Package ‘Metrics’

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Title Evaluation Metrics for Machine Learning

Description An implementation of evaluation metrics in R that are commonly used in supervised machine learning. It implements metrics for regression, time series, binary classification, classification, and information retrieval problems. It has zero dependencies and a consistent, simple interface for all functions.

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Suggests testthat

URL https://github.com/mfrasco/Metrics

BugReports https://github.com/mfrasco/Metrics/issues

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**Description**

accuracy is defined as the proportion of elements in actual that are equal to the corresponding element in predicted.

**Usage**

```
accuracy(actual, predicted)
```

**Arguments**

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>actual</td>
<td>The ground truth vector, where elements of the vector can be any variable type.</td>
</tr>
<tr>
<td>predicted</td>
<td>The predicted vector, where elements of the vector represent a prediction for the corresponding value in actual.</td>
</tr>
</tbody>
</table>
**Description**

`ae` computes the elementwise absolute difference between two numeric vectors.

**Usage**

```r
ae(actual, predicted)
```

**Arguments**

- `actual` The ground truth numeric vector.
- `predicted` The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in `actual`.

**Examples**

```r
actual <- c('a', 'a', 'c', 'b', 'c')
predicted <- c('a', 'b', 'c', 'b', 'a')
accuracy(actual, predicted)
```
ape  

**Absolute Percent Error**

**Description**

ape computes the elementwise absolute percent difference between two numeric vectors.

**Usage**

ape(actual, predicted)

**Arguments**

- **actual**: The ground truth numeric vector.
- **predicted**: The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in actual.

**Details**

ape is calculated as \((\text{actual} - \text{predicted}) / \text{abs(actual)}\). This means that the function will return \(-\infty, \infty, \text{or NaN}\) if actual is zero.

**See Also**

mape smape

**Examples**

```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
ape(actual, predicted)
```

**apk  

**Average Precision at k**

**Description**

apk computes the average precision at k, in the context of information retrieval problems.

**Usage**

apk(k, actual, predicted)
Arguments

- **k**: The number of elements of `predicted` to consider in the calculation.
- **actual**: The ground truth vector of relevant documents. The vector can contain any numeric or character values, order does not matter, and the vector does not need to be the same length as `predicted`.
- **predicted**: The predicted vector of retrieved documents. The vector can contain any numeric or character values. However, unlike `actual`, order does matter, with the most documents deemed most likely to be relevant at the beginning.

Details

apk loops over the first k values of `predicted`. For each value, if the value is contained within `actual` and has not been predicted before, we increment the number of successes by one and increment our score by the number of successes divided by k. Then, we return our final score divided by the number of relevant documents (i.e. the length of `actual`).

apk will return NaN if `length(actual)` equals 0.

See Also

- `apk f1`

Examples

```r
actual <- c('a', 'b', 'd')
predicted <- c('b', 'c', 'a', 'e', 'f')
apk(3, actual, predicted)
```

---

**auc**

*Area under the ROC curve (AUC)*

**Description**

`auc` computes the area under the receiver-operator characteristic curve (AUC).

**Usage**

`auc(actual, predicted)`

**Arguments**

- **actual**: The ground truth binary numeric vector containing 1 for the positive class and 0 for the negative class.
- **predicted**: A numeric vector of predicted values, where the smallest values correspond to the observations most believed to be in the negative class and the largest values indicate the observations most believed to be in the positive class. Each element represents the prediction for the corresponding element in `actual`. 
Details

`auc` uses the fact that the area under the ROC curve is equal to the probability that a randomly chosen positive observation has a higher predicted value than a randomly chosen negative value. In order to compute this probability, we can calculate the Mann-Whitney U statistic. This method is very fast, since we do not need to compute the ROC curve first.

Examples

```r
actual <- c(1, 1, 0, 0, 0)
predicted <- c(0.9, 0.8, 0.4, 0.5, 0.3, 0.2)
auc(actual, predicted)
```

---

### bias

**Description**

`bias` computes the average amount by which `actual` is greater than `predicted`.

