

Description

`lsglmolve(A, B, d, x, y)` finds the solution to a general Gauss–Markov linear model: $\min_x \|y\|_2$ subject to $d = Ax + By$ with no return value.

`_lsglmolve(A, B, d, x, y)` does the same thing except that it returns 0 if a solution was found and 1 otherwise. If 1 is returned, *x* and *y* are overwritten with vectors of missing values.

`_lsglm_lapacke(A, B, d, x, y)` is the interface to the LAPACK routines that do the work. It returns 0 if a solution was found and 1 otherwise. Direct use of `_lsglm_lapacke()` is not recommended.

Note that these functions can be used only when set `lapack_mk1` on is in effect on Windows or Linux or when set `lapack_openblas` on is in effect on Mac; see [M-1] [LAPACK](#).

Syntax

void `lsglmolve(A, B, d, x, y)`

real scalar `_lsglmolve(A, B, d, x, y)`

real scalar `_lsglm_lapacke(A, B, d, x, y)`

where inputs are

A: $n \times m$ numeric matrix

B: $n \times p$ numeric matrix

d: $n \times 1$ or $1 \times n$ numeric vector

and outputs are

x: $m \times 1$ numeric vector

y: $p \times 1$ numeric vector

result: *real scalar*

where $m \leq n \leq m + p$, the rank of matrix *A* is *m*, and the rank of the following matrix is *n*:

$$\begin{bmatrix} A & B \\ n \times m & n \times p \end{bmatrix}$$

Remarks and examples

Remarks are presented under the following headings:

[Introduction](#)

[Examples](#)

Introduction

The above functions solve a general Gauss–Markov linear model: $\min_x \|y\|_2$ subject to $d = Ax + By$.

To obtain a unique solution, the functions require that

1. $m \leq n \leq m + p$,
2. the rank of matrix A is m , and
3. the rank of the following matrix is n :

$$\begin{bmatrix} A & B \\ n \times m & n \times p \end{bmatrix}$$

The solution is found with the underlying LAPACK routines using a generalized QR factorization of (A, B) .

Examples

▷ Example 1: Least squares with equality constraints

Given A , B , and d , we can find x and y , satisfying $\min_x \|y\|_2$ subject to $Ax + By = d$ using

```
: A = (3, 2 \ 2, 10 \ 1, 1.5)
```

```
: B = (2, 5 \ 1, 6 \ 5, 2)
```

```
: d = (10 \ 2 \ 1)
```

```
: A
```

	1	2
1	3	2
2	2	10
3	1	1.5

```
: B
```

	1	2
1	2	5
2	1	6
3	5	2

```
: d
```

	1
1	10
2	2
3	1

```
: lsglmsolve(A, B, d, x = ., y = .)
```

```

: x
      1
1  4.00221166
2  -.5527209235

: y
      1
1  -.4315840528
2  -.0076050053

```

We can also use the `_lsjlsolve()` function to get the same solution as above and a return code of 0:

```

: A = (3, 2 \ 2, 10 \ 1, 1.5)
: B = (2, 5 \ 1, 6 \ 5, 2)
: d = (10 \ 2 \ 1)
:
: _lsjlsolve(A, B, d, x = ., y = .)
0
: x
      1
1  4.00221166
2  -.5527209235

: y
      1
1  -.4315840528
2  -.0076050053

```



Conformability

`lsglmsolve(A, B, d, x, y)`:

input:

A: $n \times m$
B: $n \times p$
d: $n \times 1$ or $1 \times n$

output:

x: $m \times 1$
y: $p \times 1$

`_lsglmsolve(A, B, d, x, y)`:

input:

A: $n \times m$
B: $n \times p$
d: $n \times 1$ or $1 \times n$

output:

x: $m \times 1$
y: $p \times 1$
result: 1×1

`_lsglm_lapacke(A, B, d, x, y)` :

input:

A: $n \times m$
B: $n \times p$
d: $n \times 1$ or $1 \times n$

output:

x: $m \times 1$
y: $p \times 1$
result: 1×1

Diagnostics

`lsglmsolve(A, B, ...)`, `_lsglmsolve(A, B, ...)`, and `_lsglm_lapacke(A, B, ...)` return a result containing missing if *A*, *B*, or *d* contains missing values. If the conditions in [Introduction](#) above are not satisfied, the functions will try to find a solution, which will either produce unstable results or abort with error. The functions abort with error if set `lapack_mkl` on is not in effect on Windows or Linux or when set `lapack_openblas` on is not in effect on Mac.

`_lsglmsolve(A, B, ...)` and `_lsglm_lapacke(A, B, ...)` abort with error if *A*, *B*, or *d* is a view.

`_lsglm_lapacke(A, B, ...)` aborts with error if *A*, *B*, and *d* are not all real or all complex.

`_lsglm_lapacke(A, B, ...)` should not be used directly; use `_lsglmsolve()`.

Also see

[M-5] **cholsolve()** — Solve $AX=B$ for X using Cholesky decomposition

[M-5] **lsesolve()** — Solve $AX=c$ for x with equality constraints using least-squares method

[M-5] **lssolve()** — Solve $AX=B$ for X using least-squares method

[M-5] **lusolve()** — Solve $AX=B$ for X using LU decomposition

[M-5] **qrsolve()** — Solve $AX=B$ for X using QR decomposition

[M-5] **_solvemmat()** — Solve $AX=B$ for X

[M-5] **svsolve()** — Solve $AX=B$ for X using singular value decomposition

[M-4] **Matrix** — Matrix functions

[M-4] **Solvers** — Functions to solve $AX=B$ and to obtain A inverse

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