

# Package: DirichletRF (via r-universe)

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**Type** Package

**Title** ``Dirichlet Random Forest''

**Version** 0.1.0

**Description** Implementation of the Dirichlet Random Forest algorithm for compositional response data. Supports maximum likelihood estimation ('MLE') and method-of-moments ('MOM') parameter estimation for the Dirichlet distribution. Provides two prediction strategies; averaging-based predictions (average of responses within terminal nodes) and parameter-based predictions (expected value derived from the estimated Dirichlet parameters within terminal nodes). For more details see Masoumifard, van der Westhuizen, and Gardner-Lubbe (2026, ISBN:9781032903910).

**License** GPL-3

**Encoding** UTF-8

**Imports** Rcpp (>= 1.0.0), parallel

**LinkingTo** Rcpp

**RoxygenNote** 7.3.3

**Suggests** testthat (>= 3.0.0)

**Config/testthat/edition** 3

**NeedsCompilation** yes

**Author** Khaled Masoumifard [aut, cre] (ORCID:

<<https://orcid.org/0000-0003-1921-2145>>), Stephan van der Westhuizen [aut] (ORCID:

<<https://orcid.org/0000-0001-9469-8427>>), Sugnet Lubbe [aut] (ORCID: <<https://orcid.org/0000-0003-2762-9944>>)

**Maintainer** Khaled Masoumifard <masoumifardk@yahoo.com>

**Repository** <https://cran.r-universe.dev>

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DirichletRF-package     *DirichletRF: Dirichlet Random Forest for Compositional Data*

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### Description

This package implements Dirichlet Random Forest for modeling and predicting compositional data using maximum likelihood estimation or method of moments.

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DirichletRF     *Build a Dirichlet Random Forest for Compositional Responses*

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### Description

Build a Dirichlet random forest for compositional responses. In compositional data analysis (CoDA), parts reside in the simplex, and this random forest ensures model output abide by CoDA principles. The implementation uses OpenMP for parallel tree building. Note that this implementation does not support out-of-bag (OOB) error estimation.

### Usage

```
DirichletRF(
  X,
  Y,
  num.trees = 100,
  max.depth = 10,
  min.node.size = 5,
  mtry = -1,
  seed = 123,
  est.method = "mom",
  num.cores = -1
)
```

**Arguments**

<code>X</code>	A numeric ( $n \times p$ ) matrix of covariates. Note that the current version only allows numeric covariates. Users may use one-hot encoding to possibly include categorical covariates.
<code>Y</code>	A numeric ( $n \times k$ ) matrix of compositional responses. Each row should sum to 1. That is, data should already be normalised if needed.
<code>num.trees</code>	Number of trees grown in the forest. Default is 100.
<code>max.depth</code>	Maximum depth of trees. Default is 10.
<code>min.node.size</code>	Minimum size of observations in each tree leaf. Default is 5. Note that nodes with sizes smaller than <code>min.node.size</code> can occur.
<code>mtry</code>	Number of covariates randomly selected as candidates at each split. Default is $\sqrt{p}$ , indicated by <code>-1</code> .
<code>seed</code>	The seed of the C++ random number generator.
<code>est.method</code>	Parameter estimation method for the Dirichlet distribution when splitting is done. Users may either use maximum likelihood (" <code>mle</code> ") or method of moments (" <code>mom</code> "). Default is " <code>mom</code> ".
<code>num.cores</code>	Number of OpenMP threads used for parallel tree building. The default is <code>-1</code> which uses all the cores on the system minus 1. Users may also specify <code>1</code> which means that the forest will be built sequentially.

**Details**

The forest provides two types of fitted values: mean-based predictions (derived from sample means at each leaf) and parameter-based predictions (derived from normalised Dirichlet alpha parameters).

**Value**

A list of the forest which contains the following:

`type` Parallelisation type used: "`openmp`" or "`sequential`".

`num.cores` Number of cores used.

`num.trees` Total number of trees in the forest.

`Y_train` Training data used (no OOB).

`fitted` A list of fitted values:

`alpha_hat` Estimated Dirichlet alpha parameters ( $n \times k$  matrix).

`mean_based` Mean-based fitted values ( $n \times k$  matrix).

`param_based` Parameter-based fitted values obtained by normalising `alpha_hat` ( $n \times k$  matrix).