**Usage**

`bias(actual, predicted)`

**Arguments**

* `actual` The ground truth numeric vector.
* `predicted` The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in `actual`.

**Details**

If a model is unbiased `bias(actual, predicted)` should be close to zero. Bias is calculated by taking the average of `(actual - predicted)`.

**See Also**

`percent_bias`

**Examples**

```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
bias(actual, predicted)
```
**Classification Error**

**Description**

Classification Error (ce) is defined as the proportion of elements in actual that are not equal to the corresponding element in predicted.

**Usage**

```r
c1(actual, predicted)
```

**Arguments**

- `actual`: The ground truth vector, where elements of the vector can be any variable type.
- `predicted`: The predicted vector, where elements of the vector represent a prediction for the corresponding value in actual.

**See Also**

- `accuracy`

**Examples**

```r
actual <- c('a', 'a', 'c', 'b', 'c')
predicted <- c('a', 'b', 'c', 'b', 'a')
c1(actual, predicted)
```

---

**F1 Score**

**Description**

F1 Score (f1) computes the F1 Score in the context of information retrieval problems.

**Usage**

```r
f1(actual, predicted)
```

**Arguments**

- `actual`: The ground truth vector of relevant documents. The vector can contain any numeric or character values, order does not matter, and the vector does not need to be the same length as predicted.
- `predicted`: The predicted vector of retrieved documents. The vector can contain any numeric or character values, order does not matter, and the vector does not need to be the same length as actual.
Details

f1 is defined as $2 \times \text{precision} \times \text{recall} / (\text{precision} + \text{recall})$. In the context of information retrieval problems, precision is the proportion of retrieved documents that are relevant to a query and recall is the proportion of relevant documents that are successfully retrieved by a query. If there are zero relevant documents that are retrieved, zero relevant documents, or zero predicted documents, f1 is defined as 0.

See Also

apk mapk

Examples

actual <- c('a', 'c', 'd')
predicted <- c('d', 'e')
f1(actual, predicted)

<table>
<thead>
<tr>
<th>fbeta_score</th>
<th>F-beta Score</th>
</tr>
</thead>
</table>

Description

fbeta_score computes a weighted harmonic mean of Precision and Recall. The beta parameter controls the weighting.

Usage

fbeta_score(actual, predicted, beta = 1)

Arguments

actual The ground truth binary numeric vector containing 1 for the positive class and 0 for the negative class.
predicted The predicted binary numeric vector containing 1 for the positive class and 0 for the negative class. Each element represents the prediction for the corresponding element in actual.
beta A non-negative real number controlling how close the F-beta score is to either Precision or Recall. When beta is at the default of 1, the F-beta Score is exactly an equally weighted harmonic mean. The F-beta score will weight toward Precision when beta is less than one. The F-beta score will weight toward Recall when beta is greater than one.

See Also

precision recall
**Examples**

```r
actual <- c(1, 1, 1, 0, 0, 0)
predicted <- c(1, 0, 1, 1, 1, 1)
recall(actual, predicted)
```

---

**Log Loss**

**Description**

Log Loss computes the elementwise log loss between two numeric vectors.

**Usage**

```r
ll(actual, predicted)
```

**Arguments**

- `actual`: The ground truth binary numeric vector containing 1 for the positive class and 0 for the negative class.
- `predicted`: A numeric vector of predicted values, where the values correspond to the probabilities that each observation in `actual` belongs to the positive class.

**See Also**

- `logloss`

---

**Examples**

```r
actual <- c(1, 1, 1, 0, 0, 0)
predicted <- c(0.9, 0.8, 0.4, 0.5, 0.3, 0.2)
ll(actual, predicted)
```

---

**logLoss**

**Mean Log Loss**

**Description**

logLoss computes the average log loss between two numeric vectors.

