

# Package: sPCR (via r-universe)

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**Type** Package

**Title** Sparse Principal Component Regression

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**Description** The sparse principal component regression is computed. The regularization parameters are optimized by cross-validation.

**License** GPL (>= 2)

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<https://doi.org/10.1016/j.csda.2018.03.008>,  
<https://sites.google.com/site/shuichikawanoen/software>

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cv.spcr

*Cross-validation for spcr***Description**

This function performs cross-validation for spcr. `cv.spcr` enables us to determine two regularization parameters  $\lambda_\beta$  and  $\lambda_\gamma$  objectively.

**Usage**

```
cv.spcr(x, y, k, w=0.1, xi=0.01, nfolds=5, adaptive=FALSE,
        center=TRUE, scale=FALSE, lambda.B.length=10, lambda.gamma.length=10,
        lambda.B=NULL, lambda.gamma=NULL)
```

**Arguments**

<code>x</code>	A data matrix.
<code>y</code>	A response vector.
<code>k</code>	The number of principal components.
<code>w</code>	Weight parameter with $0 \leq w \leq 1$ . The default is 0.1.
<code>xi</code>	The elastic net mixing parameter with $0 \leq \alpha \leq 1$ . The default is 0.01.
<code>nfolds</code>	The number of folds. The default is 5.
<code>adaptive</code>	If "TRUE", the adaptive SPCR is used.
<code>center</code>	If "TRUE", the data matrix is centered.
<code>scale</code>	If "TRUE", the data matrix is scaled.
<code>lambda.B.length</code>	The number of candidates for the parameter $\lambda_\beta$ . The default is 10.
<code>lambda.gamma.length</code>	The number of candidates for the parameter $\lambda_\gamma$ . The default is 10.
<code>lambda.B</code>	Optional user-supplied candidates for the parameter $\lambda_\beta$ . The default is NULL.
<code>lambda.gamma</code>	Optional user-supplied candidates for the parameter $\lambda_\gamma$ . The default is NULL.

**Value**

<code>lambda.gamma.seq</code>	The values of <code>lambda.gamma</code> in the fit.
<code>lambda.B.seq</code>	The values of <code>lambda.B</code> in the fit.
<code>CV.mat</code>	Matrix of the mean values of cross-validation. The row shows a sequence of <code>lambda.gamma</code> . The column shows a sequence of <code>lambda.B</code> .
<code>lambda.gamma.cv</code>	The value of <code>lambda.gamma</code> selected by cross-validation.
<code>lambda.B.cv</code>	The value of <code>lambda.B</code> selected by cross-validation.
<code>cvm</code>	The minimum of the mean cross-validated error.

**Author(s)**

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**References**

Kawano, S., Fujisawa, H., Takada, T. and Shiroishi, T. (2015). *Sparse principal component regression with adaptive loading*. *Computational Statistics & Data Analysis*, 89, 192–203.

**See Also**

spcr

**Examples**

```
#data
n <- 50
np <- 5
set.seed(1)
nu0 <- c(-1, 1)
x <- matrix( rnorm(np*n), n, np )
e <- rnorm(n)
y <- nu0[1]*x[,1] + nu0[2]*x[,2] + e

#fit
cv.spcr.fit <- cv.spcr(x=x, y=y, k=2)
cv.spcr.fit

#fit (adaptive SPCR)
cv.adaspcr.fit <- cv.spcr(x=x, y=y, k=2, adaptive=TRUE)
cv.adaspcr.fit
```

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cv.spcrglm

*Cross-validation for spcr-glm*


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**Description**

This function performs cross-validation for SPCR-glm. `cv.spcrglm` enables us to determine two regularization parameters  $\lambda_\beta$  and  $\lambda_\gamma$  objectively.

**Usage**

```
cv.spcrglm(x, y, k, family=c("binomial","poisson","multinomial"),
w=0.1, xi=0.01, nfolds=5, adaptive=FALSE, q=1, center=TRUE,
scale=FALSE, lambda.B.length=10, lambda.gamma.length=10,
lambda.B=NULL, lambda.gamma=NULL)
```

**Arguments**

x	A data matrix.
y	A response vector.
k	The number of principal components.
family	Response type.
w	Weight parameter with $w \geq 0$ . The default is 0.1.
xi	The elastic net mixing parameter with $0 \leq \alpha \leq 1$ . The default is 0.01.
nfolds	The number of folds. The default is 5.
adaptive	If "TRUE", the adaptive SPCR-glm (aSPCR-glm) is used.
q	The tuning parameter that controls weights in aSPCR-glm. The default is 1.
center	If "TRUE", the data matrix is centered.
scale	If "TRUE", the data matrix is scaled.
lambda.B.length	The number of candidates for the parameter $\lambda_\beta$ . The default is 10.
lambda.gamma.length	The number of candidates for the parameter $\lambda_\gamma$ . The default is 10.
lambda.B	Optional user-supplied candidates for the parameter $\lambda_\beta$ . The default is NULL.
lambda.gamma	Optional user-supplied candidates for the parameter $\lambda_\gamma$ . The default is NULL.

**Value**

lambda.gamma.seq	The values of lambda.gamma in the fit.
lambda.B.seq	The values of lambda.B in the fit.
CV.mat	Matrix of the mean values of cross-validation. The row shows a sequence of lambda.gamma. The column shows a sequence of lambda.B.
lambda.gamma.cv	The value of lambda.gamma selected by cross-validation.
lambda.B.cv	The value of lambda.B selected by cross-validation.
cvm	The minimum of the mean cross-validated error.

