

# Package ‘cases’

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**Type** Package

**Title** Stratified Evaluation of Subgroup Classification Accuracy

**Version** 0.1.1

**Description** Enables simultaneous statistical inference for the accuracy of multiple classifiers in multiple subgroups (strata). For instance, allows to perform multiple comparisons in diagnostic accuracy studies with co-primary endpoints sensitivity and specificity. (Westphal, Max, and Antonia Zapf. (2021). “Statistical Inference for Diagnostic Test Accuracy Studies with Multiple Comparisons.” <[arXiv:2105.13469](https://arxiv.org/abs/2105.13469)>.)

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**Encoding** UTF-8

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**RoxygenNote** 7.2.0

**VignetteBuilder** knitr, rmarkdown

**Config/testthat/edition** 3

**URL** <https://github.com/maxwestphal/cases>

**BugReports** <https://github.com/maxwestphal/cases/issues>

**Depends** R (>= 2.10)

**NeedsCompilation** no

**Author** Max Westphal [aut, cre] (<<https://orcid.org/0000-0002-8488-758X>>)

**Maintainer** Max Westphal <[max.westphal@steady.ai](mailto:max.westphal@steady.ai)>

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cases	<i>cases package</i>
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### Description

Enables simultaneous statistical inference for the accuracy of multiple classifiers in multiple sub-groups (strata). For instance, allows to perform multiple comparisons in diagnostic accuracy studies with co-primary endpoints sensitivity and specificity. (Westphal, Max, and Antonia Zapf. "Statistical Inference for Diagnostic Test Accuracy Studies with Multiple Comparisons." arXiv:2105.13469 (2021).)

### Details

See the vignettes `vignette()`

---

`categorize`*Categorize continuous values*

---

**Description**

This function allows to split continuous values, e.g. (risk) scores or (bio)markers, into two or more categories by specifying one or more cutoff values.

**Usage**

```
categorize(  
  values,  
  cutoffs = rep(0, ncol(values)),  
  map = 1:ncol(values),  
  labels = NULL  
)
```

**Arguments**

<code>values</code>	numeric matrix of continuous values to be categorized. Assume an (n x r) matrix with n observations (subjects) of r continuous values.
<code>cutoffs</code>	numeric matrix of dimension m x k. Each row of cutoffs defines a split into k+1 distinct categories. Each row must contain distinct values. In the simplest case, cutoffs is a single column matrix whereby is row defines a binary split ( $\leq t$ vs. $> t$ ). In this case (k=1), cutoffs can also be a numeric vector.
<code>map</code>	integer vector of length k with values in 1:r, whereby r = ncol(values). <code>map_l</code> gives the value which column of values should be categorized by ...
<code>labels</code>	character of length m (= number of prediction r)

**Value**

numeric (n x k) matrix with categorical outcomes after categorizing.

**Examples**

```
set.seed(123)  
M <- as.data.frame(mvtnorm::rmvnorm(20, mean=rep(0, 3), sigma=2*diag(3)))  
M  
categorize(M)  
C <- matrix(rep(c(-1, 0, 1, -2, 0, 2), 3), ncol=3, byrow = TRUE)  
C  
w <- c(1, 1, 2, 2, 3, 3)  
categorize(M, C, w)
```

compare *Compare predictions and labels*

---

### Description

Compare predictions and labels

### Usage

```
compare(  
  predictions,  
  labels,  
  partition = TRUE,  
  names = c(specificity = 0, sensitivity = 1)  
)
```

### Arguments

predictions	integer, predicted class
labels	integer, true class state (reference standard)
partition	logical, should result be split into one matrix per class (TRUE; default) or not (FALSE)
names	integer (named), values give data values, names give class names

### Value

data matrix with values 1 (correct prediction) and 0 (false prediction)

### Examples

```
pred <- matrix(c(1,1,0), 5, 3)  
labels <- c(1, 1, 0, 0, 1)  
compare(pred, labels, FALSE)  
compare(pred, labels, TRUE)
```

---

complete\_results *Complete evaluation results*

---

### Description

Complete evaluation results

### Usage

```
complete_results(results, benchmark, alpha, analysis)
```

**Arguments**

results	"cases_results" object, i.e. result of <a href="#">evaluate</a>
benchmark	numeric, vector of benchmark values
alpha	numeric, significance level
analysis	character, either "co-primary" or "full"

**Details**

Not exported, but applied at the end of evaluate by default

**Value**

"cases\_results" object

---

cormat_ar1	<i>Create an AR(1) correlation matrix</i>
------------	---

---

**Description**

Create an AR(1) correlation matrix

**Usage**

```
cormat_ar1(m, rho, d = TRUE)
```

**Arguments**

m	integer, dimension
rho	numeric, correlation parameter in (0,1)
d	binary vector of length m, whereby TRUE/FALSE (alternatively 1/0) indicate active/inactive components of underlying random vector.