**Usage**

```r
logLoss(actual, predicted)
```
Arguments

actual  The ground truth binary numeric vector containing 1 for the positive class and 0 for the negative class.

predicted  A numeric vector of predicted values, where the values correspond to the probabilities that each observation in actual belongs to the positive class.

See Also

Examples

actual <- c(1, 1, 1, 0, 0, 0)
predicted <- c(0.9, 0.8, 0.4, 0.5, 0.3, 0.2)
logLoss(actual, predicted)

mae  Mean Absolute Error

Description

mae computes the average absolute difference between two numeric vectors.

Usage

mae(actual, predicted)

Arguments

actual  The ground truth numeric vector.

predicted  The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in actual.

See Also

mdae mape

Examples

actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
mae(actual, predicted)
mape

**Mean Absolute Percent Error**

**Description**

mape computes the average absolute percent difference between two numeric vectors.

**Usage**

`mape(actual, predicted)`

**Arguments**

- `actual`: The ground truth numeric vector.
- `predicted`: The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in `actual`.

**Details**

mape is calculated as the average of `(actual - predicted) / abs(actual)`. This means that the function will return `-Inf`, `Inf`, or `NaN` if actual is zero. Due to the instability at or near zero, smape or mase are often used as alternatives.

**See Also**

- mae
- smape
- mase

**Examples**

```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
mape(actual, predicted)
```

mapk

**Mean Average Precision at k**

**Description**

mapk computes the mean average precision at k for a set of predictions, in the context of information retrieval problems.

**Usage**

`mapk(k, actual, predicted)`

**Examples**

```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
mapk(k, actual, predicted)
```
mase

Mean Absolute Scaled Error

Arguments

k
The number of elements of predicted to consider in the calculation.

actual
A list of vectors, where each vector represents a ground truth vector of relevant documents. In each vector, the elements can be numeric or character values, and the order of the elements does not matter.

predicted
A list of vectors, where each vector represents the predicted vector of retrieved documents for the corresponding element of actual. In each vector, the order of the elements does matter, with the elements believed most likely to be relevant at the beginning.

Details

mapk evaluates apk for each pair of elements from actual and predicted.

See Also

apk f1

Examples

```r
actual <- list(c('a', 'b'), c('a'), c('x', 'y', 'b'))
predicted <- list(c('a', 'c', 'd'), c('x', 'b', 'a', 'b'), c('y'))
mapk(2, actual, predicted)
```

```r
actual <- list(c(1, 5, 7, 9), c(2, 3), c(2, 5, 6))
predicted <- list(c(5, 6, 7, 8, 9), c(1, 2, 3), c(2, 4, 6, 8))
mapk(3, actual, predicted)
```

Description

mase computes the mean absolute scaled error between two numeric vectors. This function is only intended for time series data, where actual and numeric are numeric vectors ordered by time.

Usage

```r
mase(actual, predicted, step_size = 1)
```

Arguments

actual
The ground truth numeric vector ordered in time, with most recent observation at the end of the vector.

predicted
The predicted numeric vector ordered in time, where each element of the vector represents a prediction for the corresponding element of actual.
**step_size**  
A positive integer that specifies how many observations to look back in time in order to compute the naive forecast. The default is 1, which means that the naive forecast for the current time period is the actual value of the previous period. However, if actual and predictions were quarterly predictions over many years, letting step_size = 4, would mean that the naive forecast for the current time period would be the actual value from the same quarter last year. In this way, mase can account for seasonality.

**See Also**  
smape mape

**Examples**
```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
step_size <- 1
mase(actual, predicted, step_size)
```

---

**mdae**  
**Median Absolute Error**

**Description**

mdae computes the median absolute difference between two numeric vectors.

**Usage**

```r
mdae(actual, predicted)
```

**Arguments**

- `actual` The ground truth numeric vector.
- `predicted` The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in actual.

