## Package: BFpack (via r-universe)

October 12, 2024

```
Title Flexible Bayes Factor Testing of Scientific Expectations
Version 1.3.0
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Description Implementation of default Bayes factors for testing
     statistical hypotheses under various statistical models. The
     package is intended for applied quantitative researchers in the
     social and behavioral sciences, medical research, and related
     fields. The Bayes factor tests can be executed for statistical
     models such as univariate and multivariate normal linear
     models, correlation analysis, generalized linear models,
     special cases of linear mixed models, survival models,
     relational event models. Parameters that can be tested are
     location parameters (e.g., group means, regression
     coefficients), variances (e.g., group variances), and measures
     of association (e.g,.
     polychoric/polyserial/biserial/tetrachoric/product moments
     correlations), among others. The statistical underpinnings are
     described in Mulder and Xin (2019)
     <DOI:10.1080/00273171.2021.1904809>, Mulder and Gelissen (2019)
     <DOI:10.1080/02664763.2021.1992360>, Mulder (2016)
     <DOI:10.1016/j.jmp.2014.09.004>, Mulder and Fox (2019)
     <DOI:10.1214/18-BA1115>, Mulder and Fox (2013)
     <DOI:10.1007/s11222-011-9295-3>, Boeing-Messing, van Assen,
     Hofman, Hoijtink, and Mulder (2017) < DOI:10.1037/met0000116>,
     Hoijtink, Mulder, van Lissa, and Gu, (2018)
     <DOI:10.31234/osf.io/v3shc>, Gu, Mulder, and Hoijtink, (2018)
     <DOI:10.1111/bmsp.12110>, Hoijtink, Gu, and Mulder, (2018)
     <DOI:10.1111/bmsp.12145>, and Hoijtink, Gu, Mulder, and
     Rosseel, (2018) < DOI:10.1037/met0000187 >. When using the
     packages, please refer to Mulder et al. (2021)
     <DOI:10.18637/jss.v100.i18>.
License GPL (>= 3)
```

URL https://github.com/jomulder/BFpack

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BFpack-package

BFpack: Flexible Bayes factor testing of scientific expectations

#### **Description**

The R package **BFpack** provides tools for exploratory and confirmatory Bayesian hypothesis testing using Bayes factors and posterior probabilities under common statistical models. The main function 'BF' needs a fitted model 'x' as input argument. Depending on the class of the fitted model, a standard hypothesis test is executed by default. For example, if 'x' is a fitted regression model of class 'lm' then posterior probabilities are computed of whether each separate coefficient is zero, negative, or positive (assuming equal prior probabilities). If one has specific hypotheses with equality and/or order constraints on the parameters under the fitted model 'x' then these can be formulated using the 'hypothesis' argument (a character string), possibly together prior probabilities for the hypotheses via the 'prior' argument (default all hypotheses are equally likely a priori), and the 'complement' argument which is a logical stating whether the complement hypotheses should be included in the case ('TRUE' by default).

Use compilation for Fortran functions

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#### References

Mulder, J., D.R. Williams, Gu, X., A. Tomarken, F. Böing-Messing, J.A.O.C. Olsson-Collentine, Marlyne Meyerink, J. Menke, J.-P. Fox, Y. Rosseel, E.J. Wagenmakers, H. Hoijtink., and van Lissa, C. (submitted). BFpack: Flexible Bayes Factor Testing of Scientific Theories in R. https://arxiv.org/abs/1911.07728

Mulder, J., van Lissa, C., Gu, X., Olsson-Collentine, A., Boeing-Messing, F., Williams, D. R., Fox, J.-P., Menke, J., et al. (2019). BFpack: Flexible Bayes Factor Testing of Scientific Expectations. (Version 0.2.1) https://CRAN.R-project.org/package=BFpack

#### See Also

Useful links:

- https://github.com/jomulder/BFpack
- Report bugs at https://github.com/jomulder/BFpack/issues

