

# Package ‘polycor’

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**Title** Polychoric and Polyserial Correlations

**Depends** R (>= 3.3.0)

**Imports** stats, mvtnorm, Matrix

**ByteCompile** yes

**LazyLoad** yes

**Description** Computes polychoric and polyserial correlations by quick “two-step” methods or ML, optionally with standard errors; tetrachoric and biserial correlations are special cases.

**License** GPL (>= 2)

**URL** <https://r-forge.r-project.org/projects/polycor/>,  
<https://CRAN.R-project.org/package=polycor>

**NeedsCompilation** no

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 hetcor

*Heterogeneous Correlation Matrix*


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## Description

Computes a heterogenous correlation matrix, consisting of Pearson product-moment correlations between numeric variables, polyserial correlations between numeric and ordinal variables, and polychoric correlations between ordinal variables.

## Usage

```

hetcor(data, ..., ML = FALSE, std.err = TRUE,
        use=c("complete.obs", "pairwise.complete.obs"), bins=4, pd=TRUE)
## S3 method for class 'data.frame'
hetcor(data, ML = FALSE, std.err = TRUE,
        use = c("complete.obs", "pairwise.complete.obs"), bins=4, pd=TRUE, ...)
## Default S3 method:
hetcor(data, ..., ML = FALSE, std.err = TRUE,
        use=c("complete.obs", "pairwise.complete.obs"), bins=4, pd=TRUE)
## S3 method for class 'hetcor'
print(x, digits = max(3, getOption("digits") - 3), ...)
## S3 method for class 'hetcor'
as.matrix(x, ...)

```

## Arguments

<code>data</code>	a data frame consisting of factors, ordered factors, logical variables, character variables, and/or numeric variables, or the first of several variables.
<code>...</code>	variables and/or arguments to be passed down.
<code>ML</code>	if TRUE, compute maximum-likelihood estimates; if FALSE, compute quick two-step estimates.
<code>std.err</code>	if TRUE, compute standard errors.
<code>bins</code>	number of bins to use for continuous variables in testing bivariate normality; the default is 4.
<code>pd</code>	if TRUE and if the correlation matrix is not positive-definite, an attempt will be made to adjust it to a positive-definite matrix, using the <a href="#">nearPD</a> function in the <code>Matrix</code> package. Note that default arguments to <code>nearPD</code> are used (except <code>corr=TRUE</code> ); for more control call <code>nearPD</code> directly.
<code>use</code>	if <code>"complete.obs"</code> , remove observations with any missing data; if <code>"pairwise.complete.obs"</code> , compute each correlation using all observations with valid data for that pair of variables.
<code>x</code>	an object of class <code>"hetcor"</code> to be printed, or from which to extract the correlation matrix.
<code>digits</code>	number of significant digits.

**Value**

Returns an object of class "hetcor" with the following components:

correlations	the correlation matrix.
type	the type of each correlation: "Pearson", "Polychoric", or "Polyserial".
std.errors	the standard errors of the correlations, if requested.
n	the number (or numbers) of observations on which the correlations are based.
tests	p-values for tests of bivariate normality for each pair of variables.
NA.method	the method by which any missing data were handled: "complete.obs" or "pairwise.complete.obs".
ML	TRUE for ML estimates, FALSE for two-step estimates.

**Warning**

Be careful with character variables (as opposed to factors), the values of which are ordered alphabetically. Thus, e.g., the values "disagree", "neutral", "agree" are ordered "agree", "disagree", "neutral".

**Note**

Although the function reports standard errors for product-moment correlations, transformations (the most well known is Fisher's  $z$ -transformation) are available that make the approach to asymptotic normality much more rapid.

**Author(s)**

John Fox <jfox@mcmaster.ca>

**References**

- Drasgow, F. (1986) Polychoric and polyserial correlations. Pp. 68-74 in S. Kotz and N. Johnson, eds., *The Encyclopedia of Statistics, Volume 7*. Wiley.
- Olsson, U. (1979) Maximum likelihood estimation of the polychoric correlation coefficient. *Psychometrika* **44**, 443-460.
- Rodriguez, R.N. (1982) Correlation. Pp. 193-204 in S. Kotz and N. Johnson, eds., *The Encyclopedia of Statistics, Volume 2*. Wiley.
- Ghosh, B.K. (1966) Asymptotic expansion for the moments of the distribution of correlation coefficient. *Biometrika* **53**, 258-262.
- Olkin, I., and Pratt, J.W. (1958) Unbiased estimation of certain correlation coefficients. *Annals of Mathematical Statistics* **29**, 201-211.

