# Package 'apple'

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apple

Approximate Path for Penalized Likelihood Estimator

## **Description**

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

## Usage

```
apple(X, y, family="binomial", penalty = "LASSO", gamma, cha.poi = 1,
eps = 1e-15, lam.list, lambda.min.ratio, max.iter = 100, max.num,
n.lambda = 100)
```

#### **Arguments**

Χ	input matrix, of dimension nobs x nvars; each row is an observation vector.
у	response variable, of dimension nobs $x$ 1. non-negative counts for family="poisson", binary for family="binomial".
family	response type.
penalty	LASSO and MCP are provided.
gamma	the MCP concavity parameter.
cha.poi	the value used to change from Newton Raphson correction to Coordinate Descent correction, which is the $\alpha$ in the following inequality, $k > \alpha \sqrt{n}$ , where $k$ is the size of current active set, when this inequality holds, the correction

re k is the size of current active set. when this inequality holds, the correction method changes from Newton Raphson to Coordinate Descent.

eps the precision used to test the convergence.

lam.list a user supplied  $\lambda$  sequence. typical usage is to have the program compute its own lambda sequence based on lambda.min.ratio and n.lambda. supplying

a value of  $\lambda$  overrides this.

lambda.min.ratio

optional input. smallest value for lambda, as a fraction of max.lam, the (data derived) entry value. the default depends on the sample size n relative to the number of variables p. if n > p, the default is 0.0001. otherwise, the default is

0.01.

max.iter maximum number of iteration in the computation.

max.num optional input. maximum number of nonzero coefficients.

the number of lambda values. n.lambda

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## Value

a0 intercept vector of length(lambda).

beta nvar x length(lambda) matrix of coefficients.

lambda the list of lambda derived the solution path.

ebic the list of EBIC values.

ebic.loc the location of the EBIC selected solution in the path.

family the family of the supplied dataset.

## Author(s)

Yi Yu and Yang Feng

#### References

Yi Yu and Yang Feng, APPLE: Approximate Path for Penalized Likelihood Estimator, manuscript.

#### See Also

```
plot.apple, cv.apple and predict.apple
```

# **Examples**

```
p=10
n=200
d=5
coefs=c(3,1.5,0,0,2)
intercept=0
beta=rep(0,p)
beta[1:d]=coefs
X=matrix(rnorm(p*n), nrow=n)
mu=1/(1+exp(-X %*% beta-intercept))
y=rbinom(n,1,mu)

fit.apple=apple(X, y, family= "binomial")
plot(fit.apple)
```

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cv.apple cross val	lidation for apple path
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# Description

Does K-fold cross validation for apple.

# Usage

```
cv.apple(X, y, family="binomial", penalty="LASSO", gamma, K = 10,
alpha=0, seed = 1, cha.poi = 1, eps = 1e-15, lambda.min.ratio,
max.iter = 100, max.num, n.lambda = 100)
```

# Arguments

X	input matrix, of dimension nobs x nvars; each row is an observation vector.
у	response variable, of dimension nobs x 1. non-negative counts for family="poisson", binary for family="binomial".
family	response type.
penalty	LASSO and MCP are provided.
gamma	the MCP concavity parameter.
K	number of folds used in cross validation. The default it 10.
alpha	weight used in the cross validation cost function, with $Q(\lambda)=\alpha Dev(\lambda)+(1-\alpha)s(\lambda)\log n/n$ .
seed	random seed used to sample the training sets and test sets
cha.poi	the value used to change from Newton Raphson correction to Coordinate Descent correction, which is the $\alpha$ in the following inequality, $k>\alpha\sqrt{n}$ , where $k$ is the size of current active set. when this inequality holds, the correction method changes from Newton Raphson to Coordinate Descent.
eps	the precision used to test the convergence.
lambda min rati	

lambda.min.ratio

optional input. smallest value for lambda, as a fraction of max.lam, the (data derived) entry value. the default depends on the sample size n relative to the number of variables p. if n > p, the default is 0.0001. otherwise, the default is 0.01

max.iter maximum number of iteration in the computation.

max.num optional input. maximum number of nonzero coefficients.

n.lambda the number of lambda values.

