

# Package ‘configural’

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**Type** Package

**Title** Multivariate Profile Analysis

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**Description** R functions for criterion profile analysis, Davison and Davenport (2002) <doi:10.1037/1082-989X.7.4.468> and meta-analytic criterion profile analysis, Wiernik, Wilmot, Davison, and Ones (2020) <doi:10.1037/met0000305>. Sensitivity analyses to aid in interpreting criterion profile analysis results are also included.

**License** GPL (>= 3)

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**Suggests** crayon

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configural-package      **configural**: *An R package for profile analysis*

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## Description

Overview of the **configural** package.

## Details

The **configural** package provides tools for conducting configural and profile analyses. It currently supports criterion profile analysis (Davison & Davenport, 2002) and meta-analytic criterion profile analysis (Wiernik et al., 2019). Functions are provided to calculate criterion patterns and CPA variance decomposition, as well as for computing confidence intervals, shrinkage corrections, and fungible patterns.

## Author(s)

**Maintainer:** Brenton M. Wiernik <brenton@wiernik.org>

## See Also

Useful links:

- Report bugs at <https://github.com/bwiernik/configural/issues>

---

adjust_Rsq	<i>Adjust a regression model R-squared for overfitting</i>
------------	--

---

## Description

Estimate shrinkage for regression models

## Usage

```
adjust_Rsq(Rsq, n, p, adjust = c("fisher", "pop", "cv"))
```

## Arguments

Rsq	Observed model R-squared
n	Sample size
p	Number of predictors
adjust	Which adjustment to apply. Options are "fisher" for the Adjusted R-squared method used in <code>stats::lm()</code> , "pop" for the positive-part Pratt estimator of the population R-squared, and "cv" for the Browne/positive-part Pratt estimator of the cross-validity R-squared. Based on Shieh (2008), these are the estimators for the population and cross-validity R-squared values that show the least bias with a minimal increase in computational complexity.

## Value

An adjusted R-squared value.

## References

Shieh, G. (2008). Improved shrinkage estimation of squared multiple correlation coefficient and squared cross-validity coefficient. *Organizational Research Methods*, 11(2), 387–407. doi: [10.1177/1094428106292901](https://doi.org/10.1177/1094428106292901)

## Examples

```
adjust_Rsq(.55, 100, 6, adjust = "pop")
```

---

configural_news	<i>Retrieve the NEWS file for the configural package</i>
-----------------	--

---

### Description

This function gives a shortcut to the `utils::news(package = "configural")` function and displays configural's NEWS file, which contains version information, outlines additions and changes to the package, and describes other updates.

### Usage

```
configural_news()
```

### Examples

```
configural_news()
```

---

cor_covariance	<i>Calculate the asymptotic sampling covariance matrix for the unique elements of a correlation matrix</i>
----------------	--

---

### Description

Calculate the asymptotic sampling covariance matrix for the unique elements of a correlation matrix

### Usage

```
cor_covariance(r, n)
```

### Arguments

r	A correlation matrix
n	The sample size

### Value

The asymptotic sampling covariance matrix

### Author(s)

Based on an internal function from the **fungible** package by Niels Waller

### References

Nel, D. G. (1985). A matrix derivation of the asymptotic covariance matrix of sample correlation coefficients. *Linear Algebra and Its Applications*, 67, 137–145. doi: [10.1016/00243795\(85\)901910](https://doi.org/10.1016/00243795(85)901910)

**Examples**

```
cor_covariance(matrix(c(1, .2, .3, .2, 1, .3, .3, .3, 1), ncol = 3), 100)
```

---

cor_covariance_meta	<i>Estimate the asymptotic sampling covariance matrix for the unique elements of a meta-analytic correlation matrix</i>
---------------------	---

---

**Description**

Estimate the asymptotic sampling covariance matrix for the unique elements of a meta-analytic correlation matrix

**Usage**

```
cor_covariance_meta(
  r,
  n,
  sevar,
  source = NULL,
  rho = NULL,
  sevar_rho = NULL,
  n_overlap = NULL
)
```

**Arguments**

r	A meta-analytic matrix of observed correlations (can be full or lower-triangular).
n	A matrix of total sample sizes for the meta-analytic correlations in r (can be full or lower-triangular).
sevar	A matrix of estimated sampling error variances for the meta-analytic correlations in r (can be full or lower-triangular).
source	A matrix indicating the sources of the meta-analytic correlations in r (can be full or lower-triangular). Used to estimate overlapping sample size for correlations when n_overlap == NULL.
rho	A meta-analytic matrix of corrected correlations (can be full or lower-triangular).
sevar_rho	A matrix of estimated sampling error variances for the meta-analytic corrected correlations in rho (can be full or lower-triangular).
n_overlap	A matrix indicating the overlapping sample size for the unique (lower triangular) values in r (can be full or lower-triangular). Values must be arranged in the order returned by cor_labels(colnames(R)).

