# Package 'fixedpointproperty'

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| Type Package  |
|---|
| <b>Title</b> Determine and Test the Fixed-Point Property in Binary Mixture Data   |
| Version 1.0   |
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| <b>Description</b> Determine and test the fixed-point property in binary mixture data.  This package was originally developed in the context of detecting mixture of cognitive processing strategies, based on observed response time distributions.  The method is explain in more detail by Van Maanen, De Jong, Van Rijn (2014) <doi:10.1371 journal.pone.0106113=""> and Van Maanen, Couto, Lebreton, (2016) <doi:10.1371 journal.pone.0167377="">.</doi:10.1371></doi:10.1371> |
| License GPL-2   |
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| R topics documented:  |
| fpAnova fpConditionCheck  fpDensDiff  fpGet  fpp-class  normMix  plot.fpp   |

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| fpAnova Function to perform ANOVA on fp objects |  |
|---|--|
|---|--|

# Description

This function computes Bayes Factors and p-values for within-subjects ANOVA designs, encoded as fp objects.

#### Usage

```
fpAnova(object, stat = "BF", na.rm = TRUE, check = TRUE)
```

# Arguments

object a list of objects from class fpp.

stat Either "BF" (default), "p", or "both", specificying what statistic to report.

na.rm Are NAs removed?

check Should the data be checked for suitability? A warning will be provided if a

check is failed.

#### **Details**

The function expects the output of fpGet, but in a list.

# Value

A list containing the results of either the Bayesian or frequentist analysis, or both:

BF The output of anovaBF p The output of summary.aov

#### Warning

If check=TRUE, then warnings will be provided if the data are not suitable for correct inferences.

#### Author(s)

Leendert van Maanen (l.vanmaanen@uu.nl)

### References

Van Maanen, L., De Jong, R., Van Rijn, H (2014). How to assess the existence of competing strategies in cognitive tasks: A primer on the fixed-point property. PLOS One, 9, e106113

Van Maanen, L. Couto, J. & Lebetron, M. (2016). Three boundary conditions for computing the fixed-point property in binary mixture data. PLOS One, 11, e0167377.

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#### See Also

fpGet, anovaBF, fpConditionCheck, summary.aov.

#### **Examples**

```
## generate data p \leftarrow c(.1,.5,.9) rt \leftarrow sapply(1:3, function(i) {rnormMix(10000, c(1,2), c(1,1), p[i])}) dat \leftarrow data.frame(rt=c(rt), cond=rep(1:3, each=10000), pp=rep(1:50, each=200, times=3)) ## compute the list of fpp objects res \leftarrow tapply(1:nrow(dat), dat$pp, function(X) {fpGet(dat[X,], 1000, bw=.75)}) ## call fpAnova, with stat="both" to do both a Bayesian and a frequentist test fpAnova(res, stat="both")
```

fpConditionCheck

Checks preconditions before performing fpANOVA

# Description

This function checks whether two conditions are met before performing fpANOVA.

#### **Usage**

```
fpConditionCheck(object)
```

#### **Arguments**

object

a list of objects from class fpp.

#### **Details**

Finding support for the fixed-point property will be mute if there is no significant difference between experimental conditions. Whether all conditions differ can be tested using fpConditionCheck1, which performs pairwise t-tests. A warning is provided if at least one paire of conditions does not significantly differ (default settings of pairwise.t.test are used).

Finding support for the fixed-point property is difficult if the bandwidth of the density estimation is chosen too small. In that case, multiple crossing points of pairs of densities will preclude a precise estimate of the fixed point. fpConditionCheck2 tests the number of crossing points for each pair of conditions, and provides a warning if more crossing points are detected.

#### Value

No return value, called for warnings generated by fpConditionCheck1 and fpConditionCheck2

#### Author(s)

Leendert van Maanen (l.vanmaanen@uu.nl)

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#### References

Van Maanen, L., De Jong, R., Van Rijn, H (2014). How to assess the existence of competing strategies in cognitive tasks: A primer on the fixed-point property. PLOS One, 9, e106113

Van Maanen, L. Couto, J. & Lebetron, M. (2016). Three boundary conditions for computing the fixed-point property in binary mixture data. PLOS One, 11, e0167377.

