

Package ‘MVOPR’

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Type Package

Title Multi-View Orthogonal Projection Regression for Multi-Modality Integration

Version 2.0.0

Description

Implements the 'MVOPR' (Multi-View Orthogonal Projection Regression) method for robust variable selection and integration of multi-modality data.

License GPL-2 | GPL-3

Encoding UTF-8

Imports ncvreg, rrpak

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Contents

MVOPR2	2
MVOPR3	3
Index	6

Description

Fit Multi-View Orthogonal Projection Regression for two modalities with Lasso, MCP, SCAD. The function is capable for linear, logistic, and poisson regression.

Usage

```
MVOPR2(
  M1,
  M2,
  Y,
  RRR_Control = list(Sparsity = TRUE, nrank = 10, ic.type = "GIC"),
  family = "gaussian",
  penalty = "lasso"
)
```

Arguments

M1	A numeric matrix (n x p) for the first modality.
M2	A numeric matrix (n x q) for the second modality. Assumes 'M2' is correlated to 'M1' via a low-rank matrix.
Y	A numeric response vector of length 'n', connected to 'M1' and 'M2'.
RRR_Control	A list to control the fitting for reduced rank regression. Sparsity Logical. If 'TRUE', performs Sparse Orthogonal Factor Regression (SOFAR); otherwise, a reduced-rank regression model is fitted. nrank Integer. Maximum rank to be searched for the reduced-rank model. ic.type Character. Model selection criterion: "AIC", "BIC", or "GIC".
family	Either "gaussian", "binomial", or "poisson", depending on the response.
penalty	The penalty to be applied in the outcome model Y to M1 and M2. Either "MCP" (the default), "SCAD", or "lasso".

Value

A list containing:

`fitY` Results for Outcome regression ($Y \sim M1 + M2$). A fitted object from 'cv.ncvreg', which contains the penalized regression results for 'Y'.

`fitM2` Results for reduced-rank regression ($M2 \sim M1$). The fitted reduced-rank regression model from 'rrpack'.

`coefY` A vector of estimated regression coefficients for 'M1' and 'M2' on 'Y'.

`coefM2` A matrix of estimated regression coefficients for 'M1' on 'M2'.

rank An integer indicates the estimated rank of the reduced-rank regression.

P A projection matrix used to extract the orthogonal components of ‘M1’.

M1s Transformed version of ‘M1’ after projection.

M2s Transformed version of ‘M2’ after removing the effect of ‘M1’.

References

Dai, Z., Huang, Y. J., & Li, G. (2025). Multi-View Orthogonal Projection Regression with Application in Multi-omics Integration. arXiv preprint arXiv:2503.16807. Available at <<https://arxiv.org/abs/2503.16807>>

Examples

```
## Simulation.1
p = 100; q = 100; n = 200
rank = 3

beta = c(rep(c(rep(1,5),rep(0,95)),2))
M1 = matrix(rnorm(p*n),n,p)

U = matrix(rnorm(rank*p),p,rank)
V = matrix(rnorm(rank*q),rank,q)
B = U %%% V
E = matrix(rnorm(q*n),n,q)
M2 = M1 %%% B + E
Y = cbind(M1,M2) %%% matrix(beta,p+q,1)

Fit = MVOPR2(M1,M2,Y,RRR_Control = list(Sparsity = FALSE))

## Result for variable selection
print(data.frame(Trucoef = beta,estimate = Fit$CoefY[2:(p+q+1)]))

## Plot the pathway and cv error in outcome model
oldpar <- par(mfrow = c(1, 2))
on.exit(par(oldpar))
plot(Fit$fitY$fit)
plot(Fit$fitY)
```

Description

Fit Multi-View Orthogonal Projection Regression for three modalities with Lasso, MCP, SCAD. The function is capable for linear, logistic, and poisson regression.

