Package ‘glmnet’

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Type Package

Title Lasso and Elastic-Net Regularized Generalized Linear Models

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Author Jerome Friedman [aut, cre],
Trevor Hastie [aut, cre],
Noah Simon [aut, ctb],
Junyang Qian [ctb],
Rob Tibshirani [aut, cre]

Maintainer Trevor Hastie <hastie@stanford.edu>

Depends Matrix (>= 1.0-6), utils, foreach

Imports methods

Suggests survival, knitr, lars

Description Extremely efficient procedures for fitting the entire lasso or elastic-net regularization path for linear regression, logistic and multinomial regression models, Poisson regression and the Cox model. Two recent additions are the multiple-response Gaussian, and the grouped multinomial regression. The algorithm uses cyclical coordinate descent in a path-wise fashion, as described in the paper linked to via the URL below.

License GPL-2

VignetteBuilder knitr

URL http://www.jstatsoft.org/v33/i01/.

NeedsCompilation yes

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R topics documented:

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Elastic net model paths for some generalized linear models

Description

This package fits lasso and elastic-net model paths for regression, logistic and multinomial regression using coordinate descent. The algorithm is extremely fast, and exploits sparsity in the input x matrix where it exists. A variety of predictions can be made from the fitted models.

Details

Package: glmnet
Type: Package
Version: 1.0
Date: 2008-05-14
License: What license is it under?

Very simple to use. Accepts x,y data for regression models, and produces the regularization path over a grid of values for the tuning parameter lambda. Only 5 functions: glmnet
predict.glmnet
plot.glmnet
print.glmnet
coeef.glmnet

Author(s)

Jerome Friedman, Trevor Hastie and Rob Tibshirani
Maintainer: Trevor Hastie <hastie@stanford.edu>

References

http://www.jstatsoft.org/v33/i01/
Examples

```r
x = matrix(rnorm(100*20), 100, 20)
y = rnorm(100)
g2 = sample(1:2, 100, replace = TRUE)
g4 = sample(1:4, 100, replace = TRUE)
fit1 = glmnet(x, y)
predict(fit1, newx = x[1:5, ], s = c(0.01, 0.005))
predict(fit1, type = "coef")
plot(fit1, xvar = "lambda")
fit2 = glmnet(x, g2, family = "binomial")
predict(fit2, type = "response", newx = x[2:5, ])
predict(fit2, type = "nonzero")
fit3 = glmnet(x, g4, family = "multinomial")
predict(fit3, newx = x[1:3, ], type = "response", s = 0.01)
```

Description

Simple simulated data, used to demonstrate the features of glmnet.

Usage

```r
data(BinomialExample)
data(CVXResults)
data(CoxExample)
data(MultiGaussianExample)
data(MultinomialExample)
data(PoissonExample)
data(QuickStartExample)
data(SparseExample)
```

Format

Data objects used to demonstrate features in the glmnet vignette.
Details

These datasets are artificial, and are used to test out some of the features of glmnet.

Examples

data(QuickStartExample)
glmnet(x,y)

cv.glmnet

Cross-validation for glmnet

Description

Does k-fold cross-validation for glmnet, produces a plot, and returns a value for lambda

Usage

cv.glmnet(x, y, weights, offset, lambda, type.measure, nfolds, foldid, grouped, keep, parallel, ...)

Arguments

x x matrix as in glmnet.

y response y as in glmnet.

weights Observation weights; defaults to 1 per observation

offset Offset vector (matrix) as in glmnet

lambda Optional user-supplied lambda sequence; default is NULL, and glmnet chooses its own sequence

nfolds number of folds - default is 10. Although nfolds can be as large as the sample size (leave-one-out CV), it is not recommended for large datasets. Smallest value allowable is nfolds=3

foldid an optional vector of values between 1 and nfold identifying what fold each observation is in. If supplied, nfold can be missing.

