

# NAG Library Function Document

## nag\_opt\_bounds\_2nd\_deriv (e04lbc)

### 1 Purpose

nag\_opt\_bounds\_2nd\_deriv (e04lbc) is a comprehensive modified-Newton algorithm for finding:

- an unconstrained minimum of a function of several variables
- a minimum of a function of several variables subject to fixed upper and/or lower bounds on the variables.

First and second derivatives are required. nag\_opt\_bounds\_2nd\_deriv (e04lbc) is intended for objective functions which have continuous first and second derivatives (although it will usually work even if the derivatives have occasional discontinuities).

### 2 Specification

```
#include <nag.h>
#include <nage04.h>

void nag_opt_bounds_2nd_deriv (Integer n,
    void (*objfun)(Integer n, const double x[], double *objf, double g[],
        Nag_Comm *comm),
    void (*hessfun)(Integer n, const double x[], double h[], double hd[],
        Nag_Comm *comm),
    Nag_BoundType bound, double bl[], double bu[], double x[], double *objf,
    double g[], Nag_E04_Opt *options, Nag_Comm *comm, NagError *fail)
```

### 3 Description

nag\_opt\_bounds\_2nd\_deriv (e04lbc) is applicable to problems of the form:

$$\begin{array}{ll} \text{Minimize} & F(x_1, x_2, \dots, x_n) \\ \text{subject to} & l_j \leq x_j \leq u_j, \quad j = 1, 2, \dots, n. \end{array}$$

Special provision is made for unconstrained minimization (i.e., problems which actually have no bounds on the  $x_j$ ), problems which have only non-negativity bounds, and problems in which  $l_1 = l_2 = \dots = l_n$  and  $u_1 = u_2 = \dots = u_n$ . It is possible to specify that a particular  $x_j$  should be held constant. You must supply a starting point, a function **objfun** to calculate the value of  $F(x)$  and its first derivatives  $\frac{\partial F}{\partial x_j}$  at

any point  $x$ , and a function **hessfun** to calculate the second derivatives  $\frac{\partial^2 F}{\partial x_i \partial x_j}$ .

A typical iteration starts at the current point  $x$  where  $n_z$  (say) variables are free from both their bounds. The vector of first derivatives of  $F(x)$  with respect to the free variables,  $g_z$ , and the matrix of second derivatives with respect to the free variables,  $H$ , are obtained. (These both have dimension  $n_z$ .) The equations

$$(H + E)p_z = -g_z$$

are solved to give a search direction  $p_z$ . (The matrix  $E$  is chosen so that  $H + E$  is positive definite.)  $p_z$  is then expanded to an  $n$ -vector  $p$  by the insertion of appropriate zero elements;  $\alpha$  is found such that  $F(x + \alpha p)$  is approximately a minimum (subject to the fixed bounds) with respect to  $\alpha$ , and  $x$  is replaced by  $x + \alpha p$ . (If a saddle point is found, a special search is carried out so as to move away from the saddle point.) If any variable actually reaches a bound, it is fixed and  $n_z$  is reduced for the next iteration.

There are two sets of convergence criteria – a weaker and a stronger. Whenever the weaker criteria are satisfied, the Lagrange-multipliers are estimated for all active constraints. If any Lagrange-multiplier estimate is significantly negative, then one of the variables associated with a negative Lagrange-multiplier estimate is released from its bound and the next search direction is computed in the extended subspace (i.e.,  $n_z$  is increased). Otherwise, minimization continues in the current subspace until the stronger criteria are satisfied. If at this point there are no negative or near-zero Lagrange-multiplier estimates, the process is terminated.

If you specify that the problem is unconstrained, `nag_opt_bounds_2nd_deriv` (e04lbc) sets the  $l_j$  to  $-10^{10}$  and the  $u_j$  to  $10^{10}$ . Thus, provided that the problem has been sensibly scaled, no bounds will be encountered during the minimization process and `nag_opt_bounds_2nd_deriv` (e04lbc) will act as an unconstrained minimization algorithm.

## 4 References

Gill P E and Murray W (1973) Safeguarded steplength algorithms for optimization using descent methods *NPL Report NAC 37* National Physical Laboratory

Gill P E and Murray W (1974) Newton-type methods for unconstrained and linearly constrained optimization *Math. Programming* 7 311–350

Gill P E and Murray W (1976) Minimization subject to bounds on the variables *NPL Report NAC 72* National Physical Laboratory

## 5 Arguments

1: **n** – Integer *Input*

*On entry:* the number  $n$  of independent variables.

*Constraint:*  $n \geq 1$ .

2: **objfun** – function, supplied by the user *External Function*

**objfun** must evaluate the function  $F(x)$  and its first derivatives  $\frac{\partial F}{\partial x_j}$  at any point  $x$ . (However, if you do not wish to calculate  $F(x)$  or its first derivatives at a particular  $x$ , there is the option of setting an argument to cause `nag_opt_bounds_2nd_deriv` (e04lbc) to terminate immediately.)

The specification of **objfun** is:

```
void objfun (Integer n, const double x[], double *objf, double g[],
            Nag_Comm *comm)
```

1: **n** – Integer *Input*

*On entry:*  $n$ , the number of variables.

2: **x[n]** – const double *Input*

*On entry:* the point  $x$  at which the value of  $F$ , or  $F$  and  $\frac{\partial F}{\partial x_j}$ , are required.

3: **objf** – double \* *Output*

*On exit:* **objfun** must set **objf** to the value of the objective function  $F$  at the current point  $x$ . If it is not possible to evaluate  $F$  then **objfun** should assign a negative value to **comm**→**flag**; `nag_opt_bounds_2nd_deriv` (e04lbc) will then terminate.

