size n and type int, that gives the optimal status of the problem variables. If $x_{stat}(j)$ is negative, the variable $x_{j}$ most likely lies on its lower bound, if it is positive, it lies on its upper bound, and if it is zero, it lies between its bounds.
out	c_stat	is a one-dimensional array of size m and type int, that gives the optimal status of the general linear constraints. If c_stat(i) is negative, the constraint value $a_i^T x$ most likely lies on its lower bound, if it is positive, it lies on its upper bound, and if it is zero, it lies between its bounds.

# Examples

lpat.c, and lpatf.c.

# 3.1.2.6 lpa\_information()

# Provides output information

#### **Parameters**

in,out	data	holds private internal data
out	inform	is a struct containing output information (see lpa_inform_type)
out	status	is a scalar variable of type int, that gives the exit status from the package. Possible values are (currently):
		0. The values were recorded succesfully

# Examples

lpat.c, and lpatf.c.

# 3.1.2.7 lpa\_terminate()

# Deallocate all internal private storage

# Parameters

	in,out	data	holds private internal data
ſ	out	control	is a struct containing control information (see lpa_control_type)
ſ	out	inform	is a struct containing output information (see lpa_inform_type)

# Examples

lpat.c, and lpatf.c.

# **Chapter 4**

# **Example Documentation**

# 4.1 lpat.c

This is an example of how to use the package to solve a linear program. A variety of supported constraint matrix storage formats are shown.

Notice that C-style indexing is used, and that this is flaggeed by setting control.f\_indexing to false.

```
/* lpat.c */
/* Full test for the LPA C interface using C sparse matrix indexing \star/
#include <stdio.h>
#include <math.h>
#include "galahad_precision.h"
#include "galahad_cfunctions.h"
#include "galahad_lpa.h"
int main(void) {
     // Derived types
     void *data;
     struct lpa_control_type control;
     struct lpa_inform_type inform;
     // Set problem data
     int n = 3; // dimension
int m = 2; // number of general constraints
real_wp_ g[] = {0.0, 2.0, 0.0}; // linear
                                                   // linear term in the objective
     real_wp_ f = 1.0; // constant term in the objective int A_ne = 4; // Jacobian elements
     int A_row[] = {0, 0, 1, 1}; // row indices
int A_col[] = {0, 1, 1, 2}; // column indices
int A_ptr[] = {0, 2, 4}; // row pointers
real_wp_A_val[] = {2.0, 1.0, 1.0, 1.0}; // values
     real_wp_ c_1[] = {1.0, 2.0}; // constraint lower bound real_wp_ c_u[] = {2.0, 2.0}; // constraint upper bound
     real_wp_ x_1[] = {-1.0, - INFINITY, - INFINITY}; // variable lower bound real_wp_ x_u[] = {1.0, INFINITY, 2.0}; // variable upper bound
     // Set output storage
     real_wp_ c[m]; // constraint values
int x_stat[n]; // variable status
     int c_stat[m]; // constraint status
     char st;
     int status:
     printf(" C sparse matrix indexing\n\n");
     printf(" basic tests of lp storage formats\n\n");
     for( int d=1; d <= 3; d++){
           // Initialize LPA
          lpa_initialize( &data, &control, &status );
          // Set user-defined control options
control.f_indexing = false; // C sparse matrix indexing
           // Start from 0
           real_wp_ x[] = \{0.0, 0.0, 0.0\};
           real_wp_ y[] = {0.0,0.0};
           real_wp_ z[] = {0.0,0.0,0.0};
           switch(d){
                case 1: // sparse co-ordinate storage
                      lpa_import( &control, &data, &status, n, m,
```

```
"coordinate", A_ne, A_row, A_col, NULL );
                x_stat, c_stat );
                break;
            printf(" case %li break\n",d);
            case 2: // sparse by rows
                st = 'R';
                break;
            case 3: // dense
                st = 'D';
                int A_dense_ne = 6; // number of elements of A real_wp_ A_dense[] = {2.0, 1.0, 0.0, 0.0, 1.0, 1.0}; lpa_import( &control, &data, &status, n, m,
                             "dense", A_ne, NULL, NULL, NULL);
                lpa_solve_lp( &data, &status, n, m, g, f,
                               A_dense_ne, A_dense, c_1, c_u, x_1, x_u,
                               x, c, y, z, x_stat, c_stat);
                break:
        lpa_information( &data, &inform, &status );
        if(inform.status == 0){
            printf("%c:%6i iterations. Optimal objective value = %5.2f status = %1i\n",
                   st, inform.iter, inform.obj, inform.status);
        }else{
           printf("%c: LPA_solve exit status = %1i\n", st, inform.status);
        //for( int i = 0; i < n; i++) printf("%f ", x[i]);
//printf("\n");
//printf("gradient: ");</pre>
        //for( int i = 0; i < n; i++) printf("%f ", g[i]);
//printf("\n");
        // Delete internal workspace
        lpa_terminate( &data, &control, &inform );
    }
}
```

