

# Package ‘MLmetrics’

January 20, 2025

**Type** Package

**Title** Machine Learning Evaluation Metrics

**Version** 1.1.3

**Description** A collection of evaluation metrics, including loss, score and utility functions, that measure regression, classification and ranking performance.

**URL** <https://github.com/yanyachen/MLmetrics>

**BugReports** <https://github.com/yanyachen/MLmetrics/issues>

**Depends** R (>= 2.10)

**Imports** stats, utils, ROCR

**Suggests** e1071

**License** GPL-2

**RoxygenNote** 5.0.1

**NeedsCompilation** no

**Author** Yachen Yan [aut, cre]

**Maintainer** Yachen Yan <yanyachen21@gmail.com>

**Repository** CRAN

**Date/Publication** 2024-04-13 23:50:05 UTC

## Contents

Accuracy . . . . .	2
Area_Under_Curve . . . . .	3
AUC . . . . .	4
ConfusionMatrix . . . . .	4
F1_Score . . . . .	5
FBeta_Score . . . . .	6
GainAUC . . . . .	6
Gini . . . . .	7
KS_Stat . . . . .	8
LiftAUC . . . . .	8

LogLoss . . . . .	9
MAE . . . . .	9
MAPE . . . . .	10
MedianAE . . . . .	11
MedianAPE . . . . .	11
MLmetrics . . . . .	12
MSE . . . . .	12
MultiLogLoss . . . . .	13
NormalizedGini . . . . .	13
Poisson_LogLoss . . . . .	14
PRAUC . . . . .	14
Precision . . . . .	15
R2_Score . . . . .	16
RAE . . . . .	16
Recall . . . . .	17
RMSE . . . . .	18
RMSLE . . . . .	18
RMSPE . . . . .	19
RRSE . . . . .	19
Sensitivity . . . . .	20
Specificity . . . . .	21
ZeroOneLoss . . . . .	21

<b>Index</b>	<b>23</b>
--------------	-----------

---

Accuracy

*Accuracy*

---

### Description

Compute the accuracy classification score.

### Usage

Accuracy(y\_pred, y\_true)

### Arguments

y_pred	Predicted labels vector, as returned by a classifier
y_true	Ground truth (correct) 0-1 labels vector

### Value

Accuracy

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
pred <- ifelse(logreg$fitted.values < 0.5, 0, 1)
Accuracy(y_pred = pred, y_true = mtcars$vs)
```

---

Area\_Under\_Curve

*Calculate the Area Under the Curve*

---

**Description**

Calculate the area under the curve.

**Usage**

```
Area_Under_Curve(x, y, method = c("trapezoid", "step", "spline"),
                 na.rm = FALSE)
```

**Arguments**

x	the x-points of the curve
y	the y-points of the curve
method	can be "trapezoid" (default), "step" or "spline"
na.rm	a logical value indicating whether NA values should be stripped before the computation proceeds

**Value**

Area Under the Curve (AUC)

**Examples**

```
x <- seq(0, pi, length.out = 200)
plot(x = x, y = sin(x), type = "l")
Area_Under_Curve(x = x, y = sin(x), method = "trapezoid", na.rm = TRUE)
```

AUC *Area Under the Receiver Operating Characteristic Curve (ROC AUC)*

---

**Description**

Compute the Area Under the Receiver Operating Characteristic Curve (ROC AUC) from prediction scores.

**Usage**

```
AUC(y_pred, y_true)
```

**Arguments**

y_pred	Predicted probabilities vector, as returned by a classifier
y_true	Ground truth (correct) 0-1 labels vector

**Value**

Area Under the ROC Curve (ROC AUC)

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
AUC(y_pred = logreg$fitted.values, y_true = mtcars$vs)
```

---

ConfusionMatrix *Confusion Matrix*

---

**Description**

Compute confusion matrix to evaluate the accuracy of a classification.

