

GAIL—Guaranteed Automatic Integration Library in MATLAB: Documentation for Version 2.1*

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1 Introduction

Automatic and adaptive approximation, optimization, or integration of functions in a cone with guarantee of accuracy is a relatively new paradigm [7]. Our purpose is to create an open-source MATLAB package, Guaranteed Automatic Integration Library (GAIL) [5], following the philosophy of reproducible research championed by Claerbout [6] and Donoho [1], and sustainable practices of robust scientific software development [12]. For our conviction that true scholarship in computational sciences are characterized by reliable reproducibility [3, 4, 2], we employ the best practices in mathematical research and software engineering known to us and available in MATLAB.

The rest of this document describes the key features of functions in GAIL, which includes one-dimensional function approximation [7, 8] and minimization [14] using linear splines, one-dimensional numerical integration using trapezoidal rule [7], and last but not least, mean estimation and multidimensional integration by Monte Carlo methods [9, 11] or Quasi Monte Carlo methods [13, 10].

1.1 Downloads

GAIL can be downloaded from

<http://code.google.com/p/gail/>

Alternatively, you can get a local copy of the GAIL repository with this command:

```
git clone https://github.com/GailGithub/GAIL_Dev.git
```

1.2 Requirements

You will need to install MATLAB 7 or a later version.

1.3 Documentation

Detailed documentation is available at [GAIL_Matlab/Documentation](#).

1.4 General Usage Notes

GAIL Version 2.1 [5] includes the following eight algorithms:

1. `funappx_g` [7, 8]: One-dimensional function approximation on bounded interval
2. `funmin_g` [14]: global minimum value of univariate function on a closed interval
3. `integral_g` [7]: One-dimensional integration on bounded interval
4. `meanMC_g` [9]: Monte Carlo method for estimating mean of a random variable
5. `meanMCBer_g` [11]: Monte Carlo method to estimate the mean of a Bernoulli random variable
6. `cubMC_g` [9]: Monte Carlo method for numerical multiple integration
7. `cubLattice_g` [13]: Quasi-Monte Carlo method using rank-1 Lattices cubature for a d-dimensional integration
8. `cubSobol_g` [10]: Quasi-Monte Carlo method using Sobol' cubature for a d-dimensional integration

1.5 Installation Instruction

1. Unzip the contents of the zip file to a directory and maintain the existing directory and subdirectory structure. (Please note: If you install into the `toolbox` subdirectory of the MATLAB program hierarchy, you will need to click the button “Update toolbox path cache” from the File/Preferences... dialog in MATLAB.)
2. In MATLAB, add the GAIL directory to your path. This can be done by running `GAIL_Install.m`. Alternatively, this can be done by selecting “File/Set Path...” from the main or Command window menus, or with the command `pathtool`. We recommend that you select the “Save” button on this dialog so that GAIL is on the path automatically in future MATLAB sessions.
3. To check if you have installed GAIL successfully, type `help funappx_g` to see if its documentation shows up.

Alternatively, you could do this:

1. Download `DownloadInstallGail_2_1.m` and put it where you want GAIL to be installed.
2. Execute it in MATLAB.

To uninstall GAIL, execute `GAIL_Uninstall`.

To reinstall GAIL, execute `GAIL_Install`.

1.6 Tests

We provide quick doctests for each of the functions above. To run doctests in `funappx_g`, for example, issue the command `doctest funappx_g`.

We also provide unit tests for MATLAB version 8 or later. To run unit tests for `funmin_g`, for instance, execute `run(ut_funmin_g)`.

1.7 Contact Information

Please send any queries, questions, or comments to

`gail-users@googlegroups.com`

1.8 Website

For more information about GAIL, visit [GAIL Project website](#).

2 funappx_g

1-D guaranteed locally adaptive function approximation (or function recovery) on $[a,b]$

2.1 Syntax

`fappx = funappx_g(f)`

`fappx = funappx_g(f,a,b,abstol)`

`fappx = funappx_g(f,'a',a,'b',b,'abstol',abstol)`

`fappx = funappx_g(f,in_param)`

`[fappx, out_param] = funappx_g(f,...)`

2.2 Description

`fappx = funappx_g(f)` approximates function `f` on the default interval $[0,1]$ by an approximated function handle `fappx` within the guaranteed absolute error tolerance of $1e-6$. When Matlab version is higher or equal to 8.3, `fappx` is an interpolant generated by `griddedInterpolant`. When Matlab version is lower than 8.3, `fappx` is a function handle generated by `ppval` and `interp1`. Input `f` is a function handle. The statement `y = f(x)` should accept a vector argument `x` and return a vector `y` of function values that is of the same size as `x`.

`fappx = funappx_g(f,a,b,abstol)` for a given function `f` and the ordered input parameters that define the finite interval $[a,b]$, and a guaranteed absolute error tolerance `abstol`.

`fappx = funappx_g(f,'a',a,'b',b,'abstol',abstol)` approximates function `f` on the finite interval $[a,b]$, given a guaranteed absolute error tolerance `abstol`. All four field-value pairs are optional and can be supplied in different order.

`fappx = funappx_g(f,in_param)` approximates function `f` on the finite interval $[in_param.a, in_param.b]$, given a guaranteed absolute error tolerance `in_param.abstol`. If a field is not specified, the default value is used.

`[fappx, out_param] = funappx_g(f,...)` returns an approximated function `fappx` and an output structure `out_param`.

Input Arguments

- `f` — input function
- `in_param.a` — left end point of interval, default value is 0
- `in_param.b` — right end point of interval, default value is 1
- `in_param.abstol` — guaranteed absolute error tolerance, default value is $1e-6$

Optional Input Arguments

- `in_param.nlo` — lower bound of initial number of points we used, default value is 10
- `in_param.nhi` — upper bound of initial number of points we used, default value is 1000
- `in_param.nmax` — when number of points hits the value, iteration will stop, default value is $1e7$

- `in_param.maxiter` — max number of iterations, default value is 1000

Output Arguments

- `fappx` — approximated function handle (Note: When Matlab version is higher or equal to 8.3, `fappx` is an interpolant generated by `griddedInterpolant`. When Matlab version is lower than 8.3, `fappx` is a function handle generated by `ppval` and `interp1`.)
- `out_param.f` — input function
- `out_param.a` — left end point of interval
- `out_param.b` — right end point of interval
- `out_param.abstol` — guaranteed absolute error tolerance
- `out_param.nlo` — a lower bound of initial number of points we use
- `out_param.nhi` — an upper bound of initial number of points we use
- `out_param.nmax` — when number of points hits the value, iteration will stop
- `out_param.maxiter` — max number of iterations
- `out_param.ninit` — initial number of points we use for each sub interval
- `out_param.exit` — this is a number defining the conditions of success or failure satisfied when finishing the algorithm. The algorithm is considered successful (with `out_param.exit == 0`) if no other flags arise warning that the results are certainly not guaranteed. The initial value is 0 and the final value of this parameter is encoded as follows: 1 If reaching overbudget. It states whether the max budget is attained without reaching the guaranteed error tolerance.
2 If reaching overiteration. It states whether the max iterations is attained without reaching the guaranteed error tolerance.
- `out_param.iter` — number of iterations
- `out_param.npoints` — number of points we need to reach the guaranteed absolute error tolerance
- `out_param.errest` — an estimation of the absolute error for the approximation
- `out_param.nstar` — final value of the parameter defining the cone of functions for which this algorithm is guaranteed for each subinterval; `nstar = ninit-2` initially

2.3 Guarantee

For $[a, b]$, there exists a partition

$$P = \{[t_0, t_1], [t_1, t_2], \dots, [t_{L-1}, t_L]\}, a = t_0 < t_1 < \dots < t_L = b.$$