type.measure loss to use for cross-validation. Currently five options, not all available for all models. The default is type.measure="deviance", which uses squared-error for gaussian models (a.k.a type.measure="mse" there), deviance for logistic and poisson regression, and partial-likelihood for the Cox model. type.measure="class" applies to binomial and multinomial logistic regression only, and gives misclassification error. type.measure="auc" is for two-class logistic regression only, and gives area under the ROC curve. type.measure="mse" or type.measure="mae" (mean absolute error) can be used by all models except the "cox"; they measure the deviation from the fitted mean to the response.
This is an experimental argument, with default TRUE, and can be ignored by most users. For all models except the "cox", this refers to computing nfold fits, and then using their mean and estimated standard error to describe the CV curve. If grouped=FALSE, an error matrix is built up at the observation level from the predictions from the nfold fits, and then summarized (does not apply to type.measure="auc"). For the "cox" family, grouped=TRUE obtains the CV partial likelihood for the Kth fold by subtraction; by subtracting the log partial likelihood evaluated on the full dataset from that evaluated on the on the (K-1)/K dataset. This makes more efficient use of risk sets. With grouped=FALSE the log partial likelihood is computed only on the Kth fold.

If keep=TRUE, a prevalidated array is returned containing fitted values for each observation and each value of lambda. This means these fits are computed with this observation and the rest of its fold omitted. The foldid vector is also returned. Default is keep=FALSE.

If TRUE, use parallel foreach to fit each fold. Must register parallel before hand, such as doMC or others. See the example below.

Other arguments that can be passed to glmnet.

Details

The function runs glmnet n+1 times; the first to get the lambda sequence, and then the remainder to compute the fit with each of the folds omitted. The error is accumulated, and the average error and standard deviation over the folds is computed. Note that cv.glmnet does NOT search for values for alpha. A specific value should be supplied, else alpha=1 is assumed by default. If users would like to cross-validate alpha as well, they should call cv.glmnet with a pre-computed vector foldid, and then use this same fold vector in separate calls to cv.glmnet with different values of alpha. Note also that the results of cv.glmnet are random, since the folds are selected at random. Users can reduce this randomness by running cv.glmnet many times, and averaging the error curves.

Value

an object of class "cv.glmnet" is returned, which is a list with the ingredients of the cross-validation fit.

lambda
the values of lambda used in the fits.

cvm
The mean cross-validated error - a vector of length length(lambda).

cvse
estimate of standard error of cvm.

cvup
upper curve = cvm+cvse.

cvlo
lower curve = cvm-cvse.

nzero
number of non-zero coefficients at each lambda.

name
a text string indicating type of measure (for plotting purposes).

glmnet.fit
a fitted glmnet object for the full data.

lambda.min
value of lambda that gives minimum cvm.

lambda.1se
largest value of lambda such that error is within 1 standard error of the minimum.
fit.preval if keep=TRUE, this is the array of prevalidated fits. Some entries can be NA, if that and subsequent values of lambda are not reached for that fold
foldid if keep=TRUE, the fold assignments used

Author(s)
Jerome Friedman, Trevor Hastie and Rob Tibshirani
Noah Simon helped develop the 'coxnet' function.
Jeffrey Wong and B. Narasimhan helped with the parallel option
Maintainer: Trevor Hastie <hastie@stanford.edu>

References
http://www.jstatsoft.org/v33/i01/
http://www.jstatsoft.org/v39/i05/

See Also
glmnet and plot, predict, and coef methods for "cv.glmnet" object.

Examples

```r
set.seed(1010)
n=1000;p=100
nc=trunc(p/10)
x=matrix(rnorm(n*p),n,p)
beta=rnorm(nc)
fx=x[,seq(nc)] %*% beta
eps=rnorm(n))*5
y=drop(fx+eps)
px=exp(fx)
px=px/(1+px)
ly=rbetinom(n=length(px),prob=px,size=1)
set.seed(1011)
cvob=cv.glmnet(x,y)
plot(cvob)
coef(cvob)
predict(cvob1,newx=x[1:5,], s="lambda.min")
title("Gaussian Family",line=.5)
set.seed(1011)
cvob1a= CV.glmnet(x,y,type.measure="mae")
plot(cvob1a)
title("Gaussian Family",line=2.5)
set.seed(1011)
par(mfrow=c(2,2),mar=c(4,4,4,4))
cvob2=CV.glmnet(x,ly,family="binomial")
```
cv.glmnet

plot(cvob2)
title("Binomial Family", line=2.5)
frame()
set.seed(1011)
cvob2=cv.glmnet(x, ly, family="binomial", type.measure="class")
plot(cvob2)
title("Binomial Family", line=2.5)
set.seed(1011)
cvob3a=cv.glmnet(x, ly, family="binomial", type.measure="auc")
plot(cvob3a)
title("Binomial Family", line=2.5)
set.seed(1011)
mu=exp(fx/10)
y=rpois(n, mu)
cvob4=cv.glmnet(x, y, family="poisson")
plot(cvob4)
title("Poisson Family", line=2.5)