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## Value

CV	list of cross validation loss
lambda	list of lambda
a0	the list of intercept
beta	the list of coefficients
cv.loc	location of cv selected solution in the path
ebic.loc	the location of the EBIC selected solution in the path
cv.beta	cross validation selected beta
ebic.beta	ebic selected beta
cv.a0	cv selected intercept
ebic.a0	ebic selected intercept

## Author(s)

Yi Yu and Yang Feng

#### References

Yi Yu and Yang Feng, APPLE: Approximate Path for Penalized Likelihood Estimator, manuscript.

## See Also

```
plot.apple, apple and predict.apple
```

## **Examples**

```
p=10
n=200
d=5
coefs=c(3,1.5,0,0,2)
intercept=0
beta=rep(0,p)
beta[1:d]=coefs
X=matrix(rnorm(p*n), nrow=n)
mu=1/(1+exp(-X %*% beta-intercept))
y=rbinom(n,1,mu)

fit.cv=cv.apple(X, y, family="binomial", alpha=0.25, K=5)
plot(fit.cv)
```

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# Description

Plot the solution path generated from apple.

## Usage

```
## S3 method for class 'apple'
plot(x, col = "black", add = FALSE,
main = "apple", type = "l", lty = 1, ...)
```

## **Arguments**

X	the object used to plot the path. object derived by "apple" and "cv.apple" are both available here. for "cv.apple" object, ebic and cv selected solutions are also lined in the graph.
col	color used to plot the paths, for details please see the usage of col in plot.
add	whether or not to add this plot to an existing one.
main	title of the plot. default is main="apple"
type	what type of plot should be drawn, for details please see the usage of plot.
lty	the line type, for details please see the usage of par
	see matplot.

#### **Details**

if the object is a cv result, then both of ebic and cv selected result will be marked on the graph by solid and dotted vertical lines, respectively.

## Author(s)

Yi Yu and Yang Feng

#### References

Yi Yu and Yang Feng, APPLE: Approximate Path for Penalized Likelihood Estimator, manuscript.

#### See Also

```
apple, cv.apple and predict.apple
```

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predict.apple	Model prediction based on a fitted apple/cv.apple object.	
predict.appie	Model Drediction based on a titled abble/cv.abble object.	

# Description

Similar to other predict methods, this function returns predictions from a fitted "apple" or "cv.apple" object.

# Usage

```
## S3 method for class 'apple'
predict(object, X, which = 1:length(object$lambda),
type = c("link", "response", "class"),...)
```

## **Arguments**

object	fitted "apple" or "cv.apple" model object.
X	matrix of values at which predictions are to be made.
which	indices of the penalty parameter lambda at which predictions are required. by default, all indices are returned.
type	type of prediction: "link" returns the linear predictors; "response" gives the fitted values; "class" returns the binomial outcome with the highest probability.
	see matplot.

## Value

The object returned depends on type.

## Author(s)

Yi Yu and Yang Feng

## References

Yi Yu and Yang Feng, APPLE: Approximate Path for Penalized Likelihood Estimator, manuscript.

## See Also

```
apple, cv.apple and plot.apple
```

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## **Examples**

```
p=10
n=200
d=5
coefs=c(3,1.5,0,0,2)
intercept=0
beta=rep(0,p)
beta[1:d]=coefs
set.seed(2)
X=matrix(rnorm(p*n), nrow=n)
mu=1/(1+exp(-X %*% beta-intercept))
y=rbinom(n,1,mu)
fit.apple=apple(X, y, family="binomial")
set.seed(3)
testX=matrix(rnorm(p*n), nrow=n)
predict(fit.apple,testX,type="link")
predict(fit.apple,testX,type="response")
predict(fit.apple,testX,type="class")
fit=cv.apple(X, y, family="binomial", alpha=0)
predict(fit.apple,testX,type="link", which = fit$cv.loc)
predict(fit.apple,testX,type="response", which = fit$cv.loc)
predict(fit.apple,testX,type="class", which = fit$cv.loc)
```

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