

# Package: moose (via r-universe)

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**Title** Mean Squared Out-of-Sample Error Projection

**Version** 0.0.1

**Description** Projects mean squared out-of-sample error for a linear regression based upon the methodology developed in Rohlfs (2022) <[doi:10.48550/arXiv.2209.01493](https://doi.org/10.48550/arXiv.2209.01493)>. It consumes as inputs the `lm` object from an estimated OLS regression (based on the `training sample`) and a `data.frame` of out-of-sample cases (the `test sample`) that have non-missing values for the same predictors. The test sample may or may not include data on the outcome variable; if it does, that variable is not used. The aim of the exercise is to project what what mean squared out-of-sample error can be expected given the predictor values supplied in the test sample. Output consists of a list of three elements: the projected mean squared out-of-sample error, the projected out-of-sample R-squared, and a vector of out-of-sample `hat` or `leverage` values, as defined in the paper.

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**Encoding** UTF-8

**RoxygenNote** 7.2.1

**NeedsCompilation** no

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 moose

*moose: mean squared out-of-sample error projection*


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## Description

This function projects the mean squared out-of-sample error for a linear regression

## Usage

```
moose(reg, dataset)
```

## Arguments

|         |   |
|---------|---|
| reg     | an lm object containing the regression to project out-of-sample |
| dataset | a data.frame containing new cases for out-of-sample projection  |

## Value

|     |  |
|-----|--|
| mse | Projected mean squared out-of-sample error   |
| R2o | Projected out-of-sample R-squared  |
| hat | Leverage for each out-of-sample observation. For each $i$ , this is the sum of the squared elements of $x_i [X'X]^{-1} X'$ , where $X$ is the predictor matrix from the training sample. |

## Examples

```
# set the seed for reproducibility of the example
set.seed(04251978)
# randomly generate 100 observations of data
mydata <- data.frame(x1=rnorm(100),x2=rnorm(100),x3=rnorm(100))
# true outcome variable is  $y = x_1 + x_2 + x_3 + e$ 
y <- mydata$x1 + mydata$x2 + mydata$x3 + rnorm(100)
# regression with the first 25 observations from the dataset
reg <- lm(y ~ x1 + x2 + x3,data=cbind(y,mydata)[1:25,])
# using the predictor values from the first 25 observations,
# project the out-of-sample error we can expect in the case of
# "non-stochastic" predictors whose values are the same in the
# test sample as in the training sample.
# note that mydata does not include the outcome variable.
same.predictor.values.error <- moose(reg,mydata[1:25,])
# by comparison, the in-sample R-squared value observed
# in training is:
summary(reg)$r.squared
# using the predictor values from the next 75 observations,
# project the out-of-sample error we can expect in the case
# of stochastic predictors whose values potentially differ
# from those used in training.
new.predictor.values.error <- moose(reg,mydata[26:100,])
```

```
# by comparison, the actual mse and out-of-sample R-squared value
# obtained from observations 26-100 of this random sample are:
mse <- mean((y[26:100]-predict(reg,mydata[26:100,]))^2)
mse
m.total.sqs <- mean((y[26:100]-mean(y[26:100]))^2)
r2o <- 1-mse/m.total.sqs
r2o
```

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