## Not run:
# Multinomial
n=500; p=30
ncz=trunc(p/10)
x=matrix(rnorm(n*p), n, p)
beta3=matrix(rnorm(30), 10, 3)
beta3=rbind(beta3, matrix(0, p-10, 3))
f3=x%*% beta3
p3=exp(f3)
p3=p3/apply(p3, 1, sum)
g3=rmult(p3)
set.seed(10101)
cvfit=cv.glmnet(x, g3, family="multinomial")
plot(cvfit)
title("Multinomial Family", line=2.5)
# Cox
beta=runor(ncz)
fx=x[, 1:ncz]%*%beta3
hx=exp(fx)
ty=exp(n, hx)
tcens=rbinom(n, prob=.3, size=1)# censoring indicator
y=cbind(time=ty, status=1-tcens) # y=Surv(time, status) with library(survival)
foldid=sample(rep(seq(10), length=n))
fit1_cv=cv.glmnet(x, y, family="cox", foldid=foldid)
plot(fit1_cv)
title("Cox Family", line=2.5)
# Parallel
require(doMC)
registerDoMC(cores=4)
x = matrix(rnorm(1e5 * 100), 1e5, 100)
y = rnorm(1e5)
system.time(cv.glmnet(x, y))
system.time(cv.glmnet(x, y, parallel=TRUE))

## End(Not run)
deviance.glmnet  

Extract the deviance from a glmnet object

Description

Compute the deviance sequence from the glmnet object

Usage

```r
## S3 method for class 'glmnet'
deviance(object,...)
```

Arguments

- `object`: fitted glmnet object
- `...`: additional print arguments

Details

A glmnet object has components `dev.ratio` and `nulldev`. The former is the fraction of (null) deviance explained. The deviance calculations incorporate weights if present in the model. The deviance is defined to be $2*(\text{loglike}_{\text{sat}} - \text{loglike})$, where $\text{loglike}_{\text{sat}}$ is the log-likelihood for the saturated model (a model with a free parameter per observation). Null deviance is defined to be $2*(\text{loglike}_{\text{sat}} - \text{loglike}(\text{Null}))$; The NULL model refers to the intercept model, except for the Cox, where it is the 0 model. Hence dev.ratio = 1-deviance/nulldev, and this deviance method returns (1-dev.ratio)*nulldev.

Value

- $(1-\text{dev.ratio})*\text{nulldev}$

Author(s)

Jerome Friedman, Trevor Hastie and Rob Tibshirani
Maintainer: Trevor Hastie <hastie@stanford.edu>

References


See Also

glmnet, predict, print, and coef methods.
**glmnet**

**Examples**

```r
x = matrix(rnorm(100*20), 100, 20)
y = rnorm(100)
fit1 = glmnet(x, y)
deviance(fit1)
```

---

**glmnet**

*fit a GLM with lasso or elasticnet regularization*

**Description**

Fit a generalized linear model via penalized maximum likelihood. The regularization path is computed for the lasso or elasticnet penalty at a grid of values for the regularization parameter lambda. Can deal with all shapes of data, including very large sparse data matrices. Fits linear, logistic and multinomial, poisson, and Cox regression models.

**Usage**

```r
glmnet(x, y, family = c("gaussian", "binomial", "poisson", "multinomial", "cox", "mgaussian"), weights, offset = NULL, alpha = 1, nlambda = 100, lambda.min.ratio = ifelse(nobs < nvars, 0.01, 0.0001), lambda = NULL, standardize = TRUE, intercept = TRUE, thresh = 1e-07, dfmax = nvars + 1, pmax = min(dfmax * 2 + 20, nvars), exclude, penalty.factor = rep(1, nvars), lower.limits = -Inf, upper.limits = Inf, maxit = 100000, type.gaussian = ifelse(nvars < 500, "covariance", "naive"), type.logistic = c("Newton", "modified.Newton"), standardize.response = FALSE, type.multinomial = c("ungrouped", "grouped"))
```