**Author(s)**

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**References**

Kawano, S., Fujisawa, H., Takada, T. and Shiroishi, T. (2018). *Sparse principal component regression for generalized linear models*. *Computational Statistics & Data Analysis*, 124, 180–196.

**See Also**

spcrglm

**Examples**

```

# binomial
n <- 100
np <- 3
nu0 <- c(-1, 1)
set.seed(4)
x <- matrix( rnorm(np*n), n, np )
y <- rbinom(n,1,1-1/(1+exp( (nu0[1]*x[,1] + nu0[2]*x[,2] ) )))
cv.spcrglm.fit <- cv.spcrglm(x=x, y=y, k=1, family="binomial")
cv.spcrglm.fit

# Poisson
set.seed(5)
y <- rpois(n, 1)
cv.spcrglm.fit <- cv.spcrglm(x=x, y=y, k=1, family="poisson")
cv.spcrglm.fit

# multinomial
set.seed(4)
y <- sample(1:4, n, replace=TRUE)
cv.spcrglm.fit <- cv.spcrglm(x=x, y=y, k=1, family="multinomial")
cv.spcrglm.fit

```

sPCR

*Fit a sparse principal component regression (SPCR)***Description**

This function computes a principal component regression model via sparse regularization.

**Usage**

```
sPCR(x, y, k, lambda.B, lambda.gamma, w=0.1, xi=0.01,
      adaptive=FALSE, center=TRUE, scale=FALSE)
```

**Arguments**

x	A data matrix.
y	A response vector.
k	The number of principal components.
lambda.B	The regularization parameter for the parameter $B$ .
lambda.gamma	The regularization parameter for the coefficient vector $\gamma$ .
w	Weight parameter with $0 \leq w \leq 1$ . The default is 0.1.
xi	The elastic net mixing parameter with $0 \leq \alpha \leq 1$ . The default is 0.01.
adaptive	If "TRUE", the adaptive SPCR is used.
center	If "TRUE", the data matrix is centered.
scale	If "TRUE", the data matrix is scaled.

**Value**

loadings.B	the loading matrix B
gamma	the coefficient
gamma0	intercept
loadings.A	the loading matrix A

**Author(s)**

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**References**

Kawano, S., Fujisawa, H., Takada, T. and Shiroishi, T. (2015). *Sparse principal component regression with adaptive loading*. *Computational Statistics & Data Analysis*, 89, 192–203.

**See Also**

cv.spcr

**Examples**

```
#data
n <- 100
np <- 5
set.seed(4)
nu0 <- c(-1, 1)
x <- matrix( rnorm(np*n), n, np )
e <- rnorm(n)
y <- nu0[1]*x[,1] + nu0[2]*x[,2] + e

#fit
spcr.fit <- spcr(x=x, y=y, k=2, lambda.B=6, lambda.gamma=2)
spcr.fit

#fit (adaptive SPCR)
adaspcr.fit <- spcr(x=x, y=y, k=2, lambda.B=6, lambda.gamma=2, adaptive=TRUE)
adaspcr.fit
```

---

spcrglm

*Fit a sparse principal component regression for generalized linear models (SPCR-glm)*

---

**Description**

This function computes a principal component regression for generalized linear models via sparse regularization.

**Usage**

```
spcrglm(x, y, k, family=c("binomial", "poisson", "multinomial"), lambda.B,
        lambda.gamma, w=0.1, xi=0.01, adaptive=FALSE, q=1, center=TRUE, scale=FALSE)
```

**Arguments**

x	A data matrix.
y	A response data.
k	The number of principal components.
family	Response type.
lambda.B	The regularization parameter for the parameter $B$ .
lambda.gamma	The regularization parameter for the coefficient vector $\gamma$ .
w	Weight parameter with $w \geq 0$ . The default is 0.1.
xi	The elastic net mixing parameter with $0 \leq \alpha \leq 1$ . The default is 0.01.
adaptive	If "TRUE", the adaptive SPCR-glm (aSPCR-glm) is used.
q	The tuning parameter that controls weights in aSPCR-glm. The default is 1.
center	If "TRUE", the data matrix is centered.
scale	If "TRUE", the data matrix is scaled.

**Value**

loadings.B	the loading matrix $B$
gamma	the coefficient
gamma0	intercept
loadings.A	the loading matrix $A$

**Author(s)**

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**References**

Kawano, S., Fujisawa, H., Takada, T. and Shiroishi, T. (2018). *Sparse principal component regression for generalized linear models*. *Computational Statistics & Data Analysis*, 124, 180–196.

**See Also**

cv.spcrglm

**Examples**

```
# binomial
n <- 100
np <- 5
nu0 <- c(-1, 1)
set.seed(4)
x <- matrix( rnorm(np*n), n, np )
y <- rbinom(n,1,1-1/(1+exp( (nu0[1]*x[,1] + nu0[2]*x[,2] ) )))
spcrglm.fit <- spcrglm(x=x, y=y, k=2, family="binomial", lambda.B=2, lambda.gamma=1)
spcrglm.fit

# Poisson
set.seed(4)
y <- rpois(n, exp( (nu0[1]*x[,1] + nu0[2]*x[,2] ) ))
spcrglm.fit <- spcrglm(x=x, y=y, k=2, family="poisson", lambda.B=2, lambda.gamma=1)
spcrglm.fit

# multinomial
set.seed(4)
y <- sample(1:4, n, replace=TRUE)
spcrglm.fit <- spcrglm(x=x, y=y, k=2, family="multinomial", lambda.B=2, lambda.gamma=2)
spcrglm.fit
```



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