LANCELOTsimple: A Simple Interface for LANCELOT B

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LANCELOT_simple: 
A Simple Interface for LANCELOT B

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Abstract

We describe LANCELOT_simple, an interface to the LANCELOT B nonlinear optimization package within the GALAHAD library (Gould, Orban and Toint, 2003) which ignores problem structure. The result is an easy-to-use Fortran 90 subroutine, with a small number of intuitively interpretable arguments. However, since structure is ignored, the means of presenting problems to the solver limited and the choice of algorithmic parameters considerably restricted, the performance is likely to be sub-optimal. The proposed interface may nevertheless be useful for users interested in the solution of simple low-dimensional problems when ease of interface matters more than actual numerical efficiency.

Acknowledgments: Many thanks to Anke Troeltzsch (CERFACS) for her most useful suggestions and to A. Neculai and L. Watson for motivating us to provide this interface.

Résumé

Ce rapport décrit l’interface LANCELOT_simple au logiciel d’optimisation non-linéaire LANCELOT B, une composante de la librairie GALAHAD (Gould et al., 2003). Cette interface ignore la structure du problème. Le résultat est une interface Fortran 90 simple d’utilisation, comportant un petit nombre d’arguments intuitifs. Cependant, la structure étant ignorée, les possibilités de présenter des problèmes au solveur sont limitées et le choix de paramètres algorithmiques est considérablement restreint. Ainsi, la performance s’en trouvera vraisemblablement être sous-optimale. Cette interface sera néanmoins utile aux utilisateurs cherchant à résoudre des problèmes plus simples et de dimensions modestes, lorsque la simplicité d’utilisation prime sur l’efficacité numérique.
1 Introduction

This note presents LANCELOT\_simple, a simple and somewhat naive Fortran 90 interface to the sophisticated optimization package LANCELOT B, whose aim is to solve the nonlinear optimization problem

\[ \min_{x \in \mathbb{R}^n} f(x), \]

possibly subject to constraints of the one or more of the forms

\[ b_l \leq x \leq b_u, \]
\[ c_e(x) = 0, \]
\[ c_i(x) \leq 0, \]

where \( f : \mathbb{R}^n \rightarrow \mathbb{R}, c_e : \mathbb{R}^n \rightarrow \mathbb{R}^m \) and \( c_i : \mathbb{R}^n \rightarrow \mathbb{R}^q \) are twice-continuously differentiable functions, and \( b_l, b_u \) are vectors of \( \mathbb{R}^n \) whose components are allowed to be arbitrarily large in absolute value.

LANCELOT B is one of the packages contained in the GALAHAD software library for numerical nonlinear optimization\(^{(1)}\).

Why do we call this interface naive? At variance with more elaborate interfaces for LANCELOT, the one presented here completely ignores any underlying partial separability or sparsity structure, limits the forms under which the problem can be presented to the solver and drastically restricts the range of available algorithmic options. While undoubtedly simpler to use than its more elaborate counterpart, it may thus offer inferior numerical performance, especially for difficult/large problems, where structure exploitation and/or careful selection of algorithmic variants matter. Thus, be warned that

\[ \text{THE BEST PERFORMANCE OBTAINABLE WITH LANCELOT B IS MOST LIKELY NOT WITH THIS NAIVE INTERFACE.} \]

2 How to use the interface

Before using LANCELOT\_simple, its module must be made available:

```fortran
USE LANCELOT\_simple\_double
```

The complete calling sequence\(^{(2)}\) for LANCELOT\_simple is as follows:

```fortran
CALL LANCELOT\_simple( n, X, fx, exit_code [, MY\_FUN] [,MY\_GRAD] &
[,gradtol] [,feastol] [,print_level] )
```

where the arguments within brackets are optional. The following sections indicate how this calling sequence can be used in commonly occurring contexts.

\(^{(1)}\)See http://galahad.rl.ac.uk/galahad-www/index.html.

\(^{(2)}\)The reference package specification document is available in the GALAHAD installation tree in the file `../doc/lancelot\_simple.pdf`.\n
2.1 Unconstrained minimization problems

The user should provide, at the very least, suitable values for the following input arguments:

\( n \) (integer): the number \( n \) of variables,

\( \mathbf{X} \) (double precision vector of size \( n \)): the starting point for the minimization

In addition the user must provide a subroutine to compute the objective function value at any given \( x \), with the default name and interface

\[
\text{FUN}( \mathbf{X}, f x )
\]

where \( \mathbf{X}(1:n) \) contains the values of the variables \( x \) on input, and \( f x \) is a double precision scalar returning the value \( f(x) \). If no further information is given, the problem is taken to be unconstrained, and derivatives of \( f \) are unavailable (gradients will be estimated by forward finite-differences by the LANCELOT package, and Hessian approximated using the Symmetric-Rank-One quasi-Newton update).

