

Package ‘distGAMM’

May 13, 2021

Title Distributional analyses 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, pammttools, RColorBrewer, pec

Imports survival, mgcv, qgam, pammttools, RColorBrewer, pec

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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

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.

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

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.

plotBrier	<i>Plot Brier</i>
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Description

Plot the Brier score

Usage

```
plotBrier(
  pec,
  type = "BS",
  main = NA,
  xlab = NA,
  ylab = NA,
  ylim = NA,
  colors = c("#B33333", "#3333B3", "#33B333", "#B333B3", "#B37733"),
  lwd = 2,
  legend.names = NA,
  y.intersp = 1,
  ...
)
```

Arguments

pec	A pec object, as created by pec()
type	"BS" for Brier score (default), "IBS" for integrated Brier score

Examples

```

# Remove outliers
predictors = c("logFrequency", "Length", "logOLD20", "SND20")
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 = pamm(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)

# Generate prediction error curves
pec = pec(list(pam = pam_ld),
      Surv(RT, status) ~ 1,
      data = na.omit(ld),
      times = seq(510, 1080, length.out = 100),
      start = 510,
      exact = FALSE)

# Plot Brier score
plotBrier(pec)

# Plot integrated Brier score
plotBrier(pec, type = "IBS")

```

plotPAM

*Plot PAM***Description**

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

Usage

```

plotPAM(
  model,
  predictor,
  data,
  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.
predictor	The predictor to be plotted. This predictor needs to be present in the fitted model, as well as in data.
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.
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. & Scheipl, F. (2018). pammtools: Piece-wise exponential additive mixed modeling tools. arXiv:1806.01042

Examples

```

# Remove outliers
predictors = c("logFrequency", "Length", "logOLD20", "SND20")
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 = pamm(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)

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

```

plotQGAMs

*Plot quantiles***Description**

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

Usage

```

plotQGAMs(
  models,
  predictor,
  data,
  cols = c("#000080", "#1A1A9A", "#3333B3", "#4D4DCD", "#6666E6"),
  se = 2,
  xlab = NA,
  ylab = NA,
  main = NA,
  ylim = NA,
  ...
)

```

Arguments

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

References

Fasiolo M., Goude Y., Nedellec R., & 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)
predictors = c("RT", "logFrequency", "Length", "logOLD20", "SND20")
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 = mqgam(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)
```

pn	<i>Picture naming data</i>
----	----------------------------

Description

Picture naming data (Bates et al., 2003)

Usage

```
pn
```

Source

Székely, A., D'amico, S., Devescovi, A., Federmeier, K., Herron, D., Iyer, G., ... & Bates, E. (2003). Timed picture naming in seven languages. Bates, E., D'Amico, S., Jacobsen, T., Szekely, A., Andonova, E., et al. (2003). Psychonomic Bulletin & Review, 10(2), 344-380.

removeOutliers	<i>Remove outliers</i>
----------------	------------------------

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 data corresponding to the predictors that outliers should be removed for sd The number of standard deviations. Predictor values further than sd 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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