4:	<b>g[n]</b> – double	<i>Output</i>
	<i>On exit:</i> <b>objfun</b> must set <b>g[j – 1]</b> to the value of the first derivative $\frac{\partial F}{\partial x_j}$ at the current point $x$ , for $j = 1, 2, \dots, n$ . If it is not possible to evaluate the first derivatives then <b>objfun</b> should assign a negative value to <b>comm</b> → <b>flag</b> ; nag_opt_bounds_2nd_deriv (e04lbc) will then terminate.	
5:	<b>comm</b> – Nag_Comm *	
	Pointer to structure of type Nag_Comm; the following members are relevant to <b>objfun</b> .	
	<b>flag</b> – Integer	<i>Output</i>
	<i>On exit:</i> if <b>objfun</b> resets <b>comm</b> → <b>flag</b> to some negative number then nag_opt_bounds_2nd_deriv (e04lbc) will terminate immediately with the error indicator NE_USER_STOP. If <b>fail</b> is supplied to nag_opt_bounds_2nd_deriv (e04lbc), <b>fail.errnum</b> will be set to your setting of <b>comm</b> → <b>flag</b> .	
	<b>first</b> – Nag_Boolean	<i>Input</i>
	<i>On entry:</i> will be set to Nag_TRUE on the first call to <b>objfun</b> and Nag_FALSE for all subsequent calls.	
	<b>nf</b> – Integer	<i>Input</i>
	<i>On entry:</i> the number of evaluations of the objective function; this value will be equal to the number of calls made to <b>objfun</b> (including the current one).	
	<b>user</b> – double *	
	<b>iuser</b> – Integer *	
	<b>p</b> – Pointer	
	The type Pointer will be void * with a C compiler that defines void * and char * otherwise.	
	Before calling nag_opt_bounds_2nd_deriv (e04lbc) these pointers may be allocated memory and initialized with various quantities for use by <b>objfun</b> when called from nag_opt_bounds_2nd_deriv (e04lbc).	

**Note:** **objfun** should be tested separately before being used in conjunction with nag\_opt\_bounds\_2nd\_deriv (e04lbc). The array **x** must **not** be changed by **objfun**.

3: **hessfun** – function, supplied by the user *External Function*

**hessfun** must calculate the second derivatives of  $F(x)$  at any point  $x$ . (As with **objfun** there is the option of causing nag\_opt\_bounds\_2nd\_deriv (e04lbc) to terminate immediately.)

The specification of <b>hessfun</b> is:		
<code>void hessfun (Integer n, const double x[], double h[], double hd[], Nag_Comm *comm)</code>		
1:	<b>n</b> – Integer	<i>Input</i>
	<i>On entry:</i> the number $n$ of variables.	
2:	<b>x[n]</b> – const double	<i>Input</i>
	<i>On entry:</i> the point $x$ at which the second derivatives of $F$ are required.	
3:	<b>h[n × (n – 1)/2]</b> – double	<i>Output</i>
	<i>On exit:</i> <b>hessfun</b> must place the strict lower triangle of the second derivative matrix of $F$ (evaluated at the point $x$ ) in <b>h</b> , stored by rows, i.e., set	

$$\mathbf{h}[(i-1)(i-2)/2 + j - 1] = \frac{\partial^2 F}{\partial x_i \partial x_j} \Big|_{\mathbf{x}}, \quad \text{for } i = 2, 3, \dots, n \text{ and } j = 1, 2, \dots, i - 1.$$

(The upper triangle is not required because the matrix is symmetric.) If it is not possible to evaluate the elements of **h** then **hessfun** should assign a negative value to **comm**→**flag**; **nag\_opt\_bounds\_2nd\_deriv** (e04lbc) will then terminate.

4: **hd[n]** – double *Input/Output*

*On entry:* the value of  $\frac{\partial F}{\partial x_j}$  at the point  $x$ , for  $j = 1, 2, \dots, n$ . These values may be useful in the evaluation of the second derivatives.

*On exit:* unless **comm**→**flag** is reset to a negative number **hessfun** must place the diagonal elements of the second derivative matrix of  $F$  (evaluated at the point  $x$ ) in **hd**, i.e., set

$$\mathbf{hd}[j - 1] = \frac{\partial^2 F}{\partial x_j^2} \Big|_{\mathbf{x}}, \quad \text{for } j = 1, 2, \dots, n.$$

If it is not possible to evaluate the elements of **hd** then **hessfun** should assign a negative value to **comm**→**flag**; **nag\_opt\_bounds\_2nd\_deriv** (e04lbc) will then terminate.

5: **comm** – Nag\_Comm \*

Pointer to structure of type Nag\_Comm; the following members are relevant to **objfun**.

**flag** – Integer *Output*

*On exit:* if **hessfun** resets **comm**→**flag** to some negative number then **nag\_opt\_bounds\_2nd\_deriv** (e04lbc) will terminate immediately with the error indicator NE\_USER\_STOP. If **fail** is supplied to **nag\_opt\_bounds\_2nd\_deriv** (e04lbc) **fail.errnum** will be set to your setting of **comm**→**flag**.

**first** – Nag\_Boolean *Input*

*On entry:* will be set to Nag\_TRUE on the first call to **hessfun** and Nag\_FALSE for all subsequent calls.

**nf** – Integer *Input*

*On entry:* the number of calculations of the objective function; this value will be equal to the number of calls made to **hessfun** including the current one.

**user** – double \*

**iuser** – Integer \*

**p** – Pointer

The type Pointer will be `void *` with a C compiler that defines `void *` and `char *` otherwise.

Before calling **nag\_opt\_bounds\_2nd\_deriv** (e04lbc) these pointers may be allocated memory and initialized with various quantities for use by **hessfun** when called from **nag\_opt\_bounds\_2nd\_deriv** (e04lbc).

**Note:** **hessfun** should be tested separately before being used in conjunction with **nag\_opt\_bounds\_2nd\_deriv** (e04lbc). The array **x** must **not** be changed by **hessfun**.

4: **bound** – Nag\_BoundType *Input*

*On entry:* indicates whether the problem is unconstrained or bounded and, if it is bounded, whether the facility for dealing with bounds of special forms is to be used. **bound** should be set to one of the following values:

**bound** = Nag\_Bounds

If the variables are bounded and you will be supplying all the  $l_j$  and  $u_j$  individually.

**bound** = Nag\_NoBounds

If the problem is unconstrained.

**bound** = Nag\_BoundsZero

If the variables are bounded, but all the bounds are of the form  $0 \leq x_j$ .

**bound** = Nag\_BoundsEqual

If all the variables are bounded, and  $l_1 = l_2 = \dots = l_n$  and  $u_1 = u_2 = \dots = u_n$ .

*Constraint:* **bound** = Nag\_Bounds, Nag\_NoBounds, Nag\_BoundsZero or Nag\_BoundsEqual.

5: **bl[n]** – double *Input/Output*

*On entry:* the lower bounds  $l_j$ .

If **bound** = Nag\_Bounds, you must set **bl**[ $j - 1$ ] to  $l_j$ , for  $j = 1, 2, \dots, n$ . (If a lower bound is not required for any  $x_j$ , the corresponding **bl**[ $j - 1$ ] should be set to a large negative number, e.g.,  $-10^{10}$ .)

If **bound** = Nag\_BoundsEqual, you must set **bl**[0] to  $l_1$ ; nag\_opt\_bounds\_2nd\_deriv (e04lbc) will then set the remaining elements of **bl** equal to **bl**[0].

