# Package: nullabor (via r-universe)

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Description Tools for visual inference. Generate null data sets and null plots using permutation and simulation. Calculate distance metrics for a lineup, and examine the distributions of metrics. Title Tools for Graphical Inference Maintainer Di Cook <dicook@monash.edu> License GPL (>= 2) URL https://github.com/dicook/nullabor BugReports https://github.com/dicook/nullabor/issues Imports MASS, moments, fpc, ggplot2, dplyr, purrr, tidyr, tibble, magrittr Suggests forecast, viridis, knitr LazyData true Type Package LazyLoad false VignetteBuilder knitr RoxygenNote 7.3.1 **Encoding** UTF-8 Repository https://dicook.r-universe.dev RemoteUrl https://github.com/dicook/nullabor RemoteRef HEAD RemoteSha 91c02cf38b75b036f4a325f2b11ab8e9a2e38f5b

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aud

Conversion rate of 1 Australian Doller (AUD) to 1 US Dollar

# Description

The dataset consists of the daily exchange rates of 1 Australian Dollar to 1 US Dollar between Jan 9 2018 and Feb 21 2018.

bin\_dist

#### Description

Data X is binned into X.bin bins in x-direction and Y.bins in y-direction. The number of points in each cell is then counted. Same is done for data PX. An euclidean distance is calculated between the number of points in each cell between X and PX.

# Usage

```
bin_dist(X, PX, lineup.dat = lineup.dat, X.bin = 5, Y.bin = 5)
```

# Arguments

Х	a data.frame with two variables, the first two columns are used
PX	another data.frame with two variables, the first two columns are used
lineup.dat	lineup data so that the binning is done based on the lineup data and not the individual plots, by default lineup.dat = lineup.dat ; if one wishes to calculate the binned distance between two plots, one should use lineup.dat = NULL
X.bin	number of bins on the x-direction, by default $X.bin = 5$
Y.bin	number of bins on the y-direction, by default Y.bin = $5$

## Value

distance between X and PX

#### Examples

```
with(mtcars, bin_dist(data.frame(wt, mpg), data.frame(sample(wt), mpg),
lineup.dat = NULL))
```

box\_dist

Distance based on side by side Boxplots

#### Description

Assuming that data set X consists of a categorical group variable a numeric value, a summary of the first quartile, median and third quartile of this value is calculated for each group. The extent (as absolute difference) of the minimum and maximum value across groups is computed for first quartile, median and third quartile. Same is done for data PX. Finally an euclidean distance is calculated between the absolute differences of X and PX.

#### Usage

box\_dist(X, PX)

### Arguments

Х	a data.frame with one factor variable and one continuous variable
PX	a data.frame with one factor variable and one continuous variable

#### Value

distance between X and PX

# Examples

```
if(require('dplyr')) {
  with(mtcars,
    box_dist(data.frame(as.factor(am), mpg),
    data.frame(as.factor(sample(am)), mpg))
  )
}
```

calc_diff	Calculating the difference between true plot and the null plot with the
	maximum distance.

## Description

Distance metric is used to calculate the mean distance between the true plot and all the null plots in a lineup. The difference between the mean distance of the true plot and the maximum mean distance of the null plots is calculated.

# Usage

```
calc_diff(lineup.dat, var, met, pos, dist.arg = NULL, m = 20)
```

#### Arguments

lineup.dat	lineup data to get the lineup
var	a vector of names of the variables to be used to calculate the difference
met	distance metric needed to calculate the distance as a character
pos	position of the true plot in the lineup
dist.arg	a list or vector of inputs for the distance metric met; NULL by default
m	number of plots in the lineup, by default $m = 20$

## Value

difference between the mean distance of the true plot and the maximum mean distance of the null plots

#### calc\_mean\_dist

#### Examples

```
if(require('dplyr')){
lineup.dat <- lineup(null_permute('mpg'), mtcars, pos = 1)
calc_diff(lineup.dat, var = c('mpg', 'wt'), met = 'bin_dist',
dist.arg = list(lineup.dat = lineup.dat, X.bin = 5, Y.bin = 5), pos = 1, m = 8)}
if(require('dplyr')){
calc_diff(lineup(null_permute('mpg'), mtcars, pos = 1), var = c('mpg', 'wt'), met = 'reg_dist',
dist.arg = NULL, pos = 1, m = 8)}</pre>
```

calc\_mean\_dist Calculating the mean distances of each plot in the lineup.