The best value of \( x \) found by LANCELOT B is returned to the user in the vector \( \mathbf{X} \) and the associated objective function value in the double precision output argument \( f x \). In addition, the integer output argument \( \text{exit code} \) contains the exit status of the LANCELOT run, the value 0 indicating a successful run. Other values indicate errors in the input or unsuccessful runs, and are detailed in the specification documentation for LANCELOT B (with the exception of the value 19 in the constrained case discussed later, which reports a negative value for one or both input arguments \( \text{nin} \) and \( \text{neq} \)).

It is also possible to specify a different name for the objective function evaluation subroutine, replacing \( \text{FUN} \) by, say, \( \text{FUNPROB} \). This is accomplished by specifying the optional argument \( \text{MY FUN} \) to have the value \( \text{FUNPROB} \). This is, in turn, best done by including the explicit declaration \( \text{MY FUN} = \text{FUNPROB} \) in the calling sequence to LANCELOT simple after the leading four compulsory arguments, as well as the declaring \( \text{FUNPROB} \) as external. Note that \( \text{FUNPROB} \) must have the same calling sequence as \( \text{FUN} \).

If the user is able to provide the gradient of \( f \), then the optional input argument \( \text{MY GRAD} \) must be specified and given the name of the user-supplied routine for computing the gradient, say \( \text{GRADPROB} \). This subroutine must have an interface of the form

\[
\text{GRADPROB}( \mathbf{X}, G )
\]

where \( G \) is a double precision vector of size \( n \) in which the subroutine returns the value of the gradient of \( f \) at \( x = \mathbf{X} \). The calling sequence to LANCELOT simple must thus contain (in this case), \( \text{MY GRAD} = \text{GRADPROB} \), and a declaration of \( \text{GRADPROB} \) as external.

If, additionally, the user is willing to compute the second-derivative matrix of \( f \) at \( x \), the optional input argument \( \text{MY HESS} \) must be specified and given the name of the user-supplied routine computing the Hessian, say \( \text{HESSPROB} \). This must have an interface of the form

\[
\text{HESSPROB}( \mathbf{X}, H )
\]

where \( H \) is a double precision vector of size \( n \ast (n+1)/2 \) in which the subroutine returns the entries of the upper triangular part of the Hessian of \( f \) at \( x = \mathbf{X} \), stored by columns. The calling sequence to LANCELOT simple must thus contain (in this case), \( \text{MY HESS} = \text{HESSPROB} \), and a declaration of \( \text{HESSPROB} \) as external. Note that first derivatives must also be available in this case, and \( \text{MY GRAD} \) must thus be specified.

The names of the problem variables may be specified in the optional input argument
VNAMES (a size n vector of character fields of length 10)
by inserting VNAMES = mynames in the calling sequence.

One final feature is that by specifying the optional integer argument iters = lanit, the number of iterations performed by LANCELOT is recorded in lanit.

Example
Let us consider the optimization problem

$$
\min_{x_1, x_2} f(x_1, x_2) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2,
$$

which is the renowned Rosenbrock “banana” problem. The most basic way to solve the problem (but not the most efficient) is, given the starting point \( X = (/ -1.2d0, 1.0d0 /) \), to

```
CALL LANCELOT_simple( 2, X, fx, exit_code )
```

where the function FUN is given by

```
SUBROUTINE FUN( X, F )
  INTEGER, PARAMETER :: wp = KIND( 1.0D+0 )
  REAL( KIND = wp ), INTENT( IN ) :: X( : )
  REAL( KIND = wp ), INTENT( OUT ) :: F
  F = 100.0_wp*(X(2)-X(1)**2)**2 +(1.0_wp-X(1))**2
RETURN
END SUBROUTINE FUN
```

After compiling and linking (with the GALAHAD modules and library), the solution is returned in 60 iterations (with exit_code = 0). The same effect is obtained by the call

```
CALL LANCELOT_simple( 2, X, fx, exit_code, MY_FUN = FUN )
```

but this syntax allows the change of name from FUN to, say, ROSEN BROCK_FUN in the user-supplied program, as discussed above.

