MATLAB Toolbox Envelope: Reference Manual

June 20, 2012

Contents **1**

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tools

1.1 bootstrapse

Perform bootstrap to estimate actual standard errors for models in the envelope family.

Contents

Syntax

```
bootse = bootstrapse(X, Y, u, B, modelType)
bootse = bootstrapse(X, Y, u, B, modelType, Opts)
```

Input

X: Predictors. The predictors can be univariate or multivariate, discrete or continuous.

For model type for method 'env', 'envpls', 'henv', 'ienv', ' senv', 'xenv' and 'xenvpls'. X is an n by p matrix, p is the number of predictors.

For model type 'penv', X is A list containing the value of X1 and X2.

- X.X1 (only for 'penv'): Predictors of main interest. An n by p1 matrix, n is the number of observations, and p1 is the number of main predictors. The predictors can be univariate or multivariate, discrete or continuous.
- X.X2 (only for 'penv'): Covariates, or predictors not of main interest. An n by p2 matrix, p2 is the number of covariates.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

u: Dimension of the envelope subspace. The legitimate range of u depends on the model specified.

B: Number of bootstrap samples. A positive integer.

modelType: A string characters indicting the model, choices can be 'env', 'envpls', 'henv', 'ienv', 'penv', 'senv', 'xenv' and 'xenvpls'.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

bootse: For 'env', 'envpls', 'henv', 'ienv', and 'senv', an r by p matrix containing the standard errors for elements in β computed by bootstrap. For 'penv', an r by p1 matrix containing the standard errors for β_1 computed by bootstrap. For 'xenv' and 'xenvpls', a p by r matrix containing the standard errors for elements in β computed by bootstrap.

Description

This function computes the bootstrap standard errors for the regression coefficients or for partial envelope model, the main regression coefficients in the specified model by bootstrapping the residuals.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
alpha = 0.01;
u = lrt_env(X, Y, alpha);
B = 100;
modelType = 'env';
bootse = bootstrapse(X, Y, u, B, modelType)
```

```
bootse =

    0.2896
    0.4352
    0.3189
    0.5735
    0.2543
    0.5840
```

```
modelType = 'envpls';
bootse = bootstrapse(X, Y, u, B, modelType)
```

```
bootse =

    9.3899
```

```
    7.7384
    8.3303
    9.0875
   13.4416
    5.1477
```

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
alpha = 0.01;
u = lrt_penv(X, Y, alpha);
B = 100;
modelType = 'penv';
bootse = bootstrapse(X, Y, u, B, modelType)
```

```
bootse =

    0.0027
    0.0012
    0.0020
    0.0009
```

1.2 bootstrapse_OLS

Compute bootstrap standard error for ordinary least squares.

Contents

Syntax

```
bootse = bootstrapse_OLS(X, Y, B)
```

Input

X: Predictors, an n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses, an n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

B: Number of bootstrap samples. A positive integer.

Opts: A list containing the optional input parameters. If not defined, the default setting is used.

- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

bootse: The standard error for elements in β computed by bootstrap. An r by p matrix.

Description

This function computes the bootstrap standard errors for the regression coefficients in ordinary least squares by bootstrapping the residuals.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
bootse = bootstrapse_OLS(X, Y, 200)
```

```
bootse =

   10.2168
```

```
 8.3940
 9.0503
 9.9677
14.5822
 5.5874
```

1.3 mfoldcv

Select the dimension for the envelope family using m-fold cross validation.

Contents

Syntax

```
u = mfoldcv(X, Y, m, modelType)
u = mfoldcv(X, Y, m, modelType, Opts)
```

Input

X: Predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

m: A positive integer that is used to indicate m-fold cross validation.

modelType: A string characters indicting the model, choices can be 'envpls' or 'xenvpls'.

Opts: A list containing the optional input parameters. If one or several (even all) fields are not defined, the default settings are used.

- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

u: The dimension of the envelope subspace selected by m-fold cross validation.

Description

This function implements m-fold cross validation to select the dimension of the envelope space, based on prediction performance. For each u, the data is partitioned into m parts, each part is in turn used for testing for the prediction performance while the rest m-1 parts are used for training. The dimension is select as the one that minimizes the average prediction errors. If Y is multivariate, the identity inner product is used for computing the prediction errors.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
m = 5;
```

```
modelType = 'envpls';
u = mfoldcv(X, Y, m, modelType)
```

```
u =

     0
```

1.4 modelselectaic

Select the dimension for the envelope family using Akaike information criteria.

Contents

Syntax

```
u = modelselectaic(X, Y, modelType)
u = modelselectaic(X, Y, modelType, Opts)
```

Input

X: Predictors. The predictors can be univariate or multivariate, discrete or continuous.

For model type for method 'env', 'henv', 'ienv', ' senv', and 'xenv'. X is an n by p matrix, p is the number of predictors.

For model type 'penv', X is A list containing the value of X1 and X2.

- X.X1 (only for 'penv'): Predictors of main interest. An n by p1 matrix, n is the number of observations, and p1 is the number of main predictors. The predictors can be univariate or multivariate, discrete or continuous.
- X.X2 (only for 'penv'): Covariates, or predictors not of main interest. An n by p2 matrix, p2 is the number of covariates.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

modelType: A string characters indicting the model, choices can be 'env', 'henv', 'ienv', 'penv', 'senv' and 'xenv'.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and r.

Description

This function implements the Akaike information criteria (AIC) to select the dimension of the envelope subspace for method 'env', 'henv', 'ienv', 'penv', ' senv', and 'xenv'.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
modelType = 'env';
u = modelselectaic(X, Y, modelType)
```

```
u =

     1
```

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
modelType = 'penv';
u = modelselectaic(X, Y, modelType)
```

```
u =

     3
```

1.5 modelselectbic

Select the dimension for the envelope family using Bayesian information criteria.

Contents

Syntax

```
u = modelselectbic(X, Y, modelType)
u = modelselectbic(X, Y, modelType, Opts)
```

Input

X: Predictors. The predictors can be univariate or multivariate, discrete or continuous.

For model type for method 'env', 'henv', 'ienv', ' senv', and 'xenv'. X is an n by p matrix, p is the number of predictors.

For model type 'penv', X is A list containing the value of X1 and X2.

- X.X1 (only for 'penv'): Predictors of main interest. An n by p1 matrix, n is the number of observations, and p1 is the number of main predictors. The predictors can be univariate or multivariate, discrete or continuous.
- X.X2 (only for 'penv'): Covariates, or predictors not of main interest. An n by p2 matrix, p2 is the number of covariates.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

modelType: A string characters indicting the model, choices can be 'env', 'henv', 'ienv', 'penv', 'senv' and 'xenv'.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and r.

Description

This function implements the Bayesian information criteria (BIC) to select the dimension of the envelope subspace for method 'env', 'henv', 'ienv', 'penv', ' senv', and 'xenv'.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
modelType = 'env';
u = modelselectbic(X, Y, modelType)
```

```
u =

     1
```

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
modelType = 'penv';
u = modelselectbic(X, Y, modelType)
```

```
u =

     1
```

1.6 modelselectlrt

Select the dimension for the envelope family using likelihood ratio testing procedure.

Contents

Syntax

```
u = modelselectlrt(X, Y, alpha, modelType)
u = modelselectlrt(X, Y, alpha, modelType, Opts)
```

Input

X: Predictors. The predictors can be univariate or multivariate, discrete or continuous.

For model type for method 'env', 'henv', 'ienv', and 'xenv'. X is an n by p matrix, p is the number of predictors.

For model type 'penv', X is A list containing the value of X1 and X2.

- X.X1 (only for 'penv'): Predictors of main interest. An n by p1 matrix, n is the number of observations, and p1 is the number of main predictors. The predictors can be univariate or multivariate, discrete or continuous.
- X.X2 (only for 'penv'): Covariates, or predictors not of main interest. An n by p2 matrix, p2 is the number of covariates.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

alpha: Significance level for testing. A real number between 0 and 1, often taken at 0.05 or 0.01.

modelType: A string characters indicting the model, choices can be 'env', 'henv', 'ienv', 'penv' and 'xenv'.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and r.

Description

This function implements the likelihood ratio testing procedure to select the dimension of the envelope subspace for method 'env', 'henv', 'ienv', 'penv', and 'xenv'. The likelihood ratio resting procedure does not support 'senv', because the scaled envelope models are not nested with the standard model.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
alpha = 0.01;
modelType = 'env';
u = modelselectlrt(X, Y, alpha, modelType)
```

```
u =

     1
```

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
alpha = 0.01;
modelType = 'penv';
u = modelselectlrt(X, Y, alpha, modelType)
```

```
u =

     1
```

1.7 prediction

Perform estimation or prediction for models in the envelope family.

Contents

Syntax

```
PredictOutput = prediction(ModelOutput, Xnew, infType, modelType)
```

Input

ModelOutput: A list containing the model outputs from fitting the models.

Xnew: The value of X with which to estimate or predict Y.

For 'env', 'henv', 'ienv', 'senv' and 'xenv', it is a p by 1 vector.

For 'penv', it is a list containing the value of X1 and X2.

```
* Xnew.X1 (only for 'penv'): A p1 by 1 vector containing the value of X1.
* Xnew.X2 (only for 'penv'): A p2 by 1 vector containing the value of X2.
```

infType: A string of characters indicting the inference type, the choices can be 'estimation' or 'prediction'.

modelType: A string characters indicting the model, choices can be 'env', 'henv', 'ienv', 'penv', 'senv' and 'xenv'.

Output

PredictOutput: A list containing the results of the inference.

- PredictOutput.value: The fitted value or the prediction value evaluated at Xnew. An r by 1 vector.
- PredictOutput.covMatrix: The covariance matrix of PredictOutput.value. An r by r matrix.
- PredictOutput.SE: The standard error of elements in PredictOutput.value. An r by 1 vector.

Description

This function evaluates the user-specified model, could be 'env', 'henv', 'ienv', 'penv', 'senv' or 'xenv', at new value Xnew. It can perform estimation: find the fitted value when X = Xnew, or prediction: predict Y when X = Xnew. The covariance matrix and the standard errors are also provided.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
modelType = 'env';
u =  modelselectbic(X, Y, modelType);
ModelOutput = env(X, Y, u);
Xnew = X(2, :)';
PredictOutput = predict_env(ModelOutput, Xnew, 'estimation')
[PredictOutput.value, Y(2, :)'] % Compare the fitted value with
the observed value
```

```
PredictOutput =

        value: [6x1 double]
    covMatrix: [6x6 double]
           SE: [6x1 double]


ans =

  474.7135  458.0000
  127.4740  112.0000
  251.2044  236.0000
  380.8280  368.0000
  380.9473  383.0000
   -6.3287  -15.0000
```

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
modelType = 'penv';
u =  modelselectbic(X, Y, modelType);
ModelOutput = penv(X, Y, u);
Xnew.X1 = X.X1(1, :)';
Xnew.X2 = X.X2(1, :)';
PredictOutput = predict_penv(ModelOutput, Xnew, 'estimation')
PredictOutput.SE
```

```
PredictOutput =

        value: [4x1 double]
    covMatrix: [4x4 double]
```

```
           SE: [4x1 double]


ans =

    1.4680
    0.4234
    0.7145
    0.3161
```

1.8 testcoefficient

`This function tests the null hypothesis L * beta * R = A versus the alternative hypothesis L * beta * R ~= A, where beta is estimated under the model in the envelope family.`

Contents

Syntax

```
TestOutput = testcoefficient(ModelOutput, modelType)
TestOutput = testcoefficient(ModelOutput, modelType, TestInput)
```

Input

ModelOutput: A list containing the model outputs from fitting the models.

modelType: A string characters indicting the model, choices can be 'env', 'henv', 'ienv', 'penv', 'senv' and 'xenv'.

TestInput: A list that specifies the null hypothesis, including L, R, and A. If not provided by the user, default values will be used.

- TestInput.L: The matrix multiplied to β on the left. According to different model, it has different size requirement. Default value will be set if the user does not specify.
- TestInput.R: The matrix multiplied to β on the right. According to different model, it has different size requirement. Default value will be set if the user does not specify.
- TestInput.A: The matrix on the right hand side of the equation. Default value will be set if the user does not specify.

Output

TestOutput: A list containing test statistics, degrees of freedom for the reference chi-squared distribution, the p-value, and the covariance matrix of vec($L\beta R$). At the same time, a table is printed out.

- TestOutput.chisqStatistic: The test statistics. A real number.
- TestOutput.df: The degrees of freedom of the reference chi-squared distribution. A positive integer.
- TestOutput.pValue: p-value of the test. A real number in [0, 1].
- TestOutput.covMatrix: The covariance matrix of vec($L\beta R$). A d1 * d2 by d1 * d2 matrix.

Description

This function tests for hypothesis $H_0 : L\beta R = A$, versus $H_\alpha : L\beta R \neq A$. The β is estimated by a model in the envelope model. If the user does not specify the values for L, R and A, then the test is equivalent to the standard F test on if $\beta = 0$ (for 'env', 'ienv', 'penv', 'senv' and 'xenv'), or if the group main effects are all zeros (for 'henv'). The test statistics used is vec $(L\beta R - A)\ \hat{\Sigma}^{-1}$ vec $(L\beta R - A)^T$, and the reference distribution is chi-squared distribution with degrees of freedom the same as the length of vec(A).

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
alpha = 0.01;
u = lrt_env(X, Y, alpha);
ModelOutput = env(X, Y, u);
modelType = 'env';
TestOutout = testcoefficient(ModelOutput, modelType);
```

```
 Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A         116.230         6        0.0000
-------------------------------------------------------------------------
```

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
alpha = 0.01;
u = lrt_penv(X, Y, alpha);
ModelOutput = penv(X, Y, u);
r = size(Y, 2);
p1 = size(X.X1, 2);
TestInput.L = rand(2, r);
TestInput.R = rand(p1, 1);
TestInput.A = zeros(2, 1);
TestOutout = testcoefficient_penv(ModelOutput, TestInput);
```

```
 Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A          12.598          2        0.0018
-------------------------------------------------------------------------
```

2

env

2.1 aic_env

Select the dimension of the envelope subspace using Akaike information criterion.

Contents

Syntax

```
u = aic_env(X, Y)
u = aic_env(X, Y, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and r.

Description

This function implements the Akaike information criteria (AIC) to select the dimension of the envelope subspace.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
u = aic_env(X, Y)
```

```
u =

     1
```

2.2 bic_env

Select the dimension of the envelope subspace using Bayesian information criterion.

Contents

Syntax

```
u = bic_env(X, Y)
u = bic_env(X, Y, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is the number of observations. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses. The responses must be continuous variables.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and r.

Description

This function implements the Bayesian information criteria (BIC) to select the dimension of the envelope subspace.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
u = bic_env(X, Y)
```

```
u =

     1
```

2.3 bstrp_env

Compute bootstrap standard error for the envelope model.

Contents

Syntax

```
bootse = bstrp_env(X, Y, u, B)
bootse = bstrp_env(X, Y, u, B, Opts)
```

Input

X: Predictors, an n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses, an n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

u: Dimension of the envelope subspace. A positive integer between 0 and r.

B: Number of bootstrap samples. A positive integer.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

bootse: The standard error for elements in β computed by bootstrap. An r by p matrix.

Description

This function computes the bootstrap standard errors for the regression coefficients in the envelope model by bootstrapping the residuals.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
alpha = 0.01;
```

```
u = lrt_env(X, Y, alpha)
```

```
u =

     1
```

```
B = 100;
bootse = bstrp_env(X, Y, u, B)
```

```
bootse =

    0.2893
    0.4260
    0.3523
    0.5628
    0.1675
    0.6192
```

2.4 dF4env

The first derivative of the objective function for computing the envelope subspace.

Contents

Syntax

```
df = dF4env(R, DataParameter)
```

Input

R: An r by u semi orthogonal matrix, 0<u<=r.

DataParameter: A structure that contains the statistics calculated from the data.

Output

df: An r by u matrix containing the value of the derivative function evaluated at R.

Description

The objective function is derived in Section 4.3 in Cook et al. (2010) by using maximum likelihood estimation. This function is the derivative of the objective function.

2.5 env

Fit the envelope model.

Contents

Syntax

```
ModelOutput = env(X, Y, u)
ModelOutput = env(X, Y, u, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be strictly greater than p.

u: Dimension of the envelope. An integer between 0 and r.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out Grassmann manifold optimization process, logical 0 or 1. Default value: 0.
- Opts.init: The initial value for the envelope subspace. An r by u matrix. Default value is the one generated by function get_Init.