# 4.2 lpatf.c

This is the same example, but now fortran-style indexing is used.

```
/* lpatf.c */
/\star Full test for the LPA C interface using Fortran sparse matrix indexing \star/
#include <stdio.h>
#include <math.h>
#include "galahad_precision.h"
#include "galahad_cfunctions.h"
#include "galahad_lpa.h"
int main(void) {
     // Derived types
      void *data;
      struct lpa_control_type control;
      struct lpa_inform_type inform;
      // Set problem data
      int n = 3; // dimension
      int m = 2; // number of general constraints
     int m = 2; // number of general constraints
real_wp_ g[] = {0.0, 2.0, 0.0}; // linear term in the objective
real_wp_ f = 1.0; // constant term in the objective
int A_ne = 4; // Jacobian elements
int A_row[] = {1, 1, 2, 2}; // row indices
int A_col[] = {1, 2, 2, 3}; // column indices
int A_ptr[] = {1, 3, 5}; // row pointers
real_wp_ A_val[] = {2.0, 1.0, 1.0, 1.0, }; // values
      real_wp_ c_l[] = {1.0, 2.0}; // constraint lower bound real_wp_ c_u[] = {2.0, 2.0}; // constraint upper bound
      real_wp_ x_1[] = {-1.0, - INFINITY, - INFINITY}; // variable lower bound real_wp_ x_u[] = {1.0, INFINITY, 2.0}; // variable upper bound
       // Set output storage
      real_wp_ c[m]; // constraint values
      int x_stat[n]; // constraint varies
int c_stat[m]; // constraint status
      char st;
      int status:
      printf(" Fortran sparse matrix indexing\n\n");
      printf(" basic tests of lp storage formatsn\n");
```

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```
for( int d=1; d <= 3; d++) {</pre>
    // Initialize LPA
    lpa_initialize( &data, &control, &status );
    // Set user-defined control options
control.f_indexing = true; // Fortran sparse matrix indexing
// Start from 0
    real_wp_ x[] = \{0.0, 0.0, 0.0\};
    real_wp_ y[] = {0.0,0.0};
    real_wp_ z[] = \{0.0, 0.0, 0.0\};
    switch(d){
        case 1: // sparse co-ordinate storage
            st = 'C';
            lpa_import( &control, &data, &status, n, m,
            "coordinate", A_ne, A_row, A_col, NULL);
lpa_solve_lp( &data, &status, n, m, g, f,
                           A_ne, A_val, c_l, c_u, x_l, x_u, x, c, y, z,
                           x_stat, c_stat );
            break;
        printf(" case %li break\n",d);
        case 2: // sparse by rows
st = 'R';
            case 3: // dense
    st = 'D';
            A_dense_ne, A_dense, c_1, c_u, x_1, x_u,
                           x, c, y, z, x_stat, c_stat);
            break:
    lpa_information( &data, &inform, &status );
    if (inform.status == 0) {
    printf("%c:%6i iterations. Optimal objective value = %5.2f status = %1i\n",
               st, inform.iter, inform.obj, inform.status);
    }else{
       printf("%c: LPA_solve exit status = %li\n", st, inform.status);
    //printf("x: ");
//for( int i = 0; i < n; i++) printf("%f ", x[i]);
//printf("\n");
    //printf("gradient: ");
    //for( int i = 0; i < n; i++) printf("%f ", g[i]);
    //printf("\n");
    // Delete internal workspace
    lpa_terminate( &data, &control, &inform );
```