**Usage**

```
ConfusionMatrix(y_pred, y_true)
```

**Arguments**

y_pred	Predicted labels vector, as returned by a classifier
y_true	Ground truth (correct) 0-1 labels vector

**Value**

a table of Confusion Matrix

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
pred <- ifelse(logreg$fitted.values < 0.5, 0, 1)
ConfusionMatrix(y_pred = pred, y_true = mtcars$vs)
```

---

F1\_Score

*F1 Score*

---

**Description**

Compute the F1 Score.

**Usage**

```
F1_Score(y_true, y_pred, positive = NULL)
```

**Arguments**

<code>y_true</code>	Ground truth (correct) 0-1 labels vector
<code>y_pred</code>	Predicted labels vector, as returned by a classifier
<code>positive</code>	An optional character string for the factor level that corresponds to a "positive" result

**Value**

F1 Score

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
pred <- ifelse(logreg$fitted.values < 0.5, 0, 1)
F1_Score(y_pred = pred, y_true = mtcars$vs, positive = "0")
F1_Score(y_pred = pred, y_true = mtcars$vs, positive = "1")
```

---

 FBeta\_Score

*F-Beta Score*


---

**Description**

Compute the F-Beta Score

**Usage**

```
FBeta_Score(y_true, y_pred, positive = NULL, beta = 1)
```

**Arguments**

y_true	Ground truth (correct) 0-1 labels vector
y_pred	Predicted labels vector, as returned by a classifier
positive	An optional character string for the factor level that corresponds to a "positive" result
beta	Weight of precision in harmonic mean

**Value**

F-Beta Score

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
pred <- ifelse(logreg$fitted.values < 0.5, 0, 1)
FBeta_Score(y_pred = pred, y_true = mtcars$vs, positive = "0", beta = 2)
FBeta_Score(y_pred = pred, y_true = mtcars$vs, positive = "1", beta = 2)
```

---

 GainAUC

*Area Under the Gain Chart*


---

**Description**

Compute the Area Under the Gain Chart from prediction scores.

**Usage**

```
GainAUC(y_pred, y_true)
```

**Arguments**

y_pred	Predicted probabilities vector, as returned by a classifier
y_true	Ground truth (correct) 0-1 labels vector

**Value**

Area Under the Gain Chart

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
GainAUC(y_pred = logreg$fitted.values, y_true = mtcars$vs)
```

---

Gini

*Gini Coefficient*

---

**Description**

Compute the Gini Coefficient.

**Usage**

```
Gini(y_pred, y_true)
```

**Arguments**

y_pred	Predicted probabilities vector, as returned by a classifier
y_true	Ground truth (correct) 0-1 labels vector

**Value**

Gini Coefficient

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
Gini(y_pred = logreg$fitted.values, y_true = mtcars$vs)
```

---

KS_Stat	<i>Kolmogorov-Smirnov Statistic</i>
---------	-------------------------------------

---

**Description**

Compute the Kolmogorov-Smirnov statistic.

**Usage**

```
KS_Stat(y_pred, y_true)
```

**Arguments**

y_pred	Predicted probabilities vector, as returned by a classifier
y_true	Ground truth (correct) 0-1 labels vector

**Value**

Kolmogorov-Smirnov statistic

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
KS_Stat(y_pred = logreg$fitted.values, y_true = mtcars$vs)
```

---

LiftAUC	<i>Area Under the Lift Chart</i>
---------	----------------------------------

---

**Description**

Compute the Area Under the Lift Chart from prediction scores.

**Usage**

```
LiftAUC(y_pred, y_true)
```

**Arguments**

y_pred	Predicted probabilities vector, as returned by a classifier
y_true	Ground truth (correct) 0-1 labels vector

**Value**

Area Under the Lift Chart



**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
LiftAUC(y_pred = logreg$fitted.values, y_true = mtcars$vs)
```

---

LogLoss

*Log loss / Cross-Entropy Loss*

---

**Description**

Compute the log loss/cross-entropy loss.

**Usage**

```
LogLoss(y_pred, y_true)
```

**Arguments**

y\_pred            Predicted probabilities vector, as returned by a classifier  
y\_true            Ground truth (correct) 0-1 labels vector

**Value**

Log loss/Cross-Entropy Loss

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
LogLoss(y_pred = logreg$fitted.values, y_true = mtcars$vs)
```

---

MAE

*Mean Absolute Error Loss*

---

**Description**

Compute the mean absolute error regression loss.

**Usage**

```
MAE(y_pred, y_true)
```

**Arguments**

<code>y_pred</code>	Estimated target values vector
<code>y_true</code>	Ground truth (correct) target values vector

**Value**

Mean Absolute Error Loss

**Examples**

```
data(cars)
reg <- lm(log(dist) ~ log(speed), data = cars)
MAE(y_pred = exp(reg$fitted.values), y_true = cars$dist)
```

---

MAPE

*Mean Absolute Percentage Error Loss*

---

**Description**

Compute the mean absolute percentage error regression loss.