`residuals` A list of residuals ( $Y - \text{fitted}$ ):

`mean_based` Residuals from mean-based predictions.

`param_based` Residuals from parameter-based predictions.

## References

Masoumifard, K., van der Westhuizen, S., & Gardner-Lubbe, S. (In press). Dirichlet-random forest for predicting compositional data. In A. Bekker, J. Ferreira, & P. Nagar (Eds.), *Environmental Statistics: Innovative Methods and Applications*. CRC Press.

## See Also

[predict.dirichlet\\_forest](#) for making predictions on new data.

## Examples

```
# Small toy example (auto-tested)
set.seed(42)
n <- 50; p <- 2
X <- matrix(rnorm(n * p), n, p)
G <- matrix(rgamma(n * 3, shape = rep(c(2, 3, 4), each = n)), n, 3)
Y <- G / rowSums(G)
forest <- DirichletRF(X, Y, num.trees = 5, num.cores = 1)
print(forest)
Xtest <- matrix(rnorm(5 * p), 5, p)
pred <- predict(forest, Xtest)

# Larger example
n <- 500; p <- 4
X <- matrix(rnorm(n * p), n, p)
alpha <- c(2, 3, 4)
G <- matrix(rgamma(n * length(alpha), shape = rep(alpha, each = n)),
           n, length(alpha))
Y <- G / rowSums(G)

forest1 <- DirichletRF(X, Y, num.trees = 100, num.cores = 1)
forest2 <- DirichletRF(X, Y, num.trees = 100)

alpha_hat <- forest1$fitted$alpha_hat
mean_fit <- forest1$fitted$mean_based
param_fit <- forest1$fitted$param_based
resid_mean <- forest1$residuals$mean_based
resid_param <- forest1$residuals$param_based

Xtest <- matrix(rnorm(10 * p), 10, p)
pred <- predict(forest1, Xtest)
param_pred <- pred$alpha_predictions / rowSums(pred$alpha_predictions)
```

**Description**

Makes predictions using a fitted `dirichlet_forest` object returned by [DirichletRF](#).

**Usage**

```
## S3 method for class 'dirichlet_forest'
predict(object, newdata, ...)
```

**Arguments**

<code>object</code>	A <code>dirichlet_forest</code> object.
<code>newdata</code>	A numeric matrix of new covariates ( $n_{\text{new}} \times p$ ).
<code>...</code>	Currently unused.

**Value**

A list with the following elements:

`alpha_predictions` Estimated Dirichlet alpha parameters for each new observation ( $n_{\text{new}} \times k$  matrix).

`mean_predictions` Mean-based compositional predictions ( $n_{\text{new}} \times k$  matrix).

**Examples**

```
# Small toy example (auto-tested)
set.seed(42)
n <- 50; p <- 2
X <- matrix(rnorm(n * p), n, p)
G <- matrix(rgamma(n * 3, shape = rep(c(2, 3, 4), each = n)), n, 3)
Y <- G / rowSums(G)
forest <- DirichletRF(X, Y, num.trees = 5, num.cores = 1)
Xtest <- matrix(rnorm(5 * p), 5, p)
pred <- predict(forest, Xtest)
pred$mean_predictions

n <- 500; p <- 4
X <- matrix(rnorm(n * p), n, p)
alpha <- c(2, 3, 4)
G <- matrix(rgamma(n * length(alpha), shape = rep(alpha, each = n)),
           n, length(alpha))
Y <- G / rowSums(G)
forest <- DirichletRF(X, Y, num.trees = 50, num.cores = 1)
Xtest <- matrix(rnorm(10 * p), 10, p)
pred <- predict(forest, Xtest)
param_pred <- pred$alpha_predictions / rowSums(pred$alpha_predictions)
single_pred <- predict(forest, Xtest[1, , drop = FALSE])
```

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`print.dirichlet_forest`*Custom Print Method for dirichlet\_forest Objects*

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**Description**

Suppresses the display of large data matrices (Y\_train, fitted, residuals) when the object is printed, while keeping them accessible via \$.

**Usage**

```
## S3 method for class 'dirichlet_forest'  
print(x, ...)
```

**Arguments**

x	A dirichlet_forest object.
...	Further arguments passed to or from other methods.

**Value**

Invisibly returns x, the dirichlet\_forest object unchanged. Called primarily for its side effect of printing a summary of the model to the console.

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