**Details**

If both source and n\_overlap are NULL, it is assumed that all meta-analytic correlations come from the the same source.

**Value**

The estimated asymptotic sampling covariance matrix

**References**

Nel, D. G. (1985). A matrix derivation of the asymptotic covariance matrix of sample correlation coefficients. *Linear Algebra and Its Applications*, 67, 137–145. doi: [10.1016/00243795\(85\)901910](https://doi.org/10.1016/00243795(85)901910)

Wiernik, B. M. (2018). *Accounting for dependency in meta-analytic structural equations modeling: A flexible alternative to generalized least squares and two-stage structural equations modeling*. Unpublished manuscript.

**Examples**

```
cor_covariance_meta(r = mindfulness$r, n = mindfulness$n,  
                   sevar = mindfulness$sevar_r, source = mindfulness$source)
```

---

cor\_labels

*Generate labels for correlations from a vector of variable names*

---

**Description**

This function returns a vector of labels for the unique correlations between pairs of variables from a supplied vector of variable names

**Usage**

```
cor_labels(var_names)
```

**Arguments**

var\_names      A character vector of variable names

**Value**

A vector of correlation labels

**Examples**

```
cor_labels(colnames(mindfulness$r))
```

**Description**

Conduct criterion profile analysis using a correlation matrix

**Usage**

```
cpa_mat(
  formula,
  cov_mat,
  n = NULL,
  se_var_mat = NULL,
  se_beta_method = c("normal", "lm"),
  adjust = c("fisher", "pop", "cv"),
  conf_level = 0.95,
  ...
)
```

**Arguments**

formula	Regression formula with a single outcome variable on the left-hand side and one or more predictor variables on the right-hand side (e.g., $Y \sim X1 + X2$ ).
cov_mat	Correlation matrix containing the variables to be used in the regression.
n	Sample size. Used to compute adjusted R-squared and, if se_var_mat is NULL, standard errors. If NULL and se_var_mat is specified, effective sample size is computed based on se_var_mat (cf. Revelle et al., 2017).
se_var_mat	Optional. The sampling error covariance matrix among the unique elements of cov_mat. Used to calculate standard errors. If not supplied, the sampling covariance matrix is calculated using n.
se_beta_method	Method to use to estimate the standard errors of standardized regression (beta) coefficients. Current options include "normal" (use the Jones-Waller, 2015, normal-theory approach) and "lm" (estimate standard errors using conventional regression formulas).
adjust	Method to adjust R-squared for overfitting. See <a href="#">adjust_Rsq()</a> for details.
conf_level	Confidence level to use for confidence intervals.
...	Additional arguments.

**Value**

An object of class "cpa" containing the criterion pattern vector and CPA variance decomposition

## References

Jones, J. A., & Waller, N. G. (2015). The normal-theory and asymptotic distribution-free (ADF) covariance matrix of standardized regression coefficients: Theoretical extensions and finite sample behavior. *Psychometrika*, *80*(2), 365–378. doi: [10.1007/s113360139380y](https://doi.org/10.1007/s113360139380y)

Revelle, W., Condon, D. M., Wilt, J., French, J. A., Brown, A., & Elleman, L. G. (2017). Web- and phone-based data collection using planned missing designs. In N. G. Fielding, R. M. Lee, & G. Blank, *The SAGE Handbook of Online Research Methods* (pp. 578–594). SAGE Publications. doi: [10.4135/9781473957992.n33](https://doi.org/10.4135/9781473957992.n33)

Wiernik, B. M., Wilmot, M. P., Davison, M. L., & Ones, D. S. (2019). Meta-analytic criterion profile analysis. *Psychological Methods* doi: [10.1037/met0000305](https://doi.org/10.1037/met0000305)

## Examples

```
sevar <- cor_covariance_meta(mindfulness$r, mindfulness$n, mindfulness$sevar_r, mindfulness$source)
cpa_mat(mindfulness ~ ES + A + C + Ex + O,
        cov_mat = mindfulness$r,
        n = NULL,
        se_var_mat = sevar,
        adjust = "pop")
```