If **bound** = Nag\_NoBounds or Nag\_BoundsZero, **bl** will be initialized by nag\_opt\_bounds\_2nd\_deriv (e04lbc).

*On exit:* the lower bounds actually used by nag\_opt\_bounds\_2nd\_deriv (e04lbc), e.g., if **bound** = Nag\_BoundsZero, **bl**[0] = **bl**[1] =  $\dots$  = **bl**[ $n - 1$ ] = 0.0.

6: **bu[n]** – double *Input/Output*

*On entry:* the upper bounds  $u_j$ .

If **bound** = Nag\_Bounds, you must set **bu**[ $j - 1$ ] to  $u_j$ , for  $j = 1, 2, \dots, n$ . (If an upper bound is not required for any  $x_j$ , the corresponding **bu**[ $j - 1$ ] should be set to a large positive number, e.g.,  $10^{10}$ .)

If **bound** = Nag\_BoundsEqual, you must set **bu**[0] to  $u_1$ ; nag\_opt\_bounds\_2nd\_deriv (e04lbc) will then set the remaining elements of **bu** equal to **bu**[0].

If **bound** = Nag\_NoBounds or Nag\_BoundsZero, **bu** will be initialized by nag\_opt\_bounds\_2nd\_deriv (e04lbc).

*On exit:* the upper bounds actually used by nag\_opt\_bounds\_2nd\_deriv (e04lbc), e.g., if **bound** = Nag\_BoundsZero, **bu**[0] = **bu**[1] =  $\dots$  = **bu**[ $n - 1$ ] =  $10^{10}$ .

7: **x[n]** – double *Input/Output*

*On entry:* **x**[ $j - 1$ ] must be set to a guess at the  $j$ th component of the position of the minimum, for  $j = 1, 2, \dots, n$ .

*On exit:* the final point  $x^*$ . Thus, if **fail.code** = NE\_NOERROR on exit, **x**[ $j - 1$ ] is the  $j$ th component of the estimated position of the minimum.

8: **objf** – double \* *Output*

*On exit:* the function value at the final point given in **x**.

9: **g[n]** – double *Output*

*On exit:* the first derivative vector corresponding to the final point in **x**. The elements of **g** corresponding to free variables should normally be close to zero.

10: **options** – Nag\_E04\_Opt \* *Input/Output*

*On entry/exit:* a pointer to a structure of type Nag\_E04\_Opt whose members are optional parameters for nag\_opt\_bounds\_2nd\_deriv (e04lbc). These structure members offer the means of

adjusting some of the argument values of the algorithm and on output will supply further details of the results. A description of the members of **options** is given below in Section 11.

If any of these optional parameters are required then the structure **options** should be declared and initialized by a call to `nag_opt_init` (e04xxc) and supplied as an argument to `nag_opt_bounds_2nd_deriv` (e04lbc). However, if the optional parameters are not required the NAG defined null pointer, `E04_DEFAULT`, can be used in the function call.

11: **comm** – Nag\_Comm \* *Input/Output*

**Note:** **comm** is a NAG defined type (see Section 2.3.1.1 in How to Use the NAG Library and its Documentation).

*On entry/exit:* structure containing pointers for communication to user-supplied functions; see the description of **objfun** and **hessfun** for details. If you do not need to make use of this communication feature the null pointer `NAGCOMM_NULL` may be used in the call to `nag_opt_bounds_2nd_deriv` (e04lbc); **comm** will then be declared internally for use in calls to user-supplied functions.

12: **fail** – NagError \* *Input/Output*

The NAG error argument (see Section 2.7 in How to Use the NAG Library and its Documentation).

## 5.1 Description of Printed Output

Intermediate and final results are printed out by default. The level of printed output can be controlled with the structure member **options.print\_level** (see Section 11.2). The default, **options.print\_level** = `Nag_Soln_Iter` provides a single line of output at each iteration and the final result. This section describes the default printout produced by `nag_opt_bounds_2nd_deriv` (e04lbc).

The following line of output is produced at each iteration. In all cases the values of the quantities printed are those in effect *on completion* of the given iteration.

<code>Itn</code>	the iteration count, $k$ .
<code>Nfun</code>	the cumulative number of calls made to <b>objfun</b> .
<code>Objective</code>	the value of the objective function, $F(x^{(k)})$
<code>Norm g</code>	the Euclidean norm of the projected gradient vector, $\ g_z(x^{(k)})\ $ .
<code>Norm x</code>	the Euclidean norm of $x^{(k)}$ .
<code>Norm(x(k-1)-x(k))</code>	the Euclidean norm of $x^{(k-1)} - x^{(k)}$ .
<code>Step</code>	the step $\alpha^{(k)}$ taken along the computed search direction $p^{(k)}$ .
<code>Cond H</code>	the ratio of the largest to the smallest element of the diagonal factor $D$ of the projected Hessian matrix. This quantity is usually a good estimate of the condition number of the projected Hessian matrix. (If no variables are currently free, this value will be zero.)
<code>PosDef</code>	indicates whether the second derivative matrix $H$ for the current subspace is positive definite (Yes) or not (No).

The printout of the final result consists of:

<code>x</code>	the final point, $x^*$ .
<code>g</code>	the final projected gradient vector, $g_z(x^*)$ .
<code>Status</code>	the final state of the variable with respect to its bound(s).

## 6 Error Indicators and Warnings

When one of `NE_USER_STOP`, `NE_INT_ARG_LT`, `NE_BOUND`, `NE_DERIV_ERRORS`, `NE_OPT_NOT_INIT`, `NE_BAD_PARAM`, `NE_2_REAL_ARG_LT`, `NE_INVALID_INT_RANGE_1`, `NE_INVALID_REAL_RANGE_EF`, `NE_INVALID_REAL_RANGE_FF` and `NE_ALLOC_FAIL` occurs, no values will have been assigned by `nag_opt_bounds_2nd_deriv` (e04lbc) to **objf** or to the elements of **g**, **options.state**, **options.hesl**, or **options.hesd**.

An exit of **fail.code** = `NW_TOO_MANY_ITER`, `NW_LAGRANGE_MULT_ZERO` and `NW_COND_MIN` may also be caused by mistakes in **objfun**, by the formulation of the problem or by an awkward function. If there are no such mistakes, it is worth restarting the calculations from a different starting point (not the point at which the failure occurred) in order to avoid the region which caused the failure.