If the function to be approximated, f satisfies the cone condition

$$\|f''\|_\infty \leq \frac{2nstar}{t_l - t_{l-1}} \left\| f' - \frac{f(t_l) - f(t_{l-1})}{t_l - t_{l-1}} \right\|_\infty$$

for each sub interval $[t_{l-1}, t_l]$, where $1 \leq l \leq L$, then the `fappx` output by this algorithm is guaranteed to satisfy

$$\|f - fappx\|_\infty \leq abstol.$$

2.4 Examples

Example 1

```
f = @(x) x.^2; [fappx, out_param] = funappx_g(f)

% Approximate function x^2 with default input parameter to make the error
% less than 1e-6. For MATLAB version 8.3 onwards, we see:
```

```
fappx =
```

```
griddedInterpolant with properties:
```

```
    GridVectors: {[1x3169 double]}
        Values: [1x3169 double]
        Method: 'linear'
  ExtrapolationMethod: 'linear'
```

```
out_param =
```

```
    f: @(x)x.^2
    a: 0
    b: 1
  abstol: 1.0000e-06
    nlo: 10
    nhi: 1000
    nmax: 10000000
  maxiter: 1000
    ninit: 100
    exit: [2x1 logical]
    iter: 6
  npoints: 3169
  errest: 2.7429e-07
    nstar: [1x32 double]
```

For earlier versions of MATLAB, we have:

```
fappx =
```

```
@(x)ppval(pp,x)
```

```
out_param =
```

```
    f: @(x)x.^2
    a: 0
    b: 1
  abstol: 1.0000e-06
    nlo: 10
    nhi: 1000
    nmax: 10000000
```

```

maxiter: 1000
ninit: 100
exit: [2x1 logical]
iter: 6
npoints: 3169
errest: 2.7429e-07
nstar: [10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 ...
        10 10 10 10 10 10 10 10 10 10 10 10 10 10 10]

```

Example 2

```

[fappx, out_param] = funappx_g(@(x) x.^2,0,100,1e-7,10,1000,1e8)

% Approximate function x^2 on [0,100] with error tolerance 1e-7, cost
% budget 10000000, lower bound of initial number of points 10 and upper
% bound of initial number of points 100

fappx =

griddedInterpolant with properties:

    GridVectors: {[1x977921 double]}
      Values: [1x977921 double]
    Method: 'linear'
  ExtrapolationMethod: 'linear'

out_param =

    a: 0
  abstol: 1.0000e-07
    b: 100
    f: @(x)x.^2
maxiter: 1000
   nhi: 1000
   nlo: 10
   nmax: 100000000
  ninit: 956
   exit: [2x1 logical]
   iter: 11
npoints: 977921
  errest: 3.7104e-08
   nstar: [1x1024 double]

```

Example 3

```

clear in_param; in_param.a = -20; in_param.b = 20; in_param.nlo = 10;
in_param.nhi = 100; in_param.nmax = 1e8; in_param.abstol = 1e-7;
[fappx, out_param] = funappx_g(@(x) x.^2, in_param)

% Approximate function x^2 on [-20,20] with error tolerance 1e-7, cost
% budget 1000000, lower bound of initial number of points 10 and upper
% bound of initial number of points 100

```



```
fappx =

griddedInterpolant with properties:

    GridVectors: {[1x385025 double]}
      Values: [1x385025 double]
    Method: 'linear'
  ExtrapolationMethod: 'linear'
```

```
out_param =

    a: -20
  abstol: 1.0000e-07
    b: 20
    f: @(x)x.^2
maxiter: 1000
   nhi: 100
   nlo: 10
   nmax: 100000000
  ninit: 95
   exit: [2x1 logical]
   iter: 13
npoints: 385025
  errest: 2.6570e-08
   nstar: [1x4096 double]
```

Example 4

```
clear in_param; f = @(x) x.^2;
[fappx, out_param] = funappx_g(f,'a',-10,'b',50,'nmax',1e6,'abstol',1e-7)

% Approximate function x^2 with error tolerance 1e-7, cost budget 1000000,
% lower bound of initial number of points 10 and upper
% bound of initial number of points 100
```

```
fappx =

griddedInterpolant with properties:

    GridVectors: {[1x474625 double]}
      Values: [1x474625 double]
    Method: 'linear'
  ExtrapolationMethod: 'linear'
```

```
out_param =

    a: -10
  abstol: 1.0000e-07
    b: 50
    f: @(x)x.^2
maxiter: 1000
```

```
    nhi: 1000
    nlo: 10
    nmax: 1000000
    ninit: 928
    exit: [2x1 logical]
    iter: 10
    npoints: 474625
    errest: 6.0849e-08
    nstar: [1x512 double]
```

2.5 See Also

interp1, griddedInterpolant, integral_g, funmin_g, meanMC_g, cubMC_g

3 funmin_g

1-D guaranteed global minimum value on $[a,b]$ and the subset containing optimal solutions

3.1 Syntax

`fmin = funmin_g(f)`

`fmin = funmin_g(f,a,b,abstol,TolX)`

`fmin = funmin_g(f,'a',a,'b',b,'abstol',abstol,'TolX',TolX)`

`fmin = funmin_g(f,in_param)`

`[fmin, out_param] = funmin_g(f,...)`

3.2 Description

`fmin = funmin_g(f)` finds minimum value of function `f` on the default interval $[0,1]$ within the guaranteed absolute error tolerance of $1e-6$ and the X tolerance of $1e-3$. Default initial number of points is 100 and default cost budget is $1e7$. Input `f` is a function handle.

`fmin = funmin_g(f,a,b,abstol,TolX)` finds minimum value of function `f` with ordered input parameters that define the finite interval $[a,b]$, a guaranteed absolute error tolerance `abstol` and a guaranteed X tolerance `TolX`.

`fmin = funmin_g(f,'a',a,'b',b,'abstol',abstol,'TolX',TolX)` finds minimum value of function `f` on the interval $[a,b]$ with a guaranteed absolute error tolerance `abstol` and a guaranteed X tolerance `TolX`. All five field-value pairs are optional and can be supplied in different order.

`fmin = funmin_g(f,in_param)` finds minimum value of function `f` on the interval $[in_param.a, in_param.b]$ with a guaranteed absolute error tolerance `in_param.abstol` and a guaranteed X tolerance `in_param.TolX`. If a field is not specified, the default value is used.

`[fmin, out_param] = funmin_g(f,...)` returns minimum value `fmin` of function `f` and an output structure `out_param`.

Input Arguments

- `f` — input function
- `in_param.a` — left end point of interval, default value is 0
- `in_param.b` — right end point of interval, default value is 1
- `in_param.abstol` — guaranteed absolute error tolerance, default value is $1e-6$.
- `in_param.TolX` — guaranteed X tolerance, default value is $1e-3$.

Optional Input Arguments

- `in_param.nlo` — lower bound of initial number of points we used, default value is 10
- `in_param.nhi` — upper bound of initial number of points we used, default value is 1000

- `in_param.nmax` — cost budget, default value is `1e7`.

Output Arguments

- `out_param.f` — input function
- `out_param.a` — left end point of interval
- `out_param.b` — right end point of interval
- `out_param.abstol` — guaranteed absolute error tolerance
- `out_param.TolX` — guaranteed X tolerance
- `out_param.nlo` — a lower bound of initial number of points we use
- `out_param.nhi` — an upper bound of initial number of points we use
- `out_param.nmax` — cost budget
- `out_param.ninit` — initial number of points we use
- `out_param.tau` — latest value of tau
- `out_param.npoints` — number of points needed to reach the guaranteed absolute error tolerance or the guaranteed X tolerance
- `out_param.exitflag` — the state of program when exiting
 - 0 Success
 - 1 Number of points used is greater than `out_param.nmax`
- `out_param.errest` — estimation of the absolute error bound
- `out_param.volumeX` — the volume of intervals containing the point(s) where the minimum occurs
- `out_param.tauchange` — it is 1 if `out_param.tau` changes, otherwise it is 0
- `out_param.intervals` — the intervals containing point(s) where the minimum occurs. Each column indicates one interval where the first row is the left point and the second row is the right point.

