

Package ‘nlsic’

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Title Non Linear Least Squares with Inequality Constraints

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Description We solve non linear least squares problems with optional equality and/or inequality constraints. Non linear iterations are globalized with back-tracking method. Linear problems are solved by dense QR decomposition from 'LAPACK' which can limit the size of treated problems. On the other side, we avoid condition number degradation which happens in classical quadratic programming approach. Inequality constraints treatment on each non linear iteration is based on 'NNLS' method (by Lawson and Hanson). We provide an original function 'lsi_ln' for solving linear least squares problem with inequality constraints in least norm sens. Thus if Jacobian of the problem is rank deficient a solution still can be provided. However, truncation errors are probable in this case. Equality constraints are treated by using a basis of Null-space. User defined function calculating residuals must return a list having residual vector (not their squared sum) and Jacobian. If Jacobian is not in the returned list, package 'numDeriv' is used to calculate finite difference version of Jacobian. The 'NLSIC' method was first published in Sokol et al. (2012) <doi:10.1093/bioinformatics/btr716>.

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Encoding UTF-8

RoxygenNote 7.3.2

Depends nnls, dotty

Suggests numDeriv, RUnit, limSolve

URL <https://github.com/MathsCell/nlsic>

BugReports <https://github.com/MathsCell/nlsic/issues>

NeedsCompilation no

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equa2vecmat	<i>Parse linear equations/inequalities</i>
-------------	--

Description

parse a text vector of linear equations and produce a corresponding matrix and right hand side vector

Usage

```
equa2vecmat(nm_par, linear, sep = "=")
```

Arguments

- nm_par a string vector of variable names. It will be used in the symbolic derivation.
- linear string vector of linear equations like "a+2*c+3*b = 0"
- sep separator of two parts of equations. Use for example ">=" for linear inequalities

Value

an augmented matrix. Its first column is the rhs vector. Other columns are named by nm_par. If the vector linear is NULL or its content is empty a NULL is returned

Examples

```
equa2vecmat(c("a", "b", "c"), "a+2*c+3*b = 0", "=")
```

join	<i>Join elements into a string</i>
------	------------------------------------

Description

convert elements of vector *v* (and all following arguments) in strings and join them using *sep* as separator.

Usage

```
join(sep, v, ...)
```

Arguments

<i>sep</i>	A string used as a separator
<i>v</i>	A string vector to be joined
<i>...</i>	other variables to be converted to strings and joined

Value

A joined string

Examples

```
join(" ", c("Hello", "World"))
```

ldp	<i>Least Distance Problem</i>
-----	-------------------------------

Description

Solve least distance programming: find *x* satisfying $u \leq x \leq co$ and s.t. $\min(\|x\|)$ by passing to nnls (non negative least square) problem.

Usage

```
ldp(u, co, rcond = 1e+10)
```

Arguments

<i>u</i>	a dense matrix of inequality constraints
<i>co</i>	a right hand side vector of inequality constraints
<i>rcond</i>	maximal condition number for determining rank deficient matrix

Value

solution vector or NULL if constraints are unfeasible

lsi

*Linear Least Squares with Inequality constraints (LSI)***Description**

solve linear least square problem ($\min \|A\%x-b\|$) with *inequality constraints* $u\%x \geq co$

Usage

```
lsi(a, b, u = NULL, co = NULL, rcond = 1e+10, mnorm = NULL, x0 = NULL)
```

Arguments

a	dense matrix A or its QR decomposition
b	right hand side vector. Rows containing NA are dropped.
u	dense matrix of inequality constraints
co	right hand side vector of inequality constraints
rcond	maximal condition number for determining rank deficient matrix
mnorm	dummy parameter
x0	dummy parameter

Details

Method:

1. reduce the problem to ldp ($\min(xat*xa) \Rightarrow$ least distance programming)
2. solve ldp
3. change back to x
If b is all NA, then a vector of NA is returned.

mnorm, and x0 are dummy parameters which are here to make lsi() compatible with lsi_ln() argument list

Value

solution vector whose attribute 'mes' may contain a message about possible numerical problems

See Also

[lsi_ln](#), [ldp](#), [base::qr](#)

lsie_ln	<i>Linear Least Squares problem with inequality and equality constraints, least norm solution</i>
---------	---

Description

solve linear least square problem ($\min \|A\%x-b\|$) with inequality constraints $u\%x \geq co$ and equality constraints $e\%x = ce$ Method: reduce the pb to lsi_ln on the null-space of e

Usage

```
lsie_ln(
    a,
    b,
    u = NULL,
    co = NULL,
    e = NULL,
    ce = NULL,
    rcond = 1e+10,
    mnorm = NULL,
    x0 = NULL
)
```