## **Examples**

```
## Not run:
# EXAMPLE 1. One-sample t test
ttest1 <- t_test(therapeutic, mu = 5)</pre>
print(ttest1)
# confirmatory Bayesian one sample t test
BF1 <- BF(ttest1, hypothesis = "mu = 5")
summary(BF1)
# exploratory Bayesian one sample t test
BF(ttest1)
# EXAMPLE 2. ANOVA
aov1 <- aov(price ~ anchor * motivation,data = tvprices)</pre>
BF1 <- BF(aov1, hypothesis = "anchorrounded = motivationlow;
                               anchorrounded < motivationlow")</pre>
summary(BF1)
# EXAMPLE 3. Logistic regression
fit <- glm(sent ~ ztrust + zfWHR + zAfro + glasses + attract + maturity +
   tattoos, family = binomial(), data = wilson)
BF1 <- BF(fit, hypothesis = "ztrust > zfWHR > 0;
                              ztrust > 0 & zfWHR = 0")
summary(BF1)
```

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actors

Actors from a small hypothetical network

## **Description**

The related data files 'events', 'same\_location', 'same\_culture' contain information on the event sequence and the two event statistics respectively.

#### Usage

```
data(actors)
```

#### Format

dataframe (25 rows, 4 columns)

actors\$idintegerID of the employee, corresponding to the sender and receiver IDs in the events dataframeactors\$locationnumericLocation of the actor, ranging from 1-4actors\$culturecharacterCategorical variable, indicating the culture of the employee

attention

Multiple Sources of Attentional Dysfunction in Adults With Tourette's Syndrome

## **Description**

Data from a psychological study comparing attentional performances of Tourette's syndrome (TS) patients, ADHD patients, and controls. These data were simulated using the sufficient statistics from Silverstein, Como, Palumbo, West, and Osborn (1995).

#### Usage

```
data(attention)
```

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#### **Format**

A data frame with 51 rows and 2 columns.

#### **Details**

```
accuracy numeric Participant's accuracy in the attentional task
group factor Participant's group membership (TS patient, ADHD patient, or control)
```

#### References

Silverstein, S. M., Como, P. G., Palumbo, D. R., West, L. L., & Osborn, L. M. (1995). Multiple sources of attentional dysfunction in adults with Tourette's syndrome: Comparison with attention deficit-hyperactivity disorder. Neuropsychology, 9(2), 157-164. doi:10.1037/0894-4105.9.2.157

bartlett\_test

Bartlett Test of Homogeneity of Variances

## Description

Performs Bartlett's test of the null that the variances in each of the groups (samples) are the same.

## Usage

```
bartlett_test(x, g, ...)
## Default S3 method:
bartlett_test(x, g, ...)
```

## Arguments

X	a numeric vector of data values, or a list of numeric data vectors representing the
	respective samples, or fitted linear model objects (inheriting from class "lm").

g a vector or factor object giving the group for the corresponding elements of x. Ignored if x is a list.

... further arguments to be passed to or from methods.

#### **Details**

x must be a numeric data vector, and g must be a vector or factor object of the same length as x giving the group for the corresponding elements of x.

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#### Value

A list with class "bartlett\_htest" containing the following components:

statistic Bartlett's K-squared test statistic.

parameter the degrees of freedom of the approximate chi-squared distribution of the test

statistic.

p. value the p-value of the test.

conf.int a confidence interval for the mean appropriate to the specified alternative hy-

pothesis.

method the character string "Bartlett test of homogeneity of variances".

data.name a character string giving the names of the data.

vars the sample variances across groups (samples).

n the number of observations per group (sample)

#### Bain t test

In order to allow users to enjoy the functionality of bain with the familiar stats-function bartlett.test, we have had to make minor changes to the function bartlett.test.default. All rights to, and credit for, the function bartlett.test.default belong to the R Core Team, as indicated in the original license below. We make no claims to copyright and incur no liability with regard to the changes implemented in bartlett\_test.

This the original copyright notice by the R core team: File src/library/stats/R/bartlett\_test.R Part of the R package, https://www.R-project.org

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A copy of the GNU General Public License is available at https://www.R-project.org/Licenses/

#### References

Bartlett, M. S. (1937). Properties of sufficiency and statistical tests. Proceedings of the Royal Society of London Series A 160, 268–282. DOI: 10.1098/rspa.1937.0109.