Output

ModelOutput: A list that contains the maximum likelihood estimators and some statistics.

- ModelOutput.beta: The envelope estimator of the regression coefficients β. An r by p matrix.
- ModelOutput.Sigma: The envelope estimator of the error covariance matrix. An r by r matrix.
- ModelOutput.Gamma: The orthogonal basis of the envelope subspace. An r by u semi-orthogonal matrix.
- ModelOutput.Gamma0: The orthogonal basis of the complement of the envelope subspace. An r by r-u semi-orthogonal matrix.
- ModelOutput.eta: The coordinates of β with respect to Gamma. A u by p matrix.

- ModelOutput.Omega: The coordinates of Sigma with respect to Gamma. A u by u matrix.
- ModelOutput.Omega0: The coordinates of Sigma with respect to Gamma0. An r-u by r-u matrix.
- ModelOutput.alpha: The estimated intercept in the envelope model. An r by 1 vector.
- ModelOutput.l: The maximized log likelihood function. A real number.
- ModelOutput.covMatrix: The asymptotic covariance of vec(β). An rp by rp matrix. The covariance matrix returned are asymptotic. For the actual standard errors, multiply by 1/n.
- ModelOutput.asyEnv: The asymptotic standard error for elements in β under the envelope model. An r by p matrix. The standard errors returned are asymptotic, for actual standard errors, multiply by 1/sqrt(n).
- ModelOutput.ratio: The asymptotic standard error ratio of the standard multivariate linear regression estimator over the envelope estimator, for each element in β. An r by p matrix.
- ModelOutput.np: The number of parameters in the envelope model. A positive integer.
- ModelOutput.n: The number of observations in the data. A positive integer.

Description

This function fits the envelope model to the responses and predictors, using the maximum likelihood estimation. When the dimension of the envelope is between 1 and r-1, we implemented the algorithm in Cook et al. (2010). When the dimension is r, then the envelope model degenerates to the standard multivariate linear regression. When the dimension is 0, it means that X and Y are uncorrelated, and the fitting is different.

References

1. The codes are implemented based on the algorithm in Section 4.3 of Cook et al (2010).
2. The Grassmann manifold optimization step calls the package sg_min 2.4.3 by Ross Lippert (http://web.mit.edu/~ripper/www/sgmin.html).

Example

The following codes will reconstruct the results in the wheat protein data example in Cook et al. (2010).

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
alpha = 0.01;
u = lrt_env(X, Y, alpha)
```

```
u =

     1
```

```
ModelOutput = env(X, Y, u)
```

```
ModelOutput =

         beta: [6x1 double]
        Sigma: [6x6 double]
        Gamma: [6x1 double]
       Gamma0: [6x5 double]
          eta: 8.5647
        Omega: 7.8762
       Omega0: [5x5 double]
        alpha: [6x1 double]
            l: -850.7592
    covMatrix: [6x6 double]
       asyEnv: [6x1 double]
        ratio: [6x1 double]
           np: 28
            n: 50
```

```
ModelOutput.Omega
```

```
ans =

    7.8762
```

```
eig(ModelOutput.Omega0)
```

```
ans =

   1.0e+03 *

    6.5166
    0.2083
    0.0201
    0.0004
    0.0003
```

```
ModelOutput.ratio
```

```
ans =

   28.0945
```

```
18.4326
23.6384
16.3211
65.8245
 6.4668
```

2.6 F4env

Objective function for computing the envelope subspace.

Contents

Syntax

```
f = F4env(R, DataParameter)
```

Input

R: An r by u semi orthogonal matrix, 0 < u <= r.

DataParameter: A structure that contains the statistics calculated from the data.

Output

f: A scalar containing the value of the objective function evaluated at R.

Description

The objective function is derived in Section 4.3 of Cook et al. (2010) using maximum likelihood estimation. The columns of the semi-orthogonal matrix that minimizes this function span the estimated envelope subspace.

2.7 lrt_env

Select the dimension of the envelope subspace using likelihood ratio testing.

Contents

Syntax

```
u = lrt_env(X, Y, alpha)
u = lrt_env(X, Y, alpha, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

alpha: Significance level for testing. A real number between 0 and 1, often taken at 0.05 or 0.01.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and r.

Description

This function implements the likelihood ratio testing procedure to select the dimension of the envelope subspace, with pre-specified significance level α.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
alpha = 0.01;
u = lrt_env(X, Y, alpha)
```

```
u =

     1
```

2.8 predict_env

Perform estimation or prediction under the envelope model.

Contents

Syntax

```
PredictOutput = predict_env(ModelOutput, Xnew, infType)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from env.

Xnew: The value of X with which to estimate or predict Y. A p by 1 vector.

infType: A string of characters indicting the inference type, the choices can be 'estimation' or 'prediction'.

Output

PredictOutput: A list containing the results of the inference.

- PredictOutput.value: The fitted value or the prediction value evaluated at Xnew. An r by 1 vector.
- PredictOutput.covMatrix: The covariance matrix of PredictOutput.value. An r by r matrix.
- PredictOutput.SE: The standard error of elements in PredictOutput.value. An r by 1 vector.

Description

This function evaluates the envelope model at new value Xnew. It can perform estimation: find the fitted value when X = Xnew, or prediction: predict Y when X = Xnew. The covariance matrix and the standard errors are also provided.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
alpha = 0.01;
u = lrt_env(X, Y, alpha);
ModelOutput = env(X, Y, u);
Xnew = X(2, :)';
```

```
PredictOutput = predict_env(ModelOutput, Xnew, 'estimation')
[PredictOutput.value, Y(1, :)']  % Compare the fitted value with the data
PredictOutput.SE
```

```
PredictOutput =

        value: [6x1 double]
    covMatrix: [6x6 double]
           SE: [6x1 double]


ans =

  474.7135  468.0000
  127.4740  123.0000
  251.2044  246.0000
  380.8280  374.0000
  380.9473  386.0000
   -6.3287  -11.0000


ans =

    4.8892
    4.0227
    4.3237
    4.7470
    6.8186
    2.6948
```

```
PredictOutput = predict_env(ModelOutput, Xnew, 'prediction')
PredictOutput.SE
```

```
PredictOutput =

        value: [6x1 double]
    covMatrix: [6x6 double]
           SE: [6x1 double]


ans =

  474.7135
  127.4740
  251.2044
  380.8280
  380.9473
   -6.3287
```

```
ans =

   34.9161
   28.7280
   30.8775
   33.9006
   48.6945
   19.2448
```

2.9 testcoefficient_env

`This function tests the null hypothesis L * beta * R = A versus the alternative hypothesis L * beta * R ~= A, where beta is estimated under the envelope model.`

Contents

Syntax

```
TestOutput = testcoefficient_env(ModelOutput)
TestOutput = testcoefficient_env(ModelOutput, TestInput)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from env.

TestInput: A list that specifies the null hypothesis, including L, R, and A. If not provided by the user, default values will be used.

- TestInput.L: The matrix multiplied to β on the left. It is a d1 by r matrix, while d1 is less than or equal to r. Default value: identity matrix I_r.
- TestInput.R: The matrix multiplied to β on the right. It is a p by d2 matrix, while d2 is less than or equal to p. Default value: identity matrix I_p.
- TestInput.A: The matrix on the right hand side of the equation. It is a d1 by d2 matrix. Default value: d1 by d2 zero matrix.

Output

TestOutput: A list containing test statistics, degrees of freedom for the reference chi-squared distribution, the p-value, and the covariance matrix of vec($L\beta R$). At the same time, a table is printed out.

- TestOutput.chisqStatistic: The test statistics. A real number.
- TestOutput.df: The degrees of freedom of the reference chi-squared distribution. A positive integer.
- TestOutput.pValue: p-value of the test. A real number in [0, 1].
- TestOutput.covMatrix: The covariance matrix of vec($L\beta R$). A d1 * d2 by d1 * d2 matrix.

Description

This function tests for hypothesis $H_0 : L\beta R = A$, versus $H_\alpha : L\beta R \neq A$. The β is estimated by the envelope model. If the user does not specify the values for L, R and A, then the test is equivalent to the standard F test on if $\beta = 0$. The test statistics used is vec $(L\beta R - A)\ \hat{\Sigma}^{-1}$ vec $(L\beta R - A)^T$, and the reference distribution is chi-squared distribution with degrees of freedom d1 * d2.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
alpha = 0.01;
u = lrt_env(X, Y, alpha);
ModelOutput = env(X, Y, u);
TestOutout = testcoefficient_env(ModelOutput);
```

```
 Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A         116.230        6      0.0000
-------------------------------------------------------------------------
```

```
r = size(Y, 2);
p = size(X, 2);
TestInput.L = rand(2, r);
TestInput.R = rand(p, 1);
TestInput.A = zeros(2, 1);
TestOutout = testcoefficient_env(ModelOutput, TestInput);
```

```
 Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A         111.628        2      0.0000
-------------------------------------------------------------------------
```

3

envmean

3.1 aic_envmean

Select the dimension of the envelope subspace using Akaike information criterion.

Contents

Syntax

```
u = aic_envmean(X)
u = aic_envmean(X, Opts)
```

Input

X: Data matrix. An n by p matrix, p is the dimension of the variable and n is number of observations.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and p.

Description

This function implements the Akaike information criteria (AIC) to select the dimension of the envelope subspace.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
u = aic_envmean(X)
```

```
u =

     6
```

3.2 bic_envmean

Select the dimension of the envelope subspace using Bayesian information criterion.

Contents

Syntax

```
u = bic_envmean(X)
u = bic_envmean(X, Opts)
```

Input

X: Data matrix. An n by p matrix, p is the dimension of the variable and n is number of observations.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and p.

Description

This function implements the Bayesian information criteria (BIC) to select the dimension of the envelope subspace.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
u = bic_envmean(X)
```

```
u =

     5
```

3.3 bstrp_envmean

Compute bootstrap standard error for the envelope estimator of the multivariate mean.

Contents

Syntax

```
bootse = bstrp_envmean(X, u, B)
bootse = bstrp_envmean(X, u, B, Opts)
```

Input

X: Data matrix. An n by p matrix, p is the dimension of the variable and n is number of observations.

u: Dimension of the envelope subspace. A positive integer between 0 and p.

B: Number of bootstrap samples. A positive integer.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

bootse: The standard error for elements in μ computed by bootstrap. A p dimensional column vector.

Description

This function computes the bootstrap standard errors for the envelope estimator of the multivariate mean by bootstrapping the residuals.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
alpha = 0.01;
u = lrt_envmean(X, alpha)
```

```
u =

     5


       B = 100;
       bootse = bstrp_envmean(X, u, B)

bootse =

    5.6513
    4.6606
    5.0257
    5.4714
    7.7713
    3.0035
```

3.4 dF4envmean

The first derivative of the objective function for computing the envelope subspace.

Contents

Syntax

```
df = dF4envmean(R, DataParameter)
```

Input

R: A p by u semi orthogonal matrix, 0 < u <= p.

DataParameter: A structure that contains the statistics calculated from the data.

Output

df: A p by u matrix containing the value of the derivative function evaluated at R.

Description

The objective function is derived by maximum likelihood estimation. This function is the derivative of the objective function.

3.5 envmean

Provide envelope estimator for the multivariate mean.

Contents

Syntax

```
ModelOutput = envmean(X, u)
ModelOutput = envmean(X, u, Opts)
```

Input

X: Data matrix. An n by p matrix, p is the dimension of the variable and n is number of observations.

u: Dimension of the envelope. An integer between 0 and p.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out Grassmann manifold optimization process, logical 0 or 1. Default value: 0.
- Opts.init: The initial value for the envelope subspace. A p by u matrix. Default value is the one generated by function get_Init4envmean.

Output

ModelOutput: A list that contains the maximum likelihood estimators and some statistics.

- ModelOutput.mu: The envelope estimator of the multivariate mean μ. A p dimensional column vector.
- ModelOutput.Sigma: The envelope estimator of the error covariance matrix. A p by p matrix.
- ModelOutput.Gamma: The orthogonal basis of the envelope subspace. A p by u semi-orthogonal matrix.
- ModelOutput.Gamma0: The orthogonal basis of the complement of the envelope subspace. An p by p-u semi-orthogonal matrix.
- ModelOutput.eta: The coordinates of μ with respect to Gamma. A u dimensional column vector.
- ModelOutput.Omega: The coordinates of Sigma with respect to Gamma. A u by u matrix.
- ModelOutput.Omega0: The coordinates of Sigma with respect to Gamma0. A p-u by p-u matrix.

- ModelOutput.l: The maximized log likelihood function. A real number.
- ModelOutput.covMatrix: The asymptotic covariance of μ. A p by p matrix. The covariance matrix returned are asymptotic. For the actual standard errors, multiply by 1/n.
- ModelOutput.asyEnv: The asymptotic standard error for elements in μ under the envelope model. A p dimensional column vector. The standard errors returned are asymptotic, for actual standard errors, multiply by 1/sqrt(n).
- ModelOutput.ratio: The asymptotic standard error ratio of the standard multivariate linear regression estimator over the envelope estimator, for each element in μ. A p dimensional column vector.
- ModelOutput.np: The number of parameters in the envelope model. A positive integer.
- ModelOutput.n: The number of observations in the data. A positive integer.

Description

This function provides an envelope estimator for the multivariate mean, with a given dimension of the envelope subspace u. The estimator is obtained using the maximum likelihood estimation. When the dimension is p, then the envelope model degenerates to the standard sample mean. When the dimension is 0, it means that X has mean 0.

References

The Grassmann manifold optimization step calls the package sg_min 2.4.3 by Ross Lippert (http://web.mit.edu/~ripper/www.sgmin.html).

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
u = bic_envmean(X)
```

```
u =

     5
```

```
ModelOutput = envmean(X, u)
```

```
ModelOutput =

           mu: [6x1 double]
        Sigma: [6x6 double]
        Gamma: [6x5 double]
       Gamma0: [6x1 double]
          eta: [5x1 double]
        Omega: [5x5 double]
       Omega0: 214.1646
            l: -881.6301
    covMatrix: [6x6 double]
       asyEnv: [6x1 double]
```

```
        ratio: [6x1 double]
           np: 26
            n: 50
```

ModelOutput.mu

```
ans =

  474.0794
  129.7194
  253.0441
  377.6816
  381.4752
   -7.2039
```

ModelOutput.Sigma

```
ans =

   1.0e+03 *

    1.2430    1.0297    1.1037    1.1862    1.5078    0.6631
    1.0297    0.8668    0.9252    0.9733    1.2439    0.5555
    1.1037    0.9252    0.9893    1.0459    1.3413    0.5926
    1.1862    0.9733    1.0459    1.1463    1.4748    0.6382
    1.5078    1.2439    1.3413    1.4748    2.1477    0.8132
    0.6631    0.5555    0.5926    0.6382    0.8132    0.3788
```

3.6 F4envmean

Objective function for computing the envelope subspace.

Contents

Syntax

```
f = F4envmean(R, DataParameter)
```

Input

R: A p by u semi orthogonal matrix, 0 < u <= p.

DataParameter: A structure that contains the statistics calculated from the data.

Output

f: A scalar containing the value of the objective function evaluated at R.

Description

The objective function is derived by maximum likelihood estimation. The columns of the semi-orthogonal matrix that minimizes this function span the estimated envelope subspace.

3.7 lrt_envmean

Select the dimension of the envelope subspace using likelihood ratio testing.

Contents

Syntax

```
u = lrt_envmean(X, alpha)
u = lrt_envmean(X, alpha, Opts)
```

Input

X: Data matrix. An n by p matrix, p is the dimension of the variable and n is number of observations.

alpha: Significance level for testing. A real number between 0 and 1, often taken at 0.05 or 0.01.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and p.

Description

This function implements the likelihood ratio testing procedure to select the dimension of the envelope subspace, with pre-specified significance level α.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
alpha = 0.01;
u = lrt_envmean(X, alpha)
```

```
u =

     5
```

3.8 predict_envmean

Perform estimation of the multivariate mean or prediction for a new observation.