**Arguments**

- **x**
  - input matrix, of dimension nobs x nvars; each row is an observation vector. Can be in sparse matrix format (inherit from class "sparseMatrix" as in package Matrix; not yet available for family = "cox")

- **y**
  - response variable. Quantitative for family = "gaussian", or family = "poisson" (non-negative counts). For family = "binomial" should be either a factor with two levels, or a two-column matrix of counts or proportions (the second column is treated as the target class; for a factor, the last level in alphabetical order is the target class). For family = "multinomial", can be a nc>=2 level factor, or a matrix with nc columns of counts or proportions. For either "binomial" or "multinomial", if y is presented as a vector, it will be coerced into a factor. For family = "cox", y should be a two-column matrix with columns named 'time' and 'status'. The latter is a binary variable, with '1' indicating death, and '0' indicating right censored. The function `Surv()` in package `survival` produces such a matrix. For family = "mgaussian", y is a matrix of quantitative responses.

- **family**
  - Response type (see above)
weights  observation weights. Can be total counts if responses are proportion matrices. Default is 1 for each observation.

offset  A vector of length nobs that is included in the linear predictor (a nobs x nc matrix for the "multinomial" family). Useful for the "poisson" family (e.g. log of exposure time), or for refining a model by starting at a current fit. Default is NULL. If supplied, then values must also be supplied to the predict function.

alpha  The elasticnet mixing parameter, with $0 \leq \alpha \leq 1$. The penalty is defined as

$\frac{1-\alpha}{2}||\beta||_2^2 + \alpha||\beta||_1$.

alpha=1 is the lasso penalty, and alpha=0 the ridge penalty.

nlambda  The number of lambda values - default is 100.

lambda.min.ratio  Smallest value for lambda, as a fraction of lambda.max, the (data derived) entry value (i.e. the smallest value for which all coefficients are zero). The default depends on the sample size nobs relative to the number of variables nvars. If nobs > nvars, the default is 0.0001, close to zero. If nobs < nvars, the default is 0.01. A very small value of lambda.min.ratio will lead to a saturated fit in the nobs < nvars case. This is undefined for "binomial" and "multinomial" models, and glmnet will exit gracefully when the percentage deviance explained is almost 1.

lambda  A user supplied lambda sequence. Typical usage is to have the program compute its own lambda sequence based on nlambda and lambda.min.ratio. Supplying a value of lambda overrides this. WARNING: use with care. Avoid supplying a single value for lambda (for predictions after CV use predict() instead). Supply instead a decreasing sequence of lambda values. glmnet relies on its warms starts for speed, and its often faster to fit a whole path than compute a single fit.

standardize  Logical flag for x variable standardization, prior to fitting the model sequence. The coefficients are always returned on the original scale. Default is standardize=TRUE. If variables are in the same units already, you might not wish to standardize. See details below for y standardization with family="gaussian".

intercept  Should intercept(s) be fitted (default=TRUE) or set to zero (FALSE)

thresh  Convergence threshold for coordinate descent. Each inner coordinate-descent loop continues until the maximum change in the objective after any coefficient update is less than thresh times the null deviance. Defaults value is 1E-7.

dfmax  Limit the maximum number of variables in the model. Useful for very large nvars, if a partial path is desired.

pmax  Limit the maximum number of variables ever to be nonzero

exclude  Indices of variables to be excluded from the model. Default is none. Equivalent to an infinite penalty factor (next item).

penalty.factor  Separate penalty factors can be applied to each coefficient. This is a number that multiplies lambda to allow differential shrinkage. Can be 0 for some variables, which implies no shrinkage, and that variable is always included in the model. Default is 1 for all variables (and implicitly infinity for variables listed in exclude). Note: the penalty factors are internally rescaled to sum to nvars, and the lambda sequence will reflect this change.
lower.limits  Vector of lower limits for each coefficient; default -Inf. Each of these must be non-positive. Can be presented as a single value (which will then be replicated), else a vector of length nvars
upper.limits  Vector of upper limits for each coefficient; default Inf. See lower.limits
maxit        Maximum number of passes over the data for all lambda values; default is 10^5.
type.gaussian  Two algorithm types are supported for (only) family="gaussian". The default when nvar<500 is type.gaussian="covariance", and saves all inner-products ever computed. This can be much faster than type.gaussian="naive", which loops through nobs every time an inner-product is computed. The latter can be far more efficient for nvar >> nobs situations, or when nvar > 500.
type.logistic  If "Newton" then the exact hessian is used (default), while "modified.Newton" uses an upper-bound on the hessian, and can be faster.
standardize.response  This is for the family="mgaussian" family, and allows the user to standardize the response variables
type.multinomial  If "grouped" then a grouped lasso penalty is used on the multinomial coefficients for a variable. This ensures they are all in our out together. The default is "ungrouped"