#### **Examples**

```
require(graphics)
plot(count ~ spray, data = InsectSprays)
bartlett_test(InsectSprays$count, InsectSprays$spray)
```

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BF.default

Bayes factors for Bayesian exploratory and confirmatory hypothesis testing

## **Description**

The BF function can be used for hypothesis testing and model selection using the Bayes factor. By default exploratory hypothesis tests are performed of whether each model parameter equals zero, is negative, or is positive. Confirmatory hypothesis tests can be executed by specifying hypotheses with equality and/or order constraints on the parameters of interest. Depending on the class of the fitted model different Bayes factors are used as described in the output.

## Usage

```
## Default S3 method:
BF(
  hypothesis = NULL,
  prior.hyp.explo = NULL,
  prior.hyp.conf = NULL,
  prior.hyp = NULL,
  complement = TRUE,
  log = FALSE,
  Sigma,
  n,
)
## S3 method for class 'lm'
BF(
  Х,
  hypothesis = NULL,
  prior.hyp.explo = NULL,
  prior.hyp.conf = NULL,
  prior.hyp = NULL,
  complement = TRUE,
  log = FALSE,
  BF.type = 2,
  iter = 10000,
)
## S3 method for class 't_test'
BF(
  х,
  hypothesis = NULL,
  prior.hyp.explo = NULL,
```