#### Description

Distance metric is used to calculate the mean distance between the true plot and all the null plots in a lineup. The mean distances of each null plot to all the other null plots are calculated. The mean distances are returned for all the plots in the lineup.

#### Usage

```
calc_mean_dist(lineup.dat, var, met, pos, dist.arg = NULL, m = 20)
```

#### Arguments

lineup.dat	lineup data of the lineup
var	a vector of names of the variables to be used to calculate the mean distances
met	distance metric needed to calculate the distance as a character
pos	position of the true plot in the lineup
dist.arg	a list or vector of inputs for the distance metric met; NULL by default
m	number of plots in the lineup, by default $m = 20$

#### Value

the mean distances of each plot in the lineup

```
if(require('dplyr')){
  calc_mean_dist(lineup(null_permute('mpg'), mtcars, pos = 1), var = c('mpg', 'wt'),
  met = 'reg_dist', pos = 1, m = 10)}
```

decrypt

#### Description

The real data position is encrypted by the lineup function, and writes this out as a text string. Decrypt, decrypts this text string to reveal which where the real data is.

#### Usage

```
decrypt(...)
```

# Arguments

... character vector to decrypt

# Examples

```
decrypt('0uXR2p rut L202')
```

distmet

Empirical distribution of the distance

#### Description

The empirical distribution of the distance measures is calculated based on the mean distance of each of the null plots from the other null plots in a lineup. At this moment this method works only for null\_permute method. This function helps get some assessment of whether the actual data plot is very different from the null plots.

# Usage

```
distmet(
  lineup.dat,
  var,
  met,
  method,
  pos,
  repl = 1000,
  dist.arg = NULL,
  m = 20
)
```

#### distmet

#### Arguments

lineup.dat	lineup data
var	a vector of names of the variables to be used
met	distance metric needed to calculate the distance as a character
method	method for generating null data sets
pos	position of the observed data in the lineup
repl	number of sets of null plots selected to obtain the distribution; 1000 by default
dist.arg	a list or vector of inputs for the distance metric met; NULL by default
m	the number of plots in the lineup; $m = 20$ by default

# Value

lineup has the data used for the calculations

null\_values contains new null samples from which to compare nulls in lineup

diff difference in distance between nulls and actual data and that of the null that is most different from other nulls. A negative value means that the actual data plot is similar to the null plots.

closest list of the five closest nulls to the actual data plot

pos position of the actual data plot in the lineup

```
# Each of these examples uses a small number of nulls (m=8), and a small number of
# repeated sampling from the null distribution (repl=100), to make it faster to run.
# In your own examples you should think about increasing each of these, at least to the defaults.
## Not run:
if (require('dplyr')) {
  d <- lineup(null_permute('mpg'), mtcars, pos = 1)</pre>
  dd <- distmet(d, var = c('mpg', 'wt'),</pre>
    'reg_dist', null_permute('mpg'), pos = 1, repl = 100, m = 8)
  distplot(dd, m=8)
}
## End(Not run)
## Not run:
d <- lineup(null_permute('mpg'), mtcars, pos=4, n=8)</pre>
library(ggplot2)
ggplot(d, aes(mpg, wt)) + geom_point() + facet_wrap(~ .sample, ncol=4)
if (require('dplyr')) {
  dd <- distmet(d, var = c('mpg', 'wt'), 'bin_dist', null_permute('mpg'),</pre>
    pos = 4, repl = 100, dist.arg = list(lineup.dat = d, X.bin = 5,
    Y.bin = 5), m = 8)
  distplot(dd, m=8)
}
## End(Not run)
```

```
# Example using bin_dist
## Not run:
if (require('dplyr')) {
  d <- lineup(null_permute('mpg'), mtcars, pos = 1)</pre>
  library(ggplot2)
  ggplot(d, aes(mpg, wt)) + geom_point() + facet_wrap(~ .sample, ncol=5)
  dd <- distmet(d, var = c('mpg', 'wt'),</pre>
    'bin_dist', null_permute('mpg'), pos = 1, repl = 500,
    dist.arg = list(lineup.dat = d, X.bin = 5, Y.bin = 5))
  distplot(dd)
}
## End(Not run)
# Example using uni_dist
## Not run:
mod <- lm(wt ~ mpg, data = mtcars)</pre>
resid.dat <- data.frame(residual = mod$resid)</pre>
d <- lineup(null_dist('residual', dist = 'normal'), resid.dat, pos=19)</pre>
ggplot(d, aes(residual)) + geom_histogram(binwidth = 0.25) + facet_wrap(~ .sample, ncol=5)
if (require('dplyr')) {
  dd <- distmet(d, var = 'residual', 'uni_dist', null_dist('residual',</pre>
    dist = 'normal'), pos = 19, repl = 500)
  distplot(dd)
}
```

```
## End(Not run)
```