#### See Also

```
fpAnova, pairwise.t.test
```

#### **Examples**

```
N <- 200 # nr of observations per condition
M <- 50 # nr of participants
p \leftarrow seq(0.1, 0.9, 0.4) # mixture proportions
means <- c(0.3, 0.3) # means of base distributions are equal, yielding a warning if check=TRUE
sigma <- 5 # scale of base distributions
bw <- 0.01
# kernel bandwidth of the density estimation. Too small values yield a warning if check=TRUE
### generate data
rt <- NULL
for (i in 1:length(p)) {
    rt < c(rt, ifelse(sample(0:1, N * M, replace = TRUE, prob = c(p[i], 1 - p[i])),
        rnorm(N * M, means[1], sigma), rnorm(N * M, means[2], sigma)))
rt <- rt + rep(rnorm(M, sd = 0.1), times = N) # normally distributed pp random effect
dat <- data.frame(rt = rt, cond = rep(1:length(p), each = N * M), pp = rep(1:M,
    each = N)
### compute crossing points
res <- tapply(1:nrow(dat), dat$pp, function(X) {</pre>
    fpGet(dat[X, ], 1000, bw = bw)
})
### test fixed point
fpAnova(res, stat = "both", check=TRUE) # this provides both warnings
```

fpDensDiff

Function to compute the crossing point of two kernel-based distribution functions

#### **Description**

This function takes a list of fpp objects, and computes at which point the distributions cross each other.

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#### Usage

```
fpDensDiff(object)
```

#### **Arguments**

object a list of fpp objects

#### Value

An m \* n-array containing the crossing points of the pairs of distributions, with n the length of the list of fpp objects, and m the number of pairs of distributions.

#### Author(s)

Leendert van Maanen (l.vanmaanen@uu.nl)

#### References

Van Maanen, L., De Jong, R., Van Rijn, H (2014). How to assess the existence of competing strategies in cognitive tasks: A primer on the fixed-point property. PLOS One, 9, e106113

Van Maanen, L. Couto, J. & Lebetron, M. (2016). Three boundary conditions for computing the fixed-point property in binary mixture data. PLOS One, 11, e0167377.

#### See Also

fpGet,fpAnova

# **Examples**

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| fpGet | Function to compute densities and density differences for three or more data sets |
|-------|---|
|       |   |

# Description

This function computes standard kernel-based density functions for a response time data set with three or more conditions. In addition, it computes the pairwise differences for each pair of density functions.

# Usage

```
fpGet(dat, n = 512, bw = "nrd0")
```

# **Arguments**

| dat | n*2 dataframe or matrix with in col 1: RT (the values for which to compute the density); col 2: condition (an index) |
|-----|--|
| n   | the number of equally spaced points at which the density is to be estimated. See density for details.                |
| bw  | the smoothing bandwidth to be used. See density for details.   |

# Value

an object of class fpp, with the following components.

| dens | list of objects from class density   |
|------|--------------------------------------|
| diff | dataframe of the density differences |
| dat  | dataframe with the input data        |

#### Author(s)

Leendert van Maanen (l.vanmaanen@uu.nl)

#### References

Van Maanen, L., De Jong, R., Van Rijn, H (2014). How to assess the existence of competing strategies in cognitive tasks: A primer on the fixed-point property. PLOS One, 9, e106113

Van Maanen, L. Couto, J. & Lebetron, M. (2016). Three boundary conditions for computing the fixed-point property in binary mixture data. PLOS One, 11, e0167377.

#### See Also

density.

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#### **Examples**

```
### one data set or participant
## generate data
p <- c(.1,.5,.9)
rt <- sapply(1:3, function(i) {rnormMix(1000, c(1,2), c(1,1), p[i])})
dat <- data.frame(rt=c(rt), cond=rep(1:3, each=1000))

## compute one fp object
fpobject <- fpGet(dat, 1000, bw=.75)

### multiple participants
## generate data
p <- c(.1,.5,.9)
rt <- sapply(1:3, function(i) {rnormMix(10000, c(1,2), c(1,1), p[i])})
dat <- data.frame(rt=c(rt), cond=rep(1:3, each=10000), pp=rep(1:50, each=200, times=3))

## compute the list of fpp objects
res <- tapply(1:nrow(dat), dat$pp, function(X) {fpGet(dat[X,], 1000, bw=.75)})</pre>
```

fpp-class

Class "fpp"

#### **Description**

Objects of the fpp class are used by methods and functions of the fp package, to visualize and compute the fixed-point property in response time data.

#### **Objects from the Class**

Objects can be created by calls of the form new("fpp", ...).

