

# Package ‘statr’

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**Type** Package

**Title** Matt Galloway Personal R Package

**Version** 0.1.0

**Description** This is a personal R package. It contains a number of various R functions for organization and convenience purposes.

**URL** <https://github.com/MGallow/statr>

**BugReports** <https://github.com/MGallow/statr/issues>

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**RoxygenNote** 5.0.1

**Imports** Rcpp (>= 0.12.10),  
dplyr,  
MASS,  
alr4,  
formatR,  
devtools

**LinkingTo** Rcpp

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bsearch	<i>Bisection search</i>
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**Description**

Minimizes a univariate strictly pseudoconvex function over the interval  $[a, b]$ . This is augmented code from Adam Rothman's STAT 8054 course (2017).

**Usage**

```
bsearch(df, a, b, L = 1e-07, quiet = FALSE)
```

**Arguments**

a	left endpoint of the initial interval of uncertainty.
b	right endpoint of the initial interval of uncertainty.
L	the maximum length of the final interval of uncertainty.
quiet	should the function stay quiet?
dg	the derivative of the function to minimize, where $dg(u, \dots)$ is the function evaluated at $u$ .
...	additional argument specifications for dg

**Value**

returns the midpoint of the final interval of uncertainty.

**Examples**

```
bsearch(dg, -10, 10, quiet = T)
```

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data_gen	<i>Normal Linear Data Generator</i>
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**Description**

True beta values are generated from  $p$  independent draws from  $N(0, 1/p)$  distribution.  $X_{-1}$  are  $n$  independent draws from  $(p - 1)$  multivariate normal  $N(0, \text{Sigma})$  where Sigma has  $(j, k)$  entry  $\theta^{\text{abs}(j - k)}$ .

$Y$  is then generated using the  $X = (1, X_{-1})$  and true beta values with an iid error term that follows distribution  $N(0, \text{var})$ . We can specify the desired number of replications (reps).

**Usage**

```
data_gen(n, p, theta, var = 0.5, reps = 200)
```

**Arguments**

n	desired sample size
p	desired dimension
theta	parameter used to generate covariance matrix
var	variance of generated y values
reps	number of replications

**Value**

generated design matrix (X), response values (Y)(matrix if reps > 1), true beta values

**Examples**

```
data_gen(1000, 10, 0.5)
```

---

dsearch	<i>Dichotomous search</i>
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---

**Description**

Minimizes a univariate strictly quasiconvex function over the interval [a, b]. This is augmented code from Adam Rothman's STAT 8054 course (2017).

**Usage**

```
dsearch(g, a, b, L = 1e-07, eps = (L/2.1), quiet = FALSE)
```

**Arguments**

g	the function to minimize, where $g(u, \dots)$ is the function evaluated at $u$ .
a	left endpoint of the initial interval of uncertainty.
b	right endpoint of the initial interval of uncertainty.
L	the maximum length of the final interval of uncertainty.
eps	search parameter, must be less than $L/2$
quiet	should the function stay quiet?
...	additional argument specifications for g

**Value**

returns the midpoint of the final interval of uncertainty.

**Examples**

```
dsearch(g, -10, 10, quiet = T)
```

---

gradient\_IRLS\_logistic

*Gradient of Logistic Regression (IRLS)*


---

### Description

Computes the gradient of logistic regression (optional ridge regularization term). We use this to determine if the KKT conditions are satisfied. This function is to be used with the 'IRLS' function.

### Usage

```
gradient_IRLS_logistic(betas, X, y, lam = 0, vec)
```

### Arguments

betas	beta estimates (includes intercept)
X	matrix or data frame
y	response vector of 0,1
lam	tuning parameter for ridge regularization term
vec	vector to specify which coefficients will be penalized

### Value

returns the gradient

### Examples

```
gradient_IRLS_logistic(betas, X, y, lam = 0.1, penalty = 'ridge')
```

---

gradient\_linear

*Gradient of Linear Regression*


---

### Description

Computes the gradient of linear regression (optional ridge regularization term). This function is to be used with the 'Linearrr' function.