Contents

Syntax

```
PredictOutput = predict_envmean(ModelOutput, infType)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from envmean.

infType: A string of characters indicting the inference type, the choices can be 'estimation' or 'prediction'.

Output

PredictOutput: A list containing the results of the inference.

- PredictOutput.value: The estimated multivariate mean or the prediction value. A p dimensional column vector.
- PredictOutput.covMatrix: The covariance matrix of PredictOutput.value. A p by p matrix.
- PredictOutput.SE: The standard error of elements in PredictOutput.value. A p dimensional column vector.

Description

If the inference type is prediction, this function predicts a new observation and gives its covariance matrix and standard errors of its elements. If the inference type is estimation, this function gives the estimation of the multivariate mean, its covariance matrix and standard errors of its elements.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
alpha = 0.01;
u = lrt_envmean(X, alpha);
ModelOutput = envmean(X, u);
PredictOutput = predict_envmean(ModelOutput, 'prediction')
```

```
PredictOutput =

        value: [6x1 double]
    covMatrix: [6x6 double]
           SE: [6x1 double]
```

```
PredictOutput.value
```

```
ans =

  474.0794
  129.7194
  253.0441
  377.6816
  381.4752
   -7.2039
```

```
PredictOutput.SE
```

```
ans =

   35.6077
   29.7337
   31.7655
   34.1944
   46.8043
   19.6574
```

3.9 testcoefficient_envmean

```
This function tests the null hypothesis L * mu = A versus the
alternative hypothesis L * mu ~= A, where mu is the envelope estimator
of the multivariate mean.
```

Contents

Syntax

```
TestOutput = testcoefficient_envmean(ModelOutput)
TestOutput = testcoefficient_envmean(ModelOutput, TestInput)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from envmean.

TestInput: A list that specifies the null hypothesis, including L and A. If not provided by the user, default values will be used.

- TestInput.L: The matrix multiplied to μ on the left. It is a d1 by p matrix, while d1 is less than or equal to p. Default value: identity matrix I_p.
- TestInput.A: The vector on the right hand side of the equation. It is a d1 dimensional column vector. Default value: d1 by d2 zero matrix.

Output

TestOutput: A list containing test statistics, degrees of freedom for the reference chi-squared distribution, the p-value, and the covariance matrix of Lμ. At the same time, a table is printed out.

- TestOutput.chisqStatistic: The test statistics. A real number.
- TestOutput.df: The degrees of freedom of the reference chi-squared distribution. A positive integer.
- TestOutput.pValue: p-value of the test. A real number in [0, 1].
- TestOutput.covMatrix: The covariance matrix of Lμ. A d1 dimensional column vector.

Description

This function tests for hypothesis $H_0 : L\mu = A$, versus $H_\alpha : L\mu \neq A$. The μ is estimated by the envelope model. If the user does not specify the values for L and A, then the test is equivalent to the standard F test on if $\mu = 0$. The test statistics used is $(L\mu - A)\ \hat{\Sigma}^{-1}\ (L\mu - A)^T$, and the reference distribution is chi-squared distribution with degrees of freedom d1.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
alpha = 0.01;
u = lrt_envmean(X, alpha);
ModelOutput = envmean(X, u);
TestOutout = testcoefficient_envmean(ModelOutput);
```

```
 Test Hypothesis     Chisq Statistic    DF     P-value
---------------------------------------------------------
L * mu = A          1351183.290          6      0.0000
---------------------------------------------------------
```

```
p = size(X, 2);
TestInput.L = rand(2, p);
TestInput.A = zeros(2, 1);
TestOutout = testcoefficient_envmean(ModelOutput, TestInput);
```

```
 Test Hypothesis     Chisq Statistic    DF     P-value
---------------------------------------------------------
L * mu = A            37273.324          2      0.0000
---------------------------------------------------------
```

4

envpls

4.1 bstrp_envpls

Compute bootstrap standard errors of the envelope model using a sequential algorithm.

Contents

Syntax

```
bootse = bstrp_envpls(X, Y, u, B)
bootse = bstrp_envpls(X, Y, u, B, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be strictly greater than p.

u: Dimension of the envelope subspace. A positive integer between 0 and p.

B: Number of bootstrap samples. A positive integer.

Opts: A list containing the optional input parameters. If one or several (even all) fields are not defined, the default settings are used.

- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

bootse: The standard error for elements in β computed by bootstrap. An r by p matrix.

Description

This function computes the bootstrap standard errors for the regression coefficients in the envelope model by bootstrapping the residuals. The envelope model is applied for the reduction on X, using a sequential algorithm.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
m = 5;
u = mfoldcv_envpls(X, Y, m)
```

```
u =

     0
```

```
B = 100;
bootse = bstrp_envpls(X, Y, u, B);
```

```
bootse =

     0
     0
     0
     0
     0
     0
```

4.2 envpls

Fit the envelope model using a sequential algorithm.

Contents

Syntax

```
ModelOutput = envpls(X, Y, u)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be strictly greater than p.

u: Dimension of the envelope. An integer between 0 and r.

Output

ModelOutput: A list that contains the maximum likelihood estimators and some statistics.

- ModelOutput.beta: The envelope estimator of the regression coefficients β. An r by p matrix.
- ModelOutput.Sigma: The envelope estimator of the error covariance matrix. An r by r matrix.
- ModelOutput.Gamma: The orthogonal basis of the envelope subspace. An r by u semi-orthogonal matrix.
- ModelOutput.Gamma0: The orthogonal basis of the complement of the envelope subspace. An r by r-u semi-orthogonal matrix.
- ModelOutput.eta: The coordinates of β with respect to Gamma. An u by p matrix.
- ModelOutput.Omega: The coordinates of Sigma with respect to Gamma. An u by u matrix.
- ModelOutput.Omega0: The coordinates of Sigma with respect to Gamma0. An r-u by r-u matrix.
- ModelOutput.alpha: The estimated intercept in the envelope model. An r by 1 vector.
- ModelOutput.np: The number of parameters in the envelope model. A positive integer.
- ModelOutput.n: The number of observations in the data. A positive integer.

Description

This function fits the envelope model to the responses and predictors, using the maximum likelihood estimation. When the dimension of the envelope is between 1 and r-1, we implemented the algorithm in Cook et al. (2010). When the dimension is r, then the envelope model degenerates to the standard multivariate linear regression. When the dimension is 0, it means that X and Y are uncorrelated, and the fitting is different.

References

The codes are implemented based on the sequential algorithm in the lecture notes of Cook (2012).

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
u = mfoldcv_envpls(X, Y, m)
```

```
u =

     0
```

```
ModelOutput = envpls(X, Y, u)
```

```
ModelOutput =

      beta: [6x1 double]
     Sigma: [6x6 double]
     Gamma: []
    Gamma0: [6x6 double]
       eta: []
     Omega: []
    Omega0: [6x6 double]
     alpha: [6x1 double]
        np: 27
         n: 50
```

```
ModelOutput.Sigma
```

```
ans =

   1.0e+03 *

    1.1932    0.9825    1.0568    1.1508    1.5411    0.6340
    0.9825    0.8222    0.8808    0.9391    1.2700    0.5279
```

```
1.0568    0.8808    0.9451    1.0125    1.3725    0.5652
1.1508    0.9391    1.0125    1.1239    1.5217    0.6180
1.5411    1.2700    1.3725    1.5217    2.3255    0.8365
0.6340    0.5279    0.5652    0.6180    0.8365    0.3619
```

4.3 mfoldcv_xenvpls

Select the dimension of the envelope subspace using m-fold cross validation for envelope model on the reduction on X using a sequential algorithm.

Contents

Syntax

```
u = mfoldcv_xenvpls(X, Y, m)
u = mfoldcv_xenvpls(X, Y, m, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is number of observations. The number of predictors should be greater than the number of the responses. And they must be continuous variables.

Y: Responses. An n by r matrix, r is the number of responses. The response can be univariate or multivariate and must be continuous variable.

m: A positive integer that is used to indicate m-fold cross validation.

Opts: A list containing the optional input parameters. If one or several (even all) fields are not defined, the default settings are used.

- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

`*u*: The dimension of the envelope subspace selected by m-fold cross validation.`

Description

This function implements m-fold cross validation to select the dimension of the envelope space, based on prediction performance. For each u, the data is partitioned into m parts, each part is in turn used for testing for the prediction performance while the rest m-1 parts are used for training. The dimension is select as the one that minimizes the average prediction errors. If Y is multivariate, the identity inner product is used for computing the prediction errors.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
```

```
m = 5;
u = mfoldcv_envpls(X, Y, m)
```

```
u =

     0
```

5

henv

5.1 aic_henv

Select the dimension of the envelope subspace using Akaike information criterion for the heteroscedastic envelope model.

Contents

Syntax

```
u = aic_henv(X, Y)
u = aic_henv(X, Y, Opts)
```

Input

X: Group indicators. A matrix with n rows. X can only have p unique rows, where p is the number of groups. For example, if there are two groups, X can only have 2 different kinds of rows, such as (0, 1) and (1, 0), or (1, 0, 10) and (0, 5, 6). The number of columns is not restricted, as long as X only has p unique rows.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be greater than p.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and r.

Description

This function implements the Akaike information criteria (AIC) to select the dimension of the envelope subspace for the heteroscedastic envelope model.

Example

```
load waterstrider.mat
u = aic_henv(X, Y)
```

```
u =

     6
```

5.2 bic_henv

Select the dimension of the envelope subspace using Bayesian information criterion for the heteroscedastic envelope model.

Contents

Syntax

```
u = bic_henv(X, Y)
u = bic_henv(X, Y, Opts)
```

Input

X: Group indicators. A matrix with n rows. X can only have p unique rows, where p is the number of groups. For example, if there are two groups, X can only have 2 different kinds of rows, such as (0, 1) and (1, 0), or (1, 0, 10) and (0, 5, 6). The number of columns is not restricted, as long as X only has p unique rows.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be greater than p.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and r.

Description

This function implements the Bayesian information criteria (BIC) to select the dimension of the envelope subspace for the heteroscedastic envelope model.

Example

```
load waterstrider.mat
u = bic_henv(X, Y)
```

```
u =

     4
```

5.3 bstrp_henv

Compute bootstrap standard error for the heteroscedastic envelope model.

Contents

Syntax

```
bootse = bstrp_henv(X, Y, u, B)
bootse = bstrp_henv(X, Y, u, B, Opts)
```

Input

X: Group indicators. A matrix with n rows. X can only have p unique rows, where p is the number of groups. For example, if there are two groups, X can only have 2 different kinds of rows, such as (0, 1) and (1, 0), or (1, 0, 10) and (0, 5, 6). The number of columns is not restricted, as long as X only has p unique rows.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be greater than p.

u: Dimension of the envelope subspace. A positive integer between 0 and r.

B: Number of bootstrap samples. A positive integer.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

bootse: The standard error for elements in β computed by bootstrap. An r by p matrix.

Description

This function computes the bootstrap standard errors for the regression coefficients in the heteroscedastic envelope model by bootstrapping the residuals.

Example

```
load waterstrider.mat

u = lrt_henv(X, Y, 0.01)
```

```
u =

     6
```

```
B = 100;
bootse = bstrp_henv(X, Y, u, B)
```

```
bootse =

    0.0305    0.0466    0.0647
    0.0309    0.0485    0.0682
    0.0305    0.0432    0.0638
    0.0205    0.0289    0.0425
    0.0385    0.0553    0.0799
    0.0295    0.0427    0.0618
    0.0389    0.0567    0.0819
    0.0321    0.0463    0.0665
```

5.4 dF4henv

The first derivative of the objective function for computing the envelope subspace in the heteroscedastic envelope model.

Contents

Syntax

```
df = dF4henv(R, DataParameter)
```

Input

R: An r by u semi orthogonal matrix, 0 < u <= r.

DataParameter: A structure that contains the statistics calculated from the data.

Output

df: An r by u matrix containing the value of the derivative function evaluated at R.

Description

The objective function is derived in Section 2.2 in Su and Cook (2012) by using maximum likelihood estimation. This function is the derivative of the objective function.

5.5 F4henv

Objective function for computing the envelope subspace in heteroscedastic envelope model.

Contents

Syntax

```
f = F4henv(R, DataParameter)
```

Input

R: An r by u semi orthogonal matrix, 0 < u <= r.

DataParameter: A structure that contains the statistics calculated from the data.

Output

f: A scalar containing the value of the objective function evaluated at R.

Description

The objective function is derived in Section 2.2 of Su and Cook (2012) using maximum likelihood estimation. The columns of the semi-orthogonal matrix that minimizes this function span the estimated envelope subspace in the heteroscedastic envelope model.

5.6 henv

Fit the heteroscedastic envelope model.

Contents

Syntax

```
ModelOutput = henv(X, Y, u)
ModelOutput = henv(X, Y, u, Opts)
```

Input

X: Group indicators. A matrix with n rows. X can only have p unique rows, where p is the number of groups. For example, if there are two groups, X can only have 2 different kinds of rows, such as (0, 1) and (1, 0), or (1, 0, 10) and (0, 5, 6). The number of columns is not restricted, as long as X only has p unique rows.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be greater than p.

u: Dimension of the envelope. An integer between 0 and r.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out Grassmann manifold optimization process, logical 0 or 1. Default value: 0.
- Opts.init: The initial value for the heteroscedastic envelope subspace. An r by u matrix. Default value is the one generated by function get_Init4henv.

Output

ModelOutput: A list that contains the maximum likelihood estimators and some statistics.

- ModelOutput.mu: The heteroscedastic envelope estimator of the grand mean. A r by 1 vector.
- ModelOutput.mug: The heteroscedastic envelope estimator of the group mean. A r by p matrix, the ith column of the matrix contains the mean for the ith group.
- ModelOutput.Yfit: A n by r matrix, the ith row gives the group mean of the group that the ith observation belongs to. As X is just a group indicator, and is not ordinal, ModelOutput.mug alone does not tell which group corresponds to which group mean.

- ModelOutput.Gamma: The orthogonal basis of the envelope subspace. An r by u semi-orthogonal matrix.
- ModelOutput.Gamma0: The orthogonal basis of the complement of the envelope subspace. An r by r-u semi-orthogonal matrix.
- ModelOutput.beta: The heteroscedastic envelope estimator of the group main effect. An r by p matrix, the ith column of the matrix contains the main effect for the ith group.
- ModelOutput.groupInd: A matrix containing the unique values of group indicators. The matrix has p rows. The group mean of the ith row is stored in the ith column of ModelOutput.mug.
- ModelOutput.Sigma: The heteroscedastic envelope estimator of the error covariance matrix. A three dimensional matrix with dimension r, r and p, ModelOutput.Sigma(:, :, i) contains the estimated covariance matrix for the ith group.
- ModelOutput.eta: The coordinates of β with respect to Gamma. An u by p matrix, the ith column contains the coordinates of the main effect of the ith group with respect to Gamma.
- ModelOutput.Omega: The coordinates of Sigma with respect to Gamma. An u by u by p matrix, ModelOutput.Omega(:, :, i) contains the coordinates of the covariance matrix of the ith group with respect to Gamma.
- ModelOutput.Omega0: The coordinates of Sigma with respect to Gamma0. An r - u by r - u matrix.
- ModelOutput.l: The maximized log likelihood function. A real number.
- ModelOutput.np: The number of parameters in the heteroscedastic envelope model. A positive integer.
- ModelOutput.covMatrix: The asymptotic covariance of (μ', vec(β'))'. An r(p + 1) by r(p + 1) matrix. The covariance matrix returned are asymptotic. For the actual standard errors, multiply by 1 / n.
- ModelOutput.asyHenv: The asymptotic standard errors for elements in β under the heteroscedastic envelope model. An r by p matrix. The standard errors returned are asymptotic, for actual standard errors, multiply by 1 / sqrt(n).
- ModelOutput.ratio: The asymptotic standard error ratio of the standard multivariate linear regression estimator over the heteroscedastic envelope estimator. An r by p matrix, the (i, j)th element in ModelOutput.ratio is the elementwise standard error ratio for the ith element in the jth group mean effect.
- ModelOutput.ng: The number of observations in each group. A p by 1 vector.