If we now wish to use first and second derivatives of the objective function, one could use the call

```
CALL LANCELOT_simple( 2, X, fx, exit_code, MY_FUN = FUN, &
  MY_GRAD = ROSEN BROCK_GRAD, &
  MY_HESS = ROSEN BROCK_HESS )
```

provide the additional routines

```
SUBROUTINE ROSEN BROCK_GRAD( X, G )
  INTEGER, PARAMETER :: wp = KIND( 1.0D+0 )
  REAL( KIND = wp ), INTENT( IN ) :: X( : )
  REAL( KIND = wp ), INTENT( OUT ) :: G( : )
  G( 1 ) = -400.0_wp*(X(2)-X(1)**2)*X(1)-2.0_wp*(1.0_wp-X(1))
  G( 2 ) = 200.0_wp*(X(2)-X(1)**2)
RETURN
END SUBROUTINE ROSEN BROCK_GRAD
```

and
and declare them external in the calling program.

Convergence is then obtained in 23 iterations. Note that using exact first-derivatives only is also possible: `MY_HESS` should then be absent from the calling sequence and providing the subroutine `ROSENBROCK_HESS` is unnecessary.

### 2.2 Bound constrained minimization problems

Bounds on the problem variables may be imposed by specifying one or both of

- **BL** (*double precision vector of size n*): the lower bounds \( b_l \) on the variables,
- **BU** (*double precision vector of size n*): the upper bounds \( b_u \) on the variables.

Note that infinite bounds (represented by a number larger than \( 10^{20} \) in absolute value) are acceptable, as well as equal lower and upper bounds, which amounts to fixing the corresponding variables. Except for the specification of **BL** and/or **BU**, the interface is identical to that for unconstrained problems described in the previous paragraph.

**Example**

If one now wishes to impose zero upper bounds on the variables of our unconstrained problem and give names to the variables and the objective function, one could use the following call

```fortran
CALL LANCELOT_simple( 2, X, fx, exit_code, MY_FUN = FUN, &
  MY_GRAD = ROSENBROCK_GRAD, &
  MY_HESS = ROSENBROCK_HESS, &
  BU = (/ 0.0d0, 0.0d0 /), &
  VNAMES = VNAMES )
```

in which case convergence is obtained in 6 iterations.

### 2.3 Equality constrained minimization problems

If, additionally, general equality constraints are also present in the problem, this must be declared by specifying the following additional optional input argument

- **neq** (*integer*): the number \( m \) of equality constraints.

In this case, the equality constraints are numbered from 1 to \( \text{neq} \) and the value of the \( i \)-th equality constraint must be computed by a user-supplied routine of the form

```fortran
FUNPROB( X, fx, i )   \quad ( i = 1, \ldots, \text{neq} )
```
where the $f_x$ now returns the value of the $i$-th equality constraint evaluated at $x = X$ if $i$ is specified. (This extension of the unconstrained case is best implemented by adding an optional argument $i$ to the unconstrained version of $\text{FUN}$. ) If derivatives are available, then the $\text{GRADPROB}$ and (possibly) $\text{HESSPROB}$ subroutines must be adapted as well

\[
\text{GRADPROB}( X, G, i ) \quad \text{HESSPROB}( X, H, i ) \quad (i = 1, \ldots, \text{neq})
\]

to compute the gradient and Hessian of the $i$-th constraint at $x = X$. The constraints may be assigned names in the optional input argument $\text{CNAMES}$ (a size $\text{neq}$ array of character fields of length 10)

The name of constraint $i$ should appear in $\text{CNAMES}(i)$ for each $i = 1, \ldots, \text{neq}$.

Note that, if the gradient of the objective function is provided, so must the gradients of the equality constraints. $\text{The same level of derivative availability is assumed for all problem functions (objective and constraints).}$ The final values of the constraints and the values of their associated Lagrange multipliers are optionally returned to the user in the double precision output arguments $\text{CX}$ and $\text{Y}$, respectively (both being of size $\text{neq}$).

### 2.4 Inequality constrained minimization problems

If inequality constraints are present in the problem, they are included just as for equality constraints. One needs to specify the optional input argument

$nin$ (integer): the number $q$ of inequality constraints.