**Value**

$$R_{ij} = \rho^{|i-j|}$$

cormat\_equi                      *Create an equicorrelation matrix*

---

**Description**

Create an equicorrelation matrix

**Usage**

```
cormat_equi(m, rho, d = TRUE)
```

**Arguments**

m	integer, dimension
rho	numeric, correlation parameter in (0,1)
d	binary vector of length m, whereby TRUE/FALSE (alternatively 1/0) indicate active/inactive components of underlying random vector.

**Value**
$$R_{ij} = \rho, i \neq j$$

---

data\_wdbc                      *Breast Cancer Wisconsin (Diagnostic) Data Set*

---

**Description**

Dataset documentation can be found at the source website and references below.

**Usage**

```
data_wdbc
```

**Format**

data\_wdbc:

A data frame with 569 rows (patients) and 31 columns (1 target, 30 features).

**Details**

The ID variable was removed. Diagnosis (1= malignant, 0 = benign). Feature variables have been renamed.

**Source**

[https://archive.ics.uci.edu/ml/datasets/breast+cancer+wisconsin+\(diagnostic\)](https://archive.ics.uci.edu/ml/datasets/breast+cancer+wisconsin+(diagnostic))

## References

- W.N. Street, W.H. Wolberg and O.L. Mangasarian. Nuclear feature extraction for breast tumor diagnosis. IS&T/SPIE 1993 International Symposium on Electronic Imaging: Science and Technology, volume 1905, pages 861-870, San Jose, CA, 1993.
- O.L. Mangasarian, W.N. Street and W.H. Wolberg. Breast cancer diagnosis and prognosis via linear programming. Operations Research, 43(4), pages 570-577, July-August 1995.

---

define_contrast	<i>Define a contrast (matrix) to specify exact hypothesis system</i>
-----------------	--

---

## Description

Define a contrast (matrix) to specify exact hypothesis system

## Usage

```
define_contrast(type = c("raw", "dunnett", "tukey"), comparator = NA)
```

## Arguments

type	character, either "Raw", "dunnett" or "tukey")
comparator	either integer (index of comparator) or character (name of comparator)

## Details

"raw" contrast: compare all candidates against specified benchmark values

"dunnett" (all vs. one) contrast: compare all candidates to a single comparator.

"tukey" (all vs. all) contrast: compare all candidates against each other.

## Value

cases\_contrast object to be passed to [evaluate](#)

## Examples

```
define_contrast("dunnett", 1)
```

---

draw_data	<i>Generate binary data</i>
-----------	-----------------------------

---

### Description

Generate binary data

### Usage

```
draw_data(  
  n = 200,  
  prev = c(0.5, 0.5),  
  random = FALSE,  
  m = 10,  
  method = c("roc", "lfc", "pr"),  
  pars = list(),  
  ...  
)
```

### Arguments

n	integer, overall sample size
prev	numeric, vector of class prevalences (adding up to 1)
random	logical, random sampling (TRUE) or fixed group sample sizes
m	integer, number of models
method	character, either "roc", "lfc" (multiple subgroups) or "prob" (no subgroups)
pars	list, containing further named parameters passed to <a href="#">draw_data_roc</a> , <a href="#">draw_data_lfc</a>
...	further named parameters passed

### Value

generated binary data (possibly stratified for subgroups)

### Examples

```
draw_data()
```



---

draw_data_lfc	<i>Generate binary data (LFC model)</i>
---------------	---

---

**Description**

Generate binary data (LFC model)

**Usage**

```
draw_data_lfc(
  n = 100,
  prev = c(0.5, 0.5),
  random = FALSE,
  m = 10,
  se = 0.8,
  sp = 0.8,
  B = round(m/2),
  L = 1,
  Rse = diag(rep(1, m)),
  Rsp = diag(rep(1, m)),
  modnames = paste0("model", 1:m),
  ...
)
```

**Arguments**

n	integer, total sample size
prev	numeric, disease and healthy prevalence (adds up to 1)
random	logical, random sampling (TRUE) or fixed prevalence (FALSE)
m	integer, number of models
se	numeric, sensitivity (length 1)
sp	numeric, specificity (length 1)
B	integer, between 1 and m, specifies how many sensitivity values are projected to 1
L	numeric, worst alternative is computed under side condition $Acc \leq L$ (default value $L=1$ corresponds to true LFC where values are projected to 1)
Rse	matrix, correlation matrix for empirical sensitivities (m x m)
Rsp	matrix, correlation matrix for empirical specificities (m x m)
modnames	character, model names (length m)
...	further arguments (currently unused)

**Value**

Generated binary dataset

**Examples**

```
data <- draw_data_lfc()
head(data)
```

---

draw_data_prb	<i>Sample binary data (single sample)</i>
---------------	---

---

**Description**

This function is wrapper for [rmvbin](#).

