

svsolve() — Solve $AX=B$ for X using singular value decomposition

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Description

`svsolve(A, B, ...)`, uses singular value decomposition to solve $AX = B$ and return X . When A is singular, `svsolve()` computes the minimum-norm least-squares generalized solution. When *rank* is specified, in it is placed the rank of A .

`_svsolve(A, B, ...)` does the same thing, except that it destroys the contents of A and it overwrites B with the solution. Returned is the rank of A .

In both cases, *tol* specifies the tolerance for determining whether A is of full rank. *tol* is interpreted in the standard way—as a multiplier for the default if $tol > 0$ is specified and as an absolute quantity to use in place of the default if $tol \leq 0$ is specified.

Syntax

numeric matrix `svsolve(A, B)`
numeric matrix `svsolve(A, B, rank)`
numeric matrix `svsolve(A, B, rank, tol)`

real scalar `_svsolve(A, B)`
real scalar `_svsolve(A, B, tol)`

where

A: *numeric matrix*
B: *numeric matrix*
rank: irrelevant; *real scalar* returned
tol: *real scalar*

Remarks and examples

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`svsolve(A, B, ...)` is suitable for use with square or nonsquare, full-rank or rank-deficient matrix A . When A is of full rank, `svsolve()` returns the same solution as `lusolve()` (see [M-5] [lusolve\(\)](#)), ignoring roundoff error. When A is singular, `svsolve()` returns the minimum-norm least-squares generalized solution. `qrsolve()` (see [M-5] [qrsolve\(\)](#)), an alternative, returns a generalized least-squares solution that amounts to dropping rows of A .

Remarks are presented under the following headings:

[Derivation](#)
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Derivation

We wish to solve for X

$$AX = B \tag{1}$$

Perform singular value decomposition on A so that we have $A = USV'$. Then (1) can be rewritten as

$$USV'X = B$$

Premultiplying by U' and remembering that $U'U = I$, we have

$$SV'X = U'B$$

Matrix S is diagonal and thus its inverse is easily calculated, and we have

$$V'X = S^{-1}U'B$$

When we premultiply by V , remembering that $VV' = I$, the solution is

$$X = VS^{-1}U'B \tag{2}$$

See [M-5] `svd()` for more information on the SVD.

Relationship to inversion

For a general discussion, see *Relationship to inversion* in [M-5] `lusolve()`.

For an inverse based on the SVD, see [M-5] `pinv()`. `pinv(A)` amounts to `svsolve(A, I(rows(A)))`, although `pinv()` has separate code that uses less memory.

Tolerance

In (2) above, we are required to calculate the inverse of diagonal matrix S . The generalized solution is obtained by substituting zero for the i th diagonal element of S^{-1} , where the i th diagonal element of S is less than or equal to eta in absolute value. The default value of eta is

$$eta = \text{epsilon}(1) * \text{rows}(A) * \max(S)$$

If you specify $tol > 0$, the value you specify is used to multiply eta . You may instead specify $tol \leq 0$ and then the negative of the value you specify is used in place of eta ; see [M-1] **tolerance**.

Conformability

`svsolve(A, B, rank, tol):`

input:

A: $m \times n$
B: $m \times k$
tol: 1×1 (optional)

output:

rank: 1×1 (optional)
result: $n \times k$

`_svsolve(A, B, tol):`

input:

A: $m \times n$
B: $m \times k$
tol: 1×1 (optional)

output:

A: 0×0
B: $m \times k$
result: 1×1

Diagnostics

`svsolve(A, B, ...)` and `_svsolve(A, B, ...)` return missing results if A or B contain missing.

`_svsolve(A, B, ...)` aborts with error if A (but not B) is a view.

Also see

[M-5] `solverlower()` — Solve $AX=B$ for X , A triangular

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

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

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

[M-4] `matrix` — Matrix functions

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