

# Package ‘robsel’

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**Type** Package

**Title** Robust Selection Algorithm

**Version** 0.1.0

**Date** 2021-05-23

**Description** An implementation of algorithms for estimation of the graphical lasso regularization parameter described in Pedro Cisneros-Velarde, Alexander Petersen and Sang-Yun Oh (2020) <<http://proceedings.mlr.press/v108/cisneros20a.html>>.

**BugReports** <https://github.com/dddlab/robust-selection/issues>

**License** GPL-2

**Encoding** UTF-8

**Imports** glasso, Rcpp

**Suggests** knitr, rmarkdown

**VignetteBuilder** knitr

**RoxygenNote** 7.1.1

**LinkingTo** Rcpp, RcppEigen

**NeedsCompilation** yes

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**Repository** CRAN

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**robsel***Robust Selection***Description**

Robust Selection algorithm for estimation of the regularization parameter for Graphical Lasso

**Usage**

```
robsel(x, alpha = 0.9, B = 200)
```

**Arguments**

<code>x</code>	A n-by-p data matrix
<code>alpha</code>	Prespecified confidence level. Default 0.9
<code>B</code>	Number of bootstrap sample. Default 200

**Value**

`lambda` Estimation of the regularization parameter for Graphical Lasso. A vector of `lambda` will be return if more than 1 value of `alpha` is provided.

**References**

P Cisneros-Velarde, A Petersen and S-Y Oh (2020). Distributionally Robust Formulation and Model Selection for the Graphical Lasso. Proceedings of the Twenty Third International Conference on Artificial Intelligence and Statistics.

**See Also**

[robsel.glasso](#) for using Graphical Lasso algorithm with estimate regularization parameter `lambda` from Robust Selection.

**Examples**

```
set.seed(17)
library(robsel)
x <- matrix(rnorm(50*20),ncol=20)

#Compute estimation of lambda at confidence level alpha
lambda <- robsel(x = x, alpha = 0.9, B = 200)
```

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**robsel.glasso**      *Fit Graphical Lasso with RobSel*

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**Description**

Fit Graphical Lasso with estimate regularization parameter from Robust Selection

**Usage**

```
robsel.glasso(x, alpha = 0.9, B = 200, ...)
```

**Arguments**

x	A n-by-p data matrix
alpha	Prespecified confidence level. Default 0.9
B	Number of bootstrap sample. Default 200
...	Optional arguments passed on to glasso.

**Value**

A list with components:

alpha	A list of prespecified confidence level
lambda	A list of estimate regularization parameter for Graphical Lasso
Omega	A list of estimated inverse covariance matrix
Sigma	A list of estimated covariance matrix

**Note**

Each item in each component corresponds to a prespecified level alpha.

**References**

- P Cisneros-Velarde, A Petersen and S-Y Oh (2020). Distributionally Robust Formulation and Model Selection for the Graphical Lasso. Proceedings of the Twenty Third International Conference on Artificial Intelligence and Statistics.
- Friedman, Jerome, Trevor Hastie, and Robert Tibshirani. 'Sparse inverse covariance estimation with the graphical lasso.' *Biostatistics* 9.3 (2008): 432-441.
- Meinshausen, Nicolai and Buhlmann, Peter. 2006. 'High-Dimensional Graphs and Variable Selection with the Lasso.' *The Annals of Statistics*. JSTOR: 1436-1462.
- Witten, Daniela M, Friedman, Jerome H, and Simon, Noah. 2011. 'New Insights and Faster computations for the Graphical Lasso.' *Journal of Computation and Graphical Statistics*. Taylor and Francis: 892-900.

**See Also**

[robsel](#) for Robust Selection algorithm, [glasso](#) for Graphical Lasso algorithm.

**Examples**

```
set.seed(17)
library(robsel)
x <- matrix(rnorm(50*20),ncol=20)

#Use Graphical Lasso with estimate regularization parameter lambda from RobSel
fit <- robsel.glasso(x = x, alpha = 0.9, B = 200)
```

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