distplot

Plotting the distribution of the distance measure

#### Description

The permutation distribution of the distance measure is plotted with the distances for the null plots. Distance measure values for the null plots and the true plot are overlaid.

#### Usage

distplot(dat, m = 20)

#### Arguments

dat	output from distmet
m	the number of plots in the lineup; $m = 20$ by default

8

# electoral

#### Examples

```
## Not run:
if (require('dplyr')) {
    d <- lineup(null_permute('mpg'), mtcars, pos = 1)
    library(ggplot2)
    ggplot(d, aes(mpg, wt)) + geom_point() + facet_wrap(~.sample)
    distplot(distmet(d, var = c('mpg', 'wt'), 'reg_dist', null_permute('mpg'),
        pos = 1, repl = 100, m = 8), m = 8)
}
## End(Not run)
```

electoral

Polls and election results from the 2012 US Election

#### Description

Polls and election results from the 2012 US Election

#### Format

A list with two data frames: polls is a data frame of 51 rows and 4 variables

State State name

Electoral.vote Number of electoral votes in the 2012 election

**Margin** Margin between the parties with the highest number of votes and second highest number of votes. These margins are based on polls.

Democrat logical vector True, if the democratic party is the majority party in this state.

'election' is a data frame of 51 rows and 5 variables

State State name

Candidate character string of the winner: Romney or Obama

Electoral.vote Number of electoral votes in the 2012 election

**Margin** Margin between the parties with the highest number of votes and second highest number of votes. These margins are based on the actual election outcome

Democrat logical vector True, if the democratic party is the majority party in this state.

lal

Los Angeles Lakers play-by-play data.

## Description

Play by play data from all games played by the Los Angeles lakers in the 2008/2009 season.

lineup

# Description

In this protocol the plot of the real data is embedded amongst a field of plots of data generated to be consistent with some null hypothesis. If the observe can pick the real data as different from the others, this lends weight to the statistical significance of the structure in the plot. The protocol is described in Buja, Cook, Hofmann, Lawrence, Lee, Swayne, Wickham (2009) Statistical inference for exploratory data analysis and model diagnostics, Phil. Trans. R. Soc. A, 367, 4361-4383.

#### Usage

```
lineup(method, true = NULL, n = 20, pos = sample(n, 1), samples = NULL)
```

#### Arguments

method	method for generating null data sets
true	true data set. If NULL, find_plot_data will attempt to extract it from the current ggplot2 plot.
n	total number of samples to generate (including true data)
pos	position of true data. Leave missing to pick position at random. Encryped position will be printed on the command line, decrypt to understand.
samples	samples generated under the null hypothesis. Only specify this if you don't want lineup to generate the data for you.