Details

The sequence of models implied by lambda is fit by coordinate descent. For family="gaussian" this is the lasso sequence if alpha=1, else it is the elasticnet sequence. For the other families, this is a lasso or elasticnet regularization path for fitting the generalized linear regression paths, by maximizing the appropriate penalized log-likelihood (partial likelihood for the "cox" model). Sometimes the sequence is truncated before nlambda values of lambda have been used, because of instabilities in the inverse link functions near a saturated fit. glmnet(...,family="binomial") fits a traditional logistic regression model for the log-odds. glmnet(...,family="multinomial") fits a symmetric multinomial model, where each class is represented by a linear model (on the log-scale). The penalties take care of redundancies. A two-class "multinomial" model will produce the same fit as the corresponding "binomial" model, except the pair of coefficient matrices will be equal in magnitude and opposite in sign, and half the "binomial" values. Note that the objective function for "gaussian" is

\[ 1/2RSS/nobs + \lambda \ast \text{penalty}, \]

and for the other models it is

\[-loglik/nobs + \lambda \ast \text{penalty}. \]

Note also that for "gaussian", glmnet standardizes y to have unit variance (using 1/n rather than 1/(n-1) formula) before computing its lambda sequence (and then unstandardizes the resulting coefficients); if you wish to reproduce/compare results with other software, best to supply a standardized y. The coefficients for any predictor variables with zero variance are set to zero for all values of lambda. The latest two features in glmnet are the family="mgaussian" family and the type.multinomial="grouped" option for multinomial fitting. The former allows a multi-response gaussian model to be fit, using a "group-lasso" penalty on the coefficients for each variable. Typing the responses together like this is called "multi-task" learning in some domains. The grouped
multinomial allows the same penalty for the family="multinomial" model, which is also multi-responded. For both of these the penalty on the coefficient vector for variable \( j \) is

\[
(1 - \alpha)/2\|\beta_j\|_2^2 + \alpha\|\beta_j\|_1.
\]

When \( \alpha = 1 \) this is a group-lasso penalty, and otherwise it mixes with quadratic just like elastic-net.

**Value**

An object with S3 class "glmnet", "*", where "*" is "elnet", "lognet", "multnet", "fishnet" (poisson), "coxnet" or "mrelnet" for the various types of models.

- **call** the call that produced this object
- **alpha** Intercept sequence of length \( \text{length}(\text{lambda}) \)
- **beta** For "elnet", "lognet", "fishnet" and "coxnet" models, \( \text{nvars} \times \text{length}(\text{lambda}) \) matrix of coefficients, stored in sparse column format ("CsparseMatrix"). For "multnet" and "mgaussian", a list of \( \text{nc} \) such matrices, one for each class.
- **lambda** The actual sequence of \( \lambda \) values used. When \( \alpha = 0 \), the largest \( \lambda \) reported does not quite give the zero coefficients reported (\( \lambda = \text{inf} \) would in principle). Instead, the largest \( \lambda \) for \( \alpha = 0.001 \) is used, and the sequence of \( \lambda \) values is derived from this.
- **dev.ratio** The fraction of (null) deviance explained (for "elnet", this is the R-square). The deviance calculations incorporate weights if present in the model. The deviance is defined to be \( 2*(\text{loglike_sat} - \text{loglike}) \), where \( \text{loglike_sat} \) is the log-likelihood for the saturated model (a model with a free parameter per observation). Hence dev.ratio=1-dev/nulldev.
- **nulldev** Null deviance (per observation). This is defined to be \( 2*(\text{loglike_sat} - \text{loglike(Null)}) \); the NULL model refers to the intercept model, except for the Cox, where it is the 0 model.
- **df** The number of nonzero coefficients for each value of \( \lambda \). For "multnet", this is the number of variables with a nonzero coefficient for any class.
- **dim** A matrix consisting of the number of nonzero coefficients per class
- **dfmat** For "multnet" and "mrelnet" only. A matrix consisting of the number of nonzero coefficients per class.
- **nobs** Number of observations
- **npasses** Total passes over the data summed over all \( \lambda \) values
- **offset** Logical variable indicating whether an offset was included in the model
- **jerr** Error flag, for warnings and errors (largely for internal debugging).

