

# Package ‘distWorkshop’

November 4, 2019

**Title** Distributional analyses of linguistic data in the GAMM framework  
**Version** 1.0  
**Description** Data sets and code for distributional analyses of linguistic data in the GAMM framework.  
**Depends** R (>= 3.5.0), survival, mgcv, qgam, pammtools, RColorBrewer  
**Imports** survival, mgcv, qgam, pammtools, RColorBrewer  
**License** GNU General Public License v3.0  
**Encoding** UTF-8  
**LazyData** true  
**RoxygenNote** 6.1.1.9000

## R topics documented:

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| ld | <i>Lexical decision data</i> |
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### Description

Lexical decision data from the British Lexicon Project (Keuleers et al., 2012)

### Usage

ld

### Format

An object of class `data.frame` with 25401 rows and 6 columns.

**Source**

Keuleers, E., Lacey, P., Rastle, K., & Brysbaert, M. (2012). The British Lexicon Project: Lexical decision data for 28,730 monosyllabic and disyllabic English words. *Behavior Research Methods*, 44(1), 287-304.

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|-----|-------------------------|
| nam | <i>Word naming data</i> |
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**Description**

Word naming data from the English Lexicon Project (Balota et al., 2007)

**Usage**

nam

**Format**

An object of class `data.frame` with 37107 rows and 6 columns.

**Source**

Balota, D. A., Yap, M. J., Hutchison, K. A., Cortese, M. J., Kessler, B., Loftis, B., ... & Treiman, R. (2007). The English lexicon project. *Behavior Research Methods*, 39(3), 445-459.

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| plotPAM | <i>Plot PAM</i> |
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**Description**

Plot the results of a PAM model (Bender & Scheipl, 2018)

**Usage**

```
plotPAM(model, data, predictor = "logFrequency", response = "RT",
  se = 2, area = FALSE, num_grid = 100,
  pallet = colorRampPalette(rev(brewer.pal(n = 7, name =
    "RdYlBu")))(500), levs = NA, rugx = TRUE, rugy = TRUE, main = NA,
  xlab = NA, ylab = NA, ...)
```

**Arguments**

|           |   |
|-----------|---|
| model     | A PAM model.  |
| data      | The data the PAM model was fit to. Needs to include the response variable in the task, as well as all predictors in these models. Note: this is the data frame in its raw format, not the data frame converted to the piece-wise exponential data format. |
| predictor | The predictor to be plotted. This predictor needs to be present in the fitted model, as well as in data.  |

|          |  |
|----------|--|
| response | The name of the response variable in data.   |
| se       | The number of standard errors that is used for the significance test. Default: 2 (i.e., 95% confidence intervals)  |
| area     | Should the significance of the effect at different predictor values be plotted. Default: FALSE.                    |
| pallet   | A vector of color names that will be used for the contour plot.  |
| levs     | A vector of values at which the contour lines will be plotted. By default, these values are selected automatically |
| rugx     | Should a rug be plotted for the x-axis? Default: TRUE  |
| rugy     | Should a rug be plotted for the y-axis? Default: TRUE  |

## References

Bender, A. and Scheipl, F. (2018). pammttools: Piece-wise exponential additive mixed modeling tools. arXiv:1806.01042

## Examples

```
# Remove outliers
ld = removeOutliers(ld, predictors)
ld = na.omit(ld)

# Prepare data in exponential data format
ld$status = 1
cut_points = as.numeric(quantile(ld$RT[which(ld$RT <= 1085 &
  ld$RT >= 500)],seq(0, 1, by = 0.02)))
ped = split_data(Surv(RT, status)~., data = ld, id = "id",
  cut = cut_points)

# Run PAM (warning: computationally heavy)
pam_ld = gam(ped_status ~ s(tend) +
  s(logFrequency) + ti(tend, logFrequency) +
  s(Length) + ti(tend, Length) +
  s(logOLD20) + ti(tend, logOLD20) +
  s(SND20) + ti(tend, SND20),
  data = ped, offset = offset, family = poisson())

# Plot frequency effect
plotPAM(model = pam_ld, data = ld, predictor = "logFrequency")
```