Arguments

a	dense matrix A or its QR decomposition
b	right hand side vector
u	dense matrix of inequality constraints
co	right hand side vector of inequality constraints
e	dense matrix of equality constraints
ce	right hand side vector of equality constraints
rcond	maximal condition number for determining rank deficient matrix
mnorm	matrix to be used to minimize the norm $\ mnorm\%x\ $ in case of under-determined system. If mnorm is NULL (default), it is
x0	vector from which minimal norm should be searched when required

Value

solution vector whose attribute 'mes' may contain a message about possible numerical problems

See Also

[lsi_ln](#)

lsi_ln	<i>Linear Least Squares with Inequality constraints, least norm solution</i>
--------	--

Description

solve linear least square problem $\min_x ||A*x-b||$ with inequality constraints $u*x \geq co$ If A is rank deficient, least norm solution $||mnorm*(x-x0)||$ is used. If the parameter mnorm is NULL, it is treated as an identity matrix. If the vector x0 is NULL, it is treated as 0 vector.

Usage

```
lsi_ln(a, b, u = NULL, co = NULL, rcond = 1e+10, mnorm = NULL, x0 = NULL)
```

Arguments

a	dense matrix A or its QR decomposition
b	right hand side vector
u	dense matrix of inequality constraints
co	right hand side vector of inequality constraints
rcond	maximal condition number for determining rank deficient matrix
mnorm	norm matrix (can be dense or sparse) for which $%%$ operation with a dense vector is defined
x0	optional vector from which a least norm distance is searched for

Value

solution vector whose attribute 'mes' may contain a message about possible numerical problems

See Also

[lsi](#), [ldp](#), [base::qr](#)

lsi_reg	<i>Regularized Linear Least Squares</i>
---------	---

Description

solve linear least square problem $(\min_x ||a*x-b||)$ with inequality constraints $ux \geq co$ If a is rank deficient, regularization term $\lambda^2 * ||mnorm*(x-x0)||^2$ is added to $||a*x-b||^2$.

Usage

```
lsi_reg(a, b, u = NULL, co = NULL, rcond = 1e+10, mnorm = NULL, x0 = NULL)
```

Arguments

a	dense matrix A or its QR decomposition
b	right hand side vector
u	dense matrix of inequality constraints
co	right hand side vector of inequality constraints
rcond	used for calculating $\lambda = d[1]/\sqrt{rcond}$ where $d[1]$ is maximal singular value of a
mnorm	norm matrix (can be dense or sparse) for which <code>%*%</code> operation with a dense vector is defined
x0	optional vector from which a least norm distance is searched for

Details

The rank of a is estimated as number of singular values above $d[1]*1.e-10$ where $d[1]$ is the highest singular value. The scalar λ is a positive number and is calculated as $d[1]/\sqrt{rcond}$ ('rcond' parameter is preserved for compatibility with others `lsi_...` functions). At return, λ can be found in attributes of the returned vector x. NB. λ is set to NA

- if $\text{rank}(a) == 0$ or a is of full rank
- or if there is no inequality. If the matrix `mnorm` is NULL, it is supposed to be an identity matrix. If the vector `x0` is NULL, it is treated as 0 vector.

Value

solution vector whose attribute 'mes' may contain a message about possible numerical problems and 'lambda' is regularization parameter used in solution.

See Also

[lsi_ln](#)

ls_0

Particular solution of rank-deficient least squares

Description

Particular solution of rank-deficient least squares

Usage

```
ls_0(aq, b)
```

Arguments

aq	QR decomposition of a matrix
b	right-hand-side of the LS system

Value

solution vector where free variables are set to 0.

ls_ln	<i>Linear Least Squares, least norm solution</i>
-------	--

Description

Linear Least Squares, least norm solution

Usage

```
ls_ln(a, b, rcond = 1e+10, mnorm = NULL, x0 = NULL)
```

Arguments

a	matrix or its QR decomposition
b	vector (or matrix if a is full rank) of right hand side
rcond	maximal condition number for rank definition
mnorm	norm matrix (can be dense or sparse) for which %*% operation with a dense vector is defined
x0	optional vector from which a least norm distance is searched for

Value

solution vector

ls_ln_svd	<i>Linear Least Squares, least norm solution (by svd)</i>
-----------	---

Description

Least squares $a\%*\%x \approx b$ of least norm $\|x\|$ by using `svd(a)`

Usage

```
ls_ln_svd(a, b, rcond = 1e+10)
```

Arguments

a	dense matrix
b	right hand side vector
rcond	maximal condition number for determining rank deficient matrix

Value

solution vector

nlsic

*Non Linear Least Squares with Inequality Constraints***Description**

Solve non linear least squares problem $\min_{\text{par}} ||r(\text{par}, \dots) \text{\$res}||$ with optional inequality constraints $u \% \% \text{par} \geq \text{co}$ and optional equality constraints $e \% \% \text{par} = \text{eco}$