---

cpa\_scores

*Compute CPA level and pattern scores for a set of data*

---

## Description

Compute CPA level and pattern scores for a set of data

## Usage

```
cpa_scores(
  cpa_mod,
  newdata = NULL,
  augment = TRUE,
  cpa_names = c("cpa_lev", "cpa_pat"),
  scale = FALSE,
  scale_center = TRUE,
  scale_scale = TRUE
)
```

## Arguments

cpa_mod	A model returned from <code>cpa_mat()</code> (a model of class "cpa")
newdata	A data frame or matrix containing columns with the same names as the predictors in <code>cpa_mod</code> .
augment	Should CPA score columns be added to <code>newdata</code> (TRUE, default) or returned alone (FALSE)?



cpa_names	Character vector of length 2 giving the variable names to assign to the CPA score columns.
scale	Logical. Should the variables in newdata be scaled (standardized)?
scale_center	If scale is TRUE, passed to the center argument in <code>base::scale()</code> . Can be TRUE (center columns of newdata around the column means), FALSE (don't center), or a numeric vector of length equal to the number of predictors in cpa_mod containing the values to center around.
scale_scale	If scale is TRUE, passed to the scale argument in <code>base::scale()</code> . Can be TRUE (scale/standardize columns of newdata using the column standard deviations or root mean squares), FALSE (don't scale), or a numeric vector of length equal to the number of predictors in cpa_mod containing the values to scale by. See <code>base::scale()</code> for details.

### Value

A data frame containing the CPA score variables.

### Examples

```
sevar <- cor_covariance_meta(mindfulness$r, mindfulness$n, mindfulness$sevar_r, mindfulness$source)
cpa_mod <- cpa_mat(mindfulness ~ ES + A + C + Ex + O,
                  cov_mat = mindfulness$r,
                  n = NULL,
                  se_var_mat = sevar,
                  adjust = "pop")

newdata <- data.frame(ES = c(4.2, 3.2, 3.4, 4.2, 3.8, 4.0, 5.6, 2.8, 3.4, 2.8),
                    A = c(4.0, 4.2, 3.8, 4.6, 4.0, 4.6, 4.6, 2.6, 3.6, 5.4),
                    C = c(2.8, 4.0, 4.0, 3.0, 4.4, 5.6, 4.4, 3.4, 4.0, 5.6),
                    Ex = c(3.8, 5.0, 4.2, 3.6, 4.8, 5.6, 4.2, 2.4, 3.4, 4.8),
                    O = c(3.0, 4.0, 4.8, 3.2, 3.6, 5.0, 5.4, 4.2, 5.0, 5.2)
                    )

newdata_cpa <- cpa_scores(cpa_mod, newdata, augment = FALSE)
newdata_augment <- cpa_scores(cpa_mod, newdata, augment = TRUE)
```

---

disorders

*Meta-analytic correlations among Big Five personality traits and psychological disorders*

---

### Description

Big Five intercorrelations from Davies et al. (2015). Big Five–psychological disorder correlations from Kotov et al. (2010). Note that there were several duplicate or missing values in the reported data table in the published article. These results are based on corrected data values.

### Usage

```
data(disorders)
```

**Format**

list with entries `r` (mean observed correlations), `rho` (mean corrected correlations), `n` (sample sizes), `sevar_r` (sampling error variances for mean observed correlations), `sevar_rho` (sampling error variances for mean corrected correlations), and `source` (character labels indicating which meta-analytic correlations came from the same source)

**References**

Davies, S. E., Connelly, B. L., Ones, D. S., & Birkland, A. S. (2015). The general factor of personality: The “Big One,” a self-evaluative trait, or a methodological gnat that won’t go away? *Personality and Individual Differences*, *81*, 13–22. doi: [10.1016/j.paid.2015.01.006](https://doi.org/10.1016/j.paid.2015.01.006)

Kotov, R., Gamez, W., Schmidt, F., & Watson, D. (2010). Linking “big” personality traits to anxiety, depressive, and substance use disorders: A meta-analysis. *Psychological Bulletin*, *136*(5), 768–821. doi: [10.1037/a0020327](https://doi.org/10.1037/a0020327)

**Examples**

```
data(disorders)
```

---

fungible

*Locate extrema of fungible weights for regression and related models*

---

**Description**

Generates fungible regression weights (Waller, 2008) and related results using the method by Waller and Jones (2010).