**See Also**

[polychor](#), [polyserial](#), [nearPD](#)

**Examples**

```

if(require(mvtnorm)){
  set.seed(12345)
  R <- matrix(0, 4, 4)
  R[upper.tri(R)] <- runif(6)
  diag(R) <- 1
  R <- cov2cor(t(R) %*% R)
  round(R, 4) # population correlations
  data <- rmvnorm(1000, rep(0, 4), R)
  round(cor(data), 4) # sample correlations
}
if(require(mvtnorm)){
  x1 <- data[,1]
  x2 <- data[,2]
  y1 <- cut(data[,3], c(-Inf, .75, Inf))
  y2 <- cut(data[,4], c(-Inf, -1, .5, 1.5, Inf))
  data <- data.frame(x1, x2, y1, y2)
  hetcor(data) # Pearson, polychoric, and polyserial correlations, 2-step est.
}
if(require(mvtnorm)){
  hetcor(x1, x2, y1, y2, ML=TRUE) # Pearson, polychoric, polyserial correlations, ML est.
}

```

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 polychor

*Polychoric Correlation*


---

**Description**

Computes the polychoric correlation (and its standard error) between two ordinal variables or from their contingency table, under the assumption that the ordinal variables dissect continuous latent variables that are bivariate normal. Either the maximum-likelihood estimator or a (possibly much) quicker “two-step” approximation is available. For the ML estimator, the estimates of the thresholds and the covariance matrix of the estimates are also available.

**Usage**

```
polychor(x, y, ML = FALSE, control = list(), std.err = FALSE, maxcor=.9999)
```

**Arguments**

x	a contingency table of counts or an ordered categorical variable; the latter can be numeric, logical, a factor, an ordered factor, or a character variable, but if a factor, its levels should be in proper order, and the values of a character variable are ordered alphabetically.
y	if x is a variable, a second ordered categorical variable.
ML	if TRUE, compute the maximum-likelihood estimate; if FALSE, the default, compute a quicker “two-step” approximation.
control	optional arguments to be passed to the <code>optim</code> function.

std.err	if TRUE, return the estimated variance of the correlation (for the two-step estimator) or the estimated covariance matrix (for the ML estimator) of the correlation and thresholds; the default is FALSE.
maxcor	maximum absolute correlation (to insure numerical stability).

### Details

The ML estimator is computed by maximizing the bivariate-normal likelihood with respect to the thresholds for the two variables ( $\tau_i^x, i = 1, \dots, r - 1$ ;  $\tau_j^y, j = 1, \dots, c - 1$ ) and the population correlation ( $\rho$ ). Here,  $r$  and  $c$  are respectively the number of levels of  $x$  and  $y$ . The likelihood is maximized numerically using the `optim` function, and the covariance matrix of the estimated parameters is based on the numerical Hessian computed by `optim`.

The two-step estimator is computed by first estimating the thresholds ( $\tau_i^x, i = 1, \dots, r - 1$  and  $\tau_j^y, i = j, \dots, c - 1$ ) separately from the marginal distribution of each variable. Then the one-dimensional likelihood for  $\rho$  is maximized numerically, using `optim` if standard errors are requested, or `optimise` if they are not. The standard error computed treats the thresholds as fixed.

### Value

If `std.err` is TRUE, returns an object of class "polychor" with the following components:

type	set to "polychoric".
rho	the polychoric correlation.
row.cuts	estimated thresholds for the row variable (x), for the ML estimate.
col.cuts	estimated thresholds for the column variable (y), for the ML estimate.
var	the estimated variance of the correlation, or, for the ML estimate, the estimated covariance matrix of the correlation and thresholds.
n	the number of observations on which the correlation is based.
chisq	chi-square test for bivariate normality.
df	degrees of freedom for the test of bivariate normality.
ML	TRUE for the ML estimate, FALSE for the two-step estimate.

Othwise, returns the polychoric correlation.

### Author(s)

John Fox <jfox@mcmaster.ca>

### References

- Drasgow, F. (1986) Polychoric and polyserial correlations. Pp. 68–74 in S. Kotz and N. Johnson, eds., *The Encyclopedia of Statistics, Volume 7*. Wiley.
- Olsson, U. (1979) Maximum likelihood estimation of the polychoric correlation coefficient. *Psychometrika* **44**, 443–460.

### See Also

[hetcor](#), [polyserial](#), [print.polychor](#), [optim](#)

**Examples**

```

if(require(mvtnorm)){
  set.seed(12345)
  data <- rmvnorm(1000, c(0, 0), matrix(c(1, .5, .5, 1), 2, 2))
  x <- data[,1]
  y <- data[,2]
  cor(x, y) # sample correlation
}
if(require(mvtnorm)){
  x <- cut(x, c(-Inf, .75, Inf))
  y <- cut(y, c(-Inf, -1, .5, 1.5, Inf))
  polychor(x, y) # 2-step estimate
}
if(require(mvtnorm)){
  set.seed(12345)
  polychor(x, y, ML=TRUE, std.err=TRUE) # ML estimate
}

```

polyserial

*Polyserial Correlation***Description**

Computes the polyserial correlation (and its standard error) between a quantitative variable and an ordinal variables, based on the assumption that the joint distribution of the quantitative variable and a latent continuous variable underlying the ordinal variable is bivariate normal. Either the maximum-likelihood estimator or a quicker “two-step” approximation is available. For the ML estimator the estimates of the thresholds and the covariance matrix of the estimates are also available.