#### Details

Generate n - 1 null datasets and randomly position the true data. If you pick the real data as being noticeably different, then you have formally established that it is different to with p-value 1/n.

```
library(ggplot2)
ggplot(lineup(null_permute('mpg'), mtcars), aes(mpg, wt)) +
  geom_point() +
  facet_wrap(~ .sample)
ggplot(lineup(null_permute('cyl'), mtcars),
        aes(mpg, .sample, colour = factor(cyl))) +
        geom_point()
```

null\_dist

# Description

Null hypothesis: variable has specified distribution

## Usage

```
null_dist(var, dist, params = NULL)
```

#### Arguments

var	variable name
dist	distribution name. One of: beta, cauchy, chisq, exp, f, gamma, geom, lnorm, logis, nbinom, binom, norm, pois, t, unif, weibull
params	list of parameters of distribution. If NULL, will use fitdistr to estimate them.

#### Value

a function that given data generates a null data set. For use with lineup or rorschach

#### See Also

null\_permute, null\_lm

```
dframe <- data.frame(x = rnorm(150))</pre>
library(ggplot2)
# three histograms of normally distributed values
ggplot(
  data=rorschach(method=null_dist("x", "norm"), n = 3, true=dframe)
  ) +
  geom_histogram(aes(x=x, y=..density..), binwidth=0.25) +
  facet_grid(.~.sample) +
  geom_density(aes(x=x), colour="steelblue", size=1)
# uniform distributions are not as easy to recognize as such
dframe$x = runif(150)
ggplot(
  data=rorschach(method=null_dist("x", "uniform",
                 params=list(min=0, max=1)),
  n = 3, true=dframe)) +
  geom_histogram(aes(x=x, y=..density..), binwidth=0.1) +
  facet_grid(.~.sample) +
  geom_density(aes(x=x), colour="steelblue", size=1)
```

null\_lm

# Description

Null hypothesis: variable is linear combination of predictors

# Usage

```
null_lm(f, method = "rotate", ...)
```

# Arguments

f	model specification formula, as defined by lm
method	method for generating null residuals. Built in methods 'rotate', 'pboot' and 'boot' are defined by resid_rotate, resid_pboot and resid_boot respectively
	other arguments passedd onto method.

## Value

a function that given data generates a null data set. For use with lineup or rorschach

# See Also

null\_permute, null\_dist

```
data(tips)
x <- lm(tip ~ total_bill, data = tips)
tips.reg <- data.frame(tips, .resid = residuals(x), .fitted = fitted(x))
library(ggplot2)
ggplot(lineup(null_lm(tip ~ total_bill, method = 'rotate'), tips.reg)) +
geom_point(aes(x = total_bill, y = .resid)) +
facet_wrap(~ .sample)</pre>
```

null\_permute

# Description

Null hypothesis: variable is independent of others

#### Usage

```
null_permute(var)
```

### Arguments

var

name of variable to permute

## Value

a function that given data generates a null data set. For use with lineup or rorschach

#### See Also

null\_lm, null\_dist

# Examples

```
data(mtcars)
library(ggplot2)
ggplot(data=rorschach(method=null_permute("mpg"), n = 3, true=mtcars)) +
geom_boxplot(aes(x=factor(cyl), y=mpg, fill=factor(cyl))) +facet_grid(.~.sample) +
theme(legend.position="none", aspect.ratio=1)
```

null\_ts

Generate null data by simulating from a time series model.

#### Description

Null hypothesis: data follows a time series model using auto.arima from the forecast package

#### Usage

null\_ts(var, modelfn)

var	variable to model as a time series
modelfn	method for simulating from ts model.

#### Value

a function that given data generates a null data set. For use with lineup or rorschach

#### See Also

null\_model

## Examples

```
require(forecast)
require(ggplot2)
require(dplyr)
data(aud)
l <- lineup(null_ts("rate", auto.arima), aud)
ggplot(l, aes(x=date, y=rate)) + geom_line() +
facet_wrap(~.sample, scales="free_y") +
theme(axis.text = element_blank()) +
xlab("") + ylab("")
l_dif <- 1 %>%
group_by(.sample) %>%
mutate(d=c(NA,diff(rate))) %>%
ggplot(aes(x=d)) + geom_density() +
facet_wrap(~.sample)
```

opt\_bin\_diff

Finds the number of bins in x and y direction which gives the maximum binned distance.