**Author(s)**

Jerome Friedman, Trevor Hastie, Noah Simon and Rob Tibshirani

Maintainer: Trevor Hastie <hastie@stanford.edu>
References

http://www.jstatsoft.org/v33/i01/

http://www.jstatsoft.org/v39/i05/


Glmnet Vignette https://web.stanford.edu/~hastie/glmmnet/glmnet_alpha.html

See Also

print, predict, coef and plot methods, and the cv.glmnet function.

Examples

# Gaussian
x=matrix(rnorm(100*20),100,20)
y=rnorm(100)
fit1=glmnet(x,y)
print(fit1)
coef(fit1,s=0.01) # extract coefficients at a single value of lambda
predict(fit1,newx=x[,1:10],s=c(0.01,0.005)) # make predictions

#multivariate gaussian
y=matrix(rnorm(100*3),100,3)
fit1m=glmnet(x,y,family="mgaussian")
plot(fit1m,type="coef",n=3)

#binomial
y=sample(c(1,2),100,replace=TRUE)
fit2=glmnet(x,y,family="binomial")

#multinomial
y=sample(c(1:4),100,replace=TRUE)
fit3=glmnet(x,y,family="multinomial")
fit3a=glmnet(x,y,family="multinomial",type="grouped")

#poisson
y=rpois(n,p)
fit=glmnet(x,y,family="poisson")
plot(fit)
glmnet.control  

Description

View and/or change the factory default parameters in glmnet

Usage

```
glmnet.control(fdev=1.0e-5, devmax=0.999, eps=1.0e-6, big=9.9e35, mnlam=5, pmin=1.0e-9, exmx=250.0, prec=1e-10, mxit=100, factory=FALSE)
```
Arguments

- **fdev**: minimum fractional change in deviance for stopping path; factory default = 1.0e-5
- **devmax**: maximum fraction of explained deviance for stopping path; factory default = 0.999
- **eps**: minimum value of lambda.min.ratio (see glmnet); factory default = 1.0e-6
- **big**: large floating point number; factory default = 9.9e35. Inf in definition of upper.limit is set to big
- **mnlam**: minimum number of path points (lambda values) allowed; factory default = 5
- **pmin**: minimum probability for any class. factory default = 1.0e-9. Note that this implies a pmax of 1-pmin.
- **exmx**: maximum allowed exponent. factory default = 250.0
- **prec**: convergence threshold for multi response bounds adjustment solution. factory default = 1.0e-10
- **mxit**: maximum iterations for multiresponse bounds adjustment solution. factory default = 100
- **factory**: If TRUE, reset all the parameters to the factory default; default is FALSE

Details

If called with no arguments, glmnet.control() returns a list with the current settings of these parameters. Any arguments included in the call sets those parameters to the new values, and then silently returns. The values set are persistent for the duration of the R session.

Value

A list with named elements as in the argument list

Author(s)

Jerome Friedman, Trevor Hastie
Maintainer: Trevor Hastie <hastie@stanford.edu>

See Also

glmnet

Examples

```
glmnet.control(fdev=0)#continue along path even though not much changes
glmnet.control() # view current settings
glmnet.control(factory=TRUE) # reset all the parameters to their default
```
plot.cv.glmnet

plot the cross-validation curve produced by cv.glmnet

Description

Plots the cross-validation curve, and upper and lower standard deviation curves, as a function of the lambda values used.

Usage

## S3 method for class 'cv.glmnet'
plot(x, sign.lambda, ...)

Arguments

- `x`: fitted "cv.glmnet" object
- `sign.lambda`: Either plot against log(lambda) (default) or its negative if `sign.lambda=-1`.
- `...`: Other graphical parameters to plot

Details

A plot is produced, and nothing is returned.