**See Also**

mae mape

**Examples**
```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
mdae(actual, predicted)
```
MeanQuadraticWeightedKappa

Mean Quadratic Weighted Kappa

Description

MeanQuadraticWeightedKappa computes the mean quadratic weighted kappa, which can optionally be weighted.

Usage

MeanQuadraticWeightedKappa(kappas, weights = rep(1, length(kappas)))

Arguments

kappas A numeric vector of possible kappas.
weights An optional numeric vector of ratings.

See Also

ScoreQuadraticWeightedKappa

Examples

kappas <- c(0.3, 0.2, 0.2, 0.5, 0.1, 0.2)
weights <- c(1.0, 2.5, 1.0, 1.0, 2.0, 3.0)
MeanQuadraticWeightedKappa(kappas, weights)

mse

Mean Squared Error

Description

mse computes the average squared difference between two numeric vectors.

Usage

mse(actual, predicted)

Arguments

actual The ground truth numeric vector.
predicted The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in actual.
**msle**

**See Also**

* rmse  mae

**Examples**

```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
msle(actual, predicted)
```

---

**msle**

*Mean Squared Log Error*

**Description**

`msle` computes the average of squared log error between two numeric vectors.

**Usage**

```r
msle(actual, predicted)
```

**Arguments**

- `actual` The ground truth non-negative vector
- `predicted` The predicted non-negative vector, where each element in the vector is a prediction for the corresponding element in `actual`.

**Details**

`msle` adds one to both `actual` and `predicted` before taking the natural logarithm to avoid taking the natural log of zero. As a result, the function can be used if `actual` or `predicted` have zero-valued elements. But this function is not appropriate if either are negative valued.

**See Also**

* rmsle  sle

**Examples**

```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
msle(actual, predicted)
```
Inherit Documentation for Binary Classification Metrics

Description
This object provides the documentation for the parameters of functions that provide binary classification metrics.

Arguments
- **actual**: The ground truth binary numeric vector containing 1 for the positive class and 0 for the negative class.
- **predicted**: The predicted binary numeric vector containing 1 for the positive class and 0 for the negative class. Each element represents the prediction for the corresponding element in actual.

Inherit Documentation for Classification Metrics

Description
This object provides the documentation for the parameters of functions that provide classification metrics.

Arguments
- **actual**: The ground truth vector, where elements of the vector can be any variable type.
- **predicted**: The predicted vector, where elements of the vector represent a prediction for the corresponding value in actual.

Inherit Documentation for Regression Metrics

Description
This object provides the documentation for the parameters of functions that provide regression metrics.

Arguments
- **actual**: The ground truth numeric vector.
- **predicted**: The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in actual.
percent_bias

Description

percent_bias computes the average amount that \texttt{actual} is greater than \texttt{predicted} as a percentage of the absolute value of \texttt{actual}.

Usage

\texttt{percent_bias(actual, predicted)}

Arguments

\begin{itemize}
\item \texttt{actual} \hspace{1cm} The ground truth numeric vector.
\item \texttt{predicted} \hspace{1cm} The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in \texttt{actual}.
\end{itemize}

Details

If a model is unbiased \texttt{percent_bias(actual, predicted)} should be close to zero. \texttt{Percent Bias} is calculated by taking the average of \((\texttt{actual} - \texttt{predicted}) / \text{abs(\texttt{actual})}\) across all observations. \texttt{percent_bias} will give \(-\text{Inf}, \text{Inf}, \text{or NaN}\), if any elements of \texttt{actual} are 0.