## Description

This function finds the optimal number of bins in both x and y direction which should be used to calculate the binned distance. The binned distance is calculated for each combination of provided choices of number of bins in x and y direction and finds the difference using calc\_diff for each combination. The combination for which the difference is maximum should be used.

#### Usage

```
opt_bin_diff(
    lineup.dat,
    var,
    xlow,
    xhigh,
    ylow,
    yhigh,
    pos,
    plot = FALSE,
    m = 20
)
```

#### pvisual

#### Arguments

lineup.dat	lineup data to get the lineup
var	a list of names of the variables to be used to calculate the difference
xlow	the lowest value of number of bins on the x-direction
xhigh	the highest value of number of bins on the x-direction
ylow	the lowest value of number of bins on the y-direction
yhigh	the highest value of number of bins on the y-direction
pos	position of the true plot in the lineup
plot	LOGICAL; if true, returns a tile plot for the combinations of number of bins with the differences as weights
m	number of plots in the lineup, by default $m = 20$

#### Value

a dataframe with the number of bins and differences the maximum mean distance of the null plots

# Examples

```
if(require('dplyr')){
  opt_bin_diff(lineup(null_permute('mpg'), mtcars, pos = 1), var = c('mpg', 'wt'),
  2, 5, 4, 8, pos = 1, plot = TRUE, m = 8)
}
```

pvisual

P-value calculations.

#### Description

These set of functions allow the user to calculate a p-value from the lineup after it has been evaluated by K independent observers. The different functions accommodate different lineup construction and showing to observers. Details are in the papers Majumder et al (2012) JASA, and Hofmann et al (2015). We distinguish between three different scenarios:

- Scenario I: in each of K evaluations a different data set and a different set of (m-1) null plots is shown.
- Scenario II: in each of K evaluations the same data set but a different set of (m-1) null plots is shown.
- Scenario III: the same lineup, i.e. same data and same set of null plots, is shown to K different observers.

# Usage

```
pvisual(
    x,
    K,
    m = 20,
    N = 10000,
    type = "scenario3",
    xp = 1,
    target = 1,
    upper.tail = TRUE
)
```

## Arguments

x	number of observed picks of the data plot
К	number of evaluations
m	size of the lineup
Ν	MC parameter: number of replicates on which MC probabilities are based. Higher number of replicates will decrease MC variability.
type	type of simulation used: scenario 3 assumes that the same lineup is shown in all K evaluations
хр	exponent used, defaults to 1
target	integer value identifying the location of the data plot
upper.tail	compute probabilities $P(X \ge x)$ . Be aware that the use of this parameter is not consistent with the other distribution functions in base. There, a value of $P(X \ge x)$ is computed for upper.tail=TRUE.

# Value

Vector/data frame. For comparison a p value based on a binomial distribution is provided as well.

# Examples

pvisual(15, 20, m=3) # triangle test

reg	1 S	

Distance based on the regression parameters

# Description

Dataset X is binned into 5 bins in x-direction. A regression line is fitted to the data in each bin and the regression coefficients are noted. Same is done for dataset PX. An euclidean distance is calculated between the two sets of regression parameters. If the relationship between X and PX looks linear, number of bins should be equal to 1.

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# resid\_boot

# Usage

reg\_dist(X, PX, nbins = 1, intercept = TRUE, scale = TRUE)

# Arguments

Х	a data.frame with two variables, the first column giving the explanatory variable and the second column giving the response variable
РХ	another data.frame with two variables, the first column giving the explanatory variable and the second column giving the response variable
nbins	number of bins on the x-direction, by default $nbins = 1$
intercept	include the distances between intercepts?
scale	logical value: should the variables be scaled before computing regression coefficients?

# Value

distance between X and PX

# Examples

with(mtcars, reg\_dist(data.frame(wt, mpg), data.frame(sample(wt), mpg)))

resid\_boot Bootstrap residuals.

# Description

For use with null\_lm

# Usage

```
resid_boot(model, data)
```

model	to extract residuals from
data	used to fit model

resid\_pboot

# Description

For use with null\_lm

# Usage

resid\_pboot(model, data)

# Arguments

model	to extract residuals from
data	used to fit model

resid_rotate	Rotation residuals.	

