# Package 'pql'

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Title A Partitioned Quasi-Likelihood for Distributed Statistical

Type Package

mierence
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<b>Description</b> In the big data setting, working data sets are often distributed on multiple machines. However, classical statistical methods are often developed to solve the problems of single estimation or inference. We employ a novel parallel quasi-likelihood method in generalized linear models, to make the variances between different sub-estimators relatively similar. Estimates are obtained from projection subsets of data and later combined by suitably-chosen unknown weights. The philosophy of the package is described in Guo G. (2020) <doi:10.1007 s00180-020-00974-4="">.</doi:10.1007>
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pqlBLogist

The weighted Gauss-Newton estimators of the PQL in Logistic-GLMs

#### **Description**

The average weighted estimator and the unknown weighted estimator of the PQL in Logistic-GLMs through damped Gauss-Newton updates.

## Usage

```
pqlBLogist(data,G,nk)
```

#### **Arguments**

data is a design matrix with uniform distribution and the response vector.

G is the number of subsets.

nk is the size of subsets.

#### Value

betaBW,betaBA,MSEW,MSEA

#### **Examples**

```
G <- 20;n=1000;p=5; nk=50
b=runif(p, 0, 1)
beta =matrix(b,nrow=p)
X=matrix(rnorm(n*p),nrow=n)
L=X%*%beta
prob=1/exp(-(0.48+(L))+1)
y=1/(1+exp(-X))
y=(prob>runif(n))
y= ifelse((prob>runif(n)), 1, 0)
data=cbind(y,X)
pqlBLogist(data,G,nk)
```

pqlBpoisson1

The weight Gauss-Newton estimators of the PQL in Poisson-GLMS

## Description

The average weighted estimator and the unknown weighted estimator of the PQL in Poisson-GLMS through damped Gauss-Newton

#### Usage

```
pqlBpoisson1(data,G,nk)
```

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### Arguments

data is a design matrix with uniform distribution and the response vector

G is the number of subsets.

nk is the size of subsets

#### Value

betaBA, betaBW, MSEA, MSEW

## **Examples**

```
G <- 20;n=1000;p=5; nk=50
X<- matrix(runif(1000* 5, 0, 0.5), ncol = 5)
beta =matrix(runif(p, 0, 1),nrow=p)
L=X%*%beta
y<- rpois(1000, exp(L))
data=cbind(y,X)
pqlBpoisson1(data,G,nk)</pre>
```

pqlBpoisson2

The weighted Gauss-Newton estimators of the PQL in Poisson-GLMS

#### **Description**

The average weighted estimator and the unknown weighted estimator of the PQL in Poisson-GLMS through damped Gauss-Newton

#### Usage

```
pqlBpoisson2(data,G,nk)
```

#### **Arguments**

data is a design matrix with uniform distribution and the response vector

G is the number of subsets.

nk is the size of subsets.

#### Value

```
betaBA, betaBW, MSEA, MSEW
```

pqlLogist pqlLogist

#### **Examples**

```
p<- 5;G<- 20;n<- 1000;nk=50
X<- matrix(runif(n * p, 0, 0.5), ncol = p)
beta =matrix(runif(p, 0, 1),nrow=p)
L=X%*%beta
y<- rpois(n, exp(L))
data=cbind(y,X)
pqlBpoisson2(data,G,nk)</pre>
```

pqlLogist

pqlLogist

#### **Description**

The average weighted estimator and the unknown weighted estimator of the PQL in Poisson-GLMS through damped Gauss-Newton

## Usage

```
pqlLogist(data,G,nk)
```

#### **Arguments**

data	data is a highly correlated data set
G	G is the number of nodes
nk	n1 is the length of each data subset

#### Value

betaW	estimation value of betaW
betaA	estimation value of betaA
MSEW	estimation of MSEW
MSEA	estimation of MSEA

#### **Examples**

```
p<- 5;G<- 20;n<- 1000;nk=200
X<- matrix(runif(n*p, 0, 0.5), ncol = p)
beta =matrix(runif(p, 0, 1),nrow=p)
L=X%*%beta
y<- rpois(n, exp(L))
data=cbind(y,X)
pqlLogist(data,G,nk)</pre>
```

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pqlPoisson	The weighted Gauss-Newton estimators of the PQL in Poisson-GLMs

## Description

The average weighted estimator and the unknown weighted estimator of the PQL in Poisson-GLMS through damped Gauss-Newton

#### Usage

```
pqlPoisson(data,G,nk)
```

## Arguments

data is a design matrix with uniform distribution and the response vector

G is the number of subsets

nk is the number of outer subsets.

#### Value

```
betaBA, betaBW, MSEA, MSEW
```

#### **Examples**

```
#library(parallel)
#library(numDeriv)
#library(Rmpi)
#install.packages("pracma");
#library(pracma)
p<- 5;G<- 20;n<- 1000;nk=200
X<- matrix(runif(n*p, 0, 0.5), ncol = p)
beta =matrix(runif(p, 0, 1),nrow=p)
L=X%*%beta
y<- rpois(n, exp(L))
data=cbind(y,X)
pqlPoisson(data,G,nk)</pre>
```

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