See Also

bias

Examples

\begin{verbatim}
actual <- c(1.1, 1.9, 3.8, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
percent_bias(actual, predicted)
\end{verbatim}

precision

Description

\texttt{precision} computes proportion of observations predicted to be in the positive class (i.e. the element in \texttt{predicted} equals 1) that actually belong to the positive class (i.e. the element in \texttt{actual} equals 1)

Usage

\texttt{precision(actual, predicted)}
**Arguments**

- `actual`: The ground truth binary numeric vector containing 1 for the positive class and 0 for the negative class.
- `predicted`: The predicted binary numeric vector containing 1 for the positive class and 0 for the negative class. Each element represents the prediction for the corresponding element in `actual`.

**See Also**

- `recall`
- `fbeta_score`

**Examples**

```r
actual <- c(1, 1, 0, 0, 0)
predicted <- c(1, 1, 1, 1, 1)
precision(actual, predicted)
```

---

**Description**

`rae` computes the relative absolute error between two numeric vectors.

**Usage**

```r
rae(actual, predicted)
```

**Arguments**

- `actual`: The ground truth numeric vector.
- `predicted`: The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in `actual`.

**Details**

`rae` divides \( \sum(\text{ae}(\text{actual}, \text{predicted})) \) by \( \sum(\text{ae}(\text{actual}, \text{mean(\text{actual}))}) \), meaning that it provides the absolute error of the predictions relative to a naive model that predicted the mean for every data point.

**See Also**

- `rse`
- `rrse`

**Examples**

```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
rrse(actual, predicted)
```
**recall**

**Description**

recall computes proportion of observations in the positive class (i.e. the element in actual equals 1) that are predicted to be in the positive class (i.e. the element in predicted equals 1).

**Usage**

```
recall(actual, predicted)
```

**Arguments**

- **actual**: The ground truth binary numeric vector containing 1 for the positive class and 0 for the negative class.
- **predicted**: The predicted binary numeric vector containing 1 for the positive class and 0 for the negative class. Each element represents the prediction for the corresponding element in actual.

**See Also**

precision, fbeta_score

**Examples**

```
actual <- c(1, 1, 0, 0, 0)
predicted <- c(1, 0, 1, 1, 1)
recall(actual, predicted)
```

---

**rmse**

**Root Mean Squared Error**

**Description**

rmse computes the root mean squared error between two numeric vectors.

**Usage**

```
rmse(actual, predicted)
```

**Arguments**

- **actual**: The ground truth numeric vector.
- **predicted**: The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in actual.
See Also

mse

Examples

```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
rmse(actual, predicted)
```

Description

`rmsle` computes the root mean squared log error between two numeric vectors.

Usage

`rmsle(actual, predicted)`

Arguments

- `actual`: The ground truth non-negative vector
- `predicted`: The predicted non-negative vector, where each element in the vector is a prediction for the corresponding element in `actual`.

Details

`rmsle` adds one to both `actual` and `predicted` before taking the natural logarithm to avoid taking the natural log of zero. As a result, the function can be used if `actual` or `predicted` have zero-valued elements. But this function is not appropriate if either are negative valued.

See Also

msle sle

Examples

```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
rmsle(actual, predicted)
```
\textbf{rrse} \hfill \textit{Root Relative Squared Error}

\begin{description}
  \item[Description] \textit{rrse} computes the root relative squared error between two numeric vectors.
  \item[Usage] \textit{rrse(actual, predicted)}
  \item[Arguments]
    \begin{description}
      \item[actual] The ground truth numeric vector.
      \item[predicted] The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in \textit{actual}.
    \end{description}
  \item[Details] \textit{rrse} takes the square root of \textit{sse(actual, predicted)} divided by \textit{sse(actual, mean(actual))}, meaning that it provides the squared error of the predictions relative to a naive model that predicted the mean for every data point.
\end{description}
\textbf{See Also}
\textit{rse rae}

\textbf{Examples}
\begin{verbatim}
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
rrse(actual, predicted)
\end{verbatim}

\textbf{rse} \hfill \textit{Relative Squared Error}

\begin{description}
  \item[Description] \textit{rse} computes the relative squared error between two numeric vectors.
  \item[Usage] \textit{rse(actual, predicted)}
\end{description}
ScoreQuadraticWeightedKappa

Arguments

- **actual**: The ground truth numeric vector.
- **predicted**: The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in actual.

