

Package ‘SOIL’

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Type Package

Title Sparsity Oriented Importance Learning

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Imports stats, glmnet, ncvreg, MASS, parallel, brglm2

Description Sparsity Oriented Importance Learning (SOIL) provides a new variable importance measure for high dimensional linear regression and logistic regression from a sparse penalization perspective, by taking into account the variable selection uncertainty via the use of a sensible model weighting. The package is an implementation of Ye, C., Yang, Y., and Yang, Y. (2017+).

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URL <https://github.com/emeryyi/SOIL>

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R topics documented:

| | |
|--------------|----------|
| SOIL | 2 |
| Index | 6 |

SOIL

*Sparsity Oriented Importance Learning (SOIL)***Description**

Sparsity Oriented Importance Learning (SOIL) provides a new variable importance measure for high dimensional linear regression and logistic regression from a sparse penalization perspective, by taking into account the variable selection uncertainty via the use of a sensible model weighting. The package is an implementation of Ye, C., Yang, Y., and Yang, Y. (2017+) DOI: <doi:10.1080/01621459.2017.1377080>.

Usage

```
SOIL(x, y, n_train = ceiling(n/2), no_rep = 100,
     n_train_bound = n_train - 2, n_bound = n - 2,
     psi = 1, family = c("gaussian",
                        "binomial"), method = c("lasso", "union", "customize"),
     candidate_models, weight_type = c("BIC", "AIC",
                                       "ARM"), prior = TRUE, reduce_bias = FALSE)
```

Arguments

| | |
|---------------|---|
| x | Matrix of predictors. |
| y | Response variable. |
| n_train | Size of training set when the weight function is ARM or ARM with prior=TRUE. The default value is n_train=ceiling(n/2). |
| no_rep | Number of replications when the weight function is ARM and ARM with prior=TRUE. The default value is no_rep=100. |
| n_train_bound | When computing the weights using ARM, the candidate models with the size larger than n_train_bound will be dropped. The default value is n_train-2. |
| n_bound | When computing the weights using AIC or BIC, the candidate models with the size larger than n_train_bound will be dropped. The default value is n-2. |
| psi | A positive number to control the improvement of the prior weight. The default value is 1. |
| family | Choose the family for GLM models. So far gaussian and binomial are implemented. The default is gaussian. |
| method | Users can choose lasso, union or customize. If method=="lasso", then the program automatically provides the candidate models as a union of solution paths of Lasso, Adaptive Lasso; If method=="union", then the program automatically provides the candidate models as a union of solution paths of Lasso, Adaptive Lasso, SCAD, and MCP; If method=="customize", users must provide their own set of candidate models in the input argument candidate_models as a matrix, each row of which is a 0/1 index vector representing whether each variable is included/excluded in the model. For details see Example section. The default option is method=="lasso". |

| | |
|------------------|--|
| candidate_models | Only available when method="customize". It is a matrix of candidate models, each row of which is a 0/1 index vector representing whether each variable is included/excluded in the model. For details see Example section. |
| weight_type | Options for computing weights for SOIL measure. Users can choose among ARM, AIC and BIC. The default is BIC. |
| prior | Whether to use prior in the weighting function. The default is TRUE. |
| reduce_bias | If the binomial model is used, occasionally the algorithm might has convergence issue when the problem of so-called complete separation or quasi-complete separation happens. Users can set reduce_bias=TRUE to solve the issue. The algorithm will use an adjusted-score approach when fitting the binomial model for computing the weights. This method is developed in Firth, D. (1993). Bias reduction of maximum likelihood estimates. <i>Biometrika</i> 80, 27-38. |

Details

See the paper provided in Reference section.

Value

A "SOIL" object is returned. The components are:

| | |
|--------------------------|---|
| importance | SOIL importance values for each variable. |
| weight | The weight for each candidate model. |
| candidate_models_cleaned | Cleaned candidate models: the duplicated candidate models are cleaned; When computing SOIL weights using AIC and BIC, the models with more than $n-2$ variables are removed (n is the number of observations); When computing SOIL weights using ARM, the models with more than $n_{\text{train}}-2$ variables are removed (n_{train} is the number of training observations). |

References

Ye, C., Yang, Y., and Yang, Y. (2017+). "Sparsity Oriented Importance Learning for High-dimensional Linear Regression". *Journal of the American Statistical Association*. (Accepted) DOI: 10.1080/01621459.2017.1377080

BugReport: <https://github.com/emeryyi/SOIL>

Examples

```
# REGRESSION CASE

# generate simulation data
n <- 50
p <- 8
beta <- c(3, 1.5, 0, 0, 2, 0, 0, 0)
b0 <- 1
x <- matrix(rnorm(n*p, 0, 1), nrow=n, ncol=p)
```

```

e <- rnorm(n)
y <- x %*% beta + b0 + e

# compute SOIL using ARM with prior
v_ARM <- SOIL(x, y, family = "gaussian",
weight_type = "ARM", prior = TRUE)

# compute SOIL using BIC
v_BIC <- SOIL(x, y, family = "gaussian", weight_type = "BIC")

# compute SOIL using AIC
v_AIC <- SOIL(x, y, family = "gaussian",
weight_type = "AIC", prior = TRUE)

# user supplied candidate models
candidate_models = rbind(c(0,0,0,0,0,0,0,0,1),
c(0,1,0,0,0,0,0,0,1), c(0,1,1,1,0,0,0,0,1),
c(0,1,1,0,0,0,0,0,1), c(1,1,0,1,1,0,0,0,0),
c(1,1,0,0,1,0,0,0,0))

v1_BIC <- SOIL(x, y,
psi=1,
family = "gaussian",
method = "customize",
candidate_models = candidate_models,
weight_type = "BIC", prior = TRUE)

# CLASSIFICATION CASE

# generate simulation data
n = 300
p = 8
b <- c(1,1,1,-3*sqrt(2)/2)
x=matrix(rnorm(n*p, mean=0, sd=1), n, p)
feta=x[, 1:4]%*%b
fprob=exp(feta)/(1+exp(feta))
y=rbinom(n, 1, fprob)

# compute SOIL for model_check using BIC with prior
b_BIC <- SOIL(x, y, family = "binomial", weight_type = "BIC")

candidate_models =
rbind(c(0,0,0,0,0,0,0,0,1),
c(0,1,0,0,0,0,0,0,1),
c(1,1,1,1,0,0,0,0,0),
c(0,1,1,0,0,0,0,0,1),
c(1,1,0,1,1,0,0,0,0),
c(1,1,0,0,1,0,0,0,0),
c(0,0,0,0,0,0,0,0,0),
c(1,1,1,1,1,0,0,0,0))

```

```
# compute SOIL for model_check using AIC
# user supplied candidate models
b_AIC <- SOIL(x, y, family = "binomial",
method = "customize", candidate_models = candidate_models,
weight_type = "AIC")
```

Index

SOIL, [2](#)