**Usage**

```
MAPE(y_pred, y_true)
```

**Arguments**

<code>y_pred</code>	Estimated target values vector
<code>y_true</code>	Ground truth (correct) target values vector

**Value**

Mean Absolute Percentage Error Loss

**Examples**

```
data(cars)
reg <- lm(log(dist) ~ log(speed), data = cars)
MAPE(y_pred = exp(reg$fitted.values), y_true = cars$dist)
```

---

MedianAE	<i>Median Absolute Error Loss</i>
----------	-----------------------------------

---

**Description**

Compute the median absolute error regression loss.

**Usage**

```
MedianAE(y_pred, y_true)
```

**Arguments**

y_pred	Estimated target values vector
y_true	Ground truth (correct) target values vector

**Value**

Median Absolute Error Loss

**Examples**

```
data(cars)
reg <- lm(log(dist) ~ log(speed), data = cars)
MedianAE(y_pred = exp(reg$fitted.values), y_true = cars$dist)
```

---

MedianAPE	<i>Median Absolute Percentage Error Loss</i>
-----------	--

---

**Description**

Compute the Median absolute percentage error regression loss.

**Usage**

```
MedianAPE(y_pred, y_true)
```

**Arguments**

y_pred	Estimated target values vector
y_true	Ground truth (correct) target values vector

**Value**

Median Absolute Percentage Error Loss

**Examples**

```
data(cars)
reg <- lm(log(dist) ~ log(speed), data = cars)
MedianAPE(y_pred = exp(reg$fitted.values), y_true = cars$dist)
```

---

**MLmetrics***MLmetrics: Machine Learning Evaluation Metrics*

---

**Description**

A collection of evaluation metrics, including loss, score and utility functions, that measure regression and classification performance.

---

**MSE***Mean Square Error Loss*

---

**Description**

Compute the mean squared error regression loss.

**Usage**

```
MSE(y_pred, y_true)
```

**Arguments**

y_pred	Estimated target values vector
y_true	Ground truth (correct) target values vector

**Value**

Mean Square Error Loss

**Examples**

```
data(cars)
reg <- lm(log(dist) ~ log(speed), data = cars)
MSE(y_pred = exp(reg$fitted.values), y_true = cars$dist)
```

---

MultiLogLoss	<i>Multi Class Log Loss</i>
--------------	-----------------------------

---

**Description**

Compute the multi class log loss.

**Usage**

```
MultiLogLoss(y_pred, y_true)
```

**Arguments**

y_pred	Predicted probabilities matrix, as returned by a classifier
y_true	Ground truth (correct) labels vector or a matrix of correct labels indicating by 0-1, same format as probabilities matrix

**Value**

Multi Class Log Loss

**Examples**

```
data(iris)
svm.model <- e1071::svm(Species~., data = iris, probability = TRUE)
pred <- predict(svm.model, iris, probability = TRUE)
MultiLogLoss(y_true = iris$Species, y_pred = attr(pred, "probabilities"))
```

---

NormalizedGini	<i>Normalized Gini Coefficient</i>
----------------	------------------------------------

---

**Description**

Compute the Normalized Gini Coefficient.

**Usage**

```
NormalizedGini(y_pred, y_true)
```

**Arguments**

y_pred	Predicted labels vector, as returned by a model
y_true	Ground truth (correct) labels vector

**Value**

Normalized Gini Coefficient

**Examples**

```
d_AD <- data.frame(treatment = gl(3,3), outcome = gl(3,1,9),
                  counts = c(18,17,15,20,10,20,25,13,12))
glm_poisson <- glm(counts ~ outcome + treatment,
                  family = poisson(link = "log"), data = d_AD)
NormalizedGini(y_pred = glm_poisson$fitted.values, y_true = d_AD$counts)
```

---

Poisson_LogLoss	<i>Poisson Log loss</i>
-----------------	-------------------------

---

**Description**

Compute the log loss/cross-entropy loss.