Author(s)

Jerome Friedman, Trevor Hastie and Rob Tibshirani
Maintainer: Trevor Hastie <hastie@stanford.edu>

References


See Also

glmnet and cv.glmnet.

Examples

```
set.seed(1010)
n=1000;p=100
nc=trunc(p/10)
x=matrix(rnorm(n*p),n,p)
beta=rnorm(nc)
fx= (x[,seq(nc)] %% beta)
eps=rnorm(n)*5
y=drop(fx+eps)
px=exp(fx)
```
plot.glmnet

plot coefficients from a "glmnet" object

Description

Produces a coefficient profile plot of the coefficient paths for a fitted "glmnet" object.

Usage

```r
## S3 method for class 'glmnet'
plot(x, xvar = c("norm", "lambda", "dev"), label = FALSE, ...)## S3 method for class 'multnet'
plot(x, xvar = c("norm", "lambda", "dev"), label = FALSE, type.coef = c("coef", "2norm"), ...)## S3 method for class 'mrelnet'
plot(x, xvar = c("norm", "lambda", "dev"), label = FALSE, type.coef = c("coef", "2norm"), ...)
```

Arguments

- `x` fitted "glmnet" model
- `xvar` What is on the X-axis. "norm" plots against the L1-norm of the coefficients, "lambda" against the log-lambda sequence, and "dev" against the percent deviance explained.
- `label` If TRUE, label the curves with variable sequence numbers.
- `type.coef` If `type.coef = "2norm"` then a single curve per variable, else if `type.coef = "coef"`, a coefficient plot per response
- `...` Other graphical parameters to plot

Details

A coefficient profile plot is produced. If `x` is a multinomial model, a coefficient plot is produced for each class.
predict.cv.glmnet

make predictions from a "cv.glmnet" object.

Description

This function makes predictions from a cross-validated glmnet model, using the stored "glmnet.fit" object, and the optimal value chosen for lambda.

Usage

## S3 method for class 'cv.glmnet'
predict(object, newx, s=c("lambda.1se","lambda.min"),...)
## S3 method for class 'cv.glmnet'
coefficients(object,s=c("lambda.1se","lambda.min"),...)

Arguments

object Fitted "cv.glmnet" object.
newx Matrix of new values for x at which predictions are to be made. Must be a matrix; can be sparse as in Matrix package. See documentation for predict(glmnet).
predict.glmmnet

s

Value(s) of the penalty parameter lambda at which predictions are required. Default is the value s="lambda.1se" stored on the CV object. Alternatively s="lambda.min" can be used. If s is numeric, it is taken as the value(s) of lambda to be used.

... Not used. Other arguments to predict.

Details

This function makes it easier to use the results of cross-validation to make a prediction.

Value

The object returned depends on the ... argument which is passed on to the predict method for glmnet objects.

Author(s)

Jerome Friedman, Trevor Hastie and Rob Tibshirani
Maintainer: Trevor Hastie <hastie@stanford.edu>

References

http://www.jstatsoft.org/v33/i01/

See Also

glmnet, and print, and coef methods, and cv.glmnet.

Examples

```r
x=matrix(rnorm(100*20),100,20)
y=rnorm(100)
cv.fit=cv.glmnet(x,y)
predict(cv.fit,newx=x[1:5,])
coef(cv.fit)
coef(cv.fit,s="lambda.min")
predict(cv.fit,newx=x[1:5,],s=c(0.001,0.002))
```

predict.glmnet make predictions from a "glmnet" object.

Description

Similar to other predict methods, this functions predicts fitted values, logits, coefficients and more from a fitted "glmnet" object.
Usage

## S3 method for class 'glmnet'
predict(object, newx = NULL, 
s = NULL, 
type = c("link", "response", "coefficients", "nonzero", "class"), exact = FALSE, newoffset, ...)