#### **Slots**

```
dens: Object of class "array". This is an array of objects of class "density".
diff: Object of class "data.frame". This is a dataframe with the paired differences between
densities.
```

dat: Object of class "data.frame". This is a dataframe with the input data.

#### Methods

```
plot signature(x = "fpp"): ...
```

#### Author(s)

Leendert van Maanen (l.vanmaanen@uu.nl)

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#### References

Van Maanen, L., De Jong, R., Van Rijn, H (2014). How to assess the existence of competing strategies in cognitive tasks: A primer on the fixed-point property. PLOS One, 9, e106113

Van Maanen, L. Couto, J. & Lebetron, M. (2016). Three boundary conditions for computing the fixed-point property in binary mixture data. PLOS One, 11, e0167377.

# **Examples**

```
showClass("fpp")
```

normMix

Gaussian binary mixture distribution

# **Description**

Density, distribution, function, quantile function, and random generation for the mixture of two Gaussian distributions with mixture proportion p and 1-p.

#### Usage

```
dnormMix(x, mean=c(0,1), sd=c(1,1), p=1) pnormMix(x, mean=c(0,1), sd=c(1,1), p=1) qnormMix(x, mean=c(0,1), sd=c(1,1), p=1) rnormMix(n, mean=c(0,1), sd=c(1,1), p=1)
```

#### **Arguments**

x vector of quantiles or probabilities.

n number of observations.

mean vector of two means.

sd vector of standard deviations.

p mixture proportion of the first distribution (the second has proportion 1-p).

#### Value

dnormMix gives the density, pnormMix gives the distribution function, qnormMix gives the quantile function, and rnormMix generates random deviates.

#### Author(s)

Leendert van Maanen (l.vanmaanen@uu.nl)

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#### References

Van Maanen, L., De Jong, R., Van Rijn, H (2014). How to assess the existence of competing strategies in cognitive tasks: A primer on the fixed-point property. PLOS One, 9, e106113

Van Maanen, L. Couto, J. & Lebetron, M. (2016). Three boundary conditions for computing the fixed-point property in binary mixture data. PLOS One, 11, e0167377.

#### **Examples**

```
## the default value for p reduces this function to a normal distribution
dnormMix(1) == dnorm(1)

## plot that illustrates the fixed-point property
p <- c(.8,.5,.2)
m <- c(0,1); s <- c(1,2)
plot(function(X) {dnormMix(X, m, s, p[1])}, -5,10, ylab="Density",xlab="x", bty='L')
for (i in 2:3) {
   plot(function(X) {dnormMix(X, m, s, p[i])},-5,10, add=TRUE, lty=i, lwd=2)
}
legend("topright", legend=paste("p=",p,sep=''), lty=1:3, lwd=2, bty='n')</pre>
```

plot.fpp

Function to plot the distributions and differences of multiple binary mixture distributions.

#### **Description**

This function plots kernel-based densities as well as density differences for three or more data sets (fpp objects).

#### Usage

```
## S3 method for class 'fpp'
plot(x, ylab = c("Density", "Density difference"), xlim = NULL, ...)
```

#### **Arguments**

```
    x an fpp object
    ylab y-axis labels of the two plots
    xlim the x limits (x1, x2) of the two plots
    additional arguments to pass to both plot.density and matplot
```

#### **Details**

Generates two plots, one showing the estimated densities (based on plot.density) and one showing the density differences (based on matplot)

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# Value

No return value, called for side effects

#### Author(s)

Leendert van Maanen (l.vanmaanen@uu.nl)

#### References

Van Maanen, L., De Jong, R., Van Rijn, H (2014). How to assess the existence of competing strategies in cognitive tasks: A primer on the fixed-point property. PLOS One, 9, e106113

Van Maanen, L. Couto, J. & Lebetron, M. (2016). Three boundary conditions for computing the fixed-point property in binary mixture data. PLOS One, 11, e0167377.

# See Also

```
fpGet,fpDensDiff,plot.density,matplot
```

# **Examples**

```
## generate data
p <- c(.1,.5,.9)
rt <- sapply(1:3, function(i) {rnormMix(1000, c(1,2), c(1,1), p[i])})
dat <- data.frame(rt=c(rt), cond=rep(1:3, each=1000))

## compute fpp object
fpobject <- fpGet(dat, 1000, bw=.75)

## plot it
op <- par(mfrow=c(1,2))
plot(fpobject)
par(op)</pre>
```

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