**Usage**

```
fungible(
  object,
  theta = 0.005,
  Nstarts = 1000,
  MaxMin = c("min", "max"),
  silent = FALSE,
  ...
)
```

**Arguments**

<code>object</code>	A fitted model object. Currently supported classes are: "cpa"
<code>theta</code>	A vector of values to decrement from R-squared to compute families of fungible coefficients.
<code>Nstarts</code>	Maximum number of (max) minimizations from random starting configurations.

MaxMin	Should the cosine between the observed and alternative weights be maximized ("max") to find the maximally similar coefficients or minimized ("min") to find the maximally dissimilar coefficients?
silent	Should current optimization values be printed to the console (FALSE) or suppressed (TRUE)?
...	Additional arguments

**Value**

A list containing the alternative weights and other fungible weights estimation parameters

**Author(s)**

Niels Waller, Jeff Jones, Brenton M. Wiernik. Adapted from `fungible::fungibleExtrema()`.

**References**

Waller, N. G. (2008). Fungible weights in multiple regression. *Psychometrika*, *73*(4), 691–703. doi: [10.1007/s113360089066z](https://doi.org/10.1007/s113360089066z)

Waller, N. G., & Jones, J. A. (2009). Locating the extrema of fungible regression weights. *Psychometrika*, *74*(4), 589–602. doi: [10.1007/s1133600890877](https://doi.org/10.1007/s1133600890877)

**Examples**

```
mind <- cpa_mat(mindfulness ~ ES + A + C + Ex + O,
               cov_mat = mindfulness$r,
               n = harmonic_mean(vechs(mindfulness$n)),
               se_var_mat = cor_covariance_meta(mindfulness$r,
                                                mindfulness$n,
                                                mindfulness$sevar_r,
                                                mindfulness$source),
               adjust = "pop")
mind_fung <- fungible(mind, Nstarts = 100)
```

---

fungible.cpa

*Locate extrema of fungible criterion profile patterns*

---

**Description**

Identify maximally similar or dissimilar criterion patterns in criterion profile analysis

**Usage**

```
## S3 method for class 'cpa'
fungible(
  object,
  theta = 0.005,
  Nstarts = 1000,
  MaxMin = c("min", "max"),
  silent = FALSE,
  ...
)
```

**Arguments**

object	A fitted model object of class "cpa".
theta	A vector of values to decrement from R-squared to compute families of fungible coefficients.
Nstarts	Maximum number of (max) minimizations from random starting configurations.
MaxMin	Should the cosine between the observed and alternative weights be maximized ("max") to find the maximally similar coefficients or minimized ("min") to find the maximally dissimilar coefficients?
silent	Should current optimization values be printed to the console (FALSE) or suppressed (TRUE)?
...	Additional arguments

**Value**

A list containing the alternative weights and other fungible weights estimation parameters

**References**

Wiernik, B. M., Wilmot, M. P., Davison, M. L., & Ones, D. S. (2020). Meta-analytic criterion profile analysis. *Psychological Methods*. doi: [10.1037/met0000305](https://doi.org/10.1037/met0000305)

**Examples**

```
mind <- cpa_mat(mindfulness ~ ES + A + C + Ex + 0,
  cov_mat = mindfulness$r,
  n = harmonic_mean(vechs(mindfulness$n)),
  se_var_mat = cor_covariance_meta(mindfulness$r,
    mindfulness$n,
    mindfulness$sevar_r,
    mindfulness$source),
  adjust = "pop")
mind_fung <- fungible(mind, Nstarts = 100)
```

---

fungible.lm                      *Locate extrema of fungible OLS regression weights*

---

### Description

Identify maximally similar or dissimilar sets of fungible standardized regression coefficients from an OLS regression model

### Usage

```
## S3 method for class 'lm'
fungible(
  object,
  theta = 0.005,
  Nstarts = 1000,
  MaxMin = c("min", "max"),
  silent = FALSE,
  ...
)
```

### Arguments

object	A fitted model object of class "lm" or "summary.lm".
theta	A vector of values to decrement from R-squared to compute families of fungible coefficients.
Nstarts	Maximum number of (max) minimizations from random starting configurations.
MaxMin	Should the cosine between the observed and alternative weights be maximized ("max") to find the maximally similar coefficients or minimized ("min") to find the maximally dissimilar coefficients?
silent	Should current optimization values be printed to the console (FALSE) or suppressed (TRUE)?
...	Additional arguments

### Value

A list containing the alternative weights and other fungible weights estimation parameters

### References

Waller, N. G., & Jones, J. A. (2009). Locating the extrema of fungible regression weights. *Psychometrika*, 74(4), 589–602. doi: [10.1007/s1133600890877](https://doi.org/10.1007/s1133600890877)

### Examples

```
lm_mtcars <- lm(mpg ~ cyl + disp + hp + drat + wt + qsec + vs + am + gear + carb,
  data = mtcars)
lm_mtcars_fung <- fungible(lm_mtcars, Nstarts = 100)
```

---

gre	<i>Meta-analytic correlations of Graduate Record Examination subtests with graduate grade point average</i>
-----	---

---

### Description

Correlations between GRE subtests and graduate student GPA from Kuncel et al. (2001).