The inequality constraints are then numbered from 1 to $\text{nin}$ and their values or that of their derivatives is again computed by calling, for $i = 1, \ldots, \text{nin}$,

\[
\text{FUNPROB}( X, f_x, i ) \quad \text{GRADPROB}( X, G, i ) \quad \text{HESSPROB}( X, H, i )
\]

The inequality constraints are internally converted to equations by adding slack variables, whose names are set to $\text{Slack}_i$, where the character $i$ in this string takes the integers values 1 to $\text{nin}$.) As in the equality-constrained case, the name of constraint $i$ ($i = 1, \ldots, \text{nin}$) may be specified in $\text{CNAMES}(i)$. The values of the inequality constraints at the final $X$ are returned (as for equalities) in the optional double precision output argument $\text{CX}$ of size $\text{nin}$. The values of the corresponding Lagrange multipliers are returned in the optional double precision output argument $\text{Y}$ of size $\text{nin}$.

### 2.5 Minimization problems with equality and inequality constraints

If there are both equalities and inequalities, $\text{neq}$ and $\text{nin}$ must be specified and the values and derivatives of the constraints are computed by

\[
\text{FUNPROB}( X, f_x, i ) \quad \text{GRADPROB}( X, G, i ) \quad \text{HESSPROB}( X, H, i ) \quad (i=1,\ldots,\text{neq})
\]

for the equality constraints, and

\[
\text{FUNPROB}( X, f_x, i ) \quad \text{GRADPROB}( X, G, i ) \quad \text{HESSPROB}( X, H, i ) \quad (i=\text{neq}+1,\text{neq}+\text{nin})
\]

for the inequality constraints. The components $i = 1, \ldots, \text{neq}$ of the vector $\text{CNAMES}$ now contain the names of the $\text{neq}$ equality constraints and components $i = \text{neq} + 1, \ldots, \text{neq} + \text{nin}$, the names of the $\text{nin}$ inequality constraints. Again, the same level of derivative availability is assumed for all problem functions (objective and constraints). Finally, the optional arguments $\text{CX}$ and/or $\text{Y}$, if used, now have size $\text{neq}+\text{nin}$. 
2.6 An example

Suppose we now consider the problem

\[
\min_{x_1, x_2} f(x_1, x_2) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2,
\]

(2.1)

under the constraints

\[
0 \leq x_1, \quad x_1 + 3x_2 - 3 = 0, \quad x_1^2 + x_2^2 - 4 \leq 0
\]

(2.2)

To solve this problem, we now use the call

```fortran
CALL LANCELOT_simple( 2, X, fx, exit_code, BL = (/0.0D0, -1.0D20/), &
                      MY_FUN = FUN, MY_GRAD = GRAD, MY_HESS = HESS, &
                      neq = 1, nin = 1, CX = cx, Y = y ) &
```

(assuming we wish to examine \(cx\) and \(y\)), and modify the \(FUN\), \(GRAD\) and \(HESS\) functions as follows

```fortran
SUBROUTINE FUN ( X, F, i )
  INTEGER, PARAMETER :: wp = KIND( 1.0D+0 ) ! set precision
  REAL( KIND = wp ), INTENT( IN ) :: X( : )
  REAL( KIND = wp ), INTENT( OUT ) :: F
  INTEGER, INTENT( IN ), OPTIONAL :: i
  IF ( .NOT. PRESENT( i ) ) THEN
    F = 100.0_wp*(X(2)-X(1)**2)**2 + (1.0_wp-X(1))**2
  ELSE
    SELECT CASE ( i )
    CASE ( 1 )
      F = X(1)+3.0_wp*X(2)-3.0_wp
    CASE ( 2 )
      F = X(1)**2+X(2)**2-4.0_wp
    END SELECT
  END IF
RETURN
END SUBROUTINE FUN

SUBROUTINE GRAD( X, G, i )
  INTEGER, PARAMETER :: wp = KIND( 1.0D+0 ) ! set precision
  REAL( KIND = wp ), INTENT( IN ) :: X( : )
  REAL( KIND = wp ), INTENT( OUT ) :: G( : )
```
INTEGER, INTENT(IN), OPTIONAL :: i
IF (.NOT. PRESENT(i)) THEN
!
the objective function’s gradient components (user defined)
!
G(1) = -400.0_wp*(X(2)-X(1)**2)*X(1)-2.0_wp*(1.0_wp-X(1))
G(2) = 200.0_wp*(X(2)-X(1)**2)
!
ELSE
SELECT CASE (i)
CASE (1)
!
the equality constraint’s gradient components (user defined)
!
G(1) = 1.0_wp
G(2) = 3.0_wp
!
CASE (2)
!
the inequality constraint’s gradient components (user defined)
!
G(1) = 2.0_wp*X(1)
G(2) = 2.0_wp*X(2)
!
END SELECT
END IF
RETURN
END SUBROUTINE GRAD
!
-----------
SUBROUTINE HESS(X, H, i)
-----------
INTEGER, PARAMETER :: wp = KIND(1.0D+0) ! set precision
REAL(KIND=wp), INTENT(IN) :: X(:)
REAL(KIND=wp), INTENT(OUT) :: H(:)
INTEGER, INTENT(IN), OPTIONAL :: i
IF (.NOT. PRESENT(i)) THEN
!
the entries of the upper triangle of the objective function’s Hessian matrix, stored by columns (user defined)
!
H(1) = -400.0_wp*(X(2)-3.0_wp*X(1)**2)+2.0_wp
H(2) = -400.0_wp*X(1)
H(3) = 200.0_wp
!
ELSE
SELECT CASE (i)
CASE (1)
!
the entries of the upper triangle of the equality constraint’s Hessian matrix, stored by columns (user defined)
!
H(1) = 0.0_wp
H(2) = 0.0_wp
H(3) = 0.0_wp
!
CASE (2)
the entries of the upper triangle of the inequality
constraint’s Hessian matrix, stored by columns (user defined)