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```
prior.hyp.conf = NULL,
  prior.hyp = NULL,
  complement = TRUE,
  log = FALSE,
 BF.type = 2,
  iter = 1e+06,
)
```

#### **Arguments**

Х

An R object containing the outcome of a statistical analysis. An R object containing the outcome of a statistical analysis. Currently, the following objects can be processed: t\_test(), bartlett\_test(), lm(), aov(), manova(), cor\_test(), lmer() (only for testing random intercep variances), glm(), coxph(), survreg(), polr(), zeroinfl(), rma(), ergm(), bergm(), or named vector objects. In the case x is a named vector, the arguments Sigma and n are also needed. See vignettes for elaborations.

hypothesis

A character string containing the constrained (informative) hypotheses to evaluate in a confirmatory test. The default is NULL, which will result in standard exploratory testing under the model x.

prior.hyp.explo

The prior probabilities of the hypotheses in the exploratory tests. Except for objects of class aov (for (M)ANOVA, etc.), this argument should be a vector with three elements reflecting the prior probability of a zero effect, a negative effect, and a positive effect, respectively. For objects of class aov, the argument should be a list where the first element should be a vector of length 3 specifying the prior probabilities of each parameter being zero, negative, or positive, the second element should be a vector of length 2 specifying the prior probabilities of a model where is no main effect for a factor and the full model, and the third element should be a vector of length 2 specifying the prior probabilities of a model where is no interaction effect (if present) for two factors and the full model. The default (NULL) specifies equal prior probabilities for each hypothesis per exploratory test.

prior.hyp.conf The prior probabilities of the constrained hypotheses in the confirmatory test.

prior.hyp Deprecated. Please use the argument prior.hyp.conf.

complement a logical specifying whether the complement should be added to the tested hypothesis under hypothesis.

log a logical specifying whether the Bayes factors should be computed on a log

scale. Default is FALSE.

Sigma An approximate posterior covariance matrix (e.g., error covariance matrix) of

the parameters of interest. This argument is only required when x is a named

vector.

The (effective) sample size that was used to acquire the estimates in the named vector x and the error covariance matrix Sigma. This argument is only required

when x is a named vector.

n

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... Parameters passed to and from other functions.

BF. type An integer that specified the type of Bayes factor (or prior) that is used for the

test. Currently, this argument is only used for models of class 'lm' and 't\_test', where BF.type=2 implies an adjusted fractional Bayes factor with a 'fractional prior mean' at the null value (Mulder, 2014), and BF.type=1 implies a regular fractional Bayes factor (based on O'Hagan (1995)) with a 'fractional prior mean'

at the MLE.

iter Number of iterations that are used to compute the Monte Carlo estimates (only

used for certain hypotheses under multivariate models and when testing group

variances).

#### **Details**

The function requires a fitted modeling object. Current analyses that are supported: t\_test, bartlett\_test, aov, manova, lm, mlm, glm, hetcor, lmer, coxph, survreg, ergm, bergm, zeroinfl, rma and polr.

For testing parameters from the results of t\_test(), lm(), aov(), manova(), and bartlett\_test(), hypothesis testing is done using adjusted fractional Bayes factors are computed (using minimal fractions). For testing measures of association (e.g., correlations) via cor\_test(), Bayes factors are computed using joint uniform priors under the correlation matrices. For testing intraclass correlations (random intercept variances) via lmer(), Bayes factors are computed using uniform priors for the intraclass correlations. For all other tests, approximate adjusted fractional Bayes factors (with minimal fractions) are computed using Gaussian approximations, similar as a classical Wald test.

#### Value

The output is an object of class BF. The object has elements:

- BFtu\_exploratory: The Bayes factors of the constrained hypotheses against the unconstrained hypothesis in the exploratory test.
- BFtu\_main (only for any objects with predictors of class factor): The Bayes factors of a constrained model where all levels of a factor are assumed to have the same effect on the outcome variable versus an unconstrained (full) model with no constraints.
- BFtu\_interaction (only for aov objects with interaction effects with predictors of class factor): The Bayes factors of a constrained model where the effect of the dummy variables corresponding to an interaction effects are assumed to be zero versus an unconstrained (full) model with no constraints.
- PHP\_exploratory: The posterior probabilities of the constrained hypotheses in the exploratory
  test
- PHP\_main (only for aov objects with predictors of class factor): The posterior probabilities a constrained model where all levels of a factor are assumed to have the same effect on the outcome variable versus an unconstrained (full) model with no constraints.
- PHP\_interaction (only for aov objects with interaction effects with predictors of class factor): The posterior probabilities of a constrained model where the effect of the dummy variables corresponding to an interaction effects are assumed to be zero versus an unconstrained (full) model with no constraints.

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• BFtu\_confirmatory: The Bayes factors of the constrained hypotheses against the unconstrained hypothesis in the confirmatory test using the hypothesis argument.

- PHP\_confirmatory: The posterior probabilities of the constrained hypotheses in the confirmatory test using the hypothesis argument.
- BFmatrix\_confirmatory: The evidence matrix which contains the Bayes factors between all possible pairs of hypotheses in the confirmatory test.
- BFtable\_confirmatory: The Specification table (output when printing the summary of a BF for a confirmatory test) which contains the different elements of the extended Savage Dickey density ratio where
  - The first column 'complex=' quantifies the relative complexity of the equality constraints of a hypothesis (the prior density at the equality constraints in the extended Savage Dickey density ratio).
  - The second column 'complex>' quantifies the relative complexity of the order constraints
    of a hypothesis (the prior probability of the order constraints in the extended Savage
    Dickey density ratio).
  - The third column 'fit=' quantifies the relative fit of the equality constraints of a hypothesis (the posterior density at the equality constraints in the extended Savage Dickey density ratio).
  - The fourth column 'fit>' quantifies the relative fit of the order constraints of a hypothesis (the posterior probability of the order constraints in the extended Savage Dickey density ratio)
  - The fifth column 'BF=' contains the Bayes factor of the equality constraints against the unconstrained hypothesis.
  - The sixth column 'BF>' contains the Bayes factor of the order constraints against the unconstrained hypothesis.
  - The seventh column 'BF' contains the Bayes factor of the constrained hypothesis against the unconstrained hypothesis.
  - The eighth column 'PHP' contains the posterior probabilities of the hypotheses.
- prior.hyp.explo: The prior probabilities of the constrained hypotheses in the exploratory tests
- prior.hyp.conf: The prior probabilities of the constrained hypotheses in the confirmatory test.
- hypotheses: The tested constrained hypotheses in a confirmatory test.
- estimates: The unconstrained estimates.
- model: The input model x.
- bayesfactor: The type of Bayes factor that is used for this model.
- parameter: The type of parameter that is tested.
- log: logical whether the Bayes factors were reported on a log scale.
- fraction\_number\_groupIDs (only for objects of class lm): The number of 'group identifiers' that were identified based on the number of unique combinations of levels of predictor variables of class factor in the data. These group identifiers are used to automatically specify the minimal fractions that are used to compute (adjusted) fractional Bayes factors.

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fraction\_groupID\_observations (only for objects of class lm): A vector that specifies to
which 'group identifier' an observation belongs. The group identifiers are constructed based
on the unique combination of the levels based on the predictor variables of class factor of
the observations.

• call: The call of the BF function.

#### Methods (by class)

- BF(default): S3 method for a named vector 'x'
- BF(1m): S3 method for an object of class 'lm'
- BF(t\_test): BF S3 method for an object of class 't\_test'

#### References

Mulder, J., D.R. Williams, Gu, X., A. Tomarken, F. Böing-Messing, J.A.O.C. Olsson-Collentine, Marlyne Meyerink, J. Menke, J.-P. Fox, Y. Rosseel, E.J. Wagenmakers, H. Hoijtink., and van Lissa, C. (2021). BFpack: Flexible Bayes Factor Testing of Scientific Theories in R. Journal of Statistical Software. <a href="https://doi.org/10.18637/jss.v100.i18">https://doi.org/10.18637/jss.v100.i18</a>>

#### **Examples**