### Usage

```
data(jobchar)
```

### Format

list with entries `r` (mean observed correlations), `rho` (mean corrected correlations), `n` (sample sizes), `sevar_r` (sampling error variances for mean observed correlations), `sevar_rho` (sampling error variances for mean corrected correlations), and `source` (character labels indicating which meta-analytic correlations came from the same source)

### Details

GRE–GPA correlations in `rho` are corrected for direct range restriction on the GRE and unreliability in GPA. Subtest intercorrelations in `rho` are observed correlations computed among applicant norm samples. These values are also used in `r`. Due to compensatory selection on GRE scores, these values will not accurately reflect subtest intercorrelations in selected-student (range-restricted) samples. `sevar_rho` and `sevar_r` for GRE subtest intercorrelations are computed with an assumed  $SD_{\rho}SD_{rho} = .02$ .

### References

Kuncel, N. R., Hezlett, S. A., & Ones, D. S. (2001). A comprehensive meta-analysis of the predictive validity of the graduate record examinations: Implications for graduate student selection and performance. *Psychological Bulletin*, *127*(1), 162–181. doi: [10.1037/00332909.127.1.162](https://doi.org/10.1037/00332909.127.1.162)

### Examples

```
data(gre)
```

---

harmonic_mean	<i>Find the harmonic mean of a vector, matrix, or columns of a data.frame</i>
---------------	---

---

**Description**

The harmonic mean is merely the reciprocal of the arithmetic mean of the reciprocals.

**Usage**

```
harmonic_mean(x, na.rm = TRUE, zero = TRUE)
```

**Arguments**

x	A vector, matrix, or data.frame
na.rm	Logical. If TRUE, remove NA values before processing
zero	Logical, If TRUE, if there are any zeroes, return 0, else, return the harmonic mean of the non-zero elements

**Value**

The harmonic mean of x

**Author(s)**

Adapted from `psych::harmonic.mean()` by William Revelle

**Examples**

```
harmonic_mean(1:10)
```

---

hrm	<i>Meta-analytic correlations of HRM practices with organizational financial performance</i>
-----	--

---

**Description**

Human resource management practice–organizational financial performance correlations from Combs et al. (2006). Intercorrelations among HRM practices from Guest et al. (2004).

**Usage**

```
data(hrm)
```

**Format**

list with entries `r` (mean observed correlations), `rho` (mean corrected correlations), `n` (sample sizes), `sevar_r` (sampling error variances for mean observed correlations), `sevar_rho` (sampling error variances for mean corrected correlations), and `source` (character labels indicating which meta-analytic correlations came from the same source)

**References**

Combs, J., Liu, Y., Hall, A., & Ketchen, D. (2006). How much do high-performance work practices matter? A meta-analysis of their effects on organizational performance. *Personnel Psychology*, 59(3), 501–528. doi: [10.1111/j.17446570.2006.00045.x](https://doi.org/10.1111/j.17446570.2006.00045.x)

Guest, D., Conway, N., & Dewe, P. (2004). Using sequential tree analysis to search for ‘bundles’ of HR practices. *Human Resource Management Journal*, 14(1), 79–96. doi: [10.1111/j.1748-8583.2004.tb00113.x](https://doi.org/10.1111/j.1748-8583.2004.tb00113.x)

**Examples**

```
data(hrm)
```

---

```
jobchar
```

*Meta-analytic correlations of job characteristics with performance and satisfaction*

---

**Description**

Self-rated job characteristics intercorrelations and correlations with other-rated job performance and self-rated job satisfaction from Humphrey et al. (2007).