Description

This function fits the heteroscedastic envelope model to the responses and predictors, using the maximum likelihood estimation. When the dimension of the envelope is between 1 and r-1, we implemented the algorithm in Su and Cook (2012). When the dimension is r, then the envelope model degenerates to the standard multivariate linear model for comparing group means. When the dimension is 0, it means there is not any group effect, and the fitting is different.

References

1. The codes are implemented based on the algorithm in Section 2.2 of Su and Cook (2012).
2. The Grassmann manifold optimization step calls the package sg_min 2.4.3 by Ross Lippert (http://web.mit.edu/~ripper/www/sgmin.html).

Example

The following codes produce the results of the water strider example in Su and Cook (2011).

```
load waterstrider.mat
u = lrt_henv(X, Y, 0.01)
```

```
u =

     6
```

```
ModelOutput = henv(X, Y, u)
ModelOutput.ratio
```

```
ModelOutput =

           mu: [8x1 double]
          mug: [8x3 double]
         Yfit: [90x8 double]
        Gamma: [8x6 double]
       Gamma0: [8x2 double]
         beta: [8x3 double]
     groupInd: [3x2 double]
        Sigma: [8x8x3 double]
          eta: [6x3 double]
        Omega: [6x6x3 double]
       Omega0: [2x2 double]
           np: 98
            l: 1.0051e+03
    covMatrix: [32x32 double]
      asyHenv: [8x3 double]
        ratio: [8x3 double]
           ng: [3x1 double]


ans =

    6.5439   11.2830    6.4954
    4.6325    5.3226    4.7242
    4.4456    5.0741    4.4198
    4.7338    6.2469    5.1937
    8.0377   12.5386    9.4823
    9.5067   11.5974   11.3444
   11.8632   15.6080   12.5611
    6.9792   11.1559   10.1002
```

5.7 lrt_henv

Select the dimension of the envelope subspace using likelihood ratio testing for the heteroscedastic envelope model.

Contents

Syntax

```
u = lrt_henv(X, Y, alpha)
u = lrt_henv(X, Y, alpha, Opts)
```

Input

X: Group indicators. A matrix with n rows. X can only have p unique rows, where p is the number of groups. For example, if there are two groups, X can only have 2 different kinds of rows, such as (0, 1) and (1, 0), or (1, 0, 10) and (0, 5, 6). The number of columns is not restricted, as long as X only has p unique rows.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be greater than p.

alpha: Significance level for testing. A real number between 0 and 1, often taken at 0.05 or 0.01.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and r.

Description

This function implements the likelihood ratio testing procedure to select the dimension of the envelope subspace in heteroscedastic envelope model, with pre-specified significance level α.

Example

```
load waterstrider.mat
u = lrt_henv(X, Y, 0.01)
```

```
u =

     6
```

5.8 predict_henv

Perform estimation or prediction under the heteroscedastic envelope model.

Contents

Syntax

```
PredictOutput = predict_henv(ModelOutput, Xnew, infType)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from henv.

Xnew: A group indicator. It must be a column vector, whose transpose is the same as one of the group indictors from the original data.

infType: A string of characters indicting the inference type, the choices can be 'estimation' or 'prediction'.

Output

PredictOutput: A list containing the results of the inference.

- PredictOutput.value: The fitted value or the prediction value evaluated at Xnew. An r by 1 vector.
- PredictOutput.covMatrix: The covariance matrix of PredictOutput.value. An r by r matrix.
- PredictOutput.SE: The standard error of elements in PredictOutput.value. An r by 1 vector.

Description

This function evaluates the inner envelope model at new value Xnew. It can perform estimation: find the group mean for the group indicated by Xnew, or prediction: predict Y for the group indicated by Xnew. The covariance matrix and the standard errors are also provided.

Example

```
load waterstrider.mat
u = lrt_henv(X, Y, 0.01);
ModelOutput = henv(X, Y, u);
ModelOutput.groupInd
ModelOutput.mug
Xnew = X(1, :)'
```

```
  ans =

    -1    -1
     0     1
     1     0


ans =

   -1.1417   -1.1267   -1.0845
   -1.4063   -1.4067   -1.3132
   -1.3314   -1.3336   -1.2152
   -0.3113   -0.1839   -0.1736
    0.4003    0.3847    0.3072
    0.4107    0.3753    0.3735
    0.3467    0.3271    0.3179
   -0.1954   -0.2100   -0.3488


Xnew =

     1
     0
```

```
PredictOutput = predict_henv(ModelOutput, Xnew, 'estimation')
PredictOutput.value %This is the 3rd group mean
PredictOutput.SE
```

```
PredictOutput =

        value: [8x1 double]
    covMatrix: [8x8 double]
           SE: [8x1 double]


ans =

   -1.0845
   -1.3132
   -1.2152
   -0.1736
    0.3072
    0.3735
    0.3179
   -0.3488
```

```
ans =

    0.0682
    0.0695
    0.0651
    0.0436
    0.0832
    0.0636
    0.0847
    0.0698
```

```
PredictOutput = predict_henv(ModelOutput, Xnew, 'prediction')
PredictOutput.SE
```

```
PredictOutput =

        value: [8x1 double]
    covMatrix: [8x8 double]
           SE: [8x1 double]


ans =

    0.3720
    0.3812
    0.3581
    0.2398
    0.4612
    0.3519
    0.4710
    0.3854
```

5.9 testcoefficient_henv

```
This function tests the null hypothesis L * beta * R = A versus the
alternative hypothesis L * beta * R ~= A, where beta is estimated under
the heteroscedastic envelope model.
```

Contents

Syntax

```
TestOutput = testcoefficient_henv(ModelOutput)
TestOutput = testcoefficient_henv(ModelOutput, TestInput)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from henv.

TestInput: A list that specifies the null hypothesis, including L, R, and A. If not provided by the user, default values will be used.

- TestInput.L: The matrix multiplied to β on the left. It is a d1 by r matrix, while d1 is less than or equal to r - 1. Default value: identity matrix I_r.
- TestInput.R: The matrix multiplied to β on the right. It is a p by d2 matrix, while d2 is less than or equal to p. Default value: identity matrix $(I_{p-1}, 0_{(p-1)\times 1})^T$. This is because the columns of β sum to 0. Then we cannot use I_p as default.
- TestInput.A: The matrix on the right hand side of the equation. It is a d1 by d2 matrix. Default value: d1 by d2 zero matrix.

Output

TestOutput: A list containing test statistics, degrees of freedom for the reference chi-squared distribution, the p-value, and the covariance matrix of vec($L\beta R$). At the same time, a table is printed out.

- TestOutput.chisqStatistic: The test statistics. A real number.
- TestOutput.df: The degrees of freedom of the reference chi-squared distribution. A positive integer.
- TestOutput.pValue: p-value of the test. A real number in [0, 1].
- TestOutput.covMatrix: The covariance matrix of vec($L\beta R$). A d1 * d2 by d1 * d2 matrix.

Description

This function tests for hypothesis $H_0 : L\beta R = A$, versus $H_\alpha : L\beta R \neq A$. The β is estimated by the heteroscedastic envelope model. If the user does not specify the values for L, R and A, then the test is equivalent to the standard F test on if all the main group effects are 0. The test

statistics used is vec $(L\beta R - A)\ \hat{\Sigma}^{-1}$ vec $(L\beta R - A)^T$, and the reference distribution is chi-squared distribution with degrees of freedom d1 * d2.

Example

```
load waterstrider.mat
u = lrt_henv(X, Y, 0.01);
ModelOutput = henv(X, Y, u);
TestOutout = testcoefficient_henv(ModelOutput);
```

```
Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A         226.256        16       0.0000
-------------------------------------------------------------------------
```

```
r = size(Y, 2);
p = size(ModelOutput.beta, 2);
TestInput.L = rand(2, r);
TestInput.R = rand(p, 1);
TestInput.A = zeros(2, 1);
TestOutout = testcoefficient_henv(ModelOutput, TestInput);
```

```
Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A          23.429         2        0.0000
-------------------------------------------------------------------------
```

6

ienv

6.1 aic_ienv

Select the dimension of the inner envelope subspace using Akaike information criterion.

Contents

Syntax

```
u = aic_ienv(X, Y)
u = aic_ienv(X, Y, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is the number of observations. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses. The responses must be continuous variables.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the inner envelope. An integer between 0 and p or equal to r.

Description

This function implements the Akaike information criteria (AIC) to select the dimension of the inner envelope subspace.

Example

```
load irisf.mat
u = aic_ienv(X, Y)
```

```
u =

     1
```

6.2 bic_ienv

Select the dimension of the inner envelope subspace using Bayesian information criterion.

Contents

Syntax

```
u = bic_ienv(X, Y)
u = bic_ienv(X, Y, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is the number of observations. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses. The responses must be continuous variables.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the inner envelope. An integer between 0 and p or equal to r.

Description

This function implements the Bayesian information criteria (BIC) to select the dimension of the inner envelope subspace.

Example

```
load irisf.mat
u = bic_ienv(X, Y)
```

```
u =

     1
```

6.3 bstrp_ienv

Compute bootstrap standard error for the inner envelope model.

Contents

Syntax

```
bootse = bstrp_ienv(X, Y, u, B)
bootse = bstrp_ienv(X, Y, u, B, Opts)
```

Input

X: Predictors, an n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses, an n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

u: Dimension of the inner envelope. An integer between 0 and p or equal to r.

B: Number of bootstrap samples. A positive integer.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

bootse: The standard error for elements in β computed by bootstrap. An r by p matrix.

Description

This function computes the bootstrap standard errors for the regression coefficients in the inner envelope model by bootstrapping the residuals.

Example

```
load irisf.mat

u = bic_ienv(X, Y)
```

```
u =

     1
```

```
B = 100;
bootse = bstrp_ienv(X, Y, u, B)
```

```
bootse =

   13.4695    4.9601
    7.4709    2.7315
   14.9316    5.2913
    8.7597    3.0853
```

6.4 dF4ienv

First derivative of the objective function for computing the inner envelope subspace.

Contents

Syntax

```
df = dF4ienv(R, DataParameter)
```

Input

R: An r by u semi-orthogonal matrix, 0 < u <= p.

DataParameter: A structure that contains the statistics calculated from the data.

Output

df: The first derivative of the objective function for computing the inner envelope subspace. An r by u matrix.

Description

This first derivative of F4ienv obtained by matrix calculus calculations.

6.5 F4ienv

Objective function for computing the inner envelope subspace.

Contents

Syntax

```
f = F4ienv(R, DataParameter)
```

Input

R: An r by u semi orthogonal matrix, 0 < u <= p.

DataParameter: A structure that contains the statistics calculated from the data.

Output

f: A scalar containing the value of the objective function evaluated at R.

Description

The objective function is derived in Section 3.3 in Su and Cook (2012) by using maximum likelihood estimation. The columns of the semi-orthogonal matrix that minimizes this function span the estimated inner envelope subspace.

6.6 ienv

Fit the inner envelope model.

Contents

Syntax

```
ModelOutput = ienv(X, Y, u)
ModelOutput = ienv(X, Y, u, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be strictly greater than p.

u: Dimension of the inner envelope. An integer between 0 and p or equal to r.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out Grassmann manifold optimization process, logical 0 or 1. Default value: 0.
- Opts.init: The initial value for the inner envelope subspace. An r by u matrix. Default value is the one generated by function get_Init.

Output

ModelOutput: A list that contains the maximum likelihood estimators and some statistics.

- ModelOutput.beta: The envelope estimator of the regression coefficients β. An r by p matrix.
- ModelOutput.Sigma: The envelope estimator of the error covariance matrix. An r by r matrix.
- ModelOutput.Gamma1: The orthogonal basis of the inner envelope subspace. An r by u semi-orthogonal matrix.
- ModelOutput.Gamma0: The orthogonal basis of the complement of the inner envelope subspace. An r by r-u semi-orthogonal matrix.

- ModelOutput.eta1: The transpose of the coordinates of β with respect to Gamma1. An p by u matrix.
- ModelOutput.B: An (r - u) by (p - u) semi-orthogonal matrix, so that (Gamma, Gamma0 * B) spans β.
- ModelOutput.eta2: The transpose of the coordinates of β with respect to Gamma0. An p by (p-u) matrix.
- ModelOutput.Omega1: The coordinates of Sigma with respect to Gamma1. An u by u matrix.
- ModelOutput.Omega0: The coordinates of Sigma with respect to Gamma0. An r - u by r - u matrix.
- ModelOutput.alpha: The estimated intercept in the inner envelope model. An r by 1 vector.
- ModelOutput.l: The maximized log likelihood function. A real number.
- ModelOutput.covMatrix: The asymptotic covariance of vec(β). An rp by rp matrix. The covariance matrix returned are asymptotic. For the actual standard errors, multiply by 1/n.
- ModelOutput.asyIenv: Asymptotic standard error for elements in β under the inner envelope model. An r by p matrix. The standard errors returned are asymptotic, for actual standard errors, multiply by 1 / sqrt(n).
- ModelOutput.ratio: The asymptotic standard error ratio of the standard multivariate linear regression estimator over the inner envelope estimator, for each element in β. An r by p matrix.
- ModelOutput.np: The number of parameters in the inner envelope model. A positive integer.
- ModelOutput.n: The number of observations in the data. A positive integer.

Description

This function fits the inner envelope model to the responses and predictors, using the maximum likelihood estimation. When the dimension of the envelope is between 1 and p-1, we implemented the algorithm in Su and Cook (2012). When the dimension is p, then the inner envelope model degenerates to the standard multivariate linear regression. When the dimension is 0, it means that X and Y are uncorrelated, and the fitting is different.

References

1. The codes are implemented based on the algorithm in Su and Cook (2012).
2. The Grassmann manifold optimization step calls the package sg_min 2.4.3 by Ross Lippert (http://web.mit.edu/~ripper/www/sgmin.html).

Example

The following codes gives the results of the Fisher's iris data example in Su and Cook (2012).

```
load irisf.mat

d = bic_ienv(X, Y)
```

```
d =

     1
```

```
ModelOutput = ienv(X, Y, d)
1 - 1 ./ ModelOutput.ratio
```

```
ModelOutput =

         beta: [4x2 double]
        Sigma: [4x4 double]
       Gamma1: [4x1 double]
       Gamma0: [4x3 double]
            B: [3x1 double]
         eta1: [2x1 double]
         eta2: [2x1 double]
       Omega1: 8.3751
       Omega0: [3x3 double]
        alpha: [4x1 double]
           np: 16
            l: -1.4805e+03
    covMatrix: [8x8 double]
      asyIenv: [4x2 double]
        ratio: [4x2 double]
            n: 150


ans =

    0.0049    0.2122
    0.0020    0.0806
    0.0033    0.1404
    0.0004    0.0138
```

6.7 lrt_ienv

Select the dimension of the inner envelope subspace using likelihood ratio testing.

Contents

Syntax

```
u = lrt_ienv(X, Y, alpha)
u = lrt_ienv(X, Y, alpha, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

alpha: Significance level for testing. A real number between 0 and 1, often taken at 0.05 or 0.01.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the inner envelope. An integer between 0 and p or equal to r.

Description

This function implements the likelihood ratio testing procedure to select the dimension of the inner envelope subspace, with pre-specified significance level α.

Example

```
load irisf.mat

alpha = 0.01;
u = lrt_ienv(X, Y, alpha)
```

```
u =

     1
```

6.8 predict_ienv

Perform estimation or prediction under the inner envelope model.