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plotQGAMs

*Plot quantiles*


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## Description

Plot the results of a series of QGAM models (Fasiolo et al., 2017)

## Usage

```
plotQGAMs(models, predictor, data, cols = c("#3C7A5B", "#468E6A",
  "#50A379", "#5AB788", "#65CC98"), se = 2, xlab = NA, ylab = NA,
  main = NA, ylim = NA, ...)
```

### Arguments

|           |  |
|-----------|--|
| models    | A list of QGAM models as generated by the <code>mqqam()</code> function in the <code>qgam</code> package.                |
| predictor | The predictor to be plotted. This predictor needs to be present in the fitted models, as well as in data.                |
| data      | The data the QGAM models were fit to. Needs to include the response variable, as well as all predictors in these models. |
| cols      | A vector of colors. The lines corresponding to the quantiles will be plotted in these colors.                            |
| se        | The number of standard errors for the confidence intervals. Default: 2 (i.e., 95% confidence intervals)                  |

### References

- Fasiolo M., Goude Y., Nedellec R. and Wood S. N. (2017). Fast calibrated additive quantile regression. URL: <https://arxiv.org/abs/1707.03307>.
- Keuleers, E., Lacey, P., Rastle, K., & Brysbaert, M. (2012). The British Lexicon Project: Lexical decision data for 28,730 monosyllabic and disyllabic English words. *Behavior Research Methods*, 44(1), 287-304.

### Examples

```
# Remove outliers from the ld data set, which contains lexical
# decision latencies from the British Lexicon Project (Keuleers
# et al, 2012)
ld = removeOutliers(ld, predictors)
ld = na.omit(ld)

# Tune learning rate for median
tune = tuneLearnFast(RT ~ s(logFrequency) + s(Length) +
                     s(logOLD20) + s(SND20),
                     data = ld, qu = 0.5)
sigpar = tune$lsig

# Define quantiles
quants = c(0.10, 0.25, 0.50, 0.75, 0.90)

# Run qgam models
qgams = mqqam(RT ~ s(logFrequency) + s(Length) + s(logOLD20) +
              s(SND20),
              data = ld, qu = quants, lsig = sigpar)

# Plot effect of frequency at quantiles
plotQGAMs(qgams, "logFrequency", ld)
```

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pn

*Picture naming data*

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### Description

Picture naming data (Balota et al., 2007)

**Usage**

```
pn
```

**Format**

An object of class `data.frame` with 484 rows and 6 columns.

**Source**

Székely, A., D'amico, S., Devescovi, A., Federmeier, K., Herron, D., Iyer, G., ... & Bates, E. (2003). Timed picture naming: Extended norms and validation against previous studies. *Behavior Research Methods, Instruments, & Computers*, 35(4), 621-633.

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|-----------------------------|------------------------|
| <code>removeOutliers</code> | <i>Remove outliers</i> |
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**Description**

Remove outliers from a data frame

**Usage**

```
removeOutliers(data, predictors, sd = 3)
```

**Arguments**

|                         |   |
|-------------------------|---|
| <code>data</code>       | A data frame.   |
| <code>predictors</code> | A vector of the column names in <code>data</code> corresponding to the predictors that outliers should be removed for <code>sd</code> The number of standard deviations. Predictor values further than <code>sd</code> standard deviations from the predictor mean are removed. Default: 3. |
| <code>model</code>      | A PAM model.  |

**Examples**

```
# Load data for the lexical decision latencies in the British Lexicon Project (Keuleers et al, 2012)
data(ld)

# Remove outliers
ld = removeOutliers.fnc(ld, c("logFrequency", "Length", "logOLD20", "SND20"))
ld = na.omit(ld)
```

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