```
# EXAMPLE 1. One-sample t test
ttest1 <- t_test(therapeutic, mu = 5)</pre>
print(ttest1)
# confirmatory Bayesian one sample t test
BF1 <- BF(ttest1, hypothesis = "mu = 5")
summary(BF1)
# exploratory Bayesian one sample t test
BF(ttest1)
# EXAMPLE 2. ANOVA
aov1 <- aov(price ~ anchor * motivation,data = tvprices)</pre>
BF1 <- BF(aov1, hypothesis = "anchorrounded = motivationlow;
                               anchorrounded < motivationlow")</pre>
summary(BF1)
# EXAMPLE 3. linear regression
lm1 \leftarrow lm(mpg \sim cyl + hp + wt, data = mtcars)
BF(lm1, hypothesis = "wt < cyl < hp = 0")
# EXAMPLE 4. Logistic regression
fit <- glm(sent ~ ztrust + zfWHR + zAfro + glasses + attract + maturity +
   tattoos, family = binomial(), data = wilson)
BF1 <- BF(fit, hypothesis = "ztrust > zfWHR > 0;
                              ztrust > 0 & zfWHR = 0")
summary(BF1)
# EXAMPLE 5. Correlation analysis
set.seed(123)
cor1 <- cor_test(memory[1:20,c(1,2,6)])
BF1 <- BF(cor1)
```

cor\_test 13

```
summary(BF1)
BF2 <- BF(cor1, hypothesis = "Rat_with_Im > Rat_with_Del > 0;
                              Rat_with_Im = Rat_with_Del = 0")
summary(BF2)
# correlations can also be computed between continuous/ordinal variables
memory_test \leftarrow memory[1:20,c(1,2,6)]
memory_test[,3] <- as.ordered(memory_test[,3])</pre>
cor2 <- cor_test(memory_test)</pre>
BF(cor2)
# EXAMPLE 6. Bayes factor testing on a named vector
# A Poisson regression model is used to illustrate the computation
# of Bayes factors with a named vector as input
poisson1 <- glm(formula = breaks ~ wool + tension,</pre>
  data = datasets::warpbreaks, family = poisson)
# extract estimates, error covariance matrix, and sample size:
estimates <- poisson1$coefficients</pre>
covmatrix <- vcov(poisson1)</pre>
samplesize <- nobs(poisson1)</pre>
# compute Bayes factors on equal/order constrained hypotheses on coefficients
BF1 <- BF(estimates, Sigma = covmatrix, n = samplesize, hypothesis =
"woolB > tensionM > tensionH; woolB = tensionM = tensionH")
summary(BF1)
```

cor\_test

Bayesian correlation analysis

#### **Description**

Estimate the unconstrained posterior for the correlations using a joint uniform prior.

## Usage

```
cor_test(..., formula = NULL, iter = 5000, burnin = 3000, nugget.scale = 0.995)
```

#### **Arguments**

formula

... matrices (or data frames) of dimensions *n* (observations) by *p* (variables) for different groups (in case of multiple matrices or data frames).

an object of class formula. This allows for including control variables in the

model (e.g., ~ education).

iter number of iterations from posterior (default is 5000).

burnin number of iterations for burnin (default is 3000).

nugget.scale a scalar which serves to avoid violations of positive definite correlation matrices.

It should be very close to 1 (the default is .995).

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#### Value

list of class cor\_test:

- meanF posterior means of Fisher transform correlations
- covmF posterior covariance matrix of Fisher transformed correlations
- correstimates posterior estimates of correlation coefficients
- corrdraws list of posterior draws of correlation matrices per group
- corrnames names of all correlations

## **Examples**