Usage

```
nlsic(
  par,
  r,
  u = NULL,
  co = NULL,
  control = list(),
  e = NULL,
  eco = NULL,
  flsi = lsi,
  ...
)
```

Arguments

- | | |
|---------|---|
| par | initial values for parameter vector. It can be in non feasible domain, i.e. for which $u \% \% \text{par} \geq \text{co}$ is not always respected. |
| r | a function calculating residual vector by a call to <code>r(par, cjac, ...)</code> where <code>par</code> is a current parameter vector, and <code>cjac</code> is set to <code>TRUE</code> if Jacobian is required or <code>FALSE</code> if not. A call to <code>r()</code> must return a list having a field "res" containing the residual vector and optionally a field "jacobian" when <code>cjac</code> is set to <code>TRUE</code> . |
| u | constraint matrix in $u \% \% \text{par} \geq \text{co}$ |
| co | constraint right hand side vector |
| control | a list of control parameters ('=' indicates default values): <ul style="list-style-type: none"> • <code>errx=1.e-7</code> error on l2 norm of the iteration step $\sqrt{\text{pt} \cdot \text{p}}$. • <code>maxit=100</code> maximum of newton iterations • <code>btstart=1</code> staring value for backtracking coefficient • <code>btfrac=0.5 (0;1)</code> by this value we diminish the step till tight up to the quadratic model of norm reduction in backtrack (bt) iterations • <code>btdesc=0.1 (0;1)</code> how good we have to tight up to the quadratic model 0 - we are very relaxe, 1 - we are very tight (may need many bt iterations) • <code>btmaxit=15</code> maximum of backtrack iterations • <code>btkmin=1.e-7</code> a floor value for backtracking fractioning • <code>trace=0</code> no information is printed during iterations (1 - print is active) |

- `rcond=1.e10` maximal condition number in QR decomposition starting from which a matrix is considered as numerically rank deficient. The inverse of this number is also used as a measure of very small number.
 - `ci` = list of options relative to confidence interval estimation
 - `p=0.95` p-value of confidence interval. If you need a plain sd value, set p-value to 0.6826895
 - `report=T` report (or not and hence calculate or not) confidence interval information (in the field `hci`, as 'half confidence interval' width)
 - `history=TRUE` report or not the convergence history
 - `adaptbt=FALSE` use or not adaptive backtracking
 - `mnorm=NULL` a norm matrix for a case `sln==TRUE` (cf next parameter)
 - `sln=FALSE` use or not (default) least norm of the solution (instead of least norm of the increment)
 - `maxstep=NULL` a maximal norm for increment step (if not NULL), must be positive
 - `monotone=FALSE` should or not the cost decrease be monotone. If TRUE, then at first non decrease of the cost, the iterations are stopped with a warning message.
- `e` linear equality constraint matrix in `e**par = eco` (dense)
- `eco` right hand side vector in equality constraints
- `flsi` function solving linear least squares problem. For a custom function, see interfaces in `lsi` or `lsi_ln` help pages.
- `...` parameters passed through to `r()`

Details

Solving method consist in sequential LSI problems globalized by backtracking technique. If `e`, `eco` are not NULL, reduce jacobian to basis of `e`'s kernel before `lsi()` call.

NB. If the function `r()` returns a list having a field "jacobian" it is supposed to be equal to the jacobian $dr/dpar$. If not, numerical derivation `numDeriv::jacobian()` is automatically used for its calculation.

NB2. `nlsic()` does not call `stop()` on possible errors. Instead, 'error' field is set to 1 in the returned result. This is done to allow a user to examine the current state of data and possibly take another path then to merely stop the program. So, a user must allways check this field at return from `nlsic()`.

NB3. User should test that field 'mes' is not NULL even when error is 0. It may contain a warning message.

Value

a list with following components (some components can be absent depending on 'control' parameter)

- 'par' estimated values of par as vector
- 'paro' the same but in original structure (i.e. matrix if par is a matrix)
- 'lastp' the last LSI solution during non linear iterations
- 'hci' vector of half-width confidence intervals for par

- 'ci_p' p-value for which CI was calculated
- 'ci_fdeg' freedom degree used for CI calculation
- 'sd_res' standard deviation of residuals
- 'covpar' covariance matrix for par
- 'laststep' the last increment after possible back-tracking iterations
- 'normp' the norm of lastp
- 'res' the last residual vector
- 'prevres' residual vector from previous non linear iteration
- 'jacobian' the last used jacobian
- 'retres' last returned result of `r()` call
- 'it' non linear iteration number where solution was obtained
- 'btit' back-tracking iteration number done during the last non linear iteration
- 'history' list with convergence history information
- 'error' error code: 0 - normal end, 1 - some error occurred, see message in 'mes'
- 'mes' textual message explaining what problem was in case of error