**Usage**

```
data(jobchar)
```

**Format**

list with entries `r` (mean observed correlations), `rho` (mean corrected correlations), `n` (sample sizes), `sevar_r` (sampling error variances for mean observed correlations), `sevar_rho` (sampling error variances for mean corrected correlations), and `source` (character labels indicating which meta-analytic correlations came from the same source)

**References**

Humphrey, S. E., Nahrgang, J. D., & Morgeson, F. P. (2007). Integrating motivational, social, and contextual work design features: A meta-analytic summary and theoretical extension of the work design literature. *Journal of Applied Psychology*, 92(5), 1332–1356. doi: [10.1037/0021-9010.92.5.1332](https://doi.org/10.1037/0021-9010.92.5.1332)



**Examples**

```
data(jobchar)
```

---

mindfulness	<i>Meta-analytic correlations among Big Five personality traits and trait mindfulness</i>
-------------	---

---

**Description**

Big Five intercorrelations from Davies et al. (2015). Big Five–Mindfulness correlations from Hanley and Garland (2017). Coefficient alpha for mindfulness measures taken from Giluk (2009).

**Usage**

```
data(mindfulness)
```

**Format**

list with entries `r` (mean observed correlations), `rho` (mean corrected correlations), `n` (sample sizes), `sevar_r` (sampling error variances for mean observed correlations), `sevar_rho` (sampling error variances for mean corrected correlations), and `source` (character labels indicating which meta-analytic correlations came from the same source)

**References**

Davies, S. E., Connelly, B. L., Ones, D. S., & Birkland, A. S. (2015). The general factor of personality: The “Big One,” a self-evaluative trait, or a methodological gnat that won’t go away? *Personality and Individual Differences, 81*, 13–22. doi: [10.1016/j.paid.2015.01.006](https://doi.org/10.1016/j.paid.2015.01.006)

Giluk, T. L. (2009). Mindfulness, Big Five personality, and affect: A meta-analysis. *Personality and Individual Differences, 47*(8), 805–811. doi: [10.1016/j.paid.2009.06.026](https://doi.org/10.1016/j.paid.2009.06.026)

Hanley, A. W., & Garland, E. L. (2017). The mindful personality: A meta-analysis from a cybernetic perspective. *Mindfulness, 8*(6), 1456–1470. doi: [10.1007/s1267101707368](https://doi.org/10.1007/s1267101707368)

**Examples**

```
data(mindfulness)
```

---

n_effective_R2	<i>Effective sample size</i>
----------------	------------------------------

---

### Description

Estimate an effective sample size for a statistic given the observed statistic and the estimated sampling error variance (cf. Revelle et al., 2017).

### Usage

```
n_effective_R2(R2, var_R2, p)
```

### Arguments

R2	Observed $R^2$ value
var_R2	Estimated sampling error variance for $R^2$
p	Number of predictors in the regression model

### Details

n\_effective\_R2 estimates the effective sample size for the  $R^2$  value from an OLS regression model, using the sampling error variance formula from Cohen et al. (2003).

### Value

An effective sample size.

### References

Revelle, W., Condon, D. M., Wilt, J., French, J. A., Brown, A., & Elleman, L. G. (2017). Web- and phone-based data collection using planned missing designs. In N. G. Fielding, R. M. Lee, & G. Blank, *The SAGE Handbook of Online Research Methods* (pp. 578–594). SAGE Publications. doi: [10.4135/9781473957992.n33](https://doi.org/10.4135/9781473957992.n33)

Cohen, J., Cohen, P., West, S. G., & Aiken, L. S. (2003). *Applied multiple regression/correlation analysis for the behavioral sciences* (3rd ed.). Routledge. doi: [10.4324/9780203774441](https://doi.org/10.4324/9780203774441)

### Examples

```
n_effective_R2(0.3953882, 0.0005397923, 5)
```

---

prejudice	<i>Correlations between study design moderators and effect sizes for prejudice reduction following intergroup contact</i>
-----------	---

---

**Description**

Correlations among study design moderators and study design moderator–observed prejudice reduction effect sizes from Pettigrew and Tropp (2008). Note that correlations with effect size have been reverse-coded so that a positive correlation indicates that a higher level of the moderator is associated with *larger* prejudice reduction.

**Usage**

```
data(prejudice)
```

**Format**

list with entries r (observed correlations among moderators) and k (number of samples in meta-analysis)

**References**

Pettigrew, T. F., & Tropp, L. R. (2006). A meta-analytic test of intergroup contact theory. *Journal of Personality and Social Psychology*, *90*(5), 751–783. doi: [10.1037/00223514.90.5.751](https://doi.org/10.1037/00223514.90.5.751)

**Examples**

```
data(prejudice)
```

---

team	<i>Meta-analytic correlations among team processes and team effectiveness</i>
------	---

---

**Description**

Team process intercorrelations and team process–team performance/affect correlations from LePine et al. (2008).