\[
\begin{align*}
H(1) &= 2.0 \_wp \\
H(2) &= 0.0 \_wp \\
H(3) &= 2.0 \_wp 
\end{align*}
\]

END SELECT
END IF
RETURN
END SUBROUTINE HESS

Convergence is then obtained in 8 iterations. Note that, in our example, the objective function or its derivatives is/are computed if the index \( i \) is omitted.

3 Available algorithmic options

Beyond the choice of derivative level for the problem functions, the following optional arguments allow a (very limited) control of the algorithmic choices available in LANCELOT:

- \texttt{maxit (integer)}: maximum number of iterations (default if absent: 1000)
- \texttt{gradtol (double precision)}: the threshold on the maximum infinity norm of the gradient (or of the Lagrangian’s gradient) required for declaring convergence (default if absent: 1.0d-5),
- \texttt{feastol (double precision)}: the threshold on the maximum infinity norm of the constraint violation required for declaring convergence (for constrained problems) (default if absent: 1.0d-5),
- \texttt{print_level (integer)}: a positive number proportional to the amount of output by the package: 0 corresponds to the silent mode, 1 to a single line of information per iteration (the default if absent), while higher values progressively produce more output.

Users should keep in mind that more elaborate algorithmic options are available when using more elaborate interfaces to the LANCELOT code within GALAHAD.

The full call corresponding to our constrained example could be

\[
\text{CALL LANCELOT\_simple( 2, X, fx, exit\_code, \\
FUN = FUN, GRAD = GRAD, HESS = HESS, \\
BL = BL, BU = BU, VNAMES = VNAMES, \\
CNAMES = CNAMES, neq = 1, nin = 1, \\
CX = cx, Y = y, ITERS = iters, maxit = 100, \\
gradtol = 1.0d-5, feastol = 1.0d-5, \\
print\_level = 1 )}
\]

4 Compiling and linking

Of course, the interface described above is an interface within the GALAHAD library. Its use therefore requires the \textit{a priori} installation of this library and the necessary mechanisms to access its content.

At compile time, the Fortran 90 compiler must be informed of the particular module where the interface can be found. Assuming GALAHAD is properly installed in double precision and that the
environmental variable\(^{(3)}\) GALAHAD has been set to point to the main GALAHAD directory (see Gould et al., 2003 for details), the relevant Fortran 90 module can be found in the directory

\[
\text{GALAHAD/modules/pc.lnx.195/double}
\]

where the string \text{pc.lnx.195} indicates the particular architecture under which the GALAHAD library has been installed (in this case, a PC under Linux and the Lahy lf95 Fortran 90 compiler). This information needs to be provided to the compiler via an architecture-dependent flag (typically \text{-I, -M or -module}; see the makefile variable \text{MODULES} for your architecture in the directory \text{GALAHAD/makefiles}).