3.3 Guarantee

If the function to be minimized, f satisfies the cone condition

$$\|f''\|_{\infty} \leq \frac{\tau}{b-a} \left\| f' - \frac{f(b) - f(a)}{b-a} \right\|_{\infty},$$

then the `fmin` output by this algorithm is guaranteed to satisfy

$$|\min f - \text{fmin}| \leq \text{abstol},$$

or

$$\text{volumeX} \leq \text{TolX},$$

provided the flag `exitflag = 0`.

3.4 Examples

Example 1

```
f=@(x) (x-0.3).^2+1; [fmin,out_param] = funmin_g(f)

% Minimize function (x-0.3)^2+1 with default input parameter.

fmin =

    1.0000

out_param =

    f: @(x)(x-0.3).^2+1
    a: 0
    b: 1
    abstol: 1.0000e-06
    TolX: 1.0000e-03
    nlo: 10
    nhi: 1000
    nmax: 10000000
    ninit: 100
    tau: 197
    exitflag: 0
    npoints: 6337
    errest: 6.1554e-07
    volumeX: 0.0015
    tauchange: 0
    intervals: [2x1 double]
```

Example 2

```
f=@(x) (x-0.3).^2+1;
[fmin,out_param] = funmin_g(f,-2,2,1e-7,1e-4,10,10,1000000)

% Minimize function (x-0.3)^2+1 on [-2,2] with error tolerance 1e-4, X
% tolerance 1e-2, cost budget 1000000, lower bound of initial number of
% points 10 and upper bound of initial number of points 10

fmin =

    1.0000

out_param =

    a: -2
    abstol: 1.0000e-07
    b: 2
    f: @(x)(x-0.3).^2+1
```

```

        nhi: 10
        nlo: 10
        nmax: 1000000
        TolX: 1.0000e-04
        ninit: 10
        tau: 17
    exitflag: 0
    npoints: 18433
    errest: 9.5464e-08
    volumeX: 5.4175e-04
    tauchange: 0
    intervals: [2x1 double]

```

Example 3

```

clear in_param; in_param.a = -13; in_param.b = 8;
in_param.abstol = 1e-7; in_param.TolX = 1e-4;
in_param.nlo = 10; in_param.nhi = 100;
in_param.nmax = 10^6;
[fmin,out_param] = funmin_g(f,in_param)

% Minimize function (x-0.3)^2+1 on [-13,8] with error tolerance 1e-7, X
% tolerance 1e-4, cost budget 1000000, lower bound of initial number of
% points 10 and upper bound of initial number of points 100

fmin =

    1

out_param =

        a: -13
    abstol: 1.0000e-07
         b: 8
         f: @(x)(x-0.3).^2+1
        nhi: 100
        nlo: 10
        nmax: 1000000
        TolX: 1.0000e-04
        ninit: 91
        tau: 179
    exitflag: 0
    npoints: 368641
    errest: 7.1014e-08
    volumeX: 5.2445e-04
    tauchange: 0
    intervals: [2x1 double]

```

Example 4

```
f=@(x) (x-0.3).^2+1;
[fmin,out_param] = funmin_g(f,'a',-2,'b',2,'nhi',100,'nlo',10,...
    'nmax',1e6,'abstol',1e-4,'TolX',1e-2)

% Minimize function (x-0.3)^2+1 on [-2,2] with error tolerance 1e-4, X
% tolerance 1e-2, cost budget 1000000, lower bound of initial number of
% points 10 and upper bound of initial number of points 100

fmin =

    1.0000

out_param =

    a: -2
  abstol: 1.0000e-04
    b: 2
    f: @(x)(x-0.3).^2+1
   nhi: 100
   nlo: 10
  nmax: 1000000
  TolX: 0.0100
  ninit: 64
   tau: 125
exitflag: 0
 npoints: 2017
  errest: 6.2273e-05
volumeX: 0.0146
tauchange: 0
intervals: [2x1 double]
```

3.5 See Also

fminbnd, funappx_g, integral_g

4 `integral_g`

1-D guaranteed function integration using trapezoidal rule

4.1 Syntax

`q = integral_g(f)`

`q = integral_g(f,a,b,abstol)`

`q = integral_g(f,'a',a,'b',b,'abstol',abstol)`

`q = integral_g(f,in_param)`

`[q, out_param] = integral_g(f,...)`

4.2 Description

`q = integral_g(f)` computes `q`, the definite integral of function `f` on the interval `[a,b]` by trapezoidal rule with in a guaranteed absolute error of `1e-6`. Default starting number of sample points taken is 100 and default cost budget is `1e7`. Input `f` is a function handle. The function `y = f(x)` should accept a vector argument `x` and return a vector result `y`, the integrand evaluated at each element of `x`.

`q = integral_g(f,a,b,abstol)` computes `q`, the definite integral of function `f` on the finite interval `[a,b]` by trapezoidal rule with the ordered input parameters, and guaranteed absolute error tolerance `abstol`.

`q = integral_g(f,'a',a,'b',b,'abstol',abstol)` computes `q`, the definite integral of function `f` on the finite interval `[a,b]` by trapezoidal rule within a guaranteed absolute error tolerance `abstol`. All four field-value pairs are optional and can be supplied.

`q = integral_g(f,in_param)` computes `q`, the definite integral of function `f` by trapezoidal rule within a guaranteed absolute error `in_param.abstol`. If a field is not specified, the default value is used.

`[q, out_param] = integral_g(f,...)` returns the approximated integration `q` and output structure `out_param`.

Input Arguments

- `f` — input function
- `in_param.a` — left end of the integral, default value is 0
- `in_param.b` — right end of the integral, default value is 1
- `in_param.abstol` — guaranteed absolute error tolerance, default value is `1e-6`

Optional Input Arguments

- `in_param.nlo` — lowest initial number of function values used, default value is 10
- `in_param.nhi` — highest initial number of function values used, default value is 1000
- `in_param.nmax` — cost budget (maximum number of function values), default value is `1e7`
- `in_param.maxiter` — max number of iterations, default value is 1000

Output Arguments

- `q` — approximated integral
- `out_param.f` — input function
- `out_param.a` — low end of the integral
- `out_param.b` — high end of the integral
- `out_param.abstol` — guaranteed absolute error tolerance
- `out_param.nlo` — lowest initial number of function values
- `out_param.nhi` — highest initial number of function values
- `out_param.nmax` — cost budget (maximum number of function values)
- `out_param.maxiter` — max number of iterations
- `out_param.ninit` — initial number of points we use, computed by `nlo` and `nhi`
- `out_param.tauchange` — it is true if the cone constant has been changed, false otherwise. See [1] for details. If true, you may wish to change the input `in_param.ninit` to a larger number.
- `out_param.tauchange` — it is true if the cone constant has been changed, false otherwise. See [1] for details. If true, you may wish to change the input `in_param.ninit` to a larger number.
- `out_param.iter` — number of iterations
- `out_param.npoints` — number of points we need to reach the guaranteed absolute error tolerance `abstol`.
- `out_param.errest` — approximation error defined as the differences between the true value and the approximated value of the integral.
- `out_param.nstar` — final value of the parameter defining the cone of functions for which this algorithm is guaranteed; `nstar = ninit-2` initially and is increased as necessary
- `out_param.exit` — the state of program when exiting
 - 0 Success
 - 1 Number of points used is greater than `out_param.nmax`
 - 2 Number of iterations is greater than `out_param.maxiter`