```
# Bayesian correlation analysis of the 6 variables in 'memory' object
# we consider a correlation analysis of the first three variable of the memory data.
#fit <- cor_test(BFpack::memory[,1:3])

# Bayesian correlation of variables in memory object in BFpack while controlling
# for the Cat variable
#fit <- cor_test(BFpack::memory[,c(1:4)],formula = ~ Cat)

# Example of Bayesian estimation of polyserial correlations
#memory_example <- memory[,c("Im","Rat")]
#memory_example$Rat <- as.ordered(memory_example$Rat)
#fit <- cor_test(memory_example)

# Bayesian correlation analysis of first three variables in memory data
# for two different groups
#HC <- subset(BFpack::memory[,c(1:3,7)], Group == "HC")[,-4]
#SZ <- subset(BFpack::memory[,c(1:3,7)], Group == "SZ")[,-4]
#fit <- cor_test(HC,SZ)</pre>
```

Fcor

Student t approximations of Fisher transformed correlations

#### **Description**

Approximated degrees of freedom and approximated scale of the Fisher transformed correlations depending on the dimension of the vector of dependent variables P based on a joint uniform prior.

#### Usage

```
data(Fcor)
```

#### Format

A data frame with 3 columns.

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## **Details**

nu	numeric	Approximated degrees of freedom
sigma	numeric	Approximated scale
P	integer	Dimension of vector of dependent variables

fmri fMRI data

## Description

fMRI data assessing relation between individual differences in the ability to recognize faces and cars and thickness of the superficial, middle, and deep layers of the fusiform face area, as assessed by high-resolution fMRI recognition (Williams et al, 2019, under review)

## Usage

data(fmri)

#### **Format**

A data.frame with 13 rows and 6 columns.

#### **Details**

Subject	numeric	Particicpant ID number
Face	numeric	Standardized score on face recognition battery
Vehicle	numeric	Standardized score on vehicle recognition battery
Superficial	numeric	Depth in mm of superficial layer of FFA
Middle	numeric	Depth in mm of middle layer of FFA
Bform	numeric	Depth in mm of deep layer of FFA

## References

McGuigin, R.W., Newton, A.T., Tamber-Rosenau, B., Tomarken, A.J, & Gauthier, I. (under review). Thickness of deep layers in the fusiform face area predicts face recognition.

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memory	Memory data on health and schizophrenic patients
--------	--

## Description

Data set from study assessing differences between schizophrenic patients and healthy control participants in patterns of correlations among 6 verbal memory tasks (Ichinose et al., 2019).

lm	numeric	Percent correct on immediate recall of 3 word lists
Del	numeric	Percent correct on delayed recall of 3 word lists
Wmn	numeric	Number correct on letter-number span test of auditory working memory
Cat	numeric	Number correct on category fluency task
Fas	numeric	Number correct on letter fluency task
Rat	numeric	Number correct on remote associates task
Group	factor	Participant Group (HC = Healthy Control; SZ = Schizophrenia)

## Usage

```
data(memory)
```

## **Format**

A data frame with 40 rows and 8 columns.

#### References

Ichinose, M.C., Han, G., Polyn, S., Park, S., & Tomarken, A.J. (2019). Verbal memory performance discordance in schizophrenia: A reflection of cognitive dysconnectivity. Unpublished manuscript.

	M. I	
mvt_test	Multivariate Student t test	

## **Description**

First step to performs a Bayesian multivariate one sample Student t test using the (adjusted) fractional Bayes factor using the BF() function.

## Usage

```
mvt_test(X, Y, null = NULL, paired = FALSE, ...)
```

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## **Arguments**

X a data matrix with the variables in the columns.

Y an optional data matrix with the variables in the columns.

null a vector of the null values of the variables.

paired a logical indicating whether you want a multivariate paired t-test.

... further arguments to be passed to or from methods.

#### **Details**

X must be a data matrix and null must be a vector of the assumed null values of the variables.

#### Value

An object that can be applied to the BF().

#### References

Mulder, J. and Gu, X. (2023). Bayesian Testing of Scientific Expectations under Multivariate Normal Linear Models. Multivariate Behavioral Research, 57, 767-783. DOI: 10.1080/00273171.2021.1904809.

## **Examples**

```
mvt_fmri <- mvt_test(fmri[,1:2],null = c(0,0))
BF(mvt_fmri)