Similarly, the linker must be able to include the relevant GALAHAD objects into the executable code. These objects are the GALAHAD library proper, as well as the BLAS, LAPACK, HSL and METIS libraries, which may all be found (in their public domain versions) in the directory

\[
\text{GALAHAD/objects/pc.lnx.195/double}
\]

The link command should therefore include a string of the form

\[
\text{-L/GALAHAD/objects/pc.lnx.195/double -lgalahad -lgalahad_hsl} \text{ \\ \ \text{-lgalahad_metis -lgalahad_lapack -lgalahad_blas}}
\]

for proper execution.

5 Single precision version

A single precision version is also available in the GALAHAD library. All real \text{LANCELOT_simple} should then be defined in single precision.

6 Other sources of information

The user is encouraged to consult the specsheet of the (non-naive) interface to LANCELOT within the GALAHAD software library for a better view of all possibilities offered by an intelligent use of the package. The library is described in Gould et al. (2003). The book by Conn, Gould and Toint (1992) is also a good source of additional information.

References


\(^{(3)}\)in the Bourne or C-shell.
Appendix: example calling code and makefile

In order to clarify the use of the interface as much as possible, we provide in this appendix the full Fortran 90 code of a calling program corresponding to our constrained example (2.1)-(2.2) of Section 2.6. The necessary subroutines FUN, GRAD and HESS are those of Section 2.6 and are assumed to be appended to calling program. We also give a makefile for compiling/linking this program into an executable file. These are intended as a base which can be modified to suit particular user needs.

The code of the calling program⁴ is the following.

```fortran
PROGRAM RUN_LANCELOT_simple

! This program provides a simple (naive) way to run LANCELOT B on an optimization problem without interaction with CUTEr and using a simple representation of the problem where the objective function and constraints comprise each a single group with a single nonlinear element.

! The use of this program and the accompanying script lanb_simple
! requires that the GALAHAD package has prealably be installed.

! Ph. L. Toint, November 2007.
! Copyright reserved, Gould/Orban/Toint, for GALAHAD productions

USE LANCELOT_simple_double

IMPLICIT NONE
INTEGER, PARAMETER :: wp = KIND( 1.0D+0 ) ! set precision
REAL ( KIND = wp ), PARAMETER :: infinity = 10.0_wp ** 20
INTEGER :: n, neq, nin, iters, maxit, print_level, exit_code
REAL ( KIND = wp ) :: gradtol, feastol, fx
CHARACTER ( LEN = 10 ), ALLOCATABLE, DIMENSION(:) :: VNAMES, CNAMES
REAL ( KIND = wp ), ALLOCATABLE, DIMENSION(:) :: BL, BU, X, CX, Y
EXTERNAL :: FUN, GRAD, HESS

! THE TEST PROBLEM DIMENSIONS (user defined)
!
! n = 2 ! number of variables
! neq = 1 ! number of equality constraints, excluding fixed variables
! nin = 1 ! number of inequality (<= 0) constraints, excluding bounds
!
allocate space for problem defining vectors
!
ALLOCATE( X( n ), BL( n ), BU( n ), CX( neq+nin ), Y( neq+nin ) )
ALLOCATE( VNAMES( n ),CNAMES( neq+nin ) )
!
starting point
```

⁴This code is available in the GALAHAD installation tree in the file ../src/lanb_simple/run_lancelot_simple.f90.
The makefile\(^{(5)}\) is then given for the specific default DOUBLE PRECISION case where

- the Lahay lf95 Fortran compiler is used,
- the architecture specified at the installation of the GALAHAD library is pc.linx.l95 (running on a PC under linux with lf95),
- the environment variable \$GALAHAD has been set properly (as recommended in the installation of the GALAHAD library).

\(^{(5)}\)This makefile is available in the GALAHAD installation tree in the file ../src/lanb_simple/bin/Makefile.
Modifying this makefile for another example, another compiler or another precision requires (possibly) adapting the definitions of the name of the calling program (here run_lancelot_simple.f90), the variables MAIN, USRLIBS, FORTRAN, ARCH and PREC. The options in the compiling and linking commands may also vary for other Fortran 90 compilers. This can be achieved by issuing commands of the type

```
make FORTRAN=g95 ARCH=mac.osx.g95 PREC=single MAIN=spamalot USRLIBS=-lg2c
```

In our example (which corresponds to the default settings in the makefile), the simple commands

```
make run_lancelot_simple
```

produce the following output:
Penalty parameter  $1.0000E-01$ Required projected gradient norm = $1.0000E-01$
Required constraint norm = $1.0000E-01$