**Usage**

```
Poisson_LogLoss(y_pred, y_true)
```

**Arguments**

y_pred	Predicted labels vector, as returned by a model
y_true	Ground truth (correct) labels vector

**Value**

Log loss/Cross-Entropy Loss

**Examples**

```
d_AD <- data.frame(treatment = gl(3,3), outcome = gl(3,1,9),
                  counts = c(18,17,15,20,10,20,25,13,12))
glm_poisson <- glm(counts ~ outcome + treatment,
                  family = poisson(link = "log"), data = d_AD)
Poisson_LogLoss(y_pred = glm_poisson$fitted.values, y_true = d_AD$counts)
```

---

PRAUC	<i>Area Under the Precision-Recall Curve (PR AUC)</i>
-------	---

---

**Description**

Compute the Area Under the Precision-Recall Curve (PR AUC) from prediction scores.

**Usage**

```
PRAUC(y_pred, y_true)
```

**Arguments**

`y_pred` Predicted probabilities vector, as returned by a classifier  
`y_true` Ground truth (correct) 0-1 labels vector

**Value**

Area Under the PR Curve (PR AUC)

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
PRAUC(y_pred = logreg$fitted.values, y_true = mtcars$vs)
```

---

Precision

*Precision*

---

**Description**

Compute the precision score.

**Usage**

```
Precision(y_true, y_pred, positive = NULL)
```

**Arguments**

`y_true` Ground truth (correct) 0-1 labels vector  
`y_pred` Predicted labels vector, as returned by a classifier  
`positive` An optional character string for the factor level that corresponds to a "positive" result

**Value**

Precision

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
pred <- ifelse(logreg$fitted.values < 0.5, 0, 1)
Precision(y_pred = pred, y_true = mtcars$vs, positive = "0")
Precision(y_pred = pred, y_true = mtcars$vs, positive = "1")
```

---

R2_Score	<i>R-Squared (Coefficient of Determination) Regression Score</i>
----------	--

---

**Description**

Compute the R-Squared (Coefficient of Determination) Regression Score.

**Usage**

```
R2_Score(y_pred, y_true)
```

**Arguments**

y_pred	Estimated target values vector
y_true	Ground truth (correct) target values vector

**Value**

R<sup>2</sup> Score

**Examples**

```
data(cars)
reg <- lm(log(dist) ~ log(speed), data = cars)
R2_Score(y_pred = exp(reg$fitted.values), y_true = cars$dist)
```

---

RAE	<i>Relative Absolute Error Loss</i>
-----	-------------------------------------

---

**Description**

Compute the relative absolute error regression loss.

**Usage**

```
RAE(y_pred, y_true)
```

**Arguments**

y_pred	Estimated target values vector
y_true	Ground truth (correct) target values vector

**Value**

Relative Absolute Error Loss



**Examples**

```
data(cars)
reg <- lm(log(dist) ~ log(speed), data = cars)
RAE(y_pred = exp(reg$fitted.values), y_true = cars$dist)
```

---

Recall

*Recall*


---

**Description**

Compute the recall score.

**Usage**

```
Recall(y_true, y_pred, positive = NULL)
```

**Arguments**

<code>y_true</code>	Ground truth (correct) 0-1 labels vector
<code>y_pred</code>	Predicted labels vector, as returned by a classifier
<code>positive</code>	An optional character string for the factor level that corresponds to a "positive" result

**Value**

Recall

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
pred <- ifelse(logreg$fitted.values < 0.5, 0, 1)
Recall(y_pred = pred, y_true = mtcars$vs, positive = "0")
Recall(y_pred = pred, y_true = mtcars$vs, positive = "1")
```

---

RMSE

*Root Mean Square Error Loss*

---

**Description**

Compute the root mean squared error regression loss.

**Usage**

```
RMSE(y_pred, y_true)
```

**Arguments**

y_pred	Estimated target values vector
y_true	Ground truth (correct) target values vector

**Value**

Root Mean Square Error Loss

**Examples**

```
data(cars)
reg <- lm(log(dist) ~ log(speed), data = cars)
RMSE(y_pred = exp(reg$fitted.values), y_true = cars$dist)
```

---

RMSLE

*Root Mean Squared Logarithmic Error Loss*

---

**Description**

Compute the root mean squared logarithmic error regression loss.

**Usage**

```
RMSLE(y_pred, y_true)
```

**Arguments**

y_pred	Estimated target values vector
y_true	Ground truth (correct) target values vector

**Value**

Root Mean Squared Logarithmic Error Loss

**Examples**

```
data(cars)
reg <- lm(log(dist) ~ log(speed), data = cars)
RMSLE(y_pred = exp(reg$fitted.values), y_true = cars$dist)
```

---

**RMSPE***Root Mean Square Percentage Error Loss*

---

**Description**

Compute the root mean squared percentage error regression loss.

**Usage**