# the same test can be executed via the lm() function
intercept <- rep(1,nrow(fmri))
lm1 <- lm(cbind(Face,Vehicle) ~ -1 + intercept, data=fmri)
BF(lm1,hypothesis="intercept_on_Face=intercept_on_Vehicle=0")</pre>
```

relevents

A sequence of innovation-related e-mail messages

## **Description**

A time-ordered sequence of 247 communication messages between 25 actors.

## Usage

```
data(relevents)
```

#### **Format**

```
dataframe (247 rows, 3 columns)
```

same\_location

relevents\$time	numeric	Time of the e-mail message, in seconds since onset of the observation
relevents\$sender	integer	ID of the sender, corresponding to the employee IDs in the actors dataframe
relevents\$receiver	integer	ID of the receiver

#### **Details**

The related data files 'actors', 'same\_location', 'same\_culture' contain information on the actors and three event statistics respectively.

## Description

A matrix coding whether senders of events (in the rows) and receivers of events (in the column) have the background culture. Related to the 'events' data object, that contains a relational event sequence, and the 'actors' object, that contains information on the 25 actors involved in the relational event sequence.

#### Usage

```
data(same_culture)
```

#### **Format**

```
dataframe (25 rows, 4 columns)
```

same\_culture integer Event statistic. Matrix with senders in the rows and receivers in the columns. The event statistic is

me_location Same location event statistic
---

## Description

A matrix coding whether senders of events (in the rows) and receivers of events (in the column) have the same location. Related to the 'events' data object, that contains a relational event sequence, and the 'actors' object, that contains information on the 25 actors involved in the relational event sequence.

#### Usage