## S3 method for class 'glmnet'
coef(object, s = NULL, exact = FALSE, ...)

Arguments

- **object**: Fitted "glmnet" model object.
- **newx**: Matrix of new values for \( x \) at which predictions are to be made. Must be a matrix; can be sparse as in Matrix package. This argument is not used for \( \text{type} = c("coefficients", "nonzero") \)
- **s**: Value(s) of the penalty parameter \( \lambda \) at which predictions are required. Default is the entire sequence used to create the model.
- **type**: Type of prediction required. Type "link" gives the linear predictors for "binomial", "multinomial", "poisson" or "cox" models; for "gaussian" models it gives the fitted values. Type "response" gives the fitted probabilities for "binomial" or "multinomial", fitted mean for "poisson" and the fitted relative-risk for "cox"; for "gaussian" type "response" is equivalent to type "link". Type "coefficients" computes the coefficients at the requested values for \( s \). Note that for "binomial" models, results are returned only for the class corresponding to the second level of the factor response. Type "class" applies only to "binomial" or "multinomial" models, and produces the class label corresponding to the maximum probability. Type "nonzero" returns a list of the indices of the nonzero coefficients for each value of \( s \).
- **exact**: This argument is relevant only when predictions are made at values of \( s \) (\( \lambda \)) different from those used in the fitting of the original model. If exact=FALSE (default), then the predict function uses linear interpolation to make predictions for values of \( s \) (\( \lambda \)) that do not coincide with those used in the fitting algorithm. While this is often a good approximation, it can sometimes be a bit coarse. With exact=TRUE, these different values of \( s \) are merged (and sorted) with object$\lambda \text{lambda}$, and the model is refit before predictions are made. In this case, it is required to supply the original data \( x= \) and \( y= \) as additional named arguments to predict() or coef(). The workhorse predict.glmnet() needs to update the model, and so needs the data used to create it. The same is true of weights, offset, penalty, factor, lower.limits, upper.limits if these were used in the original call. Failure to do so will result in an error.
- **newoffset**: If an offset is used in the fit, then one must be supplied for making predictions (except for type="coefficients" or type="nonzero")
- **...**: This is the mechanism for passing arguments like \( x= \) when exact=TRUE; see exact argument.

Details

The shape of the objects returned are different for "multinomial" objects. This function actually calls NextMethod(), and the appropriate predict method is invoked for each of the three model
types. `coef(...)` is equivalent to `predict(type="coefficients",...)`

**Value**

The object returned depends on type.

**Author(s)**

Jerome Friedman, Trevor Hastie and Rob Tibshirani
Maintainer: Trevor Hastie <hastie@stanford.edu>

**References**

*Journal of Statistical Software, Vol. 33(1), 1-22 Feb 2010*
[http://www.jstatsoft.org/v33/i01/](http://www.jstatsoft.org/v33/i01/)

[http://www.jstatsoft.org/v39/i05/](http://www.jstatsoft.org/v39/i05/)

**See Also**

`glmnet`, and `print`, and `coef` methods, and `cv.glmnet`.

**Examples**

```r
x=matrix(rnorm(100*20),100,20)
y=rnorm(100)
g2=sample(1:2,100,replace=TRUE)
g4=sample(1:4,100,replace=TRUE)
fit1=glmnet(x,y)
predict(fit1,newx=x[1:5,],s=c(0.01,0.005))
predict(fit1,type="coef")
fit2=glmnet(x,g2,family="binomial")
predict(fit2,type="response",newx=x[2:5,])
predict(fit2,type="nonzero")
fit3=glmnet(x,g4,family="multinomial")
predict(fit3,newx=x[1:3,],type="response",s=0.01)
```

---

**print.glmnet**

**Description**

Print a summary of the glmnet path at each step along the path.
Usage

```r
## S3 method for class 'glmnet'
print(x, digits = max(3, getOption("digits") - 3), ...)
```

Arguments

- `x`: fitted glmnet object
- `digits`: significant digits in printout
- `...`: additional print arguments

Details

The call that produced the object `x` is printed, followed by a three-column matrix with columns `Df`, `%dev` and `Lambda`. The `Df` column is the number of nonzero coefficients (`Df` is a reasonable name only for lasso fits). `%dev` is the percent deviance explained (relative to the null deviance).

Value

The matrix above is silently returned

Author(s)

Jerome Friedman, Trevor Hastie and Rob Tibshirani
Maintainer: Trevor Hastie <hastie@stanford.edu>

References


See Also

glmnet, predict and coef methods.

Examples

```r
x=matrix(rnorm(100*20),100,20)
y=rnorm(100)
fit1=glmnet(x,y)
print(fit1)
```
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