

# Package ‘fpROC’

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**Type** Package

**Title** Fast Partial Receiver Operating Characteristic (ROC) Test for Ecological Niche Modeling

**Version** 0.1.0

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**Description** Provides optimized 'C++' code for computing the partial Receiver Operating Characteristic (ROC) test used in niche and species distribution modeling. The implementation follows Peterson et al. (2008) <[doi:10.1016/j.ecolmodel.2007.11.008](https://doi.org/10.1016/j.ecolmodel.2007.11.008)>. Parallelization via 'OpenMP' was implemented with assistance from the 'DeepSeek' Artificial Intelligence Assistant (<<https://www.deepseek.com/>>).

**License** GPL-3

**Encoding** UTF-8

**NeedsCompilation** yes

**SystemRequirements** GNU make, OpenMP, TBB

**URL** <https://luismurao.github.io/fpROC/>

**BugReports** <https://github.com/luismurao/fpROC/issues>

**Imports** Rcpp (>= 1.0.7), stats, terra

**LinkingTo** Rcpp, RcppArmadillo

**RoxygenNote** 7.3.2

**Suggests** testthat (>= 3.0.0)

**Config/testthat/edition** 3

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|-------------|--|
| auc_metrics | <i>Calculate Partial and complete Area Under the Curve (AUC) Metrics</i> |
|-------------|--|

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Description

Computes partial AUC ratios between model predictions and random curves at a specified threshold, with options for sampling and iterations. Handles both numeric vectors and SpatRaster inputs.

Usage

```
auc_metrics(  
  test_prediction,  
  prediction,  
  threshold = 5,  
  sample_percentage = 50,  
  iterations = 500,  
  compute_full_auc = TRUE  
)
```

Arguments

|                   |  |
|-------------------|--|
| test_prediction   | Numeric vector of test prediction values (e.g., model outputs)           |
| prediction        | Numeric vector or SpatRaster object containing prediction values         |
| threshold         | Percentage threshold for partial AUC calculation (default = 5)           |
| sample_percentage | Percentage of test data to sample (default = 50)                         |
| iterations        | Number of iterations for estimating bootstrap statistics (default = 500) |
| compute_full_auc  | Logical. If TRUE, the complete AUC values will be computed               |

Details

Partial ROC is calculated following Peterson et al. (2008; [doi:10.1016/j.ecolmodel.2007.11.008](https://doi.org/10.1016/j.ecolmodel.2007.11.008)). The function calculates partial AUC ratios by:

1. Validating input types and completeness
2. Handling NA values and SpatRaster conversion

3. Checking for prediction variability
4. Computing AUC metrics using optimized C++ code

When prediction values have no variability (all equal), the function returns NA values with a warning.

### Value

A list containing:

- If input has no variability: List with NA values for AUC metrics
- Otherwise: Matrix of AUC results.

### References

Peterson, A.T. et al. (2008) Rethinking receiver operating characteristic analysis applications in ecological niche modeling. *Ecol. Modell.*, 213, 63–72.

### Examples

```
# With numeric vectors
test_data <- rnorm(100)
pred_data <- rnorm(100)
result <- fpROC::auc_metrics(test_prediction = test_data, prediction = pred_data)

# With SpatRaster
library(terra)
r <- terra::rast(ncol=10, nrow=10)
values(r) <- rnorm(terra::ncell(r))
result <- fpROC::auc_metrics(test_prediction = test_data, prediction = r)
```

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|              |   |
|--------------|---|
| auc_parallel | <i>Parallel AUC and partial AUC calculation with optimized memory usage</i> |
|--------------|---|

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### Description

Computes bootstrap estimates of partial and complete AUC using parallel processing and optimized binning.

### Usage

```
auc_parallel(
  test_prediction,
  prediction,
  threshold = 5,
  sample_percentage = 50,
  iterations = 500L,
```

```

    compute_full_auc = TRUE,
    n_bins = 500L
  )

```

## Arguments

|                   |  |
|-------------------|--|
| test_prediction   | Numeric vector of test prediction values                             |
| prediction        | Numeric vector of model predictions (background suitability data)    |
| threshold         | Percentage threshold for partial AUC calculation (default = 5.0)     |
| sample_percentage | Percentage of test data to sample in each iteration (default = 50.0) |
| iterations        | Number of bootstrap iterations (default = 500)                       |
| compute_full_auc  | Boolean indicating whether to compute complete AUC (default = TRUE)  |
| n_bins            | Number of bins for discretization (default = 500)                    |

## Details

This function implements a highly optimized AUC calculation pipeline: 1. Cleans input data (removes non-finite values) 2. Combines background and test predictions 3. Performs range-based binning (discretization) 4. Computes cumulative distribution of background predictions 5. Runs bootstrap iterations in parallel: - Samples test predictions - Computes sensitivity-specificity curves - Calculates partial and complete AUC

Key optimizations: - OpenMP parallelization for binning and bootstrap - Vectorized operations using Armadillo

## Value

A numeric matrix with 'iterations' rows and 4 columns containing:

- auc\_complete: Complete AUC (NA when compute\_full\_auc = FALSE)
- auc\_pmodel: Partial AUC for the model (sensitivity > 1 - threshold/100)
- auc\_prand: Partial AUC for random model (reference)
- ratio: Ratio of model AUC to random AUC (model/reference)

## Partial AUC

The partial AUC focuses on the high-sensitivity region defined by: Sensitivity > 1 - (threshold/100)

## See Also

[summarize\\_auc\\_results](#) for results processing, [trap\\_roc](#) for integration method