******** Starting optimization ********

There are 3 variables
There are 3 groups
There are 3 nonlinear elements

Objective function value $1.01000000000000E+02$
Constraint norm $2.00000000000000E+00$

<table>
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<th>c.g.it</th>
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Iteration number 6 Merit function value = $2.31890590071E-02$
No. derivative evals 7 Projected gradient norm = $2.08773257903E-02$
C.G. iterations 6 Trust-region radius = $1.73866572139E+00$
Number of updates skipped 0

There are 3 variables and 0 active bounds

Times for Cauchy, systems, products and updates 0.00 0.00 0.00 0.00

Exact Cauchy step computed
Bandsolver preconditioned C.G. ( semi-bandwidth = 5 ) used
Infinity-norm trust region used
Non-monotone descent strategy ( history = 1 ) used
Exact second derivatives used

Objective function value $2.30556773462821E-02$
Penalty parameter $1.0000E-01$
Projected gradient norm $2.0877E-02$ Required gradient norm = $1.0000E-01$
Constraint norm $5.1647E-03$ Required constraint norm = $1.0000E-01$

******** Updating multiplier estimates ********

Penalty parameter $1.0000E-01$ Required projected gradient norm = $1.0000E-02$
Required constraint norm = $1.2589E-02$

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<th>f</th>
<th>proj.g</th>
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Iteration number 7 Merit function value = $2.34466960383E-02$
No. derivative evals 8 Projected gradient norm = $3.42123908435E-04$
C.G. iterations 7 Trust-region radius = $1.00000000000E+00$
Number of updates skipped 0
There are 3 variables and 0 active bounds

Times for Cauchy, systems, products and updates 0.00 0.00 0.00 0.00

Exact Cauchy step computed
Bandsolver preconditioned C.G. ( semi-bandwidth = 5 ) used
Infinity-norm trust region used
Non-monotone descent strategy ( history = 1 ) used
Exact second derivatives used

Objective function value 2.3321093681760E-02

Penalty parameter = 1.0000E-01
Projected gradient norm = 3.4212E-04 Required gradient norm = 1.0000E-02
Constraint norm = 1.5101E-04 Required constraint norm = 1.2589E-02

******** Updating multiplier estimates ********

Penalty parameter 1.0000E-01 Required projected gradient norm = 1.0000E-03
Constraint norm = 1.5849E-03

Iter #g.ev c.g.it f proj.g rho radius step cgend #free time
7 8 7 2.34E-02 5.2E-03 - - - - 3 0.0
8 9 8 2.34E-02 2.5E-07 1.0E+00 1.0E+00 6.2E-05 CONVR 3 0.0

Iteration number 8 Merit function value = 2.34391260921E-02
No. derivative evals 9 Projected gradient norm = 2.45864887574E-07
C.G. iterations 8 Trust-region radius = 1.00000000000E+00

Number of updates skipped 0

There are 3 variables and 0 active bounds

Times for Cauchy, systems, products and updates 0.00 0.00 0.00 0.00

Exact Cauchy step computed
Bandsolver preconditioned C.G. ( semi-bandwidth = 5 ) used
Infinity-norm trust region used
Non-monotone descent strategy ( history = 1 ) used
Exact second derivatives used

Objective function value 2.33135045070784E-02

Penalty parameter = 1.0000E-01
Projected gradient norm = 2.4586E-07 Required gradient norm = 1.0000E-03
Constraint norm = 2.9716E-06 Required constraint norm = 1.5849E-03

Variable name Number Status Value Lower bound Upper bound | Dual value
------------- ------ -------- ----------- ----------- ---------- - | ----------
x1 1 FREE 8.4750E-01 0.0000E+00 1.0000E+20 | 2.4586E-07
x2 2 FREE 7.1750E-01 -1.0000E+20 3.0000E+00 | -5.0502E-08
Slack_1 3 FREE 2.7669E+00 0.0000E+00 1.0000E+20 | 2.3331E-08

Constraint name Number Value Scale factor | Lagrange multiplier
------------- ------ -------- ------------- |------------------
Equality 1 -1.2974E-06 1.0000E+00 | 5.0124E-02
Inequality 2 -2.9716E-06 1.0000E+00 | 2.3343E-08

Objective function value 2.33135045070784E-02

There are 3 variables in total.
There are 2 equality constraints.
Of these 1 are primal degenerate.
There are 0 variables on their bounds.
Of these 0 are dual degenerate.