### NE\_2\_REAL\_ARG\_LT

On entry, **options.step\_max** =  $\langle value \rangle$  while **options.optim\_tol** =  $\langle value \rangle$ . These arguments must satisfy **options.step\_max**  $\geq$  **options.optim\_tol**.

### NE\_ALLOC\_FAIL

Dynamic memory allocation failed.

### NE\_BAD\_PARAM

On entry, argument **bound** had an illegal value.

On entry, argument **options.print\_level** had an illegal value.

### NE\_BOUND

The lower bound for variable  $\langle value \rangle$  (array element **bl**[ $\langle value \rangle$ ]) is greater than the upper bound.

### NE\_DERIV\_ERRORS

Large errors were found in the derivatives of the objective function.

### NE\_INT\_ARG\_LT

On entry, **n** must not be less than 1: **n** =  $\langle value \rangle$ .

### NE\_INTERNAL\_ERROR

An internal error has occurred in this function. Check the function call and any array sizes. If the call is correct then please contact NAG for assistance.

### NE\_INVALID\_INT\_RANGE\_1

Value  $\langle value \rangle$  given to **options.max\_iter** is not valid. Correct range is **options.max\_iter**  $\geq$  0.

### NE\_INVALID\_REAL\_RANGE\_EF

Value  $\langle value \rangle$  given to **options.optim\_tol** is not valid. Correct range is  $\epsilon \leq$  **options.optim\_tol**  $<$  1.0.

### NE\_INVALID\_REAL\_RANGE\_FF

Value  $\langle value \rangle$  given to **options.linesearch\_tol** is not valid. Correct range is  $0.0 \leq$  **options.linesearch\_tol**  $<$  1.0.

### NE\_NOT\_APPEND\_FILE

Cannot open file  $\langle string \rangle$  for appending.

### NE\_NOT\_CLOSE\_FILE

Cannot close file  $\langle string \rangle$ .

**NE\_OPT\_NOT\_INIT**

Options structure not initialized.

**NE\_USER\_STOP**

User requested termination, user flag value =  $\langle value \rangle$ .

This exit occurs if you set **comm**→**flag** to a negative value in **objfun** or **hessfun**. If **fail** is supplied, the value of **fail.errnum** will be the same as your setting of **comm**→**flag**.

**NE\_WRITE\_ERROR**

Error occurred when writing to file  $\langle string \rangle$ .

**NW\_COND\_MIN**

The conditions for a minimum have not all been satisfied, but a lower point could not be found.

Provided that, on exit, the first derivatives of  $F(x)$  with respect to the free variables are sufficiently small, and that the estimated condition number of the second derivative matrix is not too large, this error exit may simply mean that, although it has not been possible to satisfy the specified requirements, the algorithm has in fact found the minimum as far as the accuracy of the machine permits. This could be because **options.optim\_tol** has been set so small that rounding error in **objfun** makes attainment of the convergence conditions impossible.

If the estimated condition number of the second derivative matrix at the final point is large, it could be that the final point is a minimum but that the smallest eigenvalue of the second derivative matrix is so close to zero that it is not possible to recognize the point as a minimum.

**NW\_LAGRANGE\_MULT\_ZERO**

All the Lagrange-multiplier estimates which are not indisputably positive lie close to zero.

However, it is impossible either to continue minimizing on the current subspace or to find a feasible lower point by releasing and perturbing any of the fixed variables. You should investigate as for **NW\_COND\_MIN**.

**NW\_TOO\_MANY\_ITER**

The maximum number of iterations,  $\langle value \rangle$ , have been performed.

If steady reductions in  $F(x)$ , were monitored up to the point where this exit occurred, then the exit probably occurred simply because **options.max\_iter** was set too small, so the calculations should be restarted from the final point held in **x**. This exit may also indicate that  $F(x)$  has no minimum.

**7 Accuracy**

A successful exit (**fail.code** = **NE\_NOERROR**) is made from `nag_opt_bounds_2nd_deriv` (e04lbc) when  $H^{(k)}$  is positive definite and when (B1, B2 and B3) or B4 hold, where

$$B1 \equiv \alpha^{(k)} \times \|p^{(k)}\| < (\mathbf{options.optim\_tol} + \sqrt{\epsilon}) \times (1.0 + \|x^{(k)}\|)$$

$$B2 \equiv |F^{(k)} - F^{(k-1)}| < (\mathbf{options.optim\_tol}^2 + \epsilon) \times (1.0 + |F^{(k)}|)$$

$$B3 \equiv \|g_z^{(k)}\| < (\epsilon^{1/3} + \mathbf{options.optim\_tol}) \times (1.0 + |F^{(k)}|)$$

$$B4 \equiv \|g_z^{(k)}\| < 0.01 \times \sqrt{\epsilon}.$$

(Quantities with superscript  $k$  are the values at the  $k$ th iteration of the quantities mentioned in Section 3;  $\epsilon$  is the *machine precision*,  $\cdot$  denotes the Euclidean norm and **options.optim\_tol** is described in Section 11.)

If **fail.code** = **NE\_NOERROR**, then the vector in **x** on exit,  $x_{\text{sol}}$ , is almost certainly an estimate of the position of the minimum,  $x_{\text{true}}$ , to the accuracy specified by **options.optim\_tol**.



If `fail.code` = `NW_COND_MIN` or `NW_LAGRANGE_MULT_ZERO`,  $x_{\text{sol}}$  may still be a good estimate of  $x_{\text{true}}$ , but the following checks should be made. Let the largest of the first  $n_z$  elements of the optional parameter `options.hesd` be `options.hesd[b]`, let the smallest be `options.hesd[s]`, and define  $\kappa = \text{options.hesd}[b]/\text{options.hesd}[s]$ . The scalar  $\kappa$  is usually a good estimate of the condition number of the projected Hessian matrix at  $x_{\text{sol}}$ . If

- (a) the sequence  $\{F(x^{(k)})\}$  converges to  $F(x_{\text{sol}})$  at a superlinear or fast linear rate,
- (b)  $\|g_z(x_{\text{sol}})\|^2 < 10.0 \times \epsilon$ , and
- (c)  $\kappa < 1.0/\|g_z(x_{\text{sol}})\|$ ,

then it is almost certain that  $x_{\text{sol}}$  is a close approximation to the position of a minimum. When (b) is true, then usually  $F(x_{\text{sol}})$  is a close approximation to  $F(x_{\text{true}})$ . The quantities needed for these checks are all available in the results printout from `nag_opt_bounds_2nd_deriv` (e04lbc); in particular the final value of `Cond H` gives  $\kappa$ .

Further suggestions about confirmation of a computed solution are given in the e04 Chapter Introduction.