4.3 Guarantee

If the function to be integrated, f satisfies the cone condition

$$\|f''\|_1 \leq \frac{nstar}{2(b-a)} \left\| f' - \frac{f(b) - f(a)}{b-a} \right\|_1,$$

then the q output by this algorithm is guaranteed to satisfy

$$\left\| \int_a^b f(x)dx - q \right\|_1 \leq abstol,$$

provided the flag `exceedbudget = 0`. And the upper bound of the cost is

$$\sqrt{\frac{nstar * (b-a)^2 \text{Var}(f')}{2 \times abstol}} + 2 \times nstar + 4.$$

4.4 Examples

Example 1

```
f = @(x) x.^2; [q, out_param] = integral_g(f)

% Integrate function x with default input parameter to make the error less
% than 1e-7.

q =

    0.3333

out_param =

    f: @(x)x.^2
    a: 0
    b: 1
    abstol: 1.0000e-06
    nlo: 10
    nhi: 1000
    nmax: 10000000
    maxiter: 1000
    ninit: 100
    tau: 197
    exceedbudget: 0
    tauchange: 0
    iter: 2
    q: 0.3333
    npoints: 3565
    errest: 9.9688e-07
```

Example 2

```
[q, out_param] = integral_g(@(x) exp(-x.^2), 'a', 1, 'b', 2, ...
    'nlo', 100, 'nhi', 10000, 'abstol', 1e-5, 'nmax', 1e7)

% Integrate function x^2 with starting number of points 52, cost budget
% 10000000 and error tolerance 1e-8

q =

    0.1353

out_param =

    a: 1
    abstol: 1.0000e-05
    b: 2
    f: @(x)exp(-x.^2)
    maxiter: 1000
```

```
      nhi: 10000
      nlo: 100
      nmax: 10000000
      ninit: 1000
      tau: 1997
exceedbudget: 0
  tauchange: 0
    iter: 2
      q: 0.1353
  npoints: 2998
    errest: 7.3718e-06
```

4.5 See Also

integral, quad, funappx_g, meanMC_g, cubMC_g, funmin_g

5 meanMC_g

Monte Carlo method to estimate the mean of a random variable

5.1 Syntax

```
tmu = meanMC_g(Yrand)
```

```
tmu = meanMC_g(Yrand,abstol,reltol,alpha)
```

```
tmu = meanMC_g(Yrand,'abstol',abstol,'reltol',reltol,'alpha',alpha)
```

```
[tmu, out_param] = meanMC_g(Yrand,in_param)
```

5.2 Description

`tmu = meanMC_g(Yrand)` estimates the mean, μ , of a random variable Y to within a specified generalized error tolerance, `tolfun:=max(abstol,reltol*| μ |)`, i.e., $|\mu - \text{tmu}| \leq \text{tolfun}$ with probability at least $1-\alpha$, where `abstol` is the absolute error tolerance, and `reltol` is the relative error tolerance. Usually the `reltol` determines the accuracy of the estimation, however, if the $|\mu|$ is rather small, the `abstol` determines the accuracy of the estimation. The default values are `abstol=1e-2`, `reltol=1e-1`, and `alpha=1%`. Input `Yrand` is a function handle that accepts a positive integer input n and returns an $n \times 1$ vector of IID instances of the random variable Y .

`tmu = meanMC_g(Yrand,abstol,reltol,alpha)` estimates the mean of a random variable Y to within a specified generalized error tolerance `tolfun` with guaranteed confidence level $1-\alpha$ using all ordered parsing inputs `abstol`, `reltol`, `alpha`.

`tmu = meanMC_g(Yrand,'abstol',abstol,'reltol',reltol,'alpha',alpha)` estimates the mean of a random variable Y to within a specified generalized error tolerance `tolfun` with guaranteed confidence level $1-\alpha$. All the field-value pairs are optional and can be supplied in different order, if a field is not supplied, the default value is used.

`[tmu, out_param] = meanMC_g(Yrand,in_param)` estimates the mean of a random variable Y to within a specified generalized error tolerance `tolfun` with the given parameters `in_param` and produce the estimated mean `tmu` and output parameters `out_param`. If a field is not specified, the default value is used.

Input Arguments

- `Yrand` — the function for generating n IID instances of a random variable Y whose mean we want to estimate. Y is often defined as a function of some random variable X with a simple distribution. The input of `Yrand` should be the number of random variables n , the output of `Yrand` should be n function values. For example, if $Y = X.^2$ where X is a standard uniform random variable, then one may define `Yrand = @(n) rand(n,1).^2`.
- `in_param.abstol` — the absolute error tolerance, which should be positive, default value is `1e-2`.
- `in_param.reltol` — the relative error tolerance, which should be between 0 and 1, default value is `1e-1`.
- `in_param.alpha` — the uncertainty, which should be a small positive percentage. default value is `1%`.

Optional Input Arguments

- `in_param.fudge` — standard deviation inflation factor, which should be larger than 1, default value is `1.2`.

- `in_param.nSig` — initial sample size for estimating the sample variance, which should be a moderate large integer at least 30, the default value is 1e4.
- `in_param.n1` — initial sample size for estimating the sample mean, which should be a moderate large positive integer at least 30, the default value is 1e4.
- `in_param.tbudget` — the time budget in seconds to do the two-stage estimation, which should be positive, the default value is 100 seconds.
- `in_param.nbudget` — the sample budget to do the two-stage estimation, which should be a large positive integer, the default value is 1e9.

Output Arguments

- `tmu` — the estimated mean of Y .
- `out_param.tau` — the iteration step.
- `out_param.n` — the sample size used in each iteration.
- `out_param.nremain` — the remaining sample budget to estimate μ . It was calculated by the sample left and time left.
- `out_param.ntot` — total sample used.
- `out_param.hmu` — estimated mean in each iteration.
- `out_param.tol` — the reliable upper bound on error for each iteration.
- `out_param.var` — the sample variance.
- `out_param.exit` — the state of program when exiting.
 - 0 Success
 - 1 Not enough samples to estimate the mean
- `out_param.kurtmax` — the upper bound on modified kurtosis.
- `out_param.time` — the time elapsed in seconds.
- `out_param.flag` — parameter checking status
 - 1 checked by meanMC_g

5.3 Guarantee

This algorithm attempts to calculate the mean, μ , of a random variable to a prescribed error tolerance, $\text{tolfun} := \max(\text{abstol}, \text{reitol} * |\mu|)$, with guaranteed confidence level $1 - \alpha$. If the algorithm terminated without showing any warning messages and provide an answer tmu , then the follow inequality would be satisfied: $\Pr(|\mu - \text{tmu}| \leq \text{tolfun}) \geq 1 - \alpha$. The cost of the algorithm, N_{tot} , is also bounded above

by N_{up} , which is defined in terms of abstol , reitol , nSig , n1 , fudge , kurtmax , β . And the following inequality holds: $\Pr(N_{\text{tot}} \leq N_{\text{up}}) \geq 1 - \beta$. Please refer to our paper for detailed arguments and proofs.