# Description

For use with null\_lm

# Usage

```
resid_rotate(model, data)
```

model	to extract residuals from
data	used to fit model

resid\_sigma

# Description

For use with null\_lm

# Usage

resid\_sigma(model, data, sigma = 1)

## Arguments

model	to extract residuals from
data	used to fit model
sigma	a specific sigma to model

rorschach

The Rorschach protocol.

# Description

This protocol is used to calibrate the eyes for variation due to sampling. All plots are typically null data sets, data that is consistent with a null hypothesis. The protocol is described in Buja, Cook, Hofmann, Lawrence, Lee, Swayne, Wickham (2009) Statistical inference for exploratory data analysis and model diagnostics, Phil. Trans. R. Soc. A, 367, 4361-4383.

#### Usage

```
rorschach(method, true = NULL, n = 20, p = 0)
```

method	method for generating null data sets
true	true data set. If NULL, find_plot_data will attempt to extract it from the current ggplot2 plot.
n	total number of samples to generate (including true data)
р	probability of including true data with null data.

sample\_size

#### Description

This function calculates a table of sample sizes for with an experiment, given a lineup size, and estimates of the detection rate.

#### Usage

sample\_size(n = 53:64, m = 20, pA = seq(1/20, 1/3, 0.01), conf = 0.95)

#### Arguments

n	range of sample sizes to check, default is 53:64
m	linup size, default 20
pA	range of estimated detection rates to consider, default is $seq(1/20, 1/3, 0.01)$
conf	confidence level to use to simulate from binomial

#### Examples

```
pow <- sample_size()
pow
library(ggplot2)
library(viridis)
ggplot(pow, aes(x=n, y=pA, fill=prob, group=pA)) +
  geom_tile() +
  scale_fill_viridis_c("power") +
  ylab("detect rate (pA)") + xlab("sample size (n)") +
  theme_bw()</pre>
```

sep\_dist

Distance based on separation of clusters

#### Description

The separation between clusters is defined by the minimum distances of a point in the cluster to a point in another cluster. The number of clusters are provided. If not, the hierarchical clustering method is used to obtain the clusters. The separation between the clusters for dataset X is calculated. Same is done for dataset PX. An euclidean distance is then calculated between these separation for X and PX.

#### Usage

```
sep_dist(X, PX, clustering = FALSE, nclust = 3, type = "separation")
```

# theme\_strip

#### Arguments

Х	a data.frame with two or three columns, the first two columns providing the dataset
PX	a data.frame with two or three columns, the first two columns providing the dataset
clustering	LOGICAL; if TRUE, the third column is used as the clustering variable, by default FALSE
nclust	the number of clusters to be obtained by hierarchical clustering, by default nclust = 3
type	character string to specify which measure to use for distance, see ?cluster.stats for details

# Value

distance between X and PX

# Examples

theme\_strip A theme to minimally strip away the context

# Description

Note this is not a complete theme hence why there are no arguments.

# Usage

theme\_strip()

```
library(ggplot2)
ggplot(cars, aes(dist, speed)) + theme_strip()
```

turk\_results

# Description

Subset of data from a Turk experiment, used to show how to compute power of a lineup

uni\_dist

# Distance for univariate data

# Description

The first four moments is calculated for data X and data PX. An euclidean distance is calculated between these moments for X and PX.

# Usage

uni\_dist(X, PX)

# Arguments

Х	a data.frame where the first column is only used
РХ	another data.frame where the first column is only used

# Value

distance between X and PX

# Examples

if(require('moments')){uni\_dist(rnorm(100), rpois(100, 2))}

visual\_power

# Description

This function simply counts the proportion of people who selected the data plot, in a set of lineups. It adjusts for multiple picks by the same individual, by weighting by the total number of choices.

#### Usage

visual\_power(data, m = 20)

#### Arguments

data	summary of the results, containing columns id, pic_id, response, detected
m	size of the lineup

## Value

vector of powers for each pic\_id

# Examples

data(turk\_results)
visual\_power(turk\_results)

wasps

Wasp gene expression data.

# Description

Data from Toth et al (2010) used in Niladri Roy et al (2015)

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