**Usage**

```
draw_data_prb(n = 100, pr = c(0.8, 0.8), R = diag(length(pr)))
```

**Arguments**

n	integer, sample size
pr	numeric, vector with marginal success probabilities
R	matrix, square correlation matrix

**Value**

a matrix with n rows and length(pr) columns of randomly generated binary (0, 1) data

---

draw_data_roc	<i>Generate binary data (ROC model)</i>
---------------	---

---

**Description**

Generate binary data (ROC model)

**Usage**

```
draw_data_roc(
  n = 100,
  prev = c(0.5, 0.5),
  random = FALSE,
  m = 10,
  auc = seq(0.85, 0.95, length.out = 5),
  rho = c(0.25, 0.25),
  dist = c("normal", "exponential"),
  e = 10,
  k = 100,
  delta = 0,
```

```

    modnames = paste0("model", 1:m),
    corplot = FALSE,
    ...
)

```

### Arguments

n	integer, total sample size
prev	numeric, disease and healthy prevalence (adds up to 1)
random	logical, random sampling (TRUE) or fixed prevalence (FALSE)
m	integer, number of models
auc	numeric, vector of AUCs of biomarkers
rho	numeric, vector (length 2) of correlations between biomarkers
dist	character, either "normal" or "exponential" specifying the subgroup biomarker distributions
e	numeric, emulates better (worse) model selection quality with higher (lower) values of e
k	integer, technical parameter which adjusts grid size
delta	numeric, specify importance of sensitivity and specificity (default 0)
modnames	character, model names (length m)
corplot	logical (default: FALSE), if TRUE do not return data but instead plot correlation matrices for final binary data
...	further arguments (currently unused)

### Value

Generated binary dataset

### Examples

```

data <- draw_data_roc()
head(data)

```

---

evaluate	<i>Evaluate the accuracy of multiple (candidate) classifiers in several subgroups</i>
----------	---

---

### Description

Assess classification accuracy of multiple classification rules stratified by subgroups, e.g. in diseased (sensitivity) and healthy (specificity) individuals.

**Usage**

```

evaluate(
  data,
  contrast = define_contrast("raw"),
  benchmark = 0.75,
  alpha = 0.05,
  alternative = c("two.sided", "greater", "less"),
  adjustment = c("none", "bonferroni", "maxt", "bootstrap", "mbeta"),
  transformation = c("none", "logit"),
  analysis = c("co-primary", "full"),
  regu = FALSE,
  pars = list(),
  ...
)

```

**Arguments**

<code>data</code>	list of <code>n_g</code> x <code>m</code> binary matrix or <code>data.frame</code> ( <code>n_g</code> observations of <code>m</code> binary decisions), <code>g</code> is the index of subgroups/classes, usually created via <a href="#">compare</a> .
<code>contrast</code>	<code>cases_contrast</code> object, specified via <a href="#">define_contrast</a>
<code>benchmark</code>	value to compare against (RHS), should have same length as <code>data</code> .
<code>alpha</code>	numeric, significance level (default: 0.05)
<code>alternative</code>	character, specify alternative hypothesis
<code>adjustment</code>	character, specify type of statistical adjustment taken to address multiplicity
<code>transformation</code>	character, define transformation to ensure results (e.g. point estimates, confidence limits) lie in unit interval ("none" (default) or "logit")
<code>analysis</code>	character, "co-primary" or "full"
<code>regu</code>	numeric vector of length 3, specify type of shrinkage. Alternatively, logical of length one (TRUE := <code>c(2, 1, 1/2)</code> , FALSE := <code>c(0, 0, 0)</code> )
<code>pars</code>	further parameters given as named list <code>list(type="pairs", nboot=10000)</code>
<code>...</code>	additional named parameters, can be used instead of (in conjunction with) <code>pars</code>

**Details**

Adjustment methods (`adjustment`) and additional parameters (`pars` or `...`):

**"none"** (default): no adjustment for multiplicity

**"bonferroni"**: Bonferroni adjustment

**"maxt"**: maxT adjustment

**"bootstrap"**: Bootstrap approach

- type: "pairs" (default) or "wild" = type (for adjustment="bootstrap")
- nboot: number of bootstrap draws (default: 5000)
- res\_tra: = 0,1,2 or 3 = type of residual transformation of wild bootstrap (default = 0: no transformation) (see <https://www.math.kth.se/matstat/gru/sf2930/papers/wild.bootstrap.pdf>)

**"mbeta"**: A heuristic Bayesian approach which is based on a multivariate beta-binomial model.

- nrep: number of posterior draws (default: 5000)
- lfc\_pr: prior probability of 'least-favorable parameter configuration' (default: 1).