5.4 Examples

Example 1

```
% Calculate the mean of  $x^2$  when  $x$  is uniformly distributed in  
% [0 1], with the absolute error tolerance =  $1e-3$  and uncertainty 5%.
```

```
in_param.reltol=0; in_param.abstol =  $1e-3$ ; in_param.reltol = 0;  
in_param.alpha = 0.05; Yrand=@(n) rand(n,1).^2;  
tmu = meanMC_g(Yrand,in_param)
```

```
tmu =
```

```
0.3331
```

Example 2

```
% Calculate the mean of  $\exp(x)$  when  $x$  is uniformly distributed in  
% [0 1], with the absolute error tolerance  $1e-3$ .
```

```
tmu = meanMC_g(@(n)exp(rand(n,1)), $1e-3$ ,0)
```

```
tmu =
```

```
1.7185
```

Example 3

```
% Calculate the mean of  $\cos(x)$  when  $x$  is uniformly distributed in  
% [0 1], with the relative error tolerance  $1e-2$  and uncertainty 0.05.
```

```
tmu = meanMC_g(@(n)cos(rand(n,1)),'reltol', $1e-2$ ,'abstol',0,...  
    'alpha',0.05)
```

```
tmu =
```

```
0.8415
```

5.5 See Also

funappx_g, integral_g, cubMC_g, meanMCBer_g, cubSobol_g, cubLattice_g

6 meanMCBer_g

Monte Carlo method to estimate the mean of a Bernoulli random variable to within a specified absolute error tolerance with guaranteed confidence level $1-\alpha$.

6.1 Syntax

```
pHat = meanMCBer_g(Yrand)
```

```
pHat = meanMCBer_g(Yrand,abstol,alpha,nmax)
```

```
pHat = meanMCBer_g(Yrand,'abstol',abstol,'alpha',alpha,'nmax',nmax)
```

```
[pHat, out_param] = meanMCBer_g(Yrand,in_param)
```

6.2 Description

`pHat = meanMCBer_g(Yrand)` estimates the mean of a Bernoulli random variable Y to within a specified absolute error tolerance with guaranteed confidence level 99%. Input `Yrand` is a function handle that accepts a positive integer input n and returns a $n \times 1$ vector of IID instances of the Bernoulli random variable Y .

`pHat = meanMCBer_g(Yrand,abstol,alpha,nmax)` estimates the mean of a Bernoulli random variable Y to within a specified absolute error tolerance with guaranteed confidence level $1-\alpha$ using all ordered parsing inputs `abstol`, `alpha` and `nmax`.

`pHat = meanMCBer_g(Yrand,'abstol',abstol,'alpha',alpha,'nmax',nmax)` estimates the mean of a Bernoulli random variable Y to within a specified absolute error tolerance with guaranteed confidence level $1-\alpha$. All the field-value pairs are optional and can be supplied in different order.

`[pHat, out_param] = meanMCBer_g(Yrand,in_param)` estimates the mean of a Bernoulli random variable Y to within a specified absolute error tolerance with the given parameters `in_param` and produce the estimated mean `pHat` and output parameters `out_param`.

Input Arguments

- `Yrand` — the function for generating IID instances of a Bernoulli random variable Y whose mean we want to estimate.
- `pHat` — the estimated mean of Y .
- `in_param.abstol` — the absolute error tolerance, the default value is $1e-2$.
- `in_param.alpha` — the uncertainty, the default value is 1%.
- `in_param.nmax` — the sample budget, the default value is $1e9$.

Output Arguments

- `out_param.n` — the total sample used.
- `out_param.time` — the time elapsed in seconds.
- `out_param.exit` — the state of program when exiting.
 - 0 Success
 - 1 Not enough samples to estimate p with guarantee

6.3 Guarantee

If the sample size is calculated according Hoeffding's inequality, which equals to $\text{ceil}(\log(2/\text{out_param.alpha})/(2*\text{out_param.abstol}^2))$, then the following inequality must be satisfied:
 $\Pr(|p - \text{pHat}| \leq \text{abstol}) \geq 1 - \alpha$. Here p is the true mean of Y_{rand} , and pHat is the output of

`MEANMCBER_G`. Also, the cost is deterministic.

6.4 Examples

Example 1

```
% Calculate the mean of a Bernoulli random variable with true p=1/90,  
% absolute error tolerance 1e-3 and uncertainty 0.01.  
  
in_param.abstol = 1e-3; in_param.alpha = 0.01; in_param.nmax = 1e9;  
p=1/9; Yrand=@(n) rand(n,1)<p;  
pHat = meanMCBer_g(Yrand,in_param)
```

```
pHat =  
  
0.1113
```

Example 2

```
% Using the same function as example 1, with the absolute error tolerance  
% 1e-4.  
  
pHat = meanMCBer_g(Yrand,1e-4)
```

```
pHat =  
  
0.1111
```

Example 3

```
% Using the same function as example 1, with the absolute error tolerance  
% 1e-2 and uncertainty 0.05.  
  
pHat = meanMCBer_g(Yrand,'abstol',1e-2,'alpha',0.05)
```

```
pHat =  
  
0.1118
```

6.5 See Also

`funappx_g`, `integral_g`, `cubMC_g`, `meanMC_g`, `cubLattice_g`, `cubSobol_g`

7 cubMC_g

Monte Carlo method to evaluate a multidimensional integral

7.1 Syntax

```
[Q,out_param] = cubMC_g(f,hyperbox)
```

```
Q = cubMC_g(f,hyperbox,measure,abstol,reltol,alpha)
```

```
Q = cubMC_g(f,hyperbox,'measure',measure,'abstol',abstol,'reltol',reltol,'alpha',alpha)
```

```
[Q out_param] = cubMC_g(f,hyperbox,in_param)
```

7.2 Description

`[Q,out_param] = cubMC_g(f,hyperbox)` estimates the integral of f over hyperbox to within a specified generalized error tolerance, $\text{tolfun} = \max(\text{abstol}, \text{reltol} * |I|)$, i.e., $|I - Q| \leq \text{tolfun}$ with probability at least $1 - \alpha$, where abstol is the absolute error tolerance, and reltol is the relative error tolerance. Usually the reltol determines the accuracy of the estimation, however, if the $|I|$ is rather small, the abstol determines the accuracy of the estimation. The default values are $\text{abstol}=1\text{e-}2$, $\text{reltol}=1\text{e-}1$, and $\alpha=1\%$. Input f is a function handle that accepts an $n \times d$ matrix input, where d is the dimension of the hyperbox, and n is the number of points being evaluated simultaneously. The input hyperbox is a $2 \times d$ matrix, where the first row corresponds to the lower limits and the second row corresponds to the upper limits.

`Q = cubMC_g(f,hyperbox,measure,abstol,reltol,alpha)` estimates the integral of function f over hyperbox to within a specified generalized error tolerance tolfun with guaranteed confidence level $1 - \alpha$ using all ordered parsing inputs f , hyperbox , measure , abstol , reltol , α , fudge , nSig , n1 , tbudget , nbudget , flag . The input f and hyperbox are required and others are optional.

`Q = cubMC_g(f,hyperbox,'measure',measure,'abstol',abstol,'reltol',reltol,'alpha',alpha)` estimates the integral of f over hyperbox to within a specified generalized error tolerance tolfun with guaranteed confidence level $1 - \alpha$. All the field-value pairs are optional and can be supplied in different order. If an input is not specified, the default value is used.

`[Q out_param] = cubMC_g(f,hyperbox,in_param)` estimates the integral of f over hyperbox to within a specified generalized error tolerance tolfun with the given parameters in_param and produce output parameters out_param and the integral Q .

Input Arguments

- f — the integrand.
- hyperbox — the integration hyperbox. The default value is `[zeros(1,d); ones(1,d)]`, the default d is 1.
- in_param.measure — the measure for generating the random variable, the default is 'uniform'. The other measure could be handled is 'normal'/'Gaussian'. The input should be a string type, hence with quotes.
- in_param.abstol — the absolute error tolerance, the default value is $1\text{e-}2$.
- in_param.reltol — the relative error tolerance, the default value is $1\text{e-}1$.
- in_param.alpha — the uncertainty, the default value is 1% .