Contents

Syntax

```
PredictOutput = predict_ienv(ModelOutput, Xnew, infType)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from ienv.

Xnew: The value of X with which to estimate or predict Y. A p by 1 vector.

infType: A string of characters indicting the inference type, the choices can be 'estimation' or 'prediction'.

Output

PredictOutput: A list containing the results of the inference.

- PredictOutput.value: The fitted value or the prediction value evaluated at Xnew. An r by 1 vector.
- PredictOutput.covMatrix: The covariance matrix of PredictOutput.value. An r by r matrix.
- PredictOutput.SE: The standard error of elements in PredictOutput.value. An r by 1 vector.

Description

This function evaluates the inner envelope model at new value Xnew. It can perform estimation: find the fitted value when X = Xnew, or prediction: predict Y when X = Xnew. The covariance matrix and the standard errors are also provided.

Example

```
load irisf.mat
d = bic_ienv(X, Y);
ModelOutput = ienv(X, Y, d);
Xnew = X(1, :)';
PredictOutput = predict_ienv(ModelOutput, Xnew, 'estimation')
[PredictOutput.value, Y(1, :)']  % Compare the fitted value with the data
PredictOutput.SE
```

```
PredictOutput =

        value: [4x1 double]
    covMatrix: [4x4 double]
           SE: [4x1 double]


ans =

   49.9458   51.0000
   34.2592   35.0000
   14.5771   14.0000
    2.4513    2.0000


ans =

    1.0978
    0.7146
    0.9265
    0.4357
```

```
PredictOutput = predict_ienv(ModelOutput, Xnew, 'prediction')
PredictOutput.SE
```

```
PredictOutput =

        value: [4x1 double]
    covMatrix: [4x4 double]
           SE: [4x1 double]


ans =

    5.2197
    3.3897
    4.3996
    2.0642
```

6.9 testcoefficient_ienv

```
This function tests the null hypothesis L * beta * R = A versus the
alternative hypothesis L * beta * R ~= A, where beta is estimated under
the inner envelope model.
```

Contents

Syntax

```
TestOutput = testcoefficient_ienv(ModelOutput)
TestOutput = testcoefficient_ienv(ModelOutput, TestInput)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from ienv.

TestInput: A list that specifies the null hypothesis, including L, R, and A. If not provided by the user, default values will be used.

- TestInput.L: The matrix multiplied to β on the left. It is a d1 by r matrix, while d1 is less than or equal to r. Default value: identity matrix I_r.
- TestInput.R: The matrix multiplied to β on the right. It is a p by d2 matrix, while d2 is less than or equal to p. Default value: identity matrix I_p.
- TestInput.A: The matrix on the right hand side of the equation. It is a d1 by d2 matrix. Default value: d1 by d2 zero matrix.

Output

TestOutput: A list containing test statistics, degrees of freedom for the reference chi-squared distribution, the p-value, and the covariance matrix of vec($L\beta R$). At the same time, a table is printed out.

- TestOutput.chisqStatistic: The test statistics. A real number.
- TestOutput.df: The degrees of freedom of the reference chi-squared distribution. A positive integer.
- TestOutput.pValue: p-value of the test. A real number in [0, 1].
- TestOutput.covMatrix: The covariance matrix of vec($L\beta R$). A d1 * d2 by d1 * d2 matrix.

Description

This function tests for hypothesis $H_0 : L\beta R = A$, versus $H_\alpha : L\beta R \neq A$. The β is estimated by the inner envelope model. If the user does not specify the values for L, R and A, then the test is equivalent to the standard F test on if $\beta = 0$. The test statistics used is vec $(L\beta R - A)\ \hat{\Sigma}^{-1}$ vec $(L\beta R - A)^T$, and the reference distribution is chi-squared distribution with degrees of freedom d1 * d2.

Example

```
load irisf.mat
d = bic_ienv(X,Y);
ModelOutput = ienv(X,Y,d);
TestOutout = testcoefficient_ienv(ModelOutput);
```

```
Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A        4642.913        8       0.0000
-------------------------------------------------------------------------
```

```
TestInput.L = rand(2, 4);
TestInput.R = rand(2, 1);
TestInput.A = zeros(2, 1);
TestOutout = testcoefficient_ienv(ModelOutput, TestInput);
```

```
Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A        1834.229        2       0.0000
-------------------------------------------------------------------------
```

7

penv

7.1 aic_penv

Select the dimension of the partial envelope subspace using Akaike information criterion.

Contents

Syntax

```
u = aic_penv(X, Y)
u = aic_penv(X, Y, Opts)
```

Input

X: A list containing the value of X1 and X2.

- X.X1: Predictors of main interest. An n by p1 matrix, n is the number of observations, and p1 is the number of main predictors. The predictors can be univariate or multivariate, discrete or continuous.
- X.X2: Covariates, or predictors not of main interest. An n by p2 matrix, p2 is the number of covariates.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and r.

Description

This function implements the Akaike information criteria (AIC) to select the dimension of the partial envelope subspace.

Example

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
u = aic_penv(X, Y)
```

```
u =

     3
```

7.2 bic_penv

Select the dimension of the partial envelope subspace using Bayesian information criterion.

Contents

Syntax

```
u = bic_penv(X, Y)
u = bic_penv(X, Y, Opts)
```

Input

X: A list containing the value of X1 and X2.

- X.X1: Predictors of main interest. An n by p1 matrix, n is the number of observations, and p1 is the number of main predictors. The predictors can be univariate or multivariate, discrete or continuous.
- X.X2: Covariates, or predictors not of main interest. An n by p2 matrix, p2 is the number of covariates.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and r.

Description

This function implements the Bayesian information criteria (BIC) to select the dimension of the partial envelope subspace.

Example

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
u = bic_penv(X, Y)
```

```
u =

     1
```

7.3 bstrp_penv

Compute bootstrap standard error for the partial envelope model.

Contents

Syntax

```
bootse = bstrp_penv(X, Y, u, B)
bootse = bstrp_penv(X, Y, u, B, Opts)
```

Input

X: A list containing the value of X1 and X2.

- X.X1: Predictors of main interest. An n by p1 matrix, n is the number of observations, and p1 is the number of main predictors. The predictors can be univariate or multivariate, discrete or continuous.
- X.X2: Covariates, or predictors not of main interest. An n by p2 matrix, p2 is the number of covariates.

Y: Multivariate responses, an n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

u: Dimension of the partial envelope subspace. A positive integer between 0 and r.

B: Number of bootstrap samples. A positive integer.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

bootse: The standard error for elements in β_1 computed by bootstrap. An r by p1 matrix.

Description

This function computes the bootstrap standard errors for the regression coefficients in the partial envelope model by bootstrapping the residuals.

Example

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
alpha = 0.01;
u = lrt_penv(X, Y, alpha)
```

```
u =

     1
```

```
B = 100;
bootse = bstrp_penv(X, Y, u, B)
```

```
bootse =

    0.0074
    0.0021
    0.0043
    0.0019
```

7.4 lrt_penv

Select the dimension of the partial envelope subspace using likelihood ratio testing.

Contents

Syntax

```
u = lrt_penv(X, Y, alpha)
u = lrt_penv(X, Y, alpha, Opts)
```

Input

X: A list containing the value of X1 and X2.

- X.X1: Predictors of main interest. An n by p1 matrix, n is the number of observations, and p1 is the number of main predictors. The predictors can be univariate or multivariate, discrete or continuous.
- X.X2: Covariates, or predictors not of main interest. An n by p2 matrix, p2 is the number of covariates.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

alpha: Significance level for testing. A real number between 0 and 1, often taken at 0.05 or 0.01.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the partial envelope subspace. An integer between 0 and r.

Description

This function implements the likelihood ratio testing procedure to select the dimension of the partial envelope subspace, with pre-specified significance level α.

Example

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
alpha = 0.01;
u = lrt_penv(X, Y, alpha)
```

```
u =

     1
```

7.5 penv

Fit the partial envelope model.

Contents

Syntax

```
ModelOutput = penv(X, Y, u)
ModelOutput = penv(X, Y, u, Opts)
```

Input

X: A list containing the value of X1 and X2.

- X.X1: Predictors of main interest. An n by p1 matrix, n is the number of observations, and p1 is the number of main predictors. The predictors can be univariate or multivariate, discrete or continuous.
- X.X2: Covariates, or predictors not of main interest. An n by p2 matrix, p2 is the number of covariates.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be strictly greater than p1.

u: Dimension of the partial envelope. An integer between 0 and r.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out Grassmann manifold optimization process, logical 0 or 1. Default value: 0.
- Opts.init: The initial value for the partial envelope subspace. An r by u matrix. Default value is the one generated by function get_Init.

Output

ModelOutput: A list that contains the maximum likelihood estimators and some statistics.

- ModelOutput.beta1: The partial envelope estimator of β_1, which is the regression coefficients for X1. An r by p1 matrix.
- ModelOutput.beta2: The partial envelope estimator of β_2, which is the regression coefficients for X2. An r by p2 matrix.

- ModelOutput.Sigma: The partial envelope estimator of the error covariance matrix. An r by r matrix.
- ModelOutput.Gamma: The orthogonal basis of the partial envelope subspace. An r by u semi-orthogonal matrix.
- ModelOutput.Gamma0: The orthogonal basis of the complement of the partial envelope subspace. An r by r - u semi-orthogonal matrix.
- ModelOutput.eta: The coordinates of β_1 with respect to Gamma. An u by p1 matrix.
- ModelOutput.Omega: The coordinates of Sigma with respect to Gamma. An u by u matrix.
- ModelOutput.Omega0: The coordinates of Sigma with respect to Gamma0. An r - u by r - u matrix.
- ModelOutput.alpha: The estimated intercept in the partial envelope model. An r by 1 vector.
- ModelOutput.l: The maximized log likelihood function. A real number.
- ModelOutput.covMatrix: The asymptotic covariance of (vec(β_2)', vec(β_1)')'. An rp by rp matrix. The covariance matrix returned are asymptotic. For the actual standard errors, multiply by 1/n.
- ModelOutput.asyPenv: Asymptotic standard error for elements in β_1 under the partial envelope model. An r by p1 matrix. The standard errors returned are asymptotic, for actual standard errors, multiply by 1/sqrt(n).
- ModelOutput.ratio: The asymptotic standard error ratio of the standard multivariate linear regression estimator over the partial envelope estimator, for each element in β_1. An r by p1 matrix.
- ModelOutput.np: The number of parameters in the envelope model. A positive integer.
- ModelOutput.n: The number of observations in the data. A positive integer.

Description

This function fits the partial envelope model to the responses Y and predictors X1 and X2, using the maximum likelihood estimation. When the dimension of the envelope is between 1 and r - 1, we implemented the algorithm in Su and Cook (2011). When the dimension is r, then the partial envelope model degenerates to the standard multivariate linear regression with Y as the responses and both X1 and X2 as predictors. When the dimension is 0, X1 and Y are uncorrelated, and the fitting is the standard multivariate linear regression with Y as the responses and X2 as the predictors.

References

1. The codes are implemented based on the algorithm in Section 3.2 of Su and Cook (2012).
2. The Grassmann manifold optimization step calls the package sg_min 2.4.3 by Ross Lippert (http://web.mit.edu/~ripper/www/sgmin.html).

Example

The following codes reconstruct the results of the paper and fiber example in Su and Cook (2012).

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
```

```
       alpha = 0.01;
       u = lrt_penv(X, Y, alpha)


u =

     1


       ModelOutput = penv(X, Y, u)
       ModelOutput.Omega
       eig(ModelOutput.Omega0)
       ModelOutput.ratio


ModelOutput =

        beta1: [4x1 double]
        beta2: [4x2 double]
        alpha: [4x1 double]
        Gamma: [4x1 double]
          eta: 0.0047
        Omega: 0.0149
       Omega0: [3x3 double]
        Sigma: [4x4 double]
            l: -35.6323
           np: 23
    covMatrix: [12x12 double]
      asyPenv: [4x1 double]
        ratio: [4x1 double]
            n: 62


ans =

    0.0149


ans =

    4.9819
    0.0999
    0.0050


ans =

   66.0742
    6.9326
   10.5048
    9.6279
```

7.6 predict_penv

Perform estimation or prediction under the partial envelope model.

Contents

Syntax

PredictOutput = predict_penv(ModelOutput, Xnew, infType)

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from penv.

Xnew: A list containing the value of X1 and X2 with which to estimate or predict Y.

```
* Xnew.X1: A p1 by 1 vector containing the value of X1.
* Xnew.X2: A p2 by 1 vector containing the value of X2.
```

infType: A string of characters indicting the inference type, the choices can be 'estimation' or 'prediction'.

Output

PredictOutput: A list containing the results of the inference.

- PredictOutput.value: The fitted value or the prediction value evaluated at Xnew. An r by 1 vector.
- PredictOutput.covMatrix: The covariance matrix of PredictOutput.value. An r by r matrix.
- PredictOutput.SE: The standard error of elements in PredictOutput.value. An r by 1 vector.

Description

This function evaluates the envelope model at new value Xnew. It can perform estimation: find the fitted value when X = Xnew, or prediction: predict Y when X = Xnew. The covariance matrix and the standard errors are also provided.

Example

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
alpha = 0.01;
u = lrt_penv(X, Y, alpha);
ModelOutput = penv(X, Y, u);
Xnew.X1 = X.X1(1, :)';
Xnew.X2 = X.X2(1, :)';
PredictOutput = predict_penv(ModelOutput, Xnew, 'estimation')
[PredictOutput.value, Y(1, :)'] % Compare the fitted value with the data
PredictOutput.SE
```

```
PredictOutput =

        value: [4x1 double]
    covMatrix: [4x4 double]
           SE: [4x1 double]


ans =

   21.1169   21.3120
    7.1173    7.0390
    5.3637    5.3260
    0.8737    0.9320


ans =

    1.4680
    0.4234
    0.7145
    0.3161
```

```
PredictOutput = predict_penv(ModelOutput, Xnew, 'prediction')
PredictOutput.SE
```

```
PredictOutput =

        value: [4x1 double]
    covMatrix: [4x4 double]
           SE: [4x1 double]


ans =
```

```
2.4277
0.6982
1.1802
0.5220
```

7.7 testcoefficient_penv

`This function tests the null hypothesis L * beta1 * R = A versus the alternative hypothesis L * beta1 * R ~= A, where beta1 is estimated under the envelope model.`

Contents

Syntax

```
TestOutput = testcoefficient_penv(ModelOutput)
TestOutput = testcoefficient_penv(ModelOutput, TestInput)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from penv.

TestInput: A list that specifies the null hypothesis, including L, R, and A. If not provided by the user, default values will be used.

- TestInput.L: The matrix multiplied to β_1 on the left. It is a d1 by r matrix, while d1 is less than or equal to r. Default value: identity matrix I_r.
- TestInput.R: The matrix multiplied to β_1 on the right. It is a p1 by d2 matrix, while d2 is less than or equal to p1. Default value: identity matrix I_{p1}.
- TestInput.A: The matrix on the right hand side of the equation. It is a d1 by d2 matrix. Default value: d1 by d2 zero matrix.

Output

TestOutput: A list containing test statistics, degrees of freedom for the reference chi-squared distribution, the p-value, and the covariance matrix of vec($L\beta R$). At the same time, a table is printed out.

- TestOutput.chisqStatistic: The test statistics. A real number.
- TestOutput.df: The degrees of freedom of the reference chi-squared distribution. A positive integer.
- TestOutput.pValue: p-value of the test. A real number in [0, 1].
- TestOutput.covMatrix: The covariance matrix of vec($L\beta_1 R$). A d1 * d2 by d1 * d2 matrix.

Description

This function tests for hypothesis $H_0 : L\beta_1 R = A$, versus $H_\alpha : L\beta_1 R \neq A$. The β_1 is estimated by the partial envelope model. If the user does not specify the values for L, R and A, then the test is equivalent to the standard F test on if $\beta_1 = 0$. The test statistics used is vec $(L\beta_1 R - A)\ \hat{\Sigma}^{-1}$ vec $(L\beta_1 R - A)^T$, and the reference distribution is chi-squared distribution with degrees of freedom d1 * d2.

Example

```
load fiberpaper.dat
Y = fiberpaper(:, 1 : 4);
Xtemp = fiberpaper(:, 5 : 7);
X.X1 = Xtemp(:, 3);
X.X2 = Xtemp(:, 1 : 2);
alpha = 0.01;
u = lrt_penv(X, Y, alpha);
ModelOutput = penv(X, Y, u);
TestOutout = testcoefficient_penv(ModelOutput);
```

```
 Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A          12.604         4        0.0134
-------------------------------------------------------------------------
```

```
r = size(Y, 2);
p1 = size(X.X1, 2);
TestInput.L = rand(2, r);
TestInput.R = rand(p1, 1);
TestInput.A = zeros(2, 1);
TestOutout = testcoefficient_penv(ModelOutput, TestInput);
```

```
Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A          11.452         2        0.0033
-------------------------------------------------------------------------
```

8

senv

8.1 aic_senv

Select the dimension of the scaled envelope subspace using Akaike information criterion.

Contents

Syntax

```
u = aic_senv(X, Y)
u = aic_senv(X, Y, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is the number of observations. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses. The responses must be continuous variables.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the inner envelope. An integer between 0 and r.