## Examples

```
# Basic usage with random data
set.seed(123)
bg_pred <- runif(1000) # bg predictions
test_pred <- runif(500) # Test predictions

# Compute only partial AUC metrics (500 iterations)
results <- auc_parallel(test_pred, bg_pred,
                        threshold = 5.0,
                        iterations = 100) # Reduced for example

# View first 5 iterations
head(results, 5)

# Summarize results (assume complete AUC was not computed)
summary <- summarize_auc_results(results, has_complete_auc = FALSE)

# Interpretation:
# - auc_pmodel: Model's partial AUC (higher is better)
# - auc_prand: Random model's partial AUC
# - ratio: Model AUC / Random AUC (>1 indicates better than random)

# Compute both partial and complete AUC
full_results <- auc_parallel(test_pred, bg_pred,
                             compute_full_auc = TRUE,
                             iterations = 100)
```

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`summarize_auc_results` *Summarize Bootstrap AUC Results*

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## Description

Computes aggregated statistics from bootstrap AUC iterations. This function processes the raw output of [auc\\_parallel](#) to produce meaningful summary metrics of the partial ROC test.

## Usage

```
summarize_auc_results(auc_results, has_complete_auc)
```

## Arguments

`auc_results`      Numeric matrix output from [auc\\_parallel](#) (dimensions: n\_iterations x 4)

`has_complete_auc`      Boolean indicating whether complete AUC was computed in the bootstrap iterations (affects first summary column)

## Details

This function: 1. Filters iterations with non-finite ratio values (handles bootstrap failures) 2. Computes means for each AUC metric across valid iterations 3. Calculates proportion of iterations where model outperforms random (ratio > 1). This way of computing the the p-value of the test.

Special handling: - Returns all NAs if no valid iterations exist - First column (complete AUC) depends on `has_complete_auc` parameter - Handles NaN/Inf values safely by filtering

## Value

A numeric matrix with 1 row and 5 columns containing:

- `mean_complete_auc`: Mean of complete AUC values (NA if not computed)
- `mean_pauc`: Mean of partial AUC values for the model
- `mean_pauc_rand`: Mean of partial AUC values for random model (reference)
- `mean_auc_ratio`: Mean of AUC ratios (model/random)
- `prop_ratio_gt1`: Proportion of iterations where ratio > 1 (performance better than random)

## Interpretation Guide

- `mean_auc_ratio > 1`: Model generally outperforms random predictions - `prop_ratio_gt1 = 1.9`: 90% - `mean_pauc`: Absolute performance measure (higher = better discrimination)

## See Also

[auc\\_parallel](#) for generating the input matrix

## Examples

```
# Basic usage with simulated results
set.seed(123)
# Simulate bootstrap output (100 iterations x 4 metrics)
auc_matrix <- cbind(
  complete = rnorm(100, 0.85, 0.05), # Complete AUC
  pmodel   = rnorm(100, 0.15, 0.03), # Partial model AUC
  prand     = rnorm(100, 0.08, 0.02), # Partial random AUC
  ratio     = rnorm(100, 1.9, 0.4)    # Ratio
)

# Summarize results (assuming complete AUC was computed)
summary <- summarize_auc_results(auc_matrix, has_complete_auc = TRUE)

# Typical output interpretation:
# - mean_complete_auc: 0.85 (good overall discrimination)
# - mean_pauc: 0.15 (absolute partial AUC)
# - mean_pauc_rand: 0.08 (random expectation)
# - mean_pAUCratio: 1.9 (model 90% better than random)
# - p_value: 0.98 (98% of iterations showed model > random)

# Real-world usage with actual AUC function output
```

```

# First run bootstrap AUC calculation
bg_pred <- runif(1000)
test_pred <- runif(500)
auc_output <- auc_parallel(
  test_prediction = test_pred,
  prediction = bg_pred,
  iterations = 100
)

# Then summarize results (complete AUC not computed in this case)
summary <- summarize_auc_results(auc_output, has_complete_auc = FALSE)

# Print summary statistics
colnames(summary) <- c("mean_complete_auc", "mean_pauc",
                      "mean_pauc_rand", "mean_pAUCratio", "p_value")
print(summary)

# Expected output structure:
#      mean_complete_auc mean_pauc mean_pauc_rand mean_pAUCratio  p_value
# [1,]                NA    0.152        0.083         1.83        0.94

```

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trap\_roc

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*Calculate Area Under Curve (AUC) using trapezoidal rule*


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## Description

Computes the area under a curve using the trapezoidal rule of numerical integration.

## Usage

```
trap_roc(x, y)
```

## Arguments

|   |  |
|---|--|
| x | Numeric vector (arma::vec) of x-coordinates (should be sorted in increasing order) |
| y | Numeric vector (arma::vec) of y-coordinates corresponding to x-coordinates         |

## Details

The trapezoidal rule approximates the area under the curve by dividing it into trapezoids. For each pair of adjacent points (x[i], y[i]) and (x[i+1], y[i+1]), it calculates the area of the trapezoid formed. The total AUC is the sum of all these individual trapezoid areas.

Special cases: - Returns 0 if there are fewer than 2 points (no area can be calculated) - Handles both increasing and decreasing x values (though typically x should be increasing for ROC curves)

**Value**

A numerical value representing the computed area under the curve as a double precision value.

**See Also**

[integrate](#) for R's built-in integration functions

**Examples**

```
# R code example:  
x <- c(0, 0.5, 1, 1.5, 2)  
y <- c(0, 0.7, 0.9, 0.95, 1)  
trap_roc(x, y)
```



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