### Usage

```
gradient_linear(betas, X, y, lam = 0, weights = NULL, vec)
```

### Arguments

betas	beta estimates (includes intercept)
X	matrix or data frame
y	response vector of 0,1
lam	tuning parameter for ridge regularization term
weights	option vector of weights for weighted least squares
vec	vector to specify which coefficients will be penalized

**Value**

returns the gradient

**Examples**

```
gradient_linear(betas, X, y, lam = 0.1)
```

---

```
gradient_MM_logistic
```

*Gradient of Logistic Regression (MM)*

---

**Description**

Computes the gradient of logistic regression (optional ridge regularization term). We use this to determine if the KKT conditions are satisfied. This function is to be used with the 'MM' function.

**Usage**

```
gradient_MM_logistic(betas, X, y, lam = 0, alpha = 1.5, gamma = 1, vec)
```

**Arguments**

betas	beta estimates (includes intercept)
X	matrix or data frame
y	response vector of 0,1
lam	tuning parameter for ridge regularization term
alpha	optional tuning parameter for bridge regularization term. Defaults to 'alpha = 1.5'
gamma	indicator function. 'gamma = 1' for ridge, 'gamma = 0' for bridge. Defaults to 'gamma = 1'
vec	vector to specify which coefficients will be penalized

**Value**

returns the gradient

**Examples**

```
gradient_MM_logistic(betas, X, y, lam = 0.1, alpha = 1.5, penalty = 'bridge')
```

---

IRLS

*Iterative Re-Weighted Least Squares*


---

**Description**

Computes the logistic regression coefficient estimates using the iterative re-weighted least squares (IRLS) algorithm. This function is to be used with the 'logisticr' function.

**Usage**

```
IRLS(X, y, lam = 0, intercept = TRUE, tol = 10^(-5), maxit = 1e+05, vec)
```

**Arguments**

X	matrix or data frame
y	matrix or vector of response 0,1
lam	tuning parameter for regularization term
intercept	Defaults to TRUE
tol	tolerance - used to determine algorithm convergence
maxit	maximum iterations
vec	optional vector to specify which coefficients will be penalized
betas	beta estimates (includes intercept)

**Value**

returns beta estimates (includes intercept), total iterations, and gradients.

**Examples**

```
IRLS(X, y, n.list = c(rep(1, n)), lam = 0.1, alpha = 1.5)
```

---

linearr

*Linear*


---

**Description**

Computes the linear regression coefficient estimates (ridge-penalization and weights, optional)

**Usage**

```
linearr(X, y, lam = 0, weights = NULL, intercept = TRUE, kernel = FALSE)
```

**Arguments**

X	matrix or data frame
y	matrix or data frame of response values
lam	optional tuning parameter for ridge regularization term. Defaults to 'lam = 0'
weights	optional vector of weights for weighted least squares
intercept	add column of ones if not already present. Defaults to TRUE
kernel	use linear kernel to compute ridge regression coefficients. Defaults to TRUE when $p \gg n$

**Value**

returns the coefficient estimates

**Examples**

```

Weighted ridge regression
library(dplyr)
X = dplyr::select(iris, -c(Species, Sepal.Length))
y = dplyr::select(iris, Sepal.Length)
linearrr(X, y, lam = 0.1, weights = rep(1:150))

Kernelized ridge regression
linearrr(X, y, lam = 0.1, kernel = T)

```

---

logisticr

*Logistic Regression*


---

**Description**

Computes the coefficient estimates for logistic regression. ridge regularization and bridge regularization optional.

**Usage**

```

logisticr(X, y, lam = 0, alpha = 1.5, penalty = "none",
  intercept = TRUE, method = "IRLS", tol = 10^(-5), maxit = 10^(5),
  vec = NULL)

```

**Arguments**

X	matrix or data frame
y	matrix or vector of response values 0,1
lam	optional tuning parameter for ridge regularization term. Defaults to 'lam = 0'
alpha	optional tuning parameter for bridge regularization term. Defaults to 'alpha = 1.5'
penalty	choose from c('none', 'ridge', 'bridge'). Defaults to 'none'
intercept	Defaults to TRUE

method	optimization algorithm. Choose from 'IRLS' or 'MM'. Defaults to 'IRLS'
tol	tolerance - used to determine algorithm convergence. Defaults to $10^{-5}$
maxit	maximum iterations. Defaults to $10^5$
vec	optional vector to specify which coefficients will be penalized

### Value

returns beta estimates (includes intercept), total iterations, and gradients.