Description

This function implements the Akaike information criteria (AIC) to select the dimension of the scaled envelope subspace.

Example

```
load('sales.txt')
Y = sales(:, 4 : 7);
X = sales(:, 1 : 3);
u = aic_senv(X, Y)
```

```
u =

     4
```

8.2 bic_senv

Select the dimension of the scaled envelope subspace using Bayesian information criterion.

Contents

Syntax

```
u = bic_senv(X, Y)
u = bic_senv(X, Y, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is the number of observations. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses. The responses must be continuous variables.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the inner envelope. An integer between 0 and r.

Description

This function implements the Bayesian information criteria (BIC) to select the dimension of the scaled envelope subspace.

Example

```
load('sales.txt')
Y = sales(:, 4 : 7);
X = sales(:, 1 : 3);
u = bic_senv(X, Y)
```

```
u =

     2
```

8.3 bstrp_senv

Compute bootstrap standard error for the scaled envelope model.

Contents

Syntax

```
bootse = bstrp_senv(X, Y, u, B)
bootse = bstrp_senv(X, Y, u, B, Opts)
```

Input

X: Predictors, an n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses, an n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

u: Dimension of the envelope subspace. A positive integer between 0 and r.

B: Number of bootstrap samples. A positive integer.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

bootse: The standard error for elements in β computed by bootstrap. An r by p matrix.

Description

This function computes the bootstrap standard errors for the regression coefficients in the scaled envelope model by bootstrapping the residuals.

Example

```
load('sales.txt')
Y = sales(:, 4 : 7);
X = sales(:, 1 : 3);
```

```
u = bic_senv(X, Y)
```

```
u =

     2
```

```
B = 20;
bootse = bstrp_senv(X, Y, u, B)
```

```
bootse =

    0.0539    0.0472    0.0554
    0.0675    0.0912    0.1178
    0.0567    0.0781    0.0791
    0.0986    0.0944    0.1283
```

8.4 dF4senv

First derivative of the objective function for computing the envelope subspace in the scaled envelope model.

Contents

Syntax

```
df = dF4senv(R, DataParameter)
```

Input

R: An r by u semi-orthogonal matrix, 0 < u <= r.

DataParameter: A structure that contains the statistics calculated from the data.

Output

df: The first derivative of the objective function for computing the envelope subspace. An r by u matrix.

Description

This first derivative of F4senv obtained by matrix calculus calculations.

8.5 F4senv

Objective function for computing the envelope subspace in scaled envelope model.

Contents

Syntax

```
f = F4senv(R, DataParameter)
```

Input

R: An r by u semi orthogonal matrix, 0 < u <= r.

DataParameter: A structure that contains the statistics calculated from the data.

Output

f: A scalar containing the value of the objective function evaluated at R.

Description

The objective function is derived in Section 4.1 in Cook and Su (2012) using maximum likelihood estimation. The columns of the semi-orthogonal matrix that minimizes this function span the estimated envelope subspace.

8.6 objfun

Objective function for computing the scales in the scaled envelope model.

Contents

Syntax

```
f = objfun(d, Gamma, DataParameter)
```

Input

d: An r - 1 dimensional column vector containing the scales for the 2nd to the rth responses. All the entries in d are positive.

Gamma: A r by u semi-orthogonal matrix that spans the envelope subspace or the estimated envelope subspace.

DataParameter: A structure that contains the statistics calculated form the data.

Output

f: A scalar containing the value of the objective function evaluated at d.

Description

The objective function is derived in Section 4.1 of Su and Cook (2012) using maximum likelihood estimation.

8.7 predict_senv

Perform estimation or prediction under the scaled envelope model.

Contents

Syntax

```
PredictOutput = predict_senv(ModelOutput, Xnew, infType)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from senv.

Xnew: The value of X with which to estimate or predict Y. A p by 1 vector.

infType: A string of characters indicting the inference type, the choices can be 'estimation' or 'prediction'.

Output

PredictOutput: A list containing the results of the inference.

- PredictOutput.value: The fitted value or the prediction value evaluated at Xnew. An r by 1 vector.
- PredictOutput.covMatrix: The covariance matrix of PredictOutput.value. An r by r matrix.
- PredictOutput.SE: The standard error of elements in PredictOutput.value. An r by 1 vector.

Description

This function evaluates the scaled envelope model at new value Xnew. It can perform estimation: find the fitted value when X = Xnew, or prediction: predict Y when X = Xnew. The covariance matrix and the standard errors are also provided.

Example

```
load('sales.txt')
Y = sales(:, 4 : 7);
X = sales(:, 1 : 3);
u = bic_senv(X, Y);
ModelOutput = senv(X, Y, u);
Xnew = X(1, :)';
PredictOutput = predict_senv(ModelOutput, Xnew, 'estimation')
```

```
       [PredictOutput.value, Y(1, :)']  % Compare the fitted value with the data
       PredictOutput.SE
```

```
PredictOutput =

        value: [4x1 double]
    covMatrix: [4x4 double]
           SE: [4x1 double]


ans =

    8.9109    9.0000
   11.5096   12.0000
    9.6063    9.0000
   19.5119   20.0000


ans =

    7.7627
    5.8952
    4.2205
    8.0134
```

```
       PredictOutput = predict_senv(ModelOutput, Xnew, 'prediction')
       PredictOutput.SE
```

```
PredictOutput =

        value: [4x1 double]
    covMatrix: [4x4 double]
           SE: [4x1 double]


ans =

    8.3024
    6.3052
    4.4254
    8.5956
```

8.8 senv

Fit the scaled envelope model.

Contents

Syntax

```
ModelOutput = senv(X, Y, u)
ModelOutput = senv(X, Y, u, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be strictly greater than p.

u: Dimension of the envelope. An integer between 0 and r.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out number of iterations, logical 0 or 1. Default value: 0.
- Opts.init: The initial value for the envelope subspace. An r by u matrix. Default value is the one generated by function get_Init.

Output

ModelOutput: A list that contains the maximum likelihood estimators and some statistics.

- ModelOutput.beta: The scaled envelope estimator of the regression coefficients β. An r by p matrix.
- ModelOutput.Sigma: The scaled envelope estimator of the error covariance matrix. An r by r matrix.
- ModelOutput.Lambda: The matrix of estimated scales. An r by r diagonal matrix with the first diagonal element equal to 1 and other diagonal elements being positive.
- ModelOutput.Gamma: The orthogonal basis of the envelope subspace. An r by u semi-orthogonal matrix.
- ModelOutput.Gamma0: The orthogonal basis of the complement of the envelope subspace. An r by r - u semi-orthogonal matrix.

- ModelOutput.eta: The coordinates of β with respect to Gamma. An u by p matrix.
- ModelOutput.Omega: The coordinates of Sigma with respect to Gamma. An u by u matrix.
- ModelOutput.Omega0: The coordinates of Sigma with respect to Gamma0. An r - u by r - u matrix.
- ModelOutput.alpha: The estimated intercept in the scaled envelope model. An r by 1 vector.
- ModelOutput.l: The maximized log likelihood function. A real number.
- ModelOutput.covMatrix: The asymptotic covariance of vec(β). An rp by rp matrix. The covariance matrix returned are asymptotic. For the actual standard errors, multiply by 1 / n.
- ModelOutput.asySenv: Asymptotic standard error for elements in β under the scaled envelope model. An r by p matrix. The standard errors returned are asymptotic, for actual standard errors, multiply by 1 / sqrt(n).
- ModelOutput.ratio: The asymptotic standard error ratio of the standard multivariate linear regression estimator over the scaled envelope estimator, for each element in β. An r by p matrix.
- ModelOutput.np: The number of parameters in the scaled envelope model. A positive integer.
- ModelOutput.n: The number of observations in the data. A positive integer.

Description

This function fits the scaled envelope model to the responses and predictors, using the maximum likelihood estimation. When the dimension of the envelope is between 1 and r - 1, we implemented the algorithm in Cook and Su (2012). When the dimension is r, then the scaled envelope model degenerates to the standard multivariate linear regression. When the dimension is 0, it means that X and Y are uncorrelated, and the fitting is different.

References

1. The codes are implemented based on the algorithm in Section 4.1 of Cook and Su (2012).
2. The Grassmann manifold optimization step calls the package sg_min 2.4.3 by Ross Lippert (http://web.mit.edu/~ripper/www/sgmin.html).

Example

The following codes produce the results of the test and performance example in Cook and Su (2012).

```
load('sales.txt')
Y = sales(:, 4 : 7);
X = sales(:, 1 : 3);
u = bic_senv(X, Y)
```

```
u =

     2
```

```
ModelOutput = senv(X, Y, u)
ModelOutput.Lambda
1 - 1 ./ ModelOutput.ratio
```

```
ModelOutput =

         beta: [4x3 double]
        Sigma: [4x4 double]
       Lambda: [4x4 double]
        Gamma: [4x2 double]
       Gamma0: [4x2 double]
          eta: [2x3 double]
        Omega: [2x2 double]
       Omega0: [2x2 double]
        alpha: [4x1 double]
           np: 23
            l: -386.1900
    covMatrix: [12x12 double]
      asySenv: [4x3 double]
        ratio: [4x3 double]
            n: 50


ans =

    1.0000         0         0         0
         0    0.9729         0         0
         0         0    0.8067         0
         0         0         0    1.7016


ans =

    0.7020    0.5218    0.6517
    0.6137    0.3219    0.5614
    0.4439    0.1858    0.3924
    0.5375    0.4639    0.5228
```

8.9 testcoefficient_senv

`This function tests the null hypothesis L * beta * R = A versus the alternative hypothesis L * beta * R ~= A, where beta is estimated under the scaled envelope model.`

Contents

Syntax

```
TestOutput = testcoefficient_senv(ModelOutput)
TestOutput = testcoefficient_senv(ModelOutput, TestInput)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from senv.

TestInput: A list that specifies the null hypothesis, including L, R, and A. If not provided by the user, default values will be used.

- TestInput.L: The matrix multiplied to β on the left. It is a d1 by r matrix, while d1 is less than or equal to r. Default value: identity matrix I_r.
- TestInput.R: The matrix multiplied to β on the right. It is a p by d2 matrix, while d2 is less than or equal to p. Default value: identity matrix I_p.
- TestInput.A: The matrix on the right hand side of the equation. It is a d1 by d2 matrix. Default value: d1 by d2 zero matrix.

Output

TestOutput: A list containing test statistics, degrees of freedom for the reference chi-squared distribution, the p-value, and the covariance matrix of vec($L\beta R$). At the same time, a table is printed out.

- TestOutput.chisqStatistic: The test statistics. A real number.
- TestOutput.df: The degrees of freedom of the reference chi-squared distribution. A positive integer.
- TestOutput.pValue: p-value of the test. A real number in [0, 1].
- TestOutput.covMatrix: The covariance matrix of vec($L\beta R$). A d1 * d2 by d1 * d2 matrix.

Description

This function tests for hypothesis $H_0 : L\beta R = A$, versus $H_\alpha : L\beta R \neq A$. The β is estimated by the scaled envelope model. If the user does not specify the values for L, R and A, then the test is equivalent to the standard F test on if $\beta = 0$. The test statistics used is vec $(L\beta R - A)\ \hat{\Sigma}^{-1}$ vec $(L\beta R - A)^T$, and the reference distribution is chi-squared distribution with degrees of freedom d1 * d2.

Example

```
load('sales.txt')
Y = sales(:,4:7);
X = sales(:,1:3);
u = bic_senv(X,Y)
ModelOutput = senv(X,Y,u);
TestOutout = testcoefficient_senv(ModelOutput);
```

```
 Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A        4827.816        12      0.0000
-------------------------------------------------------------------------
```

```
r = size(Y, 2);
p = size(X, 2);
TestInput.L = rand(2, r);
TestInput.R = rand(p, 1);
TestInput.A = zeros(2, 1);
TestOutout = testcoefficient_senv(ModelOutput, TestInput);
```

```
Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A        1025.948        2       0.0000
-------------------------------------------------------------------------
```

9

xenv

9.1 aic_xenv

Use Akaike information criterion to select the dimension of the envelope subspace for the reduction on X.

Contents

- Syntax
- Input
- Output
- Description
- Example

Syntax

```
u = aic_xenv(X, Y)
u = aic_xenv(X, Y, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is number of observations. The number of predictors should be greater than the number of the responses. And they must be continuous variables.

Y: Responses. An n by r matrix, r is the number of responses. The response can be univariate or multivariate and must be continuous variable.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and p.

Description

This function implements the Akaike information criteria (AIC) to select the dimension of the envelope subspace for the reduction on X.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
Y = wheatprotein(:, 7);
u = aic_xenv(X, Y)
```

```
u =

     4
```

9.2 bic_xenv

Use Bayesian information criterion to select the dimension of the envelope subspace for the reduction on X.

Contents

Syntax

```
u = bic_xenv(X, Y)
u = bic_xenv(X, Y, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is number of observations. The number of predictors should be greater than the number of the responses. And they must be continuous variables.

Y: Responses. An n by r matrix, r is the number of responses. The response can be univariate or multivariate and must be continuous variable.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and p.

Description

This function implements the Bayesian information criteria (BIC) to select the dimension of the envelope subspace for the reduction on X.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
Y = wheatprotein(:, 7);
u = bic_xenv(X, Y)
```

```
u =

     4
```

9.3 bstrp_xenv

Compute bootstrap standard error of the envelope model for the reduction on X.

Contents

Syntax

```
bootse = bstrp_xenv(X, Y, u, B)
bootse = bstrp_xenv(X, Y, u, B, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is number of observations. The number of predictors should be greater than the number of the responses. And they must be continuous variables.

Y: Responses. An n by r matrix, r is the number of responses. The response can be univariate or multivariate and must be continuous variable.

u: Dimension of the envelope subspace. A positive integer between 0 and p.

B: Number of bootstrap samples. A positive integer.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

bootse: The standard error for elements in β computed by bootstrap. An p by r matrix.

Description

This function computes the bootstrap standard errors for the regression coefficients in the envelope model by bootstrapping the residuals. The envelope model here is for the reduction on X.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
Y = wheatprotein(:, 7);
alpha = 0.01;
u = lrt_xenv(X, Y, alpha)
```

```
u =

     4
```

```
B = 100;
bootse = bstrp_xenv(X, Y, u, B)
```

```
bootse =

    0.0222
    0.0387
    0.0413
    0.0167
    0.0022
    0.0087
```

9.4 dF4xenv

The first derivative of the objective function for computing the envelope subspace for the reduction on X.

Contents

Syntax

```
df = dF4xenv(R, DataParameter)
```

Input

R: A p by u semi orthogonal matrix, 0 < u <= p.

DataParameter: A structure that contains the statistics calculated from the data.