## 8 Parallelism and Performance

`nag_opt_bounds_2nd_deriv` (e04lbc) is not threaded in any implementation.

## 9 Further Comments

### 9.1 Timing

The number of iterations required depends on the number of variables, the behaviour of  $F(x)$ , the accuracy demanded and the distance of the starting point from the solution. The number of multiplications performed in an iteration of `nag_opt_bounds_2nd_deriv` (e04lbc) is  $n_z^3/6 + O(n_z^2)$ . In addition, each iteration makes one call of `hessfun` and at least one call of `objfun`. So, unless  $F(x)$  and its derivatives can be evaluated very quickly, the run time will be dominated by the time spent in `objfun`.

### 9.2 Scaling

Ideally, the problem should be scaled so that, at the solution,  $F(x)$  and the corresponding values of the  $x_j$  are each in the range  $(-1, +1)$ , and so that at points one unit away from the solution,  $F(x)$  differs from its value at the solution by approximately one unit. This will usually imply that the Hessian matrix at the solution is well conditioned. It is unlikely that you will be able to follow these recommendations very closely, but it is worth trying (by guesswork), as sensible scaling will reduce the difficulty of the minimization problem, so that `nag_opt_bounds_2nd_deriv` (e04lbc) will take less computer time.

### 9.3 Unconstrained Minimization

If a problem is genuinely unconstrained and has been scaled sensibly, the following points apply:

- (a)  $n_z$  will always be  $n$ ,
- (b) the optional parameters `options.hesl` and `options.hesd` will be factors of the full approximate second derivative matrix with elements stored in the natural order,
- (c) the elements of `g` should all be close to zero at the final point,
- (d) the `Status` values given in the printout from `nag_opt_bounds_2nd_deriv` (e04lbc), and in the optional parameter `options.state` on exit are unlikely to be of interest (unless they are negative, which would indicate that the modulus of one of the  $x_j$  has reached  $10^{10}$  for some reason),
- (e) `Norm g` simply gives the norm of the first derivative vector.

## 10 Example

This example minimizes the function

$$F = (x_1 + 10x_2)^2 + 5(x_3 - x_4)^2 + (x_2 - 2x_3)^4 + 10(x_1 - x_4)^4$$

subject to the bounds

$$\begin{aligned} 1 &\leq x_1 \leq 3 \\ -2 &\leq x_2 \leq 0 \\ 1 &\leq x_4 \leq 3 \end{aligned}$$

starting from the initial guess  $(1.46, -0.82, 0.57, 1.21)^T$ .

The **options** structure is declared and initialized by `nag_opt_init` (e04xxc). One option value is read from a data file by use of `nag_opt_read` (e04xyc). The memory freeing function `nag_opt_free` (e04xzc) is used to free the memory assigned to the pointers in the option structure. You must **not** use the standard C function `free()` for this purpose.

### 10.1 Program Text

```

/* nag_opt_bounds_2nd_deriv (e04lbc) Example Program.
 *
 * NAGPRODCODE Version.
 *
 * Copyright 2016 Numerical Algorithms Group.
 *
 * Mark 26, 2016.
 *
 *
 */

#include <nag.h>
#include <stdio.h>
#include <string.h>
#include <nag_stdlib.h>
#include <math.h>
#include <nage04.h>

#ifdef __cplusplus
extern "C"
{
#endif
    static void NAG_CALL funct(Integer n, const double xc[], double *fc,
                               double gc[], Nag_Comm *comm);
    static void NAG_CALL h(Integer n, const double xc[], double fheshl[],
                            double fheshd[], Nag_Comm *comm);
#ifdef __cplusplus
}
#endif

int main(void)
{
    const char *optionsfile = "e04lbce.opt";
    static double ruser[2] = { -1.0, -1.0 };
    Integer exit_status = 0;
    Nag_Boolean print;
    Integer n = 4;
    Nag_Comm comm;
    Nag_E04_Opt options;
    double *bl = 0, *bu = 0, f, *g = 0, *x = 0;
    NagError fail;

    INIT_FAIL(fail);

    printf("nag_opt_bounds_2nd_deriv (e04lbc) Example Program Results\n");

    /* For communication with user-supplied functions: */
    comm.user = ruser;

```

```

if (n >= 1) {
    if (!(x = NAG_ALLOC(n, double)) ||
        !(bl = NAG_ALLOC(n, double)) ||
        !(bu = NAG_ALLOC(n, double)) || !(g = NAG_ALLOC(n, double)))
    {
        printf("Allocation failure\n");
        exit_status = -1;
        goto END;
    }
}
else {
    printf("Invalid n.\n");
    exit_status = 1;
    return exit_status;
}

bl[0] = 1.0;
bu[0] = 3.0;
bl[1] = -2.0;
bu[1] = 0.0;

/* x[2] is not bounded, so we set bl[2] to a large negative
 * number and bu[2] to a large positive number
 */
bl[2] = -1e6;
bu[2] = 1e6;
bl[3] = 1.0;
bu[3] = 3.0;

/* Set up starting point */
x[0] = 3.0;
x[1] = -1.0;
x[2] = 0.0;
x[3] = 1.0;

print = Nag_TRUE;
/* nag_opt_init (e04xxc).
 * Initialization function for option setting
 */
nag_opt_init(&options);
/* nag_opt_read (e04xyc).
 * Read options from a text file
 */
fflush(stdout);
nag_opt_read("e04lbc", optionsfile, &options, print, "stdout", &fail);
if (fail.code != NE_NOERROR) {
    printf("Error from nag_opt_read (e04xyc).\n%s\n", fail.message);
    exit_status = 1;
    goto END;
}
/* nag_opt_bounds_2nd_deriv (e04lbc), see above. */
nag_opt_bounds_2nd_deriv(n, funct, h, Nag_Bounds, bl, bu, x, &f, g,
                        &options, &comm, &fail);
if (fail.code != NE_NOERROR) {
    printf("Error or warning from "
           "nag_opt_bounds_2nd_deriv (e04lbc).\n%s\n", fail.message);
    if (fail.code != NW_COND_MIN)
        exit_status = 1;
}

/* Free memory allocated by nag_opt_bounds_deriv (e04kbc) to pointers hesd,
 * hesl and state.
 */
/* nag_opt_free (e04xzc).
 * Memory freeing function for use with option setting
 */
nag_opt_free(&options, "all", &fail);
if (fail.code != NE_NOERROR) {
    printf("Error from nag_opt_bounds_2nd_deriv (e04lbc).\n%s\n",
           fail.message);
}