Optional Input Arguments

- `in_param.fudge` — the standard deviation inflation factor, the default value is 1.2.
- `in_param.nSig` — initial sample size for estimating the sample variance, which should be a moderate large integer at least 30, the default value is 1e4.
- `in_param.n1` — initial sample size for estimating the sample mean, which should be a moderate large positive integer at least 30, the default value is 1e4.
- `in_param.tbudget` — the time budget to do the estimation, the default value is 100 seconds.
- `in_param.nbudget` — the sample budget to do the estimation, the default value is 1e9.
- `in_param.flag` — the value corresponds to parameter checking status.
 - 0 not checked
 - 1 checked by meanMC_g
 - 2 checked by cubMC_g

Output Arguments

- `Q` — the estimated value of the integral.
- `out_param.n` — the sample size used in each iteration.
- `out_param.ntot` — total sample used.
- `out_param.nremain` — the remaining sample budget to estimate I. It was calculated by the sample left and time left.
- `out_param.tau` — the iteration step.
- `out_param.hmu` — estimated integral in each iteration.
- `out_param.tol` — the reliable upper bound on error for each iteration.
- `out_param.kurtmax` — the upper bound on modified kurtosis.
- `out_param.time` — the time elapsed in seconds.
- `out_param.var` — the sample variance.
- `out_param.exit` — the state of program when exiting.
 - 0 success
 - 1 Not enough samples to estimate the mean
 - 10 hyperbox does not contain numbers
 - 11 hyperbox is not 2 x d
 - 12 hyperbox is only a point in one direction
 - 13 hyperbox is infinite when measure is 'uniform'
 - 14 hyperbox is not doubly infinite when measure is 'normal'

7.3 Guarantee

This algorithm attempts to calculate the integral of function f over a hyperbox to a prescribed error tolerance $\text{tolfun} := \max(\text{abstol}, \text{reltol} * |I|)$ with guaranteed confidence level $1-\alpha$. If the algorithm terminated without showing any warning messages and provide an answer Q , then the follow inequality would be satisfied:

$$\Pr(|Q - I| \leq \text{tolfun}) \geq 1-\alpha$$

The cost of the algorithm, N_{tot} , is also bounded above by N_{up} , which is a function in terms of abstol , reltol , n_{Sig} , n_1 , fudge , kurtmax , β . And the following inequality holds:

$$\Pr(N_{\text{tot}} \leq N_{\text{up}}) \geq 1-\beta$$

Please refer to our paper for detailed arguments and proofs.

7.4 Examples

Example 1

```
% Estimate the integral with integrand  $f(x) = \sin(x)$  over the interval  
% [1;2]
```

```
f = @(x) sin(x); interval = [1;2];  
Q = cubMC_g(f,interval,'uniform',1e-3,1e-2)
```

```
Q =
```

```
0.9564
```

Example 2

```
% Estimate the integral with integrand  $f(x) = \exp(-x_1^2 - x_2^2)$  over the  
% hyperbox [0 0;1 1], where  $x$  is a vector  $x = [x_1 \ x_2]$ .
```

```
f = @(x) exp(-x(:,1).^2-x(:,2).^2); hyperbox = [0 0;1 1];  
Q = cubMC_g(f,hyperbox,'measure','uniform','abstol',1e-3,...  
    'reltol',1e-13)
```

```
Q =
```

```
0.5574
```

Example 3

```
% Estimate the integral with integrand  $f(x) = 2^d \cdot \prod(x_1 x_2 \dots x_d) +$   
% 0.555 over the hyperbox  $[\text{zeros}(1,d); \text{ones}(1,d)]$ , where  $x$  is a vector  $x =$   
%  $[x_1 \ x_2 \dots x_d]$ .
```

```
d = 3; f = @(x) 2^d*prod(x,2)+0.555; hyperbox = [zeros(1,d);ones(1,d)];  
in_param.abstol = 1e-3; in_param.reltol=1e-3;  
Q = cubMC_g(f,hyperbox,in_param)
```

Q =

1.5549

Example 4

% Estimate the integral with integrand $f(x) = \exp(-x_1^2 - x_2^2)$ in the
% hyperbox $[-\inf -\inf; \inf \inf]$, where x is a vector $x = [x_1 \ x_2]$.

```
f = @(x) exp(-x(:,1).^2-x(:,2).^2); hyperbox = [-inf -inf;inf inf];  
Q = cubMC_g(f,hyperbox,'normal',0,1e-2)
```

Q =

0.3328

7.5 See Also

funappx_g, integral_g, meanMC_g, meanMCBer_g, cubLattice_g, cubSobol_g

8 cubLattice_g

Quasi-Monte Carlo method using rank-1 Lattices cubature over a d-dimensional region to integrate within a specified generalized error tolerance with guarantees under Fourier coefficients cone decay assumptions.

8.1 Syntax

```
[q,out_param] = cubLattice_g(f,hyperbox)
```

```
q = cubLattice_g(f,hyperbox,measure,abstol,reltol)
```

```
q = cubLattice_g(f,hyperbox,'measure',measure,'abstol',abstol,'reltol',reltol)
```

```
q = cubLattice_g(f,hyperbox,in_param)
```

8.2 Description

`[q,out_param] = cubLattice_g(f,hyperbox)` estimates the integral of f over the d-dimensional region described by `hyperbox`, and with an error guaranteed not to be greater than a specific generalized error tolerance, `tolfun:=max(abstol,reltol*| integral(f) |)`. Input f is a function handle. f should accept an $n \times d$ matrix input, where d is the dimension and n is the number of points being evaluated simultaneously. The input `hyperbox` is a $2 \times d$ matrix, where the first row corresponds to the lower limits and the second row corresponds to the upper limits of the integral. Given the construction of our Lattices, d must be a positive integer with $1 \leq d \leq 250$.

`q = cubLattice_g(f,hyperbox,measure,abstol,reltol)` estimates the integral of f over the hyperbox. The answer is given within the generalized error tolerance `tolfun`. All parameters should be input in the order specified above. If an input is not specified, the default value is used. Note that if an input is not specified, the remaining tail cannot be specified either. Inputs f and `hyperbox` are required. The other optional inputs are in the correct order: `measure,abstol,reltol,shift,mmin,mmax,fudge,transform,toltype` and `theta`.

`q = cubLattice_g(f,hyperbox,'measure',measure,'abstol',abstol,'reltol',reltol)` estimates the integral of f over the hyperbox. The answer is given within the generalized error tolerance `tolfun`. All the field-value pairs are optional and can be supplied in any order. If an input is not specified, the default value is used.

`q = cubLattice_g(f,hyperbox,in_param)` estimates the integral of f over the hyperbox. The answer is given within the generalized error tolerance `tolfun`.

Input Arguments

- f — the integrand whose input should be a matrix $n \times d$ where n is the number of data points and d the dimension, which cannot be greater than 250. By default f is $f=@ x.^2$.
- `hyperbox` — the integration region defined by its bounds. It must be a $2 \times d$ matrix, where the first row corresponds to the lower limits and the second row corresponds to the upper limits of the integral. The default value is `[0;1]`.
- `in_param.measure` — for $f(x)*\mu(dx)$, we can define $\mu(dx)$ to be the measure of a uniformly distributed random variable in they hyperbox or normally distributed with covariance matrix I_d . The only possible values are 'uniform' or 'normal'. For 'uniform', the hyperbox must be a finite volume while for 'normal', the hyperbox can only be defined as $(-\text{Inf},\text{Inf})^d$. By default it is 'uniform'.
- `in_param.abstol` — the absolute error tolerance, $\text{abstol} \geq 0$. By default it is $1e-4$.
- `in_param.reltol` — the relative error tolerance, which should be in $[0,1]$. Default value is $1e-2$.