Details

\[
\text{rse} \text{ divides } \text{sse(actual, predicted)} \text{ by } \text{sse(actual, mean(actual))}, \text{ meaning that it provides the squared error of the predictions relative to a naive model that predicted the mean for every data point.}
\]

See Also

- rse
- rae

Examples

```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
```

ScoreQuadraticWeightedKappa

**Quadratic Weighted Kappa**

Description

ScoreQuadraticWeightedKappa computes the quadratic weighted kappa between two vectors of integers.

Usage

`ScoreQuadraticWeightedKappa(rater.a, rater.b, min.rating = min(c(rater.a, rater.b)), max.rating = max(c(rater.a, rater.b)))`

Arguments

- **rater.a**: An integer vector of the first rater’s ratings.
- **rater.b**: An integer vector of the second rater’s ratings.
- **min.rating**: The minimum possible rating.
- **max.rating**: The maximum possible rating.

See Also

- MeanQuadraticWeightedKappa
Examples
rater.a <- c(1, 4, 5, 2, 1)
rater.b <- c(2, 2, 4, 5, 3)
ScoreQuadraticWeightedKappa(rater.a, rater.b, 1, 5)

se

Squared Error

Description
se computes the elementwise squared difference between two numeric vectors.

Usage
se(actual, predicted)

Arguments
actual
The ground truth numeric vector.
predicted
The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in actual.

See Also
mse rmse

Examples
actual <- c(1.1, 1.9, 3.0, 4.4, 5.8, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
se(actual, predicted)

sle

Squared Log Error

Description
sle computes the elementwise squares of the differences in the logs of two numeric vectors.

Usage
sle(actual, predicted)
Arguments

- `actual`: The ground truth non-negative vector.
- `predicted`: The predicted non-negative vector, where each element in the vector is a prediction for the corresponding element in `actual`.

Details

`sle` adds one to both `actual` and `predicted` before taking the natural logarithm of each to avoid taking the natural log of zero. As a result, the function can be used if `actual` or `predicted` have zero-valued elements. But this function is not appropriate if either are negative valued.

See Also

`msle`, `rmsle`

Examples

```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
predicted <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
sle(actual, predicted)
```

```
smape  Symmetric Mean Absolute Percentage Error
```

Description

`smape` computes the symmetric mean absolute percentage error between two numeric vectors.

Usage

`smape(actual, predicted)`

Arguments

- `actual`: The ground truth numeric vector.
- `predicted`: The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in `actual`.

Details

`smape` is defined as two times the average of \( \frac{\text{abs}(\text{actual} - \text{predicted})}{(\text{abs}(\text{actual}) + \text{abs}(\text{predicted}))} \). Therefore, at the elementwise level, it will provide `NaN` only if `actual` and `predicted` are both zero. It has an upper bound of 2, when either `actual` or `predicted` are zero or when `actual` and `predicted` are opposite signs.

`smape` is symmetric in the sense that \( \text{smape}(x, y) = \text{smape}(y, x) \).
**sse**

**See Also**

-mape, mase

**Examples**

```r
test <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
expected <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
smape(test, expected)
```

---

**sse**

*Sum of Squared Errors*

**Description**

sse computes the sum of the squared differences between two numeric vectors.

**Usage**

```r
sse(actual, predicted)
```

**Arguments**

- `actual` - The ground truth numeric vector.
- `predicted` - The predicted numeric vector, where each element in the vector is a prediction for the corresponding element in `actual`.

**See Also**

-mse

**Examples**

```r
actual <- c(1.1, 1.9, 3.0, 4.4, 5.0, 5.6)
expected <- c(0.9, 1.8, 2.5, 4.5, 5.0, 6.2)
nse(actual, predicted)
```
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