Usage

```
MVOPR3(
  M1,
  M2,
  M3,
  Y,
  RRR_Control = list(Sparsity = TRUE, nrank = 10, ic.type = "GIC"),
  family = "gaussian",
  penalty = "lasso"
)
```

Arguments

M1	A numeric matrix (n x p1) for the first modality.
M2	A numeric matrix (n x p2) for the second modality. Assumes 'M2' is correlated to 'M1' via a low-rank matrix.
M3	A numeric matrix (n x p3) for the third modality. Assumes 'M3' is correlated to 'M1' and 'M2' via a low-rank matrix.
Y	A numeric response vector of length 'n', connected to 'M1', 'M2', and 'M3'.
RRR_Control	A list to control the fitting for reduced rank regression. Sparsity Logical. If 'TRUE', performs Sparse Orthogonal Factor Regression (SOFAR); otherwise, a reduced-rank regression model is fitted. nrank Integer. Maximum rank to be searched for the reduced-rank model. ic.type Character. Model selection criterion: "AIC", "BIC", or "GIC".
family	Either "gaussian", "binomial", or "poisson", depending on the response.
penalty	The penalty to be applied in the outcome model Y to M1 and M2. Either "MCP" (the default), "SCAD", or "lasso".

Value

A list containing:

fitY A fitted object from 'cv.ncvreg', containing the penalized regression results for 'Y'.

fitM2 The fitted reduced-rank regression ('sofar' or 'rrr' object) for 'M2' given 'M1'.

fitM3 The fitted reduced-rank regression ('sofar' or 'rrr' object) for 'M3' given 'M1' and 'M2'.

CoefY A vector of estimated regression coefficients for 'Y'.

coefM2 A matrix of estimated regression coefficients for 'M2' given 'M1'.

coefM3 A matrix of estimated regression coefficients for 'M3' given 'M1' and 'M2'.

rank1 An integer indicating the estimated rank of the reduced-rank regression for 'M2'.

rank2 An integer indicating the estimated rank of the reduced-rank regression for 'M3'.

P1 A projection matrix used to extract the orthogonal components of 'M1'.

P2 A projection matrix used to extract the orthogonal components of 'E2', which is the error term in the regression for 'M2' given 'M1'.

M1s A transformed version of ‘M1’ after projection.

M2s A transformed version of ‘M2’ after removing the effect of ‘M1’ and projecting to the orthogonal space.

M3s A transformed version of ‘M3’ after removing the effects of ‘M1’ and ‘M2’.

#' @references Dai, Z., Huang, Y. J., & Li, G. (2025). Multi-View Orthogonal Projection Regression with Application in Multi-omics Integration. arXiv preprint arXiv:2503.16807. Available at <<https://arxiv.org/abs/2503.16807>>

Examples

```
## Simulation: three modalities
p1 = 50; p2 = 50; p3 = 50; n = 200
rank = 2

beta = c(rep(c(rep(1,5),rep(0,45)),3))
M1 = matrix(rnorm(p1*n),n,p1)

U1 = matrix(rnorm(rank*p1),p1,rank)
V1 = matrix(runif(rank*p2,-0.1,0.1),rank,p2)
B1 = U1 %>% V1

U2 = matrix(rnorm(rank*p1),p1,rank)
V2 = matrix(runif(rank*p2,-0.1,0.1),rank,p3)
B2 = U2 %>% V2

U3 = matrix(rnorm(rank*p2),p2,rank)
V3 = matrix(runif(rank*p2,-0.1,0.1),rank,p3)
B3 = U3 %>% V3

E1 = matrix(rnorm(p2*n),n,p2)
E2 = matrix(rnorm(p3*n),n,p3)

M2 = M1 %>% B1 + E1
M3 = M1 %>% B2 + M2 %>% B3 + E2
Y = cbind(M1,M2,M3) %>% matrix(beta,p1+p2+p3,1)

## Fit MVOPR with Lasso
Fit1 = MVOPR3(M1,M2,M3,Y,RRR_Control = list(Sparsity = FALSE),penalty = 'lasso')

## Fit MVOPR with MCP
Fit2 = MVOPR3(M1,M2,M3,Y,RRR_Control = list(Sparsity = FALSE),penalty = 'MCP')

## Fit MVOPR with SCAD
Fit3 = MVOPR3(M1,M2,M3,Y,RRR_Control = list(Sparsity = FALSE),penalty = 'SCAD')

## Compare the variable selection between Lasso, MCP, SCAD
print(data.frame(Lasso = Fit1$CoefY[2:151],MCP = Fit2$CoefY[2:151],SCAD = Fit3$CoefY[2:151],beta))
```

Index

MVOPR2, [2](#)

MVOPR3, [3](#)