Optional Input Arguments

- `in_param.shift` — the Rank-1 lattices can be shifted to avoid the origin or other particular points. By default we consider a uniformly $[0,1)$ random shift.
- `in_param.mmin` — the minimum number of points to start is $2^{\wedge}mmin$. The cone condition on the Fourier coefficients decay requires a minimum number of points to start. The advice is to consider at least $mmin=10$. $mmin$ needs to be a positive integer with $mmin \leq mmax$. By default it is 10.
- `in_param.mmax` — the maximum budget is $2^{\wedge}mmax$. By construction of our Lattices generator, $mmax$ is a positive integer such that $mmin \leq mmax \leq 26$. The default value is 24.
- `in_param.fudge` — the positive function multiplying the finite sum of Fast Fourier coefficients specified in the cone of functions. This input is a function handle. The fudge should accept an array of nonnegative integers being evaluated simultaneously. For more technical information about this parameter, refer to the references. By default it is $@(m) 5*2.^{-m}$.
- `in_param.transform` — the algorithm is defined for continuous periodic functions. If the input function f is not, there are 5 types of transform to periodize it without modifying the result. By default it is the Baker's transform. The options are:
 - 'id' : no transformation.
 - 'Baker' : Baker's transform or tent map in each coordinate. Preserving only continuity but simple to compute. Chosen by default.
 - 'C0' : polynomial transformation only preserving continuity.
 - 'C1' : polynomial transformation preserving the first derivative.
 - 'C1sin' : Sidi's transform with sine, preserving the first derivative. This is in general a better option than 'C1'.
- `in_param.toltype` — this is the generalized tolerance function. There are two choices, 'max' which takes $\max(abstol, reltol * | \text{integral}(f) |)$ and 'comb' which is the linear combination $\theta * abstol + (1 - \theta) * reltol * | \text{integral}(f) |$. θ is another parameter to be specified with 'comb' (see below). For pure absolute error, either choose 'max' and set $reltol = 0$ or choose 'comb' and set $\theta = 1$. For pure relative error, either choose 'max' and set $abstol = 0$ or choose 'comb' and set $\theta = 0$. Note that with 'max', the user can not input $abstol = reltol = 0$ and with 'comb', if $\theta = 1$ $abstol$ can not be 0 while if $\theta = 0$, $reltol$ can not be 0. By default `toltype` is 'max'.
- `in_param.theta` — this input is parametrizing the `toltype` 'comb'. Thus, it is only active when the `toltype` chosen is 'comb'. It establishes the linear combination weight between the absolute and relative tolerances $\theta * abstol + (1 - \theta) * reltol * | \text{integral}(f) |$. Note that for $\theta = 1$, we have pure absolute tolerance while for $\theta = 0$, we have pure relative tolerance. By default, $\theta=1$.

Output Arguments

- `q` — the estimated value of the integral.
- `out_param.d` — dimension over which the algorithm integrated.
- `out_param.n` — number of Rank-1 lattice points used for computing the integral of f .
- `out_param.bound.err` — predicted bound on the error based on the cone condition. If the function lies in the cone, the real error will be smaller than generalized tolerance.
- `out_param.time` — time elapsed in seconds when calling `cubLattice_g`.

- `out_param.exitflag` — this is a binary vector stating whether warning flags arise. These flags tell about which conditions make the final result certainly not guaranteed. One flag is considered arisen when its value is 1. The following list explains the flags in the respective vector order:
 - 1 If reaching overbudget. It states whether the max budget is attained without reaching the guaranteed error tolerance.
 - 2 If the function lies outside the cone. In this case, results are not guaranteed. Note that this parameter is computed on the transformed function, not the input function. For more information on the transforms, check the input parameter `in_param.transform`; for information about the cone definition, check the article mentioned below.

8.3 Guarantee

This algorithm computes the integral of real valued functions in dimension d with a prescribed generalized error tolerance. The Fourier coefficients of the integrand are assumed to be absolutely convergent. If the algorithm terminates without warning messages, the output is given with guarantees under the assumption that the integrand lies inside a cone of functions. The guarantee is based on the decay rate of the Fourier coefficients. For more details on how the cone is defined, please refer to the references below.

8.4 Examples

Example 1

```
% Estimate the integral with integrand f(x) = x1.*x2 in the interval
% [0,1]^2:

f = @(x) prod(x,2); hyperbox = [zeros(1,2);ones(1,2)];
q = cubLattice_g(f,hyperbox,'uniform',1e-5,0,'transform','C1sin')

q =

    0.2500
```

Example 2

```
% Estimate the integral with integrand f(x) = x1.^2.*x2.^2.*x3.^2
% in the interval R^3 where x1, x2 and x3 are normally distributed:

f = @(x) x(:,1).^2.*x(:,2).^2.*x(:,3).^2; hyperbox = [-inf(1,3);inf(1,3)];
q = cubLattice_g(f,hyperbox,'normal',1e-3,1e-3,'transform','C1sin')

q =

    1.0000
```

Example 3

```
% Estimate the integral with integrand f(x) = exp(-x1^2-x2^2) in the
% interval [-1,2]^2:

f = @(x) exp(-x(:,1).^2-x(:,2).^2); hyperbox = [-ones(1,2);2*ones(1,2)];
q = cubLattice_g(f,hyperbox,'uniform',1e-3,1e-2,'transform','C1')
```

```
q =
    2.6532
```

Example 4

```
% Estimate the price of an European call with S0=100, K=100, r=sigma^2/2,
% sigma=0.05 and T=1.

f = @(x) exp(-0.05^2/2)*max(100*exp(0.05*x)-100,0); hyperbox = [-inf(1,1);inf(1,1)];
q = cubLattice_g(f,hyperbox,'normal',1e-4,1e-2,'transform','C1sin')
```

```
q =
    2.0563
```

Example 5

```
% Estimate the integral with integrand f(x) = 8*x1.*x2.*x3.*x4.*x5 in the
% interval [0,1)^5 with pure absolute error 1e-5.

f = @(x) 8*prod(x,2); hyperbox = [zeros(1,5);ones(1,5)];
q = cubLattice_g(f,hyperbox,'uniform',1e-5,0)
```

```
q =
    0.2500
```

Example 6

```
% Estimate the integral with integrand f(x) = 3./(5-4*(cos(2*pi*x))) in the interval
% [0,1) with pure absolute error 1e-5.

f = @(x) 3./(5-4*(cos(2*pi*x))); hyperbox = [0;1];
q = cubLattice_g(f,hyperbox,'uniform',1e-5,0,'transform','id')
```

```
q =
    1.0000
```

8.5 See Also

cubSobol_g, cubMC_g, meanMC_g, meanMCBer_g, integral_g

9 cubSobol_g

Quasi-Monte Carlo method using Sobol' cubature over the d-dimensional region to integrate within a specified generalized error tolerance with guarantees under Walsh-Fourier coefficients cone decay assumptions

9.1 Syntax

```
[q,out_param] = cubSobol_g(f,hyperbox)
```

```
q = cubSobol_g(f,hyperbox,measure,abstol,reltol)
```

```
q = cubSobol_g(f,hyperbox,'measure',measure,'abstol',abstol,'reltol',reltol)
```

```
q = cubSobol_g(f,hyperbox,in_param)
```

9.2 Description

`[q,out_param] = cubSobol_g(f,hyperbox)` estimates the integral of f over the d-dimensional region described by `hyperbox`, and with an error guaranteed not to be greater than a specific generalized error tolerance, `tolfun:=max(abstol,reltol*| integral(f) |)`. Input f is a function handle. f should accept an $n \times d$ matrix input, where d is the dimension and n is the number of points being evaluated simultaneously. The input `hyperbox` is a $2 \times d$ matrix, where the first row corresponds to the lower limits and the second row corresponds to the upper limits of the integral. Given the construction of Sobol' sequences, d must be a positive integer with $1 \leq d \leq 1111$.