```
data(same_location)
```

sivan 19

#### **Format**

dataframe (25 rows, 4 columns)

same\_location integer Event statistic. Matrix with senders in the rows and receivers in the columns. The event statistic is

sivan Wason task performance and morality

## **Description**

Data from an experimental study, using the Wason selection task (Wason 1968) to examine whether humans have cognitive adaptations for detecting violations of rules in multiple moral domains. Moral domains are operationalized in terms of the five domains of the Moral Foundations Questionnaire (Graham et al. 2011). These data were simulated using the R-package synthpop, based on the characteristics of the original data.

#### Usage

data(sivan)

#### **Format**

A data.frame with 887 rows and 12 columns.

factor

#### **Details**

sex	factor	Participant sex
age	integer	Participant age
nationality	factor	Participant nationality
politics	integer	How would you define your political opinions? Likert type scale, from 1 (Liberal) to 6 (Conservation)
WasonOrder	factor	Was the Wason task presented before, or after the MFQ?
Harm	numeric	MFQ harm domain.
Fairness	numeric	MFQ fairness domain.
Loyalty	numeric	MFQ loyalty domain.
Purity	numeric	MFQ purity domain.
Tasktype	ordered	How was the Wason task framed?

Did the participant give the correct answer to the Wason task?

#### References

GotRight

Sivan, J., Curry, O. S., & Van Lissa, C. J. (2018). Excavating the Foundations: Cognitive Adaptations for Multiple Moral Domains. Evolutionary Psychological Science, 4(4), 408–419. doi:10.1007/s40806-018-0154-8

20 timssICC

therapeutic

Data come from an experimental study (Rosa, Rosa, Sarner, and Barrett, 1998) that were also used in Howell (2012, p.196). An experiment was conducted to investigate if Therapeutic Touch practitioners who were blindfolded can effectively identify which of their hands is below the experimenter; s. Twenty-eight practitioners were involved and tested 10 times in the experiment. Researchers expected an average of 5 correct answers from each practitioner as it is the number by chance if they do not outperform others.

## **Description**

correct integer How many correct answers are from each practitioner)

#### Usage

data(therapeutic)

#### **Format**

A data frame with 22 rows and 1 column.

#### References

Howell, D. (2012). Statistical methods for psychology (8th ed.). Belmont, CA: Cengage Learning.

timssICC

Trends in International Mathematics and Science Study (TIMSS) 2011-2015

## **Description**

A stratified sample was drawn by country and school to obtain a balanced sample of p = 15 grade-4 students per school for each of four countries (The Netherlands (NL), Croatia (HR), Germany (DE), and Denmark (DK)) and two measurement occasions (2011, 2015). Achievement scores (first plausible value) of overall mathematics were considered. Performances of fourth and eight graders from more than 50 participating countries around the world can be found at (https://www.iea.nl/timss) The TIMSS achievement scale is centered at 500 and the standard deviation is equal to 100 scale score points. The TIMSS data set has a three-level structure, where students are nested within classrooms/schools, and the classrooms/schools are nested within countries. Only one classroom was sampled per school. Changes in the mathematics achievement can be investigated by examining

the grouping of students in schools across countries. Changes in country-specific intraclass correlation coefficient from 2011 to 2015, representing heterogeneity in mathematic achievements within and between schools across years, can be tested. When detecting a decrease in average performance together with an increase of the intraclass correlation, a subset of schools performed worse. For a constant intraclass correlation across years the drop in performance applied to the entire population of schools. For different countries, changes in the intraclass correlation across years can be tested concurrently to examine also differences across countries.

#### Usage

data(timssICC)

#### **Format**

A data.frame with 16770 rows and 15 columns.

#### **Details**

math	numeric	math score child
groupNL11	numeric	Indicator for child from NL in 2011
groupNL15	numeric	Indicator for child from NL in 2015
groupHR11	numeric	Indicator for child from HR in 2011
groupHR15	numeric	Indicator for child from HR in 2015
groupDE11	numeric	Indicator for child from DE in 2011
groupDE15	numeric	Indicator for child from DE in 2015
groupDR11	numeric	Indicator for child from DK in 2011
groupDR15	numeric	Indicator for child from DK in 2015
gender	numeric	Female=0,Male=1
weight	numeric	Child sampling weight
yeargender	numeric	Interaction for occassion and gender
lln	numeric	total number of children in school-class
groupschool	factor	Nested indicator for school in country
schoolID	factor	Unique indicator for school

#### References

Mulder, J. & Fox, J.-P. (2019). Bayes factor testing of multiple intraclass correlations. Bayesian Analysis. 14, 2, p. 521-552.

22 wilson

tvprices

Precision of the Anchor Influences the Amount of Adjustment

## **Description**

Data from an experimental study where participants have to guess the price of a plasma tv. There were two experimental conditions. These data were simulated using the sufficient statistics from Janiszewski & Uy (2008).

#### Usage

data(tvprices)

#### **Format**

A data.frame with 59 rows and 3 columns.

#### **Details**

price	numeric	Participant z-scores of price
anchor	factor	Participant anchor
motivation	factor	motivation to change

#### References

Janiszewski, C., & Uy, D. (2008). Precision of the anchor influences the amount of adjustment. Psychological Science, 19(2), 121–127. doi:10.1111/j.1467-9280.2008.02057.x

wilson

Facial trustworthiness and criminal sentencing

## **Description**

Data from a correlational study in which the correlation between ratings of facial trustworthiness of inmates was correlated with whether they had received the death penalty or not (wilson and Rule, 2015). These data were simulated using the R-package synthpop, based on the characteristics of the original data.

#### Usage

data(wilson)

wilson 23

## **Format**

A data.frame with 742 rows and 13 columns.

## **Details**

stiı	m	integer	Stimulus Number
sei	nt	integer	Sentence: $1 = Death$ , $0 = Life$
ra	ce	integer	Race: 1 = White, -1 = Black
gla	asses	integer	Glasses: $1 = \text{Yes}, 0 = \text{No}$
tat	ttoos	integer	Tattoos: $1 = \text{Yes}, 0 = \text{No}$
ztı	rust	numeric	Trustworthiness
trı	ust_2nd	numeric	Trustworthiness ratings with 2nd control group; Death targets are same as in primary analysis, Life
afi	ro	numeric	raw Afrocentricity ratings.
zA	fro	numeric	Afrocentricity ratings normalized within target race. Analyses in paper were done with this variable
att	tract	numeric	Attractiveness
fV	VHR	numeric	facial width-to-height
af	WHR	numeric	fWHR normalized within target race. Analyses in paper were done with this variable
ma	aturity	numeric	Maturity

## References

Wilson, J. P., & Rule, N. O. (2015). Facial Trustworthiness Predicts Extreme Criminal-Sentencing Outcomes. Psychological Science, 26(8), 1325–1331. doi: 10.1177/0956797615590992

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