**Usage**

```
data(team)
```

**Format**

list with entries r (mean observed correlations), rho (mean corrected correlations), n (sample sizes), sevar\_r (sampling error variances for mean observed correlations), sevar\_rho (sampling error variances for mean corrected correlations), and source (character labels indicating which meta-analytic correlations came from the same source)

## Details

Note that LePine et al. (2008) did not report confidence intervals, sampling error variances, or heterogeneity estimates for correlations among team processes; included sampling error variances in this list are based on total sample size only and do not include uncertainty stemming from any effect size heterogeneity.

## References

LePine, J. A., Piccolo, R. F., Jackson, C. L., Mathieu, J. E., & Saul, J. R. (2008). A meta-analysis of teamwork processes: tests of a multidimensional model and relationships with team effectiveness criteria. *Personnel Psychology*, *61*(2), 273–307. doi: [10.1111/j.17446570.2008.00114.x](https://doi.org/10.1111/j.17446570.2008.00114.x)

## Examples

```
data(team)
```

---

var_error_cpa	<i>Estimate the sampling error variance for criterion profile analysis parameters</i>
---------------	---

---

## Description

Estimate the sampling error variance for criterion profile analysis parameters

## Usage

```
var_error_cpa(
  Rxx,
  rxy,
  n = NULL,
  se_var_mat = NULL,
  adjust = c("fisher", "pop", "cv")
)
```

## Arguments

Rxx	An intercorrelation matrix among the predictor variables
rxy	A vector of predictor–criterion correlations
n	The sample size. If NULL and se_var_mat is provided, n will be estimated as the effective sample size based on se_var_mat. See <a href="#">n_effective_R2()</a> .
se_var_mat	A matrix of sampling covariance values for the elements of Rxx and rxy. If NULL, generated using the Normal theory covariance matrix based on n.
adjust	Method to adjust R-squared for overfitting. See <a href="#">adjust_Rsq</a> for details.

**Value**

A list containing sampling covariance matrices or sampling error variance estimates for CPA parameters

**Examples**

```
var_error_cpa(mindfulness$rho[1:5, 1:5], mindfulness$rho[1:5, 6], n = 17060)
```

---

vech	<i>Vectorize a matrix</i>
------	---------------------------

---

**Description**

cvec returns the column-wise vectorization of an input matrix (stacking the columns on one another). rvec returns the row-wise vectorization of an input matrix (concatenating the rows after each other). vech returns the column-wise half-vectorization of an input matrix (stacking the lower triangular elements of the matrix, including the diagonal). vechs returns the strict column-wise half-vectorization of an input matrix (stacking the lower triangular elements of the matrix, excluding the diagonal). All functions return the output as a vector.

**Usage**

```
vech(x)
```

```
vechs(x)
```

```
cvec(x)
```

```
rvec(x)
```

**Arguments**

x                    A matrix

**Value**

A vector of values

**Author(s)**

Based on functions from the the **OpenMx** package

**Examples**

```
cvec(matrix(1:9, 3, 3))
rvec(matrix(1:9, 3, 3))
vech(matrix(1:9, 3, 3))
vechs(matrix(1:9, 3, 3))
vechs(matrix(1:12, 3, 4))
```

---

vech2full                      *Inverse vectorize a matrix*

---

### Description

These functions return the symmetric matrix that produces the given half-vectorization result.

### Usage

```
vech2full(x)
```

```
vechs2full(x, diagonal = 1)
```

### Arguments

x                      A vector

diagonal              A value or vector of values to enter on the diagonal for `vechs2full` (default = 1)

### Details

The input consists of a vector of the elements in the lower triangle of the resulting matrix (for `vech2full`, including the elements along the diagonal of the matrix, as a column vector), filled column-wise. For `vechs2full`, the diagonal values are filled as 1 by default, alternative values can be specified using the `diag` argument. The inverse half-vectorization takes a vector and reconstructs a symmetric matrix such that `vech2full(vech(x))` is identical to `x` if `x` is symmetric.

### Value

A symmetric matrix

### Author(s)

Based on functions from the the **OpenMx** package

### Examples

```
vech2full(c(1, 2, 3, 5, 6, 9))  
vechs2full(c(2, 3, 6), diagonal = 0)
```

---

wt_cov	<i>Compute weighted covariances</i>
--------	-------------------------------------

---

**Description**

Compute the weighted covariance among variables in a matrix or between the variables in two separate matrices/vectors.