Output

df: An p by u matrix containing the value of the derivative function evaluated at R.

Description

The objective function is derived in Section 4.5.1 of Cook et al. (2012) by using maximum likelihood estimation. This function is the derivative of the objective function.

9.5 F4xenv

Objective function for computing the envelope subspace for the reduction on X.

Contents

Syntax

```
f = F4xenv(R, DataParameter)
```

Input

R: A p by u semi orthogonal matrix, 0 < u <= p.

DataParameter: A structure that contains the statistics calculated from the data.

Output

f: A scalar containing the value of the objective function evaluated at R.

Description

The objective function is derived in Section 4.5.1 of Cook et al. (2012) using maximum likelihood estimation. The columns of the semi-orthogonal matrix that minimizes this function span the estimated envelope subspace.

9.6 lrt_xenv

Use likelihood ratio testing to select the dimension of the envelope subspace for the reduction on X.

Contents

Syntax

```
u = lrt_xenv(X, Y, alpha)
u = lrt_xenv(X, Y, alpha, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is number of observations. The number of predictors should be greater than the number of the responses. And they must be continuous variables.

Y: Responses. An n by r matrix, r is the number of responses. The response can be univariate or multivariate and must be continuous variable.

alpha: Significance level for testing. A real number between 0 and 1, often taken at 0.05 or 0.01.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out dimension selection process, logical 0 or 1. Default value: 0.

Output

u: Dimension of the envelope. An integer between 0 and p.

Description

This function implements the likelihood ratio testing procedure to select the dimension of the envelope subspace for the reduction on X, with pre-specified significance level α.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
Y = wheatprotein(:, 7);
alpha = 0.01;
u = lrt_xenv(X, Y, alpha)
```

```
u =

     4
```

9.7 predict_xenv

Perform estimation or prediction under the envelope model for the reduction on X.

Contents

Syntax

```
PredictOutput = predict_xenv(ModelOutput, Xnew, infType)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from xenv.

Xnew: The value of X with which to estimate or predict Y. A p by 1 vector.

infType: A string of characters indicting the inference type, the choices can be 'estimation' or 'prediction'.

Output

PredictOutput: A list containing the results of the inference.

- PredictOutput.value: The fitted value or the prediction value evaluated at Xnew. An r by 1 vector.
- PredictOutput.covMatrix: The covariance matrix of PredictOutput.value. An r by r matrix.
- PredictOutput.SE: The standard error of elements in PredictOutput.value. An r by 1 vector.

Description

This function evaluates the envelope model for the reduction on X at new value Xnew. It can perform estimation: find the fitted value when X = Xnew, or prediction: predict Y when X = Xnew. The covariance matrix and the standard errors are also provided.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
Y = wheatprotein(:, 7);
u = bic_xenv(X, Y);
ModelOutput = xenv(X, Y, u);
Xnew = X(1, :)';
PredictOutput = predict_xenv(ModelOutput, Xnew, 'estimation')
```

```
[PredictOutput.value, Y(1, :)']  % Compare the fitted value with the data
PredictOutput.SE
```

```
PredictOutput =

        value: 9.1751
    covMatrix: 16.8439
           SE: 4.1041


ans =

    9.1751    9.2300


ans =

    4.1041
```

```
PredictOutput = predict_xenv(ModelOutput, Xnew, 'prediction')
PredictOutput.SE
```

```
PredictOutput =

        value: 9.1751
    covMatrix: 16.8760
           SE: 4.1080


ans =

    4.1080
```

9.8 testcoefficient_xenv

```
This function tests the null hypothesis L * beta * R = A versus the
alternative hypothesis L * beta * R ~= A, where beta is estimated under
the envelope model for the reduction on X.
```

Contents

Syntax

```
TestOutput = testcoefficient_xenv(ModelOutput)
TestOutput = testcoefficient_xenv(ModelOutput, TestInput)
```

Input

ModelOutput: A list containing the maximum likelihood estimators and other statistics inherited from xenv.

TestInput: A list that specifies the null hypothesis, including L, R, and A. If not provided by the user, default values will be used.

- TestInput.L: The matrix multiplied to β on the left. It is a d1 by p matrix, while d1 is less than or equal to p. Default value: identity matrix I_p.
- TestInput.R: The matrix multiplied to β on the right. It is a r by d2 matrix, while d2 is less than or equal to r. Default value: identity matrix I_r.
- TestInput.A: The matrix on the right hand side of the equation. It is a d1 by d2 matrix. Default value: d1 by d2 zero matrix.

Output

TestOutput: A list containing test statistics, degrees of freedom for the reference chi-squared distribution, the p-value, and the covariance matrix of vec($L\beta R$). At the same time, a table is printed out.

- TestOutput.chisqStatistic: The test statistics. A real number.
- TestOutput.df: The degrees of freedom of the reference chi-squared distribution. A positive integer.
- TestOutput.pValue: p-value of the test. A real number in [0, 1].
- TestOutput.covMatrix: The covariance matrix of vec($L\beta R$). A d1 * d2 by d1 * d2 matrix.

Description

This function tests for hypothesis $H_0 : L\beta R = A$, versus $H_\alpha : L\beta R \neq A$. The β is estimated by the envelope model for the reduction on X. If the user does not specify the values for L, R and A, then the test is equivalent to the standard F test on if $\beta = 0$. The test statistics used is vec $(L\beta R - A)\ \hat{\Sigma}^{-1}$ vec $(L\beta R - A)^T$, and the reference distribution is chi-squared distribution with degrees of freedom d1 * d2.

Example

```
load wheatprotein.txt
 X=wheatprotein(:, 1 : 6);
 Y=wheatprotein(:, 7);
 u = bic_xenv(X, Y);
 ModelOutput=xenv(X, Y, u);
 TestOutout = testcoefficient_xenv(ModelOutput);
```

```
 Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A         3233.053          6         0.0000
-------------------------------------------------------------------------
```

```
 r = size(Y, 2);
 p = size(X, 2);
 TestInput.L = rand(2, p);
 TestInput.R = rand(r, 1);
 TestInput.A = zeros(2, 1);
 TestOutout = testcoefficient_xenv(ModelOutput, TestInput);
```

```
 Test Hypothesis     Chisq Statistic    DF     P-value
-------------------------------------------------------------------------
L * beta * R = A           33.578          2         0.0000
-------------------------------------------------------------------------
```

9.9 xenv

Fit the envelope model for the reduction on X.

Contents

Syntax

```
ModelOutput = xenv(X, Y, u)
ModelOutput = xenv(X, Y, u, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is number of observations. The number of predictors should be greater than the number of the responses. And they must be continuous variables.

Y: Responses. An n by r matrix, r is the number of responses. The response can be univariate or multivariate and must be continuous variable.

u: Dimension of the envelope. An integer between 0 and p.

Opts: A list containing the optional input parameters, to control the iterations in sg_min. If one or several (even all) fields are not defined, the default settings are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out Grassmann manifold optimization process, logical 0 or 1. Default value: 0.
- Opts.init: The initial value for the envelope subspace. An p by u matrix. Default value is the one generated by function get_Init.

Output

ModelOutput: A list that contains the maximum likelihood estimators and some statistics.

- ModelOutput.beta: The envelope estimator of the regression coefficients β. An p by r matrix.
- ModelOutput.SigX: The envelope estimator of the covariance matrix of X, Σ_X. A p by p matrix.
- ModelOutput.Gamma: The orthogonal basis of the envelope subspace. An p by u semi-orthogonal matrix.
- ModelOutput.Gamma0: The orthogonal basis of the complement of the envelope subspace. An p by p-u semi-orthogonal matrix.
- ModelOutput.eta: The coordinates of β with respect to Gamma. An u by r matrix.

- ModelOutput.Omega: The coordinates of Σ_X with respect to Gamma. An u by u matrix.
- ModelOutput.Omega0: The coordinates of Σ_X with respect to Gamma0. An p - u by p - u matrix.
- ModelOutput.mu: The estimated intercept. An r by 1 vector.
- ModelOutput.sigYcX: The estimated conditional covariance matrix of Y given X. An r by r matrix.
- ModelOutput.l: The maximized log likelihood function. A real number.
- ModelOutput.covMatrix: The asymptotic covariance of vec(β). An pr by pr matrix. The covariance matrix returned are asymptotic. For the actual standard errors, multiply by 1 / n.
- ModelOutput.asyXenv: Asymptotic standard error for elements in β under the envelope model. An r by p matrix. The standard errors returned are asymptotic, for actual standard errors, multiply by 1 / sqrt(n).
- ModelOutput.ratio: The asymptotic standard error ratio of the standard multivariate linear regression estimator over the envelope estimator, for each element in β. An p by r matrix.
- ModelOutput.np: The number of parameters in the envelope model. A positive integer.
- ModelOutput.n: The number of observations in the data. A positive integer.

Description

This function fits the envelope model in the predictor's space, using the maximum likelihood estimation. When the dimension of the envelope is between 1 and r - 1, we implemented the algorithm in Cook et al. (2012). When the dimension is r, then the envelope model degenerates to the standard multivariate linear regression. When the dimension is 0, it means that X and Y are uncorrelated, and the fitting is different.

References

1. The codes are implemented based on the algorithm in Section 4.5.1 of Cook et al (2012).
2. The Grassmann manifold optimization step calls the package sg_min 2.4.3 by Ross Lippert (http://web.mit.edu/~ripper/www/sgmin.html).

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
Y = wheatprotein(:, 7);

p = size(X, 2);
ModelOutput = xenv(X, Y, p);

% When u = p, the envelope model reduces to the ordinary least squares
% regression

temp = fit_OLS(X, Y);
temp.SigmaOLS
ModelOutput.sigYcX
```

```
ans =

    0.0321
```

```
ans =

    0.0321
```

```
temp.betaOLS'
ModelOutput.beta
```

```
ans =

   -0.0416
   -0.0490
    0.3368
   -0.1981
    0.0020
   -0.0480
```

```
ans =

   -0.0416
   -0.0490
    0.3368
   -0.1981
    0.0020
   -0.0480
```

```
u = bic_xenv(X, Y);
ModelOutput = xenv(X, Y, u)
```

```
ModelOutput =

         beta: [6x1 double]
         SigX: [6x6 double]
        Gamma: [6x4 double]
       Gamma0: [6x2 double]
          eta: [4x1 double]
        Omega: [4x4 double]
       Omega0: [2x2 double]
           mu: 24.8863
       sigYcX: 0.0321
            l: -865.6407
    covMatrix: [6x6 double]
      asyXenv: [6x1 double]
        ratio: [6x1 double]
```

```
           np: 27
            n: 50


      %  To compare with the results obtained by Partial Least Squares, use the
      %  plsregress command
      [XL, YL, XS, YS, BETA, PCTVAR, MSE, stats] = plsregress(X, Y, u);
      ModelOutput.beta
      BETA(2 : end, :)


ans =

   -0.0443
   -0.0481
    0.3377
   -0.1963
    0.0019
   -0.0487



ans =

   -0.0199
    0.1373
    0.1309
   -0.1827
    0.0056
   -0.0708
```

10

xenvpls

10.1 bstrp_xenvpls

Compute bootstrap standard error of the envelope model for the reduction on X using partial least squares algorithm.

Contents

Syntax

```
bootse = bstrp_xenvpls(X, Y, u, B)
bootse = bstrp_xenvpls(X, Y, u, B, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is number of observations. The number of predictors should be greater than the number of the responses. And they must be continuous variables.

Y: Responses. An n by r matrix, r is the number of responses. The response can be univariate or multivariate and must be continuous variable.

u: Dimension of the envelope subspace. A positive integer between 0 and p.

B: Number of bootstrap samples. A positive integer.

Opts: A list containing the optional input parameters. If one or several (even all) fields are not defined, the default settings are used.

- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

bootse: The standard error for elements in β computed by bootstrap. A p by r matrix.

Description

This function computes the bootstrap standard errors for the regression coefficients in the envelope model by bootstrapping the residuals. The envelope model is applied for the reduction on X, using the partial least squares algorithm.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
Y = wheatprotein(:, 7);
m = 5;
u = mfoldcv_xenvpls(X, Y, m)
```

```
u =

     6
```

```
B = 100;
bootse = bstrp_xenvpls(X, Y, u, B, Opts)
```

```
bootse =

    0.0287
    0.0383
    0.0409
    0.0207
    0.0026
    0.0097
```

10.2 mfoldcv_xenvpls

Select the dimension of the envelope subspace using m-fold cross validation for envelope model on the reduction on X using partial least squares algorithm.

Contents

Syntax

```
u = mfoldcv_xenvpls(X, Y, m)
u = mfoldcv_xenvpls(X, Y, m, Opts)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is number of observations. The number of predictors should be greater than the number of the responses. And they must be continuous variables.

Y: Responses. An n by r matrix, r is the number of responses. The response can be univariate or multivariate and must be continuous variable.

m: A positive integer that is used to indicate m-fold cross validation.

Opts: A list containing the optional input parameters. If one or several (even all) fields are not defined, the default settings are used.

- Opts.verbose: Flag for print out the number of bootstrap samples, logical 0 or 1. Default value: 0.

Output

u: The dimension of the envelope subspace selected by m-fold cross validation.

Description

This function implements m-fold cross validation to select the dimension of the envelope space, based on prediction performance. For each u, the data is partitioned into m parts, each part is in turn used for testing for the prediction performance while the rest m-1 parts are used for training. The dimension is select as the one that minimizes the average prediction errors. If Y is multivariate, the identity inner product is used for computing the prediction errors.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
Y = wheatprotein(:, 7);
```

```
m = 5;
u = mfoldcv_xenvpls(X, Y, m)
```

```
u =

     6
```

10.3 xenvpls

Fit the envelope model for the reduction on X using partial least squares algorithm.