```

```

        exit_status = 1;
        goto END;
    }

END:
    NAG_FREE(x);
    NAG_FREE(bl);
    NAG_FREE(bu);
    NAG_FREE(g);

    return exit_status;
}

static void NAG_CALL funct(Integer n, const double xc[], double *fc,
                          double gc[], Nag_Comm *comm)
{
    /* Function to evaluate objective function and its 1st derivatives. */
    double term1, term1_sq;
    double term2, term2_sq;
    double term3, term3_sq, term3_cu;
    double term4, term4_sq, term4_cu;

    if (comm->user[0] == -1.0) {
        printf("(User-supplied callback funct, first invocation.)\n");
        fflush(stdout);
        comm->user[0] = 0.0;
    }
    term1 = xc[0] + 10.0 * xc[1];
    term1_sq = term1 * term1;

    term2 = xc[2] - xc[3];
    term2_sq = term2 * term2;

    term3 = xc[1] - 2.0 * xc[2];
    term3_sq = term3 * term3;
    term3_cu = term3 * term3_sq;

    term4 = xc[0] - xc[3];
    term4_sq = term4 * term4;
    term4_cu = term4_sq * term4;

    *fc = term1_sq + 5.0 * term2_sq
          + term3_sq * term3_sq + 10.0 * term4_sq * term4_sq;

    gc[0] = 2.0 * term1 + 40.0 * term4_cu;
    gc[1] = 20.0 * term1 + 4.0 * term3_cu;
    gc[2] = 10.0 * term2 - 8.0 * term3_cu;
    gc[3] = -10.0 * term2 - 40.0 * term4_cu;
}

/* funct */

static void NAG_CALL h(Integer n, const double xc[], double fhed[],
                      double fhed[], Nag_Comm *comm)
{
    /* Routine to evaluate 2nd derivatives */
    double term3_sq;
    double term4_sq;

    if (comm->user[1] == -1.0) {
        printf("(User-supplied callback h, first invocation.)\n");
        fflush(stdout);
        comm->user[1] = 0.0;
    }
    term3_sq = (xc[1] - 2.0 * xc[2]) * (xc[1] - 2.0 * xc[2]);
    term4_sq = (xc[0] - xc[3]) * (xc[0] - xc[3]);

    fhed[0] = 2.0 + 120.0 * term4_sq;
    fhed[1] = 200.0 + 12.0 * term3_sq;
    fhed[2] = 10.0 + 48.0 * term3_sq;
    fhed[3] = 10.0 + 120.0 * term4_sq;
}

```

```

fhesl[0] = 20.0;
fhesl[1] = 0.0;
fhesl[2] = -24.0 * term3_sq;
fhesl[3] = -120.0 * term4_sq;
fhesl[4] = 0.0;
fhesl[5] = -10.0;
}

/* h */

```

## 10.2 Program Data

nag\_opt\_bounds\_2nd\_deriv (e04lbc) Example Program Optional Parameters

```

begin e04lbc
  print_level = Nag_Soln
end

```

## 10.3 Program Results

nag\_opt\_bounds\_2nd\_deriv (e04lbc) Example Program Results

Optional parameter setting for e04lbc.

-----

Option file: e04lbce.opt

print\_level set to Nag\_Soln

Parameters to e04lbc

-----

Number of variables..... 4

optim_tol.....	1.05e-07	linesearch_tol.....	9.00e-01
step_max.....	1.00e+05	max_iter.....	200
print_level.....	Nag_Soln	machine precision.....	1.11e-16
deriv_check.....	Nag_TRUE		
outfile.....	stdout		

Memory allocation:

state.....	Nag		
hesl.....	Nag	hesd.....	Nag

(User-supplied callback funct, first invocation.)  
 (User-supplied callback h, first invocation.)

Final solution:

Itn	Nfun	Objective	Norm g	Norm x	Norm step	Step	CondH	PosDef
10	14	2.4338e+00	1.3e-09	1.5e+00	2.4e-11	1.0e+00	4.4e+00	Yes

Variable	x	g	Status
1	1.0000e+00	2.9535e-01	Lower Bound
2	-8.5233e-02	-5.8675e-10	Free
3	4.0930e-01	1.1735e-09	Free
4	1.0000e+00	5.9070e+00	Lower Bound

Error or warning from nag\_opt\_bounds\_2nd\_deriv (e04lbc).

NW\_COND\_MIN:

The conditions for a minimum have not all been satisfied but a lower point could not be found.

## 11 Optional Parameters

A number of optional input and output arguments to nag\_opt\_bounds\_2nd\_deriv (e04lbc) are available through the structure argument **options**, type Nag\_E04\_Opt. An argument may be selected by assigning an appropriate value to the relevant structure member; those arguments not selected will be assigned default values. If no use is to be made of any of the optional parameters you should use the NAG

defined null pointer, `E04_DEFAULT`, in place of **options** when calling `nag_opt_bounds_2nd_deriv` (e04lbc); the default settings will then be used for all arguments.

Before assigning values to **options** directly the structure **must** be initialized by a call to the function `nag_opt_init` (e04xxc). Values may then be assigned to the structure members in the normal C manner.

Option settings may also be read from a text file using the function `nag_opt_read` (e04xyc) in which case initialization of the **options** structure will be performed automatically if not already done. Any subsequent direct assignment to the **options** structure must **not** be preceded by initialization.

If assignment of functions and memory to pointers in the **options** structure is required, then this must be done directly in the calling program; they cannot be assigned using `nag_opt_read` (e04xyc).

### 11.1 Optional Parameter Checklist and Default Values

For easy reference, the following list shows the members of **options** which are valid for `nag_opt_bounds_2nd_deriv` (e04lbc) together with their default values where relevant. The number  $\epsilon$  is a generic notation for *machine precision* (see `nag_machine_precision` (X02AJC)).

Boolean list	Nag_TRUE
Nag_PrintType print_level	Nag_Soln_Iter
char outfile[80]	stdout
void (*print_fun)()	<b>NULL</b>
Boolean deriv_check	Nag_TRUE
Integer max_iter	<b>50n</b>
double optim_tol	$10\sqrt{\epsilon}$
double linesearch_tol	0.9 (0.0 if $n = 1$ )
double step_max	100000.0
Integer *state	size <b>n</b>
double *hesl	size $\max(n(n-1)/2, 1)$
double *hesd	size <b>n</b>
Integer iter	
Integer nf	

### 11.2 Description of the Optional Parameters

**list** – Nag\_Boolean Default = Nag\_TRUE

*On entry:* if **options.list** = Nag\_TRUE the argument settings in the call to `nag_opt_bounds_2nd_deriv` (e04lbc) will be printed.