## Value

cases\_results object, which is a list of analysis results

## Examples

```
#
data <- draw_data_roc()
evaluate(data)
```

---

generate\_instance\_lfc *Generate data sets under least favorable parameter configurations*

---

## Description

Generates a (simulation) instance, a list of multiple datasets to be processed (analyzed) with [process\\_instance](#). Ground truth parameters (Sensitivity & Specificity) are least-favorable in the sense that the type-I error rate of the subsequently applied multiple test procedures is maximized.

## Usage

```
generate_instance_lfc(
  nrep = 10,
  n = 100,
  prev = 0.5,
  random = FALSE,
  m = 10,
  se = 0.8,
  sp = 0.8,
  L = 1,
  rhose = 0,
  rhosp = 0,
  cortype = "equi",
  ...,
  data = NULL,
  job = NULL
)
```

**Arguments**

nrep	integer, number of instances
n	integer, total sample size
prev	numeric, disease prevalence
random	logical, fixed prevalence (FALSE) or simple random sampling (TRUE)
m	integer, number of candidates
se	numeric
sp	numeric
L	numeric
rhose	numeric
rhosp	numeric
cortype	character, "equi" or "ak1"
...	further arguments
data	ignored (for batchtools compatibility)
job	ignored (for batchtools compatibility)

**Details**

Utilizes same arguments as [draw\\_data\\_lfc](#) unless mentioned above.

**Value**

a list, a single (LFC) simulation instance

---

generate\_instance\_roc *Generate data sets under realistic parameter configurations*

---

**Description**

Generates a (simulation) instance, a list of multiple datasets to be processed (analyzed) with [process\\_instance](#). Ground truth parameters (Sensitivity & Specificity) are initially generated according to a generative model whereby multiple decision rules (with different parameter values) are derived by thresholding multiple biomarkers.

**Usage**

```
generate_instance_roc(
  nrep = 10,
  n = 100,
  prev = 0.5,
  random = FALSE,
  m = 10,
  auc = "seq(0.85, 0.95, length.out = 5)",
```

```
rhose = 0.5,  
rhosp = 0.5,  
dist = "normal",  
e = 10,  
k = 100,  
delta = 0,  
...,  
data = NULL,  
job = NULL  
)
```

**Arguments**

nrep	integer, number of instances
n	integer, total sample size
prev	numeric, disease prevalence
random	logical, fixed prevalence (FALSE) or simple random sampling (TRUE)
m	integer, number of candidates
auc	numeric
rhose	numeric
rhosp	numeric
dist	character
e	numeric
k	numeric
delta	numeric
...	further arguments
data	ignored (for batchtools compatibility)
job	ignored (for batchtools compatibility)

**Details**

Utilizes same arguments as [draw\\_data\\_roc](#) unless mentioned above.

**Value**

a list, a single (ROC) simulation instance

---

process\_instance      *Analyze simulated synthetic datasets.*

---

### Description

Process data instances, a list of multiple datasets generated via [generate\\_instance\\_lfc](#) or [generate\\_instance\\_roc](#). This function applies [evaluate](#) to all datasets.

### Usage

```
process_instance(
  instance = NULL,
  contrast = "cases::define_contrast('raw', NA)",
  benchmark = 0.5,
  alpha = 0.05,
  alternative = "greater",
  adjustment = "none",
  transformation = "none",
  analysis = "co-primary",
  regu = "c(1,1/2,1/4)",
  pars = "list()",
  ...,
  data = NULL,
  job = list(id = NA)
)
```

### Arguments

instance	generated via <a href="#">generate_instance_lfc</a> or <a href="#">generate_instance_roc</a> .
contrast	cases_contrast object, specified via <a href="#">define_contrast</a>
benchmark	value to compare against (RHS), should have same length as data.
alpha	numeric, significance level (default: 0.05)
alternative	character, specify alternative hypothesis
adjustment	character, specify type of statistical adjustment taken to address multiplicity
transformation	character, define transformation to ensure results (e.g. point estimates, confidence limits) lie in unit interval ("none" (default) or "logit")
analysis	character, "co-primary" (default; only option currently)
regu	numeric vector of length 3, specify type of shrinkage. Alternatively, logical of length one (TRUE := c(2, 1, 1/2), FALSE := c(0, 0, 0))
pars	further parameters given as named list
...	additional named parameters
data	ignored (for batchtools compatibility)
job	for batchtools compatibility, do not change



**Details**

Utilizes same arguments as [evaluate](#) unless mentioned above.

**Value**

standardized evaluation results

---

visualize	<i>Visualize evaluation results</i>
-----------	-------------------------------------

---

**Description**

Visualize evaluation results

**Usage**

```
visualize(x, ...)
```

**Arguments**

x,	a cases_results object, see <a href="#">evaluate</a>
...	further arguments (currently ignored)

**Details**

+++ early development version (only alternative = "greater" is supported) +++

**Value**

a ggplot

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