

# Package ‘mvrSquared’

June 1, 2022

**Type** Package

**Title** Compute the Coefficient of Determination for Vector or Matrix Outcomes

**Version** 0.1.4

**Description** Compute the coefficient of determination for outcomes in n-dimensions. May be useful for multidimensional predictions (such as a multinomial model) or calculating goodness of fit from latent variable models such as probabilistic topic models like latent Dirichlet allocation or deterministic topic models like latent semantic analysis. Based on Jones (2019) <[arXiv:1911.11061](https://arxiv.org/abs/1911.11061)>.

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**Encoding** UTF-8

**URL** <https://github.com/TommyJones/mvrSquared>

**BugReports** <https://github.com/TommyJones/mvrSquared/issues>

**Depends** R (>= 3.0.2)

**Imports** Matrix, methods, Rcpp (>= 1.0.2)

**Suggests** dplyr, furr, knitr, MASS, nnet, parallel, rmarkdown, stats, stringr, testthat, textmineR, tidytext, spelling

**LinkingTo** Rcpp, RcppArmadillo, RcppThread (>= 2.1.3)

**RoxygenNote** 7.2.0

**VignetteBuilder** knitr

**Language** en-US

**NeedsCompilation** yes

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calc_rsquared	<i>Calculate R-Squared.</i>
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### Description

Calculate R-Squared for univariate or multivariate outcomes.

### Usage

```
calc_rsquared(y, yhat, ybar = NULL, return_ss_only = FALSE, threads = 1)
```

### Arguments

y	The true outcome. This must be a numeric vector, numeric matrix, or coercible to a sparse matrix of class <code>dgCMatrix</code> . See 'Details' below for more information.
yhat	The predicted outcome or a list of two matrices whose dot product makes the predicted outcome. See 'Details' below for more information.
ybar	Numeric scalar or vector; the mean of y. Useful for parallel computation in batches.
return_ss_only	Logical. Do you want to forego calculating R-squared and only return the sums of squares?
threads	Integer number of threads for parallelism; defaults to 1.

### Details

There is some flexibility in what you can pass as y and yhat. In general, y can be a numeric vector, numeric matrix, a sparse matrix of class `dgCMatrix` from the [Matrix](#) package, or any object that can be coerced into a `dgCMatrix`.

yhat can be a numeric vector, numeric matrix, or a list of two matrices whose dot product has the same dimensionality as y. If yhat is a list of two matrices you may optionally name them x and w indicating the order of multiplication (x left multiplies w). If unnamed or ambiguously named, then it is assumed that `yhat[[1]]` left multiplies `yhat[[2]]`.

### Value

If `return_ss_only = FALSE`, `calc_rsquared` returns a numeric scalar R-squared. If `return_ss_only = TRUE`, `calc_rsquared` returns a vector; the first element is the error sum of squares (SSE) and the second element is the total sum of squares (SST). R-squared may then be calculated as  $1 - \text{SSE} / \text{SST}$ .

**Note**

On some Linux systems, setting threads greater than 1 for parallelism may introduce some imprecision in the calculation. As of this writing, the cause is still under investigation. In the meantime setting `threads = 1` should fix the issue.

Setting `return_ss_only` to `TRUE` is useful for parallel or distributed computing for large data sets, particularly when `y` is a large matrix. However if you do parallel execution you **MUST** pre-calculate 'ybar' and pass it to the function. If you do not, SST will be calculated based on means of each batch independently. The resulting r-squared will be incorrect.

See example below for parallel computation with `future_map` from the `furr` package.

**Examples**

```
# standard r-squared with y and yhat as vectors
f <- stats::lm(mpg ~ cyl + disp + hp + wt, data = datasets::mtcars)

y <- f$model$mpg

yhat <- f$fitted.values

calc_rsquared(y = y, yhat = yhat)

# standard r-squared with y as a matrix and yhat containing 'x' and linear coefficients
s <- summary(f)

x <- cbind(1, as.matrix(f$model[, -1]))

w <- matrix(s$coefficients[, 1], ncol = 1)

calc_rsquared(y = matrix(y, ncol = 1), yhat = list(x, w))

# multivariate r-squared with y and yhat as matrices
calc_rsquared(y = cbind(y, y), yhat = cbind(yhat, yhat))

# multivariate r-squared with yhat as a linear reconstruction of two matrices
calc_rsquared(y = cbind(y, y), yhat = list(x, cbind(w,w)))
```

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mvrSquared

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**Description**

Compute the Coefficient of Determination for Vector or Matrix Outcomes

## Details

Welcome to the mvrsquared package! This package does one thing: calculate the coefficient of determination or "R-squared". However, this implementation is different from what you may be familiar with. In addition to the standard R-squared used frequently in linear regression, 'mvrsquared' calculates R-squared for multivariate outcomes. (This is why there is an 'mv' in mvrsquared).

mvrsquared implements R-squared based on a derivation in this paper (<https://arxiv.org/abs/1911.11061>). It's the same definition of R-squared you're probably familiar with, i.e.  $1 - \text{SSE}/\text{SST}$  but generalized to n-dimensions.

In the standard case, your outcome and prediction are vectors. In other words, each observation is a single number. This is fine if you are predicting a single variable. But what if you are predicting multiple variables at once? In that case, your outcome and prediction are matrices. This situation occurs frequently in topic modeling or simultaneous equation modeling.

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