

# Package ‘PoisBinOrdNor’

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**Type** Package

**Title** Data Generation with Poisson, Binary, Ordinal and Normal Components

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**Description** Generation of multiple count, binary, ordinal and normal variables simultaneously given the marginal characteristics and association structure.

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PoisBinOrdNor-package *Data Generation with Count, Binary, Ordinal and Normal Components*

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## Description

Generation of multiple count, binary, ordinal and normal variables simultaneously given the marginal characteristics and association structure based on the methodologies proposed in Demirtas et al. (2012), Demirtas and Yavuz (2015), Amatya and Demirtas (2016), Demirtas and Hedeker (2016).

## Details

Package: PoisBinOrdNor  
Type: Package  
Version: 1.6.1  
Date: 2019-05-15  
License: GPL-2 | GPL-3

PoisBinOrdNor package consists of ten functions. The function `validation.specs` validates the specified quantities to avoid obvious specification errors. The functions `corr.nn4bb`, `corr.nn4bn`, `corr.nn4on`, `corr.nn4pbo`, `corr.nn4pn`, and `corr.nn4pp` each computes the intermediate correlation coefficient for binary-binary combinations, binary-normal combinations, ordinal-normal combinations, count-binary/ordinal combinations, count-normal and count-count combinations, respectively. The function `intermat` assembles the intermediate correlation matrix for the multivariate data based on input from functions `corr.nn4bb`, `corr.nn4bn`, `corr.nn4on`, `corr.nn4pbo`, `corr.nn4pn` and `corr.nn4pp`. The engine function `genPBONdata` computes the final correlation matrix and generates mixed data in accordance with the specified marginal and correlational quantities.

## Author(s)

Hakan Demirtas, Yiran Hu, Rawan Allozi, Ran Gao

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## References

- Amatya, A. & Demirtas, H. (2015). Simultaneous generation of multivariate mixed data with Poisson and normal marginals. *Journal of Statistical Computation and Simulation*, 85(15), 3129-3139.
- Demirtas, H. & Doganay, B. (2012). Simultaneous generation of binary and normal data with specified marginal and association structures. *Journal of Biopharmaceutical Statistics*, 22(2), 223-236.
- Demirtas, H. & Hedeker, D. (2011). A practical way for computing approximate lower and upper correlation bounds. *The American Statistician*, 65(2), 104-109.

Demirtas, H. & Hedeker, D. (2016). Computing the point-biserial correlation under any underlying continuous distribution. *Communications in Statistics–Simulation and Computation*, 45(8), 2744-2751.

Demirtas, H., Hedeker, D. & Mermelstein, R. J. (2012). Simulation of massive public health data by power polynomials. *Statistics in Medicine*, 31(27), 3337-3346.

Demirtas, H. & Yavuz, Y. (2015). Concurrent generation of ordinal and normal data. *Journal of Biopharmaceutical Statistics*, 25(4), 635-650.

Ferrari, P.A. and Barberio, A. (2012). Simulating ordinal data. *Multivariate Behavioral Research*, 47(4), 566-589.

Yahav, I. & Shmueli, G. (2012). On generating multivariate Poisson data in management science applications. *Applied Stochastic Models in Business and Industry*, 28(1), 91-102.

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corr . nn4bb	<i>Finds the tetrachoric correlation based on user-specified correlation between binary variables.</i>
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## Description

This function computes the tetrachoric correlation given the correlation for a pair of binary variables (phi coefficient).

## Usage

```
corr.nn4bb(p1, p2, BB.cor)
```

## Arguments

p1	Probability parameter for the first binary variable.
p2	Probability parameter for the second binary variable.
BB.cor	Pre-specified correlation for a pair of binary variables.

## Value

A tetrachoric correlation coefficient.

## References

Demirtas, H. & Doganay, B. (2012). Simultaneous generation of binary and normal data with specified marginal and association structures. *Journal of Biopharmaceutical Statistics*, 22(2), 223-236.

## Examples

```
## Not run:
corr.nn4bb(0.43, 0.7, 0.129)
```

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

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corr.nn4bn	<i>Finds the biserial correlation given the correlation for a binary-normal pair.</i>
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### Description

This function computes the biserial correlation given the specified correlation for a pair of binary and normal variables (point-biserial correlation).

### Usage

```
corr.nn4bn(p, BN.cor)
```

### Arguments

p	Probability parameter for the binary variable.
BN.cor	Pre-specified correlation for a pair of binary and normal variables.

### Value

A biserial correlation coefficient.

### Examples

```
## Not run:
corr.nn4bn(0.43, 0.12)

## End(Not run)
```

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corr.nn4on	<i>Finds polyserial correlation for given the correlation for an ordinal-normal pair.</i>
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### Description

This function computes the polyserial correlation given the specified correlation for a pair of ordinal and normal variables (point-polyserial correlation).

### Usage

```
corr.nn4on(p, ON.cor)
```

**Arguments**

p	A vector of probabilities for an ordinal variable. The i-th element of the pvec is the cumulative probability defining the marginal distribution of the ordinal variable. If the variable has k categories, the i-th element of p will contain k-1 probabilities. The k-th element is implicitly 1.
ON.cor	Pre-specified correlation for a pair of ordinal-normal variables.

**Value**

A tetrachoric correlation coefficient.

**Examples**

```
## Not run:
corr.nn4on(c(0.33, 0.66), 0.22)

## End(Not run)
```

---

corr.nn4pbo	<i>Finds the underlying bivariate normal correlation given the correlation for a count-binary or count-ordinal pair.</i>
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**Description**

This function computes the underlying bivariate normal correlation given the correlation for a pair of count and binary variables or a pair of count and ordinal variables.

**Usage**

```
corr.nn4pbo(lam, p, P0.cor)
```

**Arguments**

lam	Rate parameter for the count variable.
p	A vector of probabilities for an ordinal variable. The i-th element of the pvec is the cumulative probability defining the marginal distribution of the ordinal variable. If the variable has k categories, the i-th element of p will contain k-1 probabilities. The k-th element is implicitly 1.
P0.cor	Pre-specified correlation for a pair of count and binary, or count and ordinal, variables.

**Value**

A tetrachoric correlation coefficient.

## References

Amatya, A. & Demirtas, H. (2015). Simultaneous generation of multivariate mixed data with Poisson and normal marginals. *Journal of Statistical Computation and Simulation*, 85(15), 3129-3139.

Yahav, I. & Shmueli, G. (2012). On generating multivariate Poisson data in management science applications. *Applied Stochastic Models in Business and Industry*, 28(1), 91-102.

## