Fortran routines for testing unconstrained optimization software with derivatives up to third-order^{*}

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1 Introduction

This document gives details about the implementation and usage of a Fortran package that implements the computation of objective function and its first-, second-, and third-order derivatives for the well-known 35 problems proposed by Moré, Garbow and Hillstrom [2, 3].

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Originally, Moré, Garbow, and Hilltrom proposed 35 test problems for unconstrained optimization and code for computing the objective function and its first-derivative. The problems were divided into three categories: (a) 14 systems of nonlinear equations, cases where m = n and one searches for x^* such that $f_i(x^*) = 0$, i = 1, 2, ..., m; (b) 18 nonlinear least-squares problems, cases where $m \ge n$ and one is interested in solving the problem

$$\underset{x \in \mathbb{R}^n}{\operatorname{Minimize}} f(x) = \sum_{i=1}^m f_i^2(x) \tag{1}$$

by exploring its particular structure; and (c) 18 *unconstrained minimization* problems, where one is interested in solving (1) just by applying a general unconstrained optimization solver.

In 1994, Averbukh, Figueroa, and Schlick [1] added code to compute the second-order derivative for the 18 unconstrained minimization problems.

We now propose a package that considers (1) for all the 35 test problems and implements its first-, second-, and third-order derivatives.

2 Getting Started

When unzipping the code, the user must get the following directories and files:

\$(MGH)	root directory.
Makefile	
mgh_doc.pdf	documentation PDF file.
driver1.f08	driver with new routines.
driver2.f08	driver with alg 566 routines.
mgh.f08	all the new routines.
mgh_wrapper.f08	wrapper with alg 566 routines.
set_precision.f08	precision definitions file.

After compiling the code, the user must get the binaries driver1 and driver2 and the object files inside \$(MGH).

3 Compiling the code

To compile the main code,

- 1. The user must have a Fortran compiler installed and must configure in \$(MGH)/Makefile the variables FC with the Fortran compiler command-line and FFLAGS with the desired flags for the chosen Fortran compiler. We tested gfortran and nagfor compilers. Other Fortran compilers were not tested, but they may work as well.
- 2. Using the terminal, type make in the root directory.

To clean the compiled code, use make clean.

4 Using the module and compiling code

To use the module, the user must make the following modifications in the code.

- 1. Choose the precision you want to use in module \$(MGH)/set_precision.f08. For this, set the parameter rk as kind_s for single-precision, kind_d for double-precision, and kind_q for quad-precision.
- 2. Implement the desired routines (see Section 6.1).
- 3. Compile your code

replacing \$(FC) by the Fortran compiler you are using. You may need to ajust commandline option -I, that stands for the directory where .mod files are.

If the user prefer, it is possible to use the classic Algorithm 566 routines, to which we added a new one to compute third-order derivatives. In this case,

- 1. Choose the precision you want to use in module \$(MGH)/set_precision.f08. For this, set the parameter rk as kind_s for single-precision, kind_d for double-precision, and kind_q for quad-precision.
- 2. Implement the desired routines in the code (see Section 6.2).
- 3. Compile your code

replacing \$(FC) by the Fortran compiler you are using. You may need to ajust commandline option -I, that stands for the directory where .mod files are.

5 Using the drivers

Two drivers are available to test and validate the code:

- 1. \$(MGH)/driver1 implements the new routines module. It runs all the 35 problems from the test set. The output is given in driver1.out file.
- 2. \$(MGH)/driver2 implements the algorithm 566 routines. It runs all the 18 unconstrained minimization problems (see [3]). The output is given in driver2.out file.

6 Routines description

6.1 New routines

In order to use the new routines, first of all the user must

- 1. set the number of problem to work with, between 1 and 35, using mgh_set_problem,
- 2. customize m and n using mgh_set_dims or retrieve default values using mgh_get_dims,

After that, the user is able to retrieve the initial point using mgh_get_x0 and compute the objective function and its first-, second-, and third-order derivatives using mgh_evalf, mgh_evalg, mgh_evalh, and mgh_evalt, respectively. A detailed description of each routine follows.

subroutine mgh_set_problem(user_problem, flag)

Sets the problem number. When the user set the problem number, default dimensions (n and m) for it are automatically set. The subroutines arguments are

- user_problem is an input integer argument that should contain the problem number between 1 and 35.
- flag is an output integer argument that contains 0 if the problem number was successfully set or -1 if the user_problem is out of the range.

subroutine mgh_set_dims(n, m, flag)

Sets the dimensions for the problem.

- n is an input optional integer argument, sets the number of variables for the problem set.
- m is an input optional integer argument, sets the number of equations for the problem set.
- flag is an output optional integer, in the return contains 0 if the dimensions were set successfully, -1 if n is not valid, -2 if m is not valid, or -3 if both are not valid.

subroutine mgh_get_dims(n, m)

Gets the dimension for the problem.

- **n** is an output optional integer argument with the number of variables for the problem.
- m is an output optional integer argument with the number of equations for the problem.

subroutine mgh_get_x0(x0, factor)

Gets the initial point for the problem.

x0 is an output array of length **n** that contains the initial point.

factor is an optional real scalar that scales the initial point returned at x0.

subroutine mgh_evalf(x, f, flag)

Computes the objective function evaluated at x.

- \mathbf{x} is an input real array of length \mathbf{n} , contains the point in which the objective function must be evaluated.
- f is an output real that contains the objective function evaluated at x.
- flag is an output integer that contains 0 is the computation was made successfully, -1 if problem is not between 1 and 35, or -3 if a division by zero was made.

subroutine mgh_evalg(x, g, flag)

Computes the gradient of the objective function evaluated at x.

- x is an input real array of length n, contains the point in which the gradient must be evaluated.
- g is an output real array of length n that contains the gradient evaluated at x.
- flag is an output integer that contains 0 is the computation was made successfully, -1 if problem is not between 1 and 35, or -3 if a division by zero was made.

subroutine mgh_evalh(x, h, flag)

Computes the Hessian of the objective function evaluated at x.

- \mathbf{x} is an input real array of length \mathbf{n} , contains the point in which the Hessian must be evaluated.
- h is an output real array of length $n \times n$ that contains the upper triangle of the Hessian evaluated at x.
- flag is an output integer that contains 0 is the computation was made successfully, -1 if problem is not between 1 and 35, or -3 if a division by zero was made.

subroutine mgh_evalt(x, t, flag)

Computes the third-order derivative tensor of the objective function evaluated at x.

- x is an input real array of length n, contains the point in which the third-order derivative must be evaluated.
- t is an output real array of length $n \times n \times n$ that contains the upper part of the third-derivative evaluated at x.
- flag is an output integer that contains 0 is the computation was made successfully, -1 if problem is not between 1 and 35, or -3 if a division by zero was made.

subroutine mgh_get_name(name)

Returns the problem name

name is a character(len=60) output parameter that contains the problem name.

6.2 Algorithm 566 Routines + Third derivative computation

subroutine initpt(n, x, nprob, factor)

Returns the initial point for a given problem.

n is an integer input argument, should contain the dimension of the problem.

x is a real output array of length **n**, contains the initial point.

- nprob is an integer input, contains the number of the problem between 1 and 18.
- factor is a real input, contains the factor by which the initial point will be scaled.

subroutine objfcn(n, x, f, nprob)

Compute the objective function value for a given problem at x.

- **n** is an integer input argument, should contain the dimension of the problem.
- \mathbf{x} is a real input array of length \mathbf{n} , contains the point in which the objective function will be evaluated.
- f is a real output argument that contains the objective function value.
- nprob is an integer input, contains the number of the problem between 1 and 18.

subroutine grdfcn(n, x, g, nprob)

Compute the gradient of the objective function, for a given problem, evaluated at x.

n is an integer input argument, should contain the dimension of the problem.

- \mathbf{x} is a real input array of length \mathbf{n} , contains the point in which the objective function will be evaluated.

nprob is an integer input, contains the number of the problem between 1 and 18.

subroutine hesfcn(n, x, hesd, hesu, nprob)

Compute the Hessian of the objective function, for a given problem, evaluated at x.

- **n** is an integer input argument, should contain the dimension of the problem.
- \mathbf{x} is a real input array of length \mathbf{n} , contains the point in which the objective function will be evaluated.
- hesd is a real output array of length n, contains the diagonal of the Hessian.
- hesu is a real output array of length

$$\frac{n(n-1)}{2},$$

contains the strict upper triangle of the Hessian stored by columns. The i, j term of the Hessian, i < j, is located at the position

$$\frac{(j-1)(j-2)}{2} + i$$

at hesu.

nprob is an integer input, contains the number of the problem between 1 and 18.

subroutine trdfcn(n, x, td, tu, nprob)

Compute the third-order derivative tensor of the objective function, for a given problem, evaluated at x.

- **n** is an integer input argument, should contain the dimension of the problem.
- \mathbf{x} is a real input array of length \mathbf{n} , contains the point in which the objective function will be evaluated.
- td is a real output array of length **n**, contains the diagonal of the tensor.

tu is a real output array of length

$$\frac{n-1}{6}((n-2)(n-3)+9(n-2)+12),$$

contains the strict upper part of the tensor stored by columns. The i, j, k term of the tensor, $i \leq j \leq k$ but not i = j = k, is located at the position

$$\frac{k-2}{6}((k-3)(k-4)+9(k-3)+12)+\frac{j(j-1)}{2}+i$$

at tu.

nprob is an integer input, contains the number of the problem between 1 and 18.

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