Contents

Syntax

```
ModelOutput = xenvpls(X, Y, u)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors and n is number of observations. The number of predictors should be greater than the number of the responses. And they must be continuous variables.

Y: Responses. An n by r matrix, r is the number of responses. The response can be univariate or multivariate and must be continuous variable.

u: Dimension of the envelope. An integer between 0 and p.

Output

ModelOutput: A list that contains the maximum likelihood estimators and some statistics.

- ModelOutput.beta: The envelope estimator of the regression coefficients β. An p by r matrix.
- ModelOutput.SigX: The envelope estimator of the covariance matrix of X, Σ_X. A p by p matrix.
- ModelOutput.Gamma: The orthogonal basis of the envelope subspace. An p by u semi-orthogonal matrix.
- ModelOutput.Gamma0: The orthogonal basis of the complement of the envelope subspace. An p by p-u semi-orthogonal matrix.
- ModelOutput.eta: The coordinates of β with respect to Gamma. An u by r matrix.
- ModelOutput.Omega: The coordinates of Σ_X with respect to Gamma. An u by u matrix.
- ModelOutput.Omega0: The coordinates of Σ_X with respect to Gamma0. An p - u by p - u matrix.
- ModelOutput.mu: The estimated intercept. An r by 1 vector.
- ModelOutput.sigYcX: The estimated conditional covariance matrix of Y given X. An r by r matrix.
- ModelOutput.np: The number of parameters in the envelope model. A positive integer.
- ModelOutput.n: The number of observations in the data. A positive integer.

Description

This function fits the envelope model in the predictor's space, by the partial least squares algorithm in Cook et al. (2012). In the population level, this algorithm is equivalent to that in xenv.m, which uses the maximum likelihood estimation. In the sample version, the two algorithms are different. And this algorithm is much faster, which provides a root n consistent starting value for the one in xenv.m.

Reference

The codes are implemented based on the algorithm in Section 4.3 of Cook et al (2012).

Example

```
load wheatprotein.txt
X = wheatprotein(:, 1 : 6);
Y = wheatprotein(:, 7);
m = 5;
u = mfoldcv_xenvpls(X, Y, m)
```

```
u =

     6
```

```
ModelOutput = xenvpls(X, Y, u)
```

```
ModelOutput =

      beta: [6x1 double]
      SigX: [6x6 double]
     Gamma: [6x6 double]
    Gamma0: []
       eta: [6x1 double]
     Omega: [6x6 double]
    Omega0: []
        mu: 24.5781
    sigYcX: 0.0321
        np: 29
         n: 50
```

```
ModelOutput.beta
```

```
ans =

   -0.0416
   -0.0490
    0.3368
   -0.1981
```

```
 0.0020
-0.0480
```

11

auxiliary

11.1 center

Subtract the mean of each column.

Contents

Syntax

```
XC = center(X)
```

Input

X: A matrix or a column vector.

Output

XC: A matrix or a column vector with the mean for each column equal to 0.

Description

This function centerizes a matrix or a vector, by subtracting each column by its column mean.

11.2 Contr

Compute the contraction matrix of dimension r.

Contents

Syntax

```
C = Contr(r)
```

Input

r: Dimension of the contraction matrix. A positive integer.

Output

C: Contraction matrix of dimension r. C is an r(r + 1) / 2 by r ^ 2 matrix.

Description

The contraction and expansion matrices are links between the "vec" operator and "vech"operator: for an r by r symmetric matrix A, vech(A) = Contr(r) * vec(A), and vec(A) = Expan(r) * vech(A). The "vec" operator stacks the matrix A into an r ^ 2 by 1 vector columnwise. The "vech" operator stacks the lower triangle or the upper triangle of a symmetric matrix into an r(r+1)/2 vector. For more details of "vec", "vech", contraction and expansion matrix, refer to Henderson and Searle (1979).

11.3 Expan

Compute the expansion matrix of dimension r.

Contents

Syntax

```
E = Expan(r)
```

Input

r: Dimension of the expansion matrix. A positive integer.

Output

E: Expansion matrix of dimension r. E is an r ^ 2 by r(r + 1) / 2 matrix.

Description

The contraction and expansion matrices are links between the "vec" operator and "vech" operator: for an r by r symmetric matrix A, vech(A) = Contr(r) * vec(A), and vec(A) = Expan(r) * vech(A). The "vec" operator stacks the matrix A into an r ^ 2 by 1 vector columnwise. The "vech" operator stacks the lower triangle or the upper triangle of a symmetric matrix into an r(r + 1) / 2 vector. For more details of "vec", "vech", contraction and expansion matrix, refer to Henderson and Searle (1979).

11.4 fit_OLS

Multivariate linear regression.

Contents

Syntax

```
ModelOutput = fit_OLS(X, Y)
```

Input

X: Predictors, an n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses, an n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables.

Output

ModelOutput: A list that contains the maximum likelihood estimators of regression coefficients and error covariance matrix.

- ModelOutput.betaOLS: An r by p matrix containing estimate of the regression coefficients β.
- ModelOutput.SigmaOLS: An r by r matrix containing estimate of the error covariance matrix.
- ModelOutput.alpha: An r by 1 vector containing estimate of the intercept.
- ModelOutput.n: The number of observations in the data. A positive integer.

Description

In a multivariate linear model, Y and X follows the following relationship: $Y = \alpha + \beta X + \varepsilon$, where ε contains the errors. This function performs the ordinary least squares fit to the inputs, and returns the estimates of β and the covariance matrix of ε.

Example

```
load wheatprotein.txt
X = wheatprotein(:, 8);
Y = wheatprotein(:, 1 : 6);
ModelOutput = fit_OLS(X, Y)
ModelOutput.betaOLS
ModelOutput.SigmaOLS
```

```
ModelOutput =

     betaOLS: [6x1 double]
    SigmaOLS: [6x6 double]
       alpha: [6x1 double]
           n: 50


ans =

    3.2724
    8.0288
    7.5224
   -2.0609
    3.2244
    0.6538


ans =

   1.0e+03 *

    1.1905    0.9759    1.0506    1.1524    1.5384    0.6335
    0.9759    0.8061    0.8657    0.9432    1.2636    0.5266
    1.0506    0.8657    0.9310    1.0164    1.3664    0.5640
    1.1524    0.9432    1.0164    1.1228    1.5234    0.6183
    1.5384    1.2636    1.3664    1.5234    2.3229    0.8360
    0.6335    0.5266    0.5640    0.6183    0.8360    0.3618
```

11.5 get_envelope

Construct the envelope subspace using a sequential algorithm.

Contents

Syntax

```
W = get_envelope(S, M, u)
```

Input

S: An r by p matrix whose columns span the subspace, the rank of S cannot be greater than u.

M: An r by r positive semi-definite matrix.

u: Dimension of the envelope. An integer between 0 and r.

Output

W: An r by u semi-orthogonal matrix that spans the M-envelope of span(S).

Description

This function constructs the M-envelope of span(S) using a sequential algorithm similar to partial least squares.

Reference

The codes are implemented based on the algorithm in the lecture notes of Cook (2012).

Example

```
S = [1 2 3]';
S0 = grams(nulbasis(S'));
M = S * S' + S0 * S0';
u = 1;
W = get_envelope(S, M, u)
```

```
W =

    0.2673
    0.5345
    0.8018
```

11.6 get_Init

Starting value for the envelope subspace.

Contents

Syntax

```
WInit = get_Init(F, u, DataParameter)
```

Input

F: Objective function of the envelope subspace.

u: Dimension of the envelope. An integer between 1 and r - 1.

DataParameter: A list containing commonly used statistics computed from the data.

Output

WInit: The initial estimate of the orthogonal basis of the envelope subspace. An r by u orthogonal matrix.

Description

We compute the eigenvectors for the covariance matrices of Y and the estimated errors, and get 2r vectors. Then we get all the combinations of u vectors out of the 2r vectors. If the number of 2r choose u is small(<= 50), we search over all the combinations and find out the one that minimizes the objective function F. If that number is large, then we do it iteratively: we pick up any u eigenvectors, fix all of them except the first one. Then we search over all the vectors orthogonal to the fixed ones, and record the one that minimizes F. Next, we fix the first u eigenvectors again but this time search for the second one, then we record the vector. This goes on and on until the last one. We do it for 5 rounds and use the final set as our starting value.

Reference

The codes are implemented based on the algorithm in Section 3.5 of Su and Cook (2011).

11.7 get_Init4envmean

Starting value for the envelope subspace in estimating the multiavriate mean.

Contents

Syntax

```
WInit = get_Init4envmean(F, u, DataParameter)
```

Input

F: Objective function to get the envelope subspace.

u: Dimension of the envelope. An integer between 1 and p - 1.

DataParameter: A list containing commonly used statistics computed from the data.

Output

WInit: The initial estimate of the orthogonal basis of the envelope subspace. A p by u orthogonal matrix.

Description

We compute the eigenvectors for the estimated error covariance matrix, and get p vectors. Then we get all the combinations of u vectors out of the p vectors. If the number of p choose u is small(<=50), we search over all the combinations and find out the one that minimizes the objective function F. If that number is large, then we do it iteratively: we pick up any u eigenvectors, fix all of them except the first one. Then we search over all the vectors orthogonal to the fixed ones, and record the one that minimizes F. Next, we fix the first u eigenvectors again but this time search for the second one, then we record the vector. This goes on and on until the last one. We do it for 3 rounds and use the final set as our starting value.

Reference

The codes are implemented based on the algorithm in Section 3.5 of Su and Cook (2011).

11.8 get_Init4henv

Starting value for the heteroscedastic envelope subspace.

Contents

Syntax

```
WInit = get_Init4henv(F, u, DataParameter)
```

Input

F: Objective function to get the heteroscedastic envelope subspace.

u: Dimension of the envelope. An integer between 1 and r - 1.

DataParameter: A list containing commonly used statistics computed from the data.

Output

WInit: The initial estimate of the orthogonal basis of the heteroscedastic envelope subspace. An r by u orthogonal matrix.

Description

We compute the eigenvectors for the estimated error covariance matrix, and get r vectors. Then we get all the combinations of u vectors out of the r vectors. If the number of r choose u is small(<=50), we search over all the combinations and find out the one that minimizes the objective function F. If that number is large, then we do it iteratively: we pick up any u eigenvectors, fix all of them except the first one. Then we search over all the vectors orthogonal to the fixed ones, and record the one that minimizes F. Next, we fix the first u eigenvectors again but this time search for the second one, then we record the vector. This goes on and on until the last one. We do it for 3 rounds and use the final set as our starting value.

Reference

The codes are implemented based on the algorithm in Section 3.5 of Su and Cook (2011).

11.9 Kpd

Compute the communication matrix Kpd.

Contents

Syntax

```
k = Kpd(p, d)
```

Input

p, **d**: two positive integers represent the dimension parameters for the communication matrix.

Output

k: The communication matrix Kpd. An p * d by p * d matrix.

Description

For a p by d matrix A, vec(A') = Kpd * vec(A), and Kpd is called a communication matrix.

Reference

The codes are implemented based on Definition 3.1 in Magnus and Neudecker (1979).

11.10 Lmatrix

Extract the 2nd to the last diagonal element of a matrix into a vector.

Contents

Syntax

```
L = Lmatrix(r)
```

Input

r: The dimension of the matrix being extracted. The matrix should be an r by r matrix.

Output

L: An r - 1 dimensional vector that contains all the diagonal elements but the first one of the matrix.

Description

Let A be an r by r matrix, and vec be the vector operator, then Lmatrix(r) * vec(A) will give the 2nd to the rth diagonal elements of A, arranged in a column vector.

11.11 make_dF

Generic function to generate the derivative function of the objective function F.

Contents

Syntax

dF = make_dF(dfun_method_handle, FParameters)

Input

- dfun_method_handle: A specific model derivative function of the objective function.
- FParameters: A structure that contains data parameters as input for the function dfun_method_handle.

Output

- dF: The generic derivative function of the objective function for computing the envelope subspace.

Description

Generic function to generate the derivative function of the objective function F. The function first sets a handle to the specific model function and fixes the data parameters from the sample needed for its computation. The handle fixed with those parameters is then evaluated at a given value for argument W. A generic derivative function dF is returned.

11.12 make_F

Generic function to generate the objective function F.

Contents

Syntax

F = make_F(fun_method_handle, FParameters)

Input

- fun_method_handle: A specific model objective function.
- FParameters: A structure that contains data parameters as input for the function fun_method_handle.

Output

- F: The generic objective function for computing the envelope subspace.

Description

Generic function to generate the objective function F. The function first sets a handle to the specific model function and fixes the data parameters from the sample needed for its computation. The handle fixed with those parameters is then evaluated at a given value for argument W. A generic objective function F is returned.

11.13 make_opts

Make optional input parameters for running the sg_min package.

Contents

Syntax

```
Opts = make_opts(Opts)
```

Input

Opts: A list containing optional input parameters for sg_min.m specified by users. One or several (even all) fields could be empty.

- Opts.maxIter: Maximum number of iterations.
- Opts.ftol: Tolerance parameter for F.
- Opts.gradtol: Tolerance parameter for dF.
- Opts.verbose: Flag for print out output, logical 0 or 1.

Output:

Opts: A list containing optional input parameters for sg_min.m, specified by users or the default values are used.

- Opts.maxIter: Maximum number of iterations. Default value: 300.
- Opts.ftol: Tolerance parameter for F. Default value: 1e-10.
- Opts.gradtol: Tolerance parameter for dF. Default value: 1e-7.
- Opts.verbose: Flag for print out output, logical 0 or 1. Default value: 0.

Description

The sg_min function has some optional input parameters that control the iteration process. These parameters include maximum number of iteration, tolerance parameters for convergence of the objective function F and the derivative of the objective function dF, and the print out of the iteration process. The user can set one or all of parameters, if not, default values will be used.

11.14 make_parameter

Compute summary statistics from the data.

Contents

Syntax

```
DataParameter = make_parameter(X, Y, method)
```

Input

X: Predictors. An n by p matrix, p is the number of predictors. The predictors can be univariate or multivariate, discrete or continuous.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be strictly greater than p.

method: A string of characters indicating which member of the envelope family to be used, the choices can be 'env', 'ienv', 'henv', 'senv' or 'xenv'.

Output

DataParameter: A list that contains summary statistics computed from the data. The output list can vary from method to method.

- DataParameter.n: The number of observations in the data. A positive integer.
- DataParameter.ng: A p by 1 vector containing the number of observations in each group. p is the number of groups. Only for 'henv'.
- DataParameter.ncum: A p by 1 vector containing the total number of observations till this group. Only for 'henv'.
- DataParameter.ind: An n by 1 vector indicating the sequence of the observations after sorted by groups.
- DataParameter.p: The number of predictors or number of groups for 'henv'. A positive integer.
- DataParameter.r: The number of responses. A positive integer.
- DataParameter.XC: Centered predictors. An n by p matrix with the ith row being the ith observation of X subtracted by the mean of X. Only for 'env' and 'ienv'.
- DataParameter.YC: Centered responses. An n by r matrix with the ith row being the ith observation of Y subtracted by the mean of Y. Only for 'env' and 'ienv'.
- DataParameter.mX: The mean of predictors. A p by 1 vector. For all method except 'henv'.
- DataParameter.mY: The mean of responses. An r by 1 vector.
- DataParameter.mYg: An r by p matrix with the ith column being the sample mean of the ith group.
- DataParameter.sigX: The sample covariance matrix of X. A p by p matrix.
- DataParameter.sigY: The sample covariance matrix of Y. An r by r matrix.

- DataParameter.sigRes: For 'env', 'senv', 'ienv': The sample covariance matrix of the residuals from the ordinary least squares regression of Y on X. An r by r matrix. For 'henv', an r by r by p three dimensional matrix with the ith depth is the ith sample covariance matrix for the ith group.
- DataParameter.sigFit: The sample covariance matrix of the fitted value from the ordinary least squares regression of Y on X. An r by r matrix. Only for method 'ienv'.
- DataParameter.betaOLS: The regression coefficients from the ordinary least squares regression of Y on X. An r by p matrix. For all method except 'henv'.

Description

This function computes statistics that will be used frequently in the estimation for each method.

11.15 mtest

Perform Box's M test to check the homogeneity of the covariance matrices.

Contents

Syntax

```
TestOutput = mtest(X, Y, alpha)
```

Input

X: Group indicators. A matrix with n rows. X can only have p unique rows, where p is the number of groups. For example, if there are two groups, X can only have 2 different kinds of rows, such as (0, 1) and (1, 0), or (1, 0, 10) and (0, 5, 6). The number of columns is not restricted, as long as X only has p unique rows.

Y: Multivariate responses. An n by r matrix, r is the number of responses and n is number of observations. The responses must be continuous variables, and r should be greater than p.

alpha: Significance level for testing. A real number between 0 and 1, often taken at 0.05 or 0.01.

Output

TestOutput: A list containing the Box's M statistic, the approximation test statistic, degrees of freedom for the approximation statistic test, and the p-value. At the same time, a table is printed out.

- TestOutput.mStatistic: The Box's M statistic. A real number.
- TestOutput.approxStatistic: The approximation test statistic.
- TestOutput.df: The degrees of freedom of the approximation statistic test. A positive integer.
- TestOutput.pValue: p-value of the test. A real number in [0, 1].

Description

This function performs the Box's M test for homegeneity of the covariance matrices for different groups, indicated by X. If the groups sample-size is at least 20 (sufficiently large), Box's M test takes a Chi-square approximation; otherwise it takes an F approximation.

References

The codes are implemented based on
Trujillo-Ortiz, A., R. Hernandez-Walls, K. Castro-Morales, A. Espinoza-Tenorio, A. Guia-Ramirez and R. Carmona-Pina. (2002). MBoxtest: Multivariate Statistical Testing for the Homogeneity of Covariance Matrices by the Box's M. A MATLAB file. [WWW document]. URL: http://www.mathworks.com/matlabcentral/fileexchange /loadFile.do ?objectId=2733&objectType=FILE

Example

```
load waterstrider.mat
alpha = 0.01;
TestOutput = mtest(X, Y, alpha);
```

```
-----------------------------------------------------------
     MBox     Chi-sqr.          df          P
-----------------------------------------------------------
  157.5977   137.3361          72       0.0000
-----------------------------------------------------------
Covariance matrices are significantly different.
```