

# Package: BioMoR (via r-universe)

May 14, 2026

**Type** Package

**Title** Bioinformatics Modeling with Recursion and Autoencoder-Based Ensemble

**Version** 0.1.0

**Description** Provides tools for bioinformatics modeling using recursive transformer-inspired architectures, autoencoders, random forests, XGBoost, and stacked ensemble models. Includes utilities for cross-validation, calibration, benchmarking, and threshold optimization in predictive modeling workflows.

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**Encoding** UTF-8

**LazyData** false

**Roxygen** list(markdown = TRUE)

**RoxygenNote** 7.3.3

**Depends** R (>= 4.2.0)

**Imports** caret, recipes, themis, xgboost, magrittr, dplyr, pROC

**Suggests** randomForest, testthat (>= 3.0.0), PRROC, ggplot2, purrr, tibble, yardstick, knitr, rmarkdown

**VignetteBuilder** knitr

**Config/testthat/edition** 3

**Config/pak/sysreqs** make libicu-dev

**Repository** <https://sulkysubject37.r-universe.dev>

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biomor_benchmark	<i>Benchmark a trained model</i>
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### Description

Evaluates a trained caret model on test data, returning Accuracy, F1 score, and ROC-AUC. If only one class is present in the test set, ROC-AUC is returned as NA.

### Usage

```
biomor_benchmark(model, test_data, outcome_col)
```

### Arguments

model	A trained caret model
test_data	Dataframe containing predictors and outcome
outcome_col	Name of outcome column

### Value

A named list of metrics

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biomor_run_pipeline	<i>Run full BioMoR pipeline</i>
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**Description**

Run full BioMoR pipeline

**Usage**

```
biomor_run_pipeline(data, feature_cols = NULL, epochs = 50)
```

**Arguments**

data	dataframe with Label + descriptors
feature_cols	optional feature set
epochs	autoencoder epochs

**Value**

list of trained models + benchmark reports

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brier_score	<i>Compute Brier Score</i>
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**Description**

The Brier score is the mean squared error between predicted probabilities and the true binary outcome (0/1). Lower is better.

**Usage**

```
brier_score(y_true, y_prob, positive = "Active")
```

**Arguments**

y_true	True factor labels.
y_prob	Predicted probabilities for the positive class.
positive	Name of the positive class (default "Active").

**Value**

Numeric Brier score.

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calibrate_model	<i>Calibrate model probabilities</i>
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**Description**

Calibrate model probabilities

**Usage**

```
calibrate_model(model, test_data, method = "platt")
```

**Arguments**

model	caret or xgboost model
test_data	test dataframe
method	"platt" or "isotonic"

**Value**

calibrated probs

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compute_f1_threshold	<i>Compute optimal threshold for maximum F1 score</i>
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**Description**

Sweeps thresholds between 0 and 1 to find the one that maximizes F1.

**Usage**

```
compute_f1_threshold(y_true, y_prob, positive = "Active")
```

**Arguments**

y_true	True factor labels.
y_prob	Predicted probabilities for the positive class.
positive	Name of the positive class (default "Active").

**Value**

A list with elements:

**threshold** Best probability cutoff.

**best\_f1** Maximum F1 score achieved.

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get_cv_control	<i>Get caret cross-validation control</i>
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**Description**

Creates a `caret::trainControl` object for cross-validation, configured for two-class problems, ROC-based performance, and optional sampling strategies such as SMOTE or ROSE.

**Usage**

```
get_cv_control(cv = 5, sampling = NULL)
```

**Arguments**

cv	Number of folds (default 5).
sampling	Sampling method (e.g., "smote", "rose", or NULL).

**Value**

A `caret::trainControl` object.

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get_embeddings	<i>Get Embeddings from Autoencoder (stub)</i>
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**Description**

Placeholder for extracting embeddings from a trained autoencoder.

**Usage**

```
get_embeddings(ae_obj, data, feature_cols = NULL)
```

**Arguments**

ae_obj	Autoencoder object
data	Input data
feature_cols	Columns to use as features

**Value**

Matrix of embeddings (currently NULL since this is a stub)

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prepare_model_data	<i>Prepare dataset for modeling</i>
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### Description

Prepare dataset for modeling

### Usage

```
prepare_model_data(df, outcome_col = "Label")
```

### Arguments

df	A data.frame
outcome_col	Name of the outcome column

### Value

A processed data.frame with factor outcome

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train_autoencoder	<i>Train Autoencoder (stub)</i>
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### Description

Placeholder for future autoencoder integration in BioMoR.

### Usage

```
train_autoencoder(
  data,
  feature_cols = NULL,
  epochs = 10,
  batch_size = 32,
  lr = 0.001
)
```

### Arguments

data	Input data (matrix or data frame)
feature_cols	Columns to use as features
epochs	Number of training epochs
batch_size	Mini-batch size
lr	Learning rate

**Value**

A placeholder list with class "autoencoder"

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train_biomor	<i>Train BioMoR Autoencoder</i>
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**Description**

Train BioMoR Autoencoder

**Usage**

```
train_biomor(data, feature_cols, epochs = 100, batch_size = 50, lr = 0.001)
```

**Arguments**

data	Dataframe with numeric features + Label
feature_cols	Character vector of feature columns
epochs	Number of training epochs
batch_size	Batch size
lr	Learning rate

**Value**

list(model, dataset, embeddings)

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train_rf	<i>Train a Random Forest model with caret</i>
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**Description**

Train a Random Forest model with caret

**Usage**

```
train_rf(df, outcome_col = "Label", ctrl)
```

**Arguments**

df	A data.frame containing predictors and outcome
outcome_col	Name of the outcome column (binary factor)
ctrl	A caret::trainControl object

**Value**

A caret train object

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train_xgb_caret	<i>Train an XGBoost model with caret</i>
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**Description**

Train an XGBoost model with caret

**Usage**

```
train_xgb_caret(df, outcome_col = "Label", ctrl)
```

**Arguments**

df	A data.frame containing predictors and outcome
outcome_col	Name of the outcome column (binary factor)
ctrl	A caret::trainControl object

**Value**

A caret train object

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