### Examples

```

Logistic Regression
library(dplyr)
X = dplyr::select(iris, -Species)
y = dplyr::select(iris, Species)
y$Species = ifelse(y$Species == 'setosa', 1, 0)
logisticr(X, y)

ridge Logistic Regression with IRLS
logistir(X, y, lam = 0.1, penalty = 'ridge')

ridge Logistic Regression with MM
logisticr(X, y, lam = 0.1, penalty = 'ridge', method = 'MM')

bridge Logistic Regression
(Defaults to MM -- IRLS will return error)
logisticr(X, y, lam = 0.1, alpha = 1.5, penalty = 'bridge')

```

---

logitr	<i>Logit</i>
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---

### Usage

```
logitr(u)
```

### Arguments

u	some number. Ex: X
---	--------------------

returns the logit of u

Computes the logit for u

$\text{logit}(X \%*\% \text{beta})$



---

MM	<i>Majorize-Minimization function</i>
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**Description**

This function utilizes the MM algorithm. It will be used to compute the logistic regression coefficient estimates. This function is to be used with the 'logisticr' function.

**Usage**

```
MM(X, y, lam = 0, alpha = 1.5, gamma = 1, intercept = TRUE,
    tol = 10-5, maxit = 1e+05, vec = NULL)
```

**Arguments**

X	matrix or data frame
y	matrix or vector of response 0,1
lam	optional tuning parameter for ridge regularization term. Defaults to 'lam = 0'
alpha	optional tuning parameter for bridge regularization term. Defaults to 'alpha = 1.5'
gamma	gamma indicator function. 'gamma = 1' for ridge, 'gamma = 0' for bridge. Defaults to 'gamma = 1'
intercept	defaults to TRUE
tol	tolerance - used to determine algorithm convergence
maxit	maximum iterations
vec	optional vector to specify which coefficients will be penalized

**Value**

returns beta estimates (includes intercept), total iterations, and gradients.

**Examples**

```
MM(X, y)
```

---

predict_linearr	<i>Predict Linear Regression</i>
-----------------	----------------------------------

---

**Description**

Generates prediction for linear regression. Note that one can either input a 'linearr' object or a matrix of beta coefficients.

**Usage**

```
predict_linearr(object, X, y = NULL)
```

**Arguments**

object	'linearr' object or matrix of betas
X	matrix or data frame of (new) observations
y	optional, matrix or vector of response values

**Value**

predictions and loss metrics

**Examples**

```
fitted = linearr(X, y, lam = 0.1)
predict_linearr(fitted, X)
```

---

predict_logisticr	<i>Predict Logistic Regression</i>
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---

**Description**

Generates prediction for logistic regression. Note that one can either input a 'logisticr' object or a matrix of beta coefficients.

**Usage**

```
predict_logisticr(object, X, y = NULL)
```

**Arguments**

object	'logisticr' object or matrix of betas
X	matrix or data frame of (new) observations
y	optional, matrix or vector of response values 0,1

**Value**

predictions and loss metrics

**Examples**

```
fitted = logisticr(X, y, lam = 0.1, penalty = 'ridge', method = 'MM')
predict_logisticr(fitted, X)
```

---

`tidy`*Tidy*

---

**Description**

tidys package R code and updates package documentation. Directly uses Yihui Xie's 'formatR' package.

**Usage**

```
tidy()
```

**Examples**

```
tidy()
```

---

`timeit`*Time-It*

---

**Description**

Simple function that prints the computation time of a function

**Usage**

```
timeit(f)
```

**Arguments**

`f` the function to time

**Value**

returns the elapsed time

**Examples**

```
timeit(lm(dist ~ speed, cars))
```

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