**print\_level** – Nag\_PrintType Default = Nag\_Soln\_Iter

*On entry:* the level of results printout produced by `nag_opt_bounds_2nd_deriv` (e04lbc). The following values are available:

Nag_NoPrint	No output.
Nag_Soln	The final solution.
Nag_Iter	One line of output for each iteration.
Nag_Soln_Iter	The final solution and one line of output for each iteration.
Nag_Soln_Iter_Full	The final solution and detailed printout at each iteration.

Details of each level of results printout are described in Section 11.3.

*Constraint:* **options.print\_level** = Nag\_NoPrint, Nag\_Soln, Nag\_Iter, Nag\_Soln\_Iter or Nag\_Soln\_Iter\_Full.

**outfile** – const char[80] Default = stdout

*On entry:* the name of the file to which results should be printed. If **options.outfile**[0] = '\0' then the stdout stream is used.

**print\_fun** – pointer to function Default = NULL

*On entry:* printing function defined by you; the prototype of **options.print\_fun** is

```
void (*print_fun)(const Nag_Search_State *st, Nag_COmm *comm);
```

See Section 11.3.1 below for further details.

**deriv\_check** – Nag\_Boolean Default = Nag\_TRUE

*On entry:* if **options.deriv\_check** = Nag\_TRUE a check of the derivatives defined by **objfun** and **hessfun** will be made at the starting point **x**. A starting point of  $x = 0$  or  $x = 1$  should be avoided if this test is to be meaningful.

**max\_iter** – Integer Default = 50n

*On entry:* the limit on the number of iterations allowed before termination.

*Constraint:* **options.max\_iter**  $\geq 0$ .

**optim\_tol** – double Default =  $10\sqrt{\epsilon}$

*On entry:* the accuracy in  $x$  to which the solution is required. If  $x_{\text{true}}$  is the true value of  $x$  at the minimum, then  $x_{\text{sol}}$ , the estimated position prior to a normal exit, is such that

$$\|x_{\text{sol}} - x_{\text{true}}\| < \mathbf{options.optim\_tol} \times (1.0 + \|x_{\text{true}}\|),$$

where  $\|y\| = \left(\sum_{j=1}^n y_j^2\right)^{1/2}$ . For example, if the elements of  $x_{\text{sol}}$  are not much larger than 1.0 in modulus and if **options.optim\_tol** is set to  $10^{-5}$ , then  $x_{\text{sol}}$  is usually accurate to about five decimal places. (For further details see Section 9.) If the problem is scaled roughly as described in Section 9 and  $\epsilon$  is the *machine precision*, then  $\sqrt{\epsilon}$  is probably the smallest reasonable choice for **options.optim\_tol**. (This is because, normally, to machine accuracy,  $F(x + \sqrt{\epsilon}e_j) = F(x)$  where  $e_j$  is any column of the identity matrix.)

*Constraint:*  $\epsilon \leq \mathbf{options.optim\_tol} < 1.0$ .

**linesearch\_tol** – double Default = 0.9 if  $n > 1$ , and 0.0 otherwise

*On entry:* every iteration of nag\_opt\_bounds\_2nd\_deriv (e04lbc) involves a linear minimization (i.e., minimization of  $F(x + \alpha p)$  with respect to  $\alpha$ ). **options.linesearch\_tol** specifies how accurately these linear minimizations are to be performed. The minimum with respect to  $\alpha$  will be located more accurately for small values of **options.linesearch\_tol** (say 0.01) than for large values (say 0.9).

Although accurate linear minimizations will generally reduce the number of iterations performed by nag\_opt\_bounds\_2nd\_deriv (e04lbc), they will increase the number of function evaluations required for each iteration. On balance, it is usually more efficient to perform a low accuracy linear minimization.

A smaller value such as 0.01 may be worthwhile:

- (a) if **objfun** takes so little computer time that it is worth using extra calls of **objfun** to reduce the number of iterations and associated matrix calculations
- (b) if calls to **hessfun** are expensive compared with calls to **objfun**.
- (c) if  $F(x)$  is a penalty or barrier function arising from a constrained minimization problem (since such problems are very difficult to solve).

If  $n = 1$ , the default for **options.linesearch\_tol** = 0.0 (if the problem is effectively one-dimensional then **options.linesearch\_tol** should be set to 0.0 even though  $n > 1$ ; i.e., if for all except one of the variables the lower and upper bounds are equal).

*Constraint:*  $0.0 \leq \mathbf{options.linesearch\_tol} < 1.0$ .

**step\_max** – double

Default = 100000.0

*On entry:* an estimate of the Euclidean distance between the solution and the starting point supplied by you. (For maximum efficiency a slight overestimate is preferable.) `nag_opt_bounds_2nd_deriv` (e04lbc) will ensure that, for each iteration,

$$\left( \sum_{j=1}^n [x_j^{(k)} - x_j^{(k-1)}]^2 \right)^{1/2} \leq \mathbf{options.step\_max},$$

where  $k$  is the iteration number. Thus, if the problem has more than one solution, `nag_opt_bounds_2nd_deriv` (e04lbc) is most likely to find the one nearest the starting point. On difficult problems, a realistic choice can prevent the sequence of  $x^{(k)}$  entering a region where the problem is ill-behaved and can also help to avoid possible overflow in the evaluation of  $F(x)$ . However, an underestimate of **options.step\_max** can lead to inefficiency.

*Constraint:* **options.step\_max**  $\geq$  **options.optim\_tol**.

**state** – Integer \*Default memory = **n**

*On exit:* **options.state** contains information about which variables are on their bounds and which are free at the final point given in **x**. If  $x_j$  is:

- (a) fixed on its upper bound, **options.state**[ $j - 1$ ] is  $-1$ ;
- (b) fixed on its lower bound, **options.state**[ $j - 1$ ] is  $-2$ ;
- (c) effectively a constant (i.e.,  $l_j = u_j$ ), **options.state**[ $j - 1$ ] is  $-3$ ;
- (d) free, **options.state**[ $j - 1$ ] gives its position in the sequence of free variables.

**hesl** – double \*Default memory =  $\max(\mathbf{n}(\mathbf{n} - 1)/2, 1)$ **hesd** – double \*Default memory = **n**

*On exit:* during the determination of a direction  $p_z$  (see Section 3),  $H + E$  is decomposed into the product  $LDL^T$ , where  $L$  is a unit lower triangular matrix and  $D$  is a diagonal matrix. (The matrices  $H$ ,  $E$ ,  $L$  and  $D$  are all of dimension  $n_z$ , where  $n_z$  is the number of variables free from their bounds.  $H$  consists of those rows and columns of the full second derivative matrix which relate to free variables.  $E$  is chosen so that  $H + E$  is positive definite.)