**Usage**

```
polyserial(x, y, ML = FALSE, control = list(), std.err = FALSE, maxcor=.9999, bins=4)
```

**Arguments**

x	a numerical variable.
y	an ordered categorical variable; can be numeric, logical, a factor, an ordered factor, or a character variables, but if a factor, its levels should be in proper order, and the values of a character variable are ordered alphabetically.
ML	if TRUE, compute the maximum-likelihood estimate; if FALSE, the default, compute a quicker “two-step” approximation.
control	optional arguments to be passed to the <code>optim</code> function.
std.err	if TRUE, return the estimated variance of the correlation (for the two-step estimator) or the estimated covariance matrix of the correlation and thresholds (for the ML estimator); the default is FALSE.
maxcor	maximum absolute correlation (to insure numerical stability).
bins	the number of bins into which to dissect x for a test of bivariate normality; the default is 4.

## Details

The ML estimator is computed by maximizing the bivariate-normal likelihood with respect to the thresholds for  $y$  ( $\tau_j^y, i = 1, \dots, c - 1$ ) and the population correlation ( $\rho$ ). The likelihood is maximized numerically using the `optim` function, and the covariance matrix of the estimated parameters is based on the numerical Hessian computed by `optim`.

The two-step estimator is computed by first estimating the thresholds ( $\tau_j^y, i = 1, \dots, c - 1$ ) from the marginal distribution of  $y$ . Then if the standard error of  $\hat{\rho}$  is requested, the one-dimensional likelihood for  $\rho$  is maximized numerically, using `optim` if standard errors are requested; the standard error computed treats the thresholds as fixed. If the standard error isn't request,  $\hat{\rho}$  is computed directly.

## Value

If `std.err` is TRUE, returns an object of class "polycor" with the following components:

<code>type</code>	set to "polyserial".
<code>rho</code>	the polyserial correlation.
<code>cuts</code>	estimated thresholds for the ordinal variable ( $y$ ), for the ML estimator.
<code>var</code>	the estimated variance of the correlation, or, for the ML estimator, \ the estimated covariance matrix of the correlation and thresholds.
<code>n</code>	the number of observations on which the correlation is based.
<code>chisq</code>	chi-square test for bivariate normality.
<code>df</code>	degrees of freedom for the test of bivariate normality.
<code>ML</code>	TRUE for the ML estimate, FALSE for the two-step estimate.

Otherwise, returns the polyserial correlation.

## Author(s)

John Fox <jfox@mcmaster.ca>

## References

Drasgow, F. (1986) Polychoric and polyserial correlations. Pp. 68–74 in S. Kotz and N. Johnson, eds., *The Encyclopedia of Statistics, Volume 7*. Wiley.

## See Also

[hetcor](#), [polychor](#), [print.polycor](#), [optim](#)

## Examples

```
if(require(mvtnorm)){
  set.seed(12345)
  data <- rmvnorm(1000, c(0, 0), matrix(c(1, .5, .5, 1), 2, 2))
  x <- data[,1]
  y <- data[,2]
  cor(x, y) # sample correlation
```

```

    }
  if(require(mvtnorm)){
    y <- cut(y, c(-Inf, -1, .5, 1.5, Inf))
    polyserial(x, y) # 2-step estimate
  }
  if(require(mvtnorm)){
    polyserial(x, y, ML=TRUE, std.err=TRUE) # ML estimate
  }

```

---

print.polycor

*Print Method for polycor Objects*


---

### Description

print method for objects of class polycor, produced by polychor and polyserial.

### Usage

```

## S3 method for class 'polycor'
print(x, digits = max(3, getOption("digits") - 3), ...)

```

### Arguments

x	an object of class polycor, as returned by polychor or polyserial.
digits	number of significant digits to be printed.
...	not used.

### Value

Invisibly returns x; used for its side effect — i.e., printing.

### Author(s)

John Fox <jfox@mcmaster.ca>

### See Also

[polychor](#), [polyserial](#)

### Examples

```

if(require(mvtnorm)){
  set.seed(12345)
  data <- rmvnorm(1000, c(0, 0), matrix(c(1, .5, .5, 1), 2, 2))
  x <- data[,1]
  y <- data[,2]
  cor(x, y) # sample correlation
}
if(require(mvtnorm)){

```



```
x <- cut(x, c(-Inf, .75, Inf))
y <- cut(y, c(-Inf, -1, .5, 1.5, Inf))
polychor(x, y, ML=TRUE, std.err=TRUE) # polychoric correlation, ML estimate
}
```

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