See Also

[lsi](#), [lsi_ln](#), [uplo2uco](#)

Examples

```
# solve min_{a,b} ||exp(a*x+b)-meas||, a,b>=1
a_true=1; b_true=2; x=0:5
# simulation function
sim=function(par, x) exp(par[["a"]]*x+par[["b"]])
# residual function to be passed to nlsic()
resid=function(par, cjac, ...) {
  dots=list(...)
  s=sim(par, dots$x)
  result=list(res=s-dots$meas)
  if (cjac) {
    result$jacobian=cbind(a=s*dots$x, b=s)
  }
  result
}
# simulated noised measurements for true parameters
set.seed(7) # for reproducible results
meas=sim(c(a=a_true, b=b_true), x)+rnorm(x)
# starting values for par
par=c(a=0, b=0)
# prepare constraints
uco=uplo2uco(par, lower=c(a=1, b=1))
# main call: solve the problem
fit=nlsic(par, resid, uco$u, uco$co, control=list(trace=1), x=x, meas=meas)
if (fit$error == 1) {
  stop(fit$mes)
```

```
} else {
  print(fit$par) # a=1.001590, b=1.991194
  if (! is.null(fit$mes)) {
    warning(fit$mes)
  }
}
```

Nulla	<i>Null-space basis</i>
-------	-------------------------

Description

use Lapack for null space basis (derived from MASS::Null)

Usage

```
Nulla(M, rcond = 1e+10)
```

Arguments

- M matrix such that $t(M) \%*\% B = 0$ where B is a basis of t(M)'s kernel (aka Null-space)
- rcond maximal condition number for rank definition

Value

numeric matrix whose columns are basis vectors. Its attribute 'qr' contains QR decomposition of M.

See Also

[MASS::Null](#)

Examples

```
Nulla(1:3)
```

pnull

*Particular least-squares solution and Null-space basis***Description**

use Lapack to find a particular solution x_p of under-determined least squares system $Ax=b$ and build null space basis B of A (derived from `MASS::Null`). In such a way, a general solution is given by $x=x_p+Bz$ where z is an arbitrary vector of size $\text{ncol}(A)-\text{rank}(A)$.

Usage

```
pnull(A, b = NULL, qrat = NULL, rcond = 1e+10, keepqr = FALSE)
```

Arguments

<code>A</code>	matrix (or its QR decomposition) such that $A\%*\%B=0$ where B is a basis of $\text{Kern}(A)$ (aka Null-space).
<code>b</code>	is the right hand side of the least squares system $Ax=b$.
<code>qrat</code>	is QR decomposition of $t(A)$.
<code>rcond</code>	maximal condition number for rank definition
<code>keepqr</code>	if TRUE store <code>qr</code> and <code>qrat</code> as attribute of <code>B</code>

Value

alist with `xp` and `B`, particular solution and numeric matrix whose columns are basis vectors. If requested, attributes `'qr'` and `'qrat'` of `B` contain QR decomposition of A and $t(A)$ respectively.

See Also

[MASS::Null](#)

Examples

```
A=diag(nrow=3L)[1:2,,drop=FALSE]
b=A%*(1:3)
res=pnull(A, b)
stopifnot(all.equal(res$xp, c(1:2,0)))
stopifnot(all.equal(c(res$B), c(0,0,1)))
```

tls	<i>Total Least Squares $a\%*x \sim b$</i>
-----	--

Description

Total Least Squares $a\%*x \sim b$

Usage

tls(a, b)

Arguments

- | | |
|---|------------------------|
| a | matrix |
| b | right hand side vector |

Value

solution vector

uplo2uco	<i>Transform box-type inequalities into matrix and vector form</i>
----------	--

Description

Transform a set of inequalities `param["name"] >= lower["name"]` and `param["name"] <= upper["name"]` into a list with matrix `u` and vector `co` such that `u%*%param>=co`. In addition to box inequalities above, user can provide linear inequalities in a form like `"a+2*c+3*b >= 0"` where `'a'`, `'b'` and `'c'` must be names of param components. Numeric and symbolic coefficients and right hand sides are allowed in these expressions. However, symbols must be defined at the moment of calling `uplo2uco()` so that expressions containing such symbols could be `eval()`-ed to numerical values. All inequalities must be written with `'>='` sign (not with `'<='`, `'>'`, ...).

Usage

uplo2uco(param, upper = NULL, lower = NULL, linear = NULL)

Arguments

- | | |
|--------|--|
| param | a named vector whose names will be used for parsing inequalities |
| upper | a named numeric vector of upper limits |
| lower | a named numeric vector of lower limits |
| linear | a string vector of linear inequalities |

Value

a list with numeric matrix 'u' and vector 'co' such that $u \% \% \text{param-co} \geq 0$

See Also

[equa2vecmat](#) for parsing linear expressions

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