**options.hesl** and **options.hesd** are used to store the factors  $L$  and  $D$ . The elements of the strict lower triangle of  $L$  are stored row by row in the first  $n_z(n_z - 1)/2$  positions of **options.hesl**. The diagonal elements of  $D$  are stored in the first  $n_z$  positions of **options.hesd**.

In the last factorization before a normal exit, the matrix  $E$  will be zero, so that **options.hesl** and **options.hesd** will contain, on exit, the factors of the final second derivative matrix  $H$ . The elements of **options.hesd** are useful for deciding whether to accept the result produced by `nag_opt_bounds_2nd_deriv` (e04lbc) (see Section 9).

**iter** – Integer

*On exit:* the number of iterations which have been performed in `nag_opt_bounds_2nd_deriv` (e04lbc).

**nf** – Integer

*On exit:* the number of times the residuals have been evaluated (i.e., number of calls of **objfun**).

### 11.3 Description of Printed Output

The level of printed output can be controlled with the structure members **options.list** and **options.print\_level** (see Section 11.2). If **options.list** = Nag\_TRUE then the argument values to `nag_opt_bounds_2nd_deriv` (e04lbc) are listed, whereas the printout of results is governed by the value of **options.print\_level**. The default of **options.print\_level** = Nag\_Soln\_Iter provides a single line of output at each iteration and the final result. This section describes all of the possible levels of results printout available from `nag_opt_bounds_2nd_deriv` (e04lbc).



When **options.print\_level** = Nag\_Iter or Nag\_Soln\_Iter the following line of output is produced on completion of each iteration.

Itn	the iteration count, $k$ .
Nfun	the cumulative number of calls made to <b>objfun</b> .
Objective	the value of the objective function, $F(x^{(k)})$
Norm g	the Euclidean norm of the projected gradient vector, $\ g_z(x^{(k)})\ $ .
Norm x	the Euclidean norm of $x^{(k)}$ .
Norm(x(k-1)-x(k))	the Euclidean norm of $x^{(k-1)} - x^{(k)}$ .
Step	the step $\alpha^{(k)}$ taken along the computed search direction $p^{(k)}$ .
Cond H	the ratio of the largest to the smallest element of the diagonal factor $D$ of the projected Hessian matrix. This quantity is usually a good estimate of the condition number of the projected Hessian matrix. (If no variables are currently free, this value will be zero.)
PosDef	indicates whether the second derivative matrix for the current subspace, $H$ , is positive definite (Yes) or not (No).

When **options.print\_level** = Nag\_Soln\_Iter\_Full more detailed results are given at each iteration. Additional values output are

x	the current point $x^{(k)}$ .
g	the current projected gradient vector, $g_z(x^{(k)})$ .
Status	the current state of the variable with respect to its bound(s).

If **options.print\_level** = Nag\_Soln, Nag\_Soln\_Iter or Nag\_Soln\_Iter\_Full the final result is printed out. This consists of:

x	the final point, $x^*$ .
g	the final projected gradient vector, $g_z(x^*)$ .
Status	the final state of the variable with respect to its bound(s).

If **options.print\_level** = Nag\_NoPrint then printout will be suppressed; you can print the final solution when nag\_opt\_bounds\_2nd\_deriv (e04lbc) returns to the calling program.

### 11.3.1 Output of results via a user-defined printing function

You may also specify your own print function for output of iteration results and the final solution by use of the **options.print\_fun** function pointer, which has prototype

```
void (*print_fun)(const Nag_Search_State *st, Nag_Comm *comm);
```

The rest of this section can be skipped if the default printing facilities provide the required functionality.

When a user-defined function is assigned to **options.print\_fun** this will be called in preference to the internal print function of nag\_opt\_bounds\_2nd\_deriv (e04lbc). Calls to the user-defined function are again controlled by means of the **options.print\_level** member. Information is provided through **st** and **comm**, the two structure arguments to **options.print\_fun**.

If **comm**→**it\_prt** = Nag\_TRUE then the results on completion of an iteration of nag\_opt\_bounds\_2nd\_deriv (e04lbc) are contained in the members of **st**. If **comm**→**sol\_prt** = Nag\_TRUE then the final results from nag\_opt\_bounds\_2nd\_deriv (e04lbc), including details of the final iteration, are contained in the members of **st**. In both cases, the same members of **st** are set, as follows:

**iter** – Integer

The current iteration count,  $k$ , if **comm**→**it\_prt** = Nag\_TRUE; the final iteration count,  $k$ , if **comm**→**sol\_prt** = Nag\_TRUE.

**n** – Integer

The number of variables.

**x** – double \*

The coordinates of the point  $x^{(k)}$ .

**f** – double \*

The value of the objective function at  $x^{(k)}$ .

**g** – double \*

The value of  $\frac{\partial F}{\partial x_j}$  at  $x^{(k)}$ ,  $j = 1, 2, \dots, n$ .

**gpj\_norm** – double

The Euclidean norm of the projected gradient  $g_z$  at  $x^{(k)}$ .

**step** – double

The step  $\alpha^{(k)}$  taken along the search direction  $p^{(k)}$ .

**cond** – double

The estimate of the condition number of the projected Hessian matrix, see Section 11.3.

**xk\_norm** – double

The Euclidean norm of  $x^{(k-1)} - x^{(k)}$ .

**state** – Integer \*

The status of variables  $x_j$ , for  $j = 1, 2, \dots, n$ , with respect to their bounds. See Section 11.2 for a description of the possible status values.

**posdef** – Nag\_Boolean

Will be Nag\_TRUE if the second derivative matrix  $H$  for the current subspace is positive definite, and Nag\_FALSE otherwise.

**nf** – Integer

The cumulative number of calls made to **objfun**.

The relevant members of the structure **comm** are:

**it\_prt** – Nag\_Boolean

Will be Nag\_TRUE when the print function is called with the results of the current iteration.

**sol\_prt** – Nag\_Boolean

Will be Nag\_TRUE when the print function is called with the final result.

**user** – double \*

**iuser** – Integer \*

**p** – Pointer

Pointers for communication of user information. If used they must be allocated memory either before entry to `nag_opt_bounds_2nd_deriv (e04lbc)` or during a call to **objfun** or **options.print\_fun**. The type Pointer will be `void *` with a C compiler that defines `void *` and `char *` otherwise.