**Usage**

```
wt_cov(
  x,
  y = NULL,
  wt = NULL,
  as_cor = FALSE,
  use = c("everything", "listwise", "pairwise"),
  unbiased = TRUE,
  df_type = c("count", "sum_wts")
)
```

```
wt_cor(x, y = NULL, wt = NULL, use = "everything")
```

**Arguments**

x	Vector or matrix of x variables.
y	Vector or matrix of y variables
wt	Vector of weights
as_cor	Logical scalar that determines whether the covariances should be standardized (TRUE) or unstandardized (FALSE).
use	Method for handling missing values. "everything" uses all values and does not account for missingness, "listwise" uses only complete cases, and "pairwise" uses pairwise deletion.
unbiased	Logical scalar determining whether variance should be unbiased (TRUE) or maximum-likelihood (FALSE).
df_type	Character scalar determining whether the degrees of freedom for unbiased estimates should be based on numbers of cases (n - 1; "count"; default) or squared sums of weights (1 - sum(w^2); "sum_wts").

**Value**

Scalar, vector, or matrix of covariances.

**Author(s)**

Jeffrey A. Dahlke

**Examples**

```

wt_cov(x = c(1, 0, 2), y = c(1, 2, 3), wt = c(1, 2, 2), as_cor = FALSE, use = "everything")
wt_cov(x = c(1, 0, 2), y = c(1, 2, 3), wt = c(1, 2, 2), as_cor = TRUE, use = "everything")
wt_cov(x = cbind(c(1, 0, 2), c(1, 2, 3)), wt = c(1, 2, 2), as_cor = FALSE, use = "everything")
wt_cov(x = cbind(c(1, 0, 2), c(1, 2, 3)), wt = c(1, 2, 2), as_cor = TRUE, use = "everything")
wt_cov(x = cbind(c(1, 0, 2, NA), c(1, 2, 3, 3)),
       wt = c(1, 2, 2, 1), as_cor = FALSE, use = "listwise")
wt_cov(x = cbind(c(1, 0, 2, NA), c(1, 2, 3, 3)),
       wt = c(1, 2, 2, 1), as_cor = TRUE, use = "listwise")

```

wt\_dist

*Weighted descriptive statistics for a vector of numbers***Description**

Compute the weighted mean and variance of a vector of numeric values. If no weights are supplied, defaults to computing the unweighted mean and the unweighted maximum-likelihood variance.

**Usage**

```

wt_dist(
  x,
  wt = rep(1, length(x)),
  unbiased = TRUE,
  df_type = c("count", "sum_wts")
)

wt_mean(x, wt = rep(1, length(x)))

wt_var(
  x,
  wt = rep(1, length(x)),
  unbiased = TRUE,
  df_type = c("count", "sum_wts")
)

```

**Arguments**

x	Vector of values to be analyzed.
wt	Weights associated with the values in x.
unbiased	Logical scalar determining whether variance should be unbiased (TRUE) or maximum-likelihood (FALSE).
df_type	Character scalar determining whether the degrees of freedom for unbiased estimates should be based on numbers of cases ("count"; default) or sums of weights ("sum_wts").



**Details**

The weighted mean is computed as

$$\bar{x}_w = \frac{\sum_{i=1}^k x_i w_i}{\sum_{i=1}^k w_i}$$

where  $x$  is a numeric vector and  $w$  is a vector of weights.

The weighted variance is computed as

$$var_w(x) = \frac{\sum_{i=1}^k (x_i - \bar{x}_w)^2 w_i}{\sum_{i=1}^k w_i}$$

and the unbiased weighted variance is estimated by multiplying  $var_w(x)$  by  $\frac{k}{k-1}$ .

**Value**

A weighted mean and variance if weights are supplied or an unweighted mean and variance if weights are not supplied.

**Author(s)**

Jeffrey A. Dahlke

**Examples**

```
wt_dist(x = c(.1, .3, .5), wt = c(100, 200, 300))
wt_mean(x = c(.1, .3, .5), wt = c(100, 200, 300))
wt_var(x = c(.1, .3, .5), wt = c(100, 200, 300))
```

---

 %%%

*Quadratic form matrix product*

---

**Description**

Calculate the quadratic form

$$Q = x'Ax$$

**Usage**

A %%% x

**Arguments**

A	A square matrix
x	A vector or matrix

**Value**

The quadratic product

**Examples**

```
diag(5) %&% 1:5
```

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