`q = cubSobol_g(f,hyperbox,measure,abstol,reltol)` estimates the integral of f over the hyperbox. The answer is given within the generalized error tolerance `tolfun`. All parameters should be input in the order specified above. If an input is not specified, the default value is used. Note that if an input is not specified, the remaining tail cannot be specified either. Inputs f and `hyperbox` are required. The other optional inputs are in the correct order: `measure,abstol,reltol,mmin,mmax,fudge,toltype` and `theta`.

`q = cubSobol_g(f,hyperbox,'measure',measure,'abstol',abstol,'reltol',reltol)` estimates the integral of f over the hyperbox. The answer is given within the generalized error tolerance `tolfun`. All the field-value pairs are optional and can be supplied in any order. If an input is not specified, the default value is used.

`q = cubSobol_g(f,hyperbox,in_param)` estimates the integral of f over the hyperbox. The answer is given within the generalized error tolerance `tolfun`.

Input Arguments

- f — the integrand whose input should be a matrix $n \times d$ where n is the number of data points and d the dimension, which cannot be greater than 1111. By default f is $f=@ x.^2$.
- `hyperbox` — the integration region defined by its bounds. It must be a $2 \times d$ matrix, where the first row corresponds to the lower limits and the second row corresponds to the upper limits of the integral. The default value is `[0;1]`.
- `in_param.measure` — for $f(x)*\mu(dx)$, we can define $\mu(dx)$ to be the measure of a uniformly distributed random variable in the hyperbox or normally distributed with covariance matrix I_d . The only possible values are 'uniform' or 'normal'. For 'uniform', the hyperbox must be a finite volume while for 'normal', the hyperbox can only be defined as $(-\text{Inf},\text{Inf})^d$. By default it is 'uniform'.
- `in_param.abstol` — the absolute error tolerance, $\text{abstol} \geq 0$. By default it is $1e-4$.
- `in_param.reltol` — the relative error tolerance, which should be in $[0,1]$. Default value is $1e-2$.

Optional Input Arguments

- `in_param.mmin` — the minimum number of points to start is 2^{mmin} . The cone condition on the Fourier coefficients decay requires a minimum number of points to start. The advice is to consider at least `mmin=10`. `mmin` needs to be a positive integer with `mmin ≤ mmax`. By default it is 10.
- `in_param.mmax` — the maximum budget is 2^{mmax} . By construction of the Sobol' generator, `mmax` is a positive integer such that `mmin ≤ mmax ≤ 53`. The default value is 24.
- `in_param.fudge` — the positive function multiplying the finite sum of Fast Walsh Fourier coefficients specified in the cone of functions. This input is a function handle. The fudge should accept an array of nonnegative integers being evaluated simultaneously. For more technical information about this parameter, refer to the references. By default it is $\frac{1}{5} 2^{-m}$.
- `in_param.toltype` — this is the generalized tolerance function. There are two choices, 'max' which takes $\max(\text{abstol}, \text{reitol} * |\text{integral}(f)|)$ and 'comb' which is the linear combination $\text{theta} * \text{abstol} + (1 - \text{theta}) * \text{reitol} * |\text{integral}(f)|$. `theta` is another parameter to be specified with 'comb' (see below). For pure absolute error, either choose 'max' and set `reitol = 0` or choose 'comb' and set `theta = 1`. For pure relative error, either choose 'max' and set `abstol = 0` or choose 'comb' and set `theta = 0`. Note that with 'max', the user can not input `abstol = reitol = 0` and with 'comb', if `theta = 1` `abstol` can not be 0 while if `theta = 0`, `reitol` can not be 0. By default `toltype` is 'max'.
- `in_param.theta` — this input is parametrizing the `toltype` 'comb'. Thus, it is only active when the `toltype` chosen is 'comb'. It establishes the linear combination weight between the absolute and relative tolerances $\text{theta} * \text{abstol} + (1 - \text{theta}) * \text{reitol} * |\text{integral}(f)|$. Note that for `theta = 1`, we have pure absolute tolerance while for `theta = 0`, we have pure relative tolerance. By default, `theta=1`.

Output Arguments

- `q` — the estimated value of the integral.
- `out_param.d` — dimension over which the algorithm integrated.
- `out_param.n` — number of Sobol' points used for computing the integral of `f`.
- `out_param.bound.err` — predicted bound on the error based on the cone condition. If the function lies in the cone, the real error will be smaller than generalized tolerance.
- `out_param.time` — time elapsed in seconds when calling `cubSobol_g`.
- `out_param.exitflag` — this is a binary vector stating whether warning flags arise. These flags tell about which conditions make the final result certainly not guaranteed. One flag is considered arisen when its value is 1. The following list explains the flags in the respective vector order:
 - 1 If reaching overbudget. It states whether the max budget is attained without reaching the guaranteed error tolerance.
 - 2 If the function lies outside the cone. In this case, results are not guaranteed. For more information about the cone definition, check the article mentioned below.

9.3 Guarantee

This algorithm computes the integral of real valued functions in dimension `d` with a prescribed generalized error tolerance. The Walsh-Fourier coefficients of the integrand are assumed to be absolutely convergent. If the algorithm terminates without warning messages, the output is given with guarantees under the assumption that the integrand lies inside a cone of functions. The guarantee is based on the decay rate of the Walsh-Fourier coefficients. For more details on how the cone is defined, please refer to the references below.

9.4 Examples

Example 1

```
% Estimate the integral with integrand  $f(x) = x_1 \cdot x_2$  in the interval  
%  $[0,1]^2$ :
```

```
f = @(x) prod(x,2); hyperbox = [zeros(1,2);ones(1,2)];  
q = cubSobol_g(f,hyperbox,'uniform',1e-5,0)
```

```
q =
```

```
0.2500
```

Example 2

```
% Estimate the integral with integrand  $f(x) = x_1^2 \cdot x_2^2 \cdot x_3^2$   
% in the interval  $\mathbb{R}^3$  where  $x_1$ ,  $x_2$  and  $x_3$  are normally distributed:
```

```
f = @(x) x(:,1).^2.*x(:,2).^2.*x(:,3).^2; hyperbox = [-inf(1,3);inf(1,3)];  
q = cubSobol_g(f,hyperbox,'normal',1e-3,1e-3)
```

```
q =
```

```
1.0004
```

Example 3

```
% Estimate the integral with integrand  $f(x) = \exp(-x_1^2 - x_2^2)$  in the  
% interval  $[-1,2]^2$ :
```

```
f = @(x) exp(-x(:,1).^2-x(:,2).^2); hyperbox = [-ones(1,2);2*ones(1,2)];  
q = cubSobol_g(f,hyperbox,'uniform',1e-3,1e-2)
```

```
q =
```

```
2.6532
```

Example 4

```
% Estimate the price of an European call with  $S_0=100$ ,  $K=100$ ,  $r=\sigma^2/2$ ,  
%  $\sigma=0.05$  and  $T=1$ .
```

```
f = @(x) exp(-0.05^2/2)*max(100*exp(0.05*x)-100,0); hyperbox = [-inf(1,1);inf(1,1)];  
q = cubSobol_g(f,hyperbox,'normal',1e-4,1e-2)
```

```
q =
```

```
2.0552
```

Example 5

```
% Estimate the integral with integrand  $f(x) = 8x_1 \cdot x_2 \cdot x_3 \cdot x_4 \cdot x_5$  in the  
% interval  $[0,1]^5$  with pure absolute error  $1e-5$ .
```

```
f = @(x) 8*prod(x,2); hyperbox = [zeros(1,5);ones(1,5)];  
q = cubSobol_g(f,hyperbox,'uniform',1e-5,0)
```

```
q =
```

```
0.2500
```

9.5 See Also

cubLattice_g, cubMC_g, meanMC_g, meanMCBer_g, integral_g

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