

# Package ‘QuanDA’

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**Title** Quantile-Based Discriminant Analysis for High-Dimensional Imbalanced Classification

**Version** 1.0.0

**Description** Implements quantile-based discriminant analysis (QuanDA) for imbalanced classification in high-dimensional, low-sample-size settings. The method fits penalized quantile regression directly on discrete class labels and tunes the quantile level to reflect class imbalance.

**Depends** R (>= 3.5.0)

**Imports** hdqr, pROC, stats, methods

**License** GPL-2

**NeedsCompilation** yes

**RoxygenNote** 7.2.3

**Encoding** UTF-8

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## Contents

breast . . . . .	2
predict.quanda . . . . .	2
quanda . . . . .	3

## Index

5

breast	<i>Example breast cancer data</i>
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**Description**

A list containing predictor matrix X and binary response y.

**Usage**

```
data(breast)
```

**Value**

This data frame contains the following:

- |   |   |
|---|---|
| x | gene expression levels.                 |
| y | Disease state that is coded as 1 and -1 |

**Examples**

```
data(breast)
```

predict.quanda	<i>Make Predictions from a ‘quanda’ Object</i>
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**Description**

Produces fitted values for new predictor data using a fitted ‘quanda()‘ object.

**Usage**

```
## S3 method for class 'quanda'
predict(object, newx, type = c("class", "loss"), ...)
```

**Arguments**

- |        |   |
|--------|---|
| object | Fitted ‘quanda()‘ object from which predictions are to be derived.  |
| newx   | Matrix of new predictor values for which predictions are desired. This must be a matrix and is a required argument.                                       |
| type   | Type of prediction required. Type ““class”“ produces the predicted binary class labels and type ““loss”“ returns the fitted values. Default is ““class”“. |
| ...    | Not used.   |

**Value**

Numeric vector of length n\_new.

**See Also**[quanda](#)**Examples**

```
data(breast)
X <- as.matrix(X)
y <- as.numeric(as.character(y))
y[y== -1] = 0
fit <- quanda(X, y)
```

quanda

*Fit QuanDA for imbalanced binary classification***Description**

QuanDA fits a quantile-regression-based discriminant with label jittering. For each candidate quantile level  $\tau$ , the binary labels are jittered (adding  $U(0, 1)$ ), a penalized quantile regression is fit multiple times, and the coefficient vectors are averaged. The best  $\tau$  is selected by AUC.

**Usage**

```
quanda(
  x,
  y,
  lambda = 10^(seq(1, -4, length.out = 30)),
  lam2 = 0.01,
  n_rep = 10,
  tau_window = 0.05,
  nfolds = 5,
  maxit = 10000,
  eps = 1e-07,
  maxit_cv = 10000,
  eps_cv = 1e-05
)
```

**Arguments**

- |                     |   |
|---------------------|---|
| <code>x</code>      | A numeric matrix of predictors with $n$ rows (observations) and $p$ columns (features).   |
| <code>y</code>      | A binary response vector of length $n$ with values 0 or 1.  |
| <code>lambda</code> | Optional numeric vector of penalty values (largest <code>lambda[1]</code> ). If NULL, a default sequence will be generated from the data. |
| <code>lam2</code>   | Numeric, secondary penalty (ridge/elastic term) passed to <code>hdqr</code> . Default 0.01.   |
| <code>n_rep</code>  | Integer, number of jittering repetitions (averaged). Default 10.  |

<b>tau_window</b>	Width around the class rate to explore quantiles. Candidate $\tau$ are $b + \{-w, \dots, w\}$ in steps of 0.01, clipped to $[0, 1]$ , where $b$ is the class rate and $w$ is <b>tau_window</b> . Default 0.1.
<b>nfolds</b>	Integer, number of CV folds used by <b>cv_z()</b> . Default 5.
<b>maxit, maxit_cv, eps, eps_cv</b>	Controls for inner optimizers and CV helper.

## Details

We jitter labels via  $z_i = y_i + U_i$ , where  $U_i \sim \text{Unif}(0, 1)$ , fit penalized quantile regression at multiple  $\tau$ , average coefficients over **n\_rep** jitters, compute AUCs on the original  $(x, y)$ , and pick the  $\tau$  that maximizes AUC.

## Value

An object of class "quanda" with elements:

- beta** Numeric vector of length  $p + 1$  (intercept first).
- tau\_grid** Numeric vector of candidate  $\tau$  values.
- tau\_best** Chosen  $\tau$ .
- auc** Vector of AUCs across  $\tau$ .
- call** The matched call.

## Examples

```
data(breast)
X <- as.matrix(X)
y <- as.numeric(as.character(y))
y[y== -1] = 0
fit <- quanda(X, y)
pred <- predict(fit, tail(X))
```

# Index

- \* **binary-classification**

- quanda, [3](#)

- \* **datasets**

- breast, [2](#)

- \* **imbalanced-learning**

- quanda, [3](#)

- \* **quantile**

- quanda, [3](#)

- \* **regression**

- quanda, [3](#)

- breast, [2](#)

- `predict.quanda`, [2](#)

- quanda, [3](#), [3](#)

- `X (breast)`, [2](#)

- `y (breast)`, [2](#)