```
RMSPE(y_pred, y_true)
```

**Arguments**

y_pred	Estimated target values vector
y_true	Ground truth (correct) target values vector

**Value**

Root Mean Squared Percentage Error Loss

**Examples**

```
data(cars)
reg <- lm(log(dist) ~ log(speed), data = cars)
RMSPE(y_pred = exp(reg$fitted.values), y_true = cars$dist)
```

---

**RRSE***Root Relative Squared Error Loss*

---

**Description**

Compute the root relative squared error regression loss.

**Usage**

```
RRSE(y_pred, y_true)
```

**Arguments**

y_pred	Estimated target values vector
y_true	Ground truth (correct) target values vector

**Value**

Root Relative Squared Error Loss

**Examples**

```
data(cars)
reg <- lm(log(dist) ~ log(speed), data = cars)
RRSE(y_pred = exp(reg$fitted.values), y_true = cars$dist)
```

---

Sensitivity

*Sensitivity*

---

**Description**

Compute the sensitivity score.

**Usage**

```
Sensitivity(y_true, y_pred, positive = NULL)
```

**Arguments**

<code>y_true</code>	Ground truth (correct) 0-1 labels vector
<code>y_pred</code>	Predicted labels vector, as returned by a classifier
<code>positive</code>	An optional character string for the factor level that corresponds to a "positive" result

**Value**

Sensitivity

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
pred <- ifelse(logreg$fitted.values < 0.5, 0, 1)
Sensitivity(y_pred = pred, y_true = mtcars$vs, positive = "0")
Sensitivity(y_pred = pred, y_true = mtcars$vs, positive = "1")
```

---

Specificity	<i>Specificity</i>
-------------	--------------------

---

**Description**

Compute the specificity score.

**Usage**

```
Specificity(y_true, y_pred, positive = NULL)
```

**Arguments**

<code>y_true</code>	Ground truth (correct) 0-1 labels vector
<code>y_pred</code>	Predicted labels vector, as returned by a classifier
<code>positive</code>	An optional character string for the factor level that corresponds to a "positive" result

**Value**

Specificity

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
pred <- ifelse(logreg$fitted.values < 0.5, 0, 1)
Specificity(y_pred = pred, y_true = mtcars$vs, positive = "0")
Specificity(y_pred = pred, y_true = mtcars$vs, positive = "1")
```

---

ZeroOneLoss	<i>Normalized Zero-One Loss (Classification Error Loss)</i>
-------------	---

---

**Description**

Compute the normalized zero-one classification loss.

**Usage**

```
ZeroOneLoss(y_pred, y_true)
```

**Arguments**

<code>y_pred</code>	Predicted labels vector, as returned by a classifier
<code>y_true</code>	Ground truth (correct) 0-1 labels vector

**Value**

Zero-One Loss

**Examples**

```
data(cars)
logreg <- glm(formula = vs ~ hp + wt,
              family = binomial(link = "logit"), data = mtcars)
pred <- ifelse(logreg$fitted.values < 0.5, 0, 1)
ZeroOneLoss(y_pred = pred, y_true = mtcars$vs)
```

# Index

Accuracy, [2](#)  
Area\_Under\_Curve, [3](#)  
AUC, [4](#)  
  
ConfusionMatrix, [4](#)  
  
F1\_Score, [5](#)  
FBeta\_Score, [6](#)  
  
GainAUC, [6](#)  
Gini, [7](#)  
  
KS\_Stat, [8](#)  
  
LiftAUC, [8](#)  
LogLoss, [9](#)  
  
MAE, [9](#)  
MAPE, [10](#)  
MedianAE, [11](#)  
MedianAPE, [11](#)  
MLmetrics, [12](#)  
MLmetrics-package (MLmetrics), [12](#)  
MSE, [12](#)  
MultiLogLoss, [13](#)  
  
NormalizedGini, [13](#)  
  
Poisson\_LogLoss, [14](#)  
PRAUC, [14](#)  
Precision, [15](#)  
  
R2\_Score, [16](#)  
RAE, [16](#)  
Recall, [17](#)  
RMSE, [18](#)  
RMSLE, [18](#)  
RMSPE, [19](#)  
RRSE, [19](#)  
  
Sensitivity, [20](#)  
Specificity, [21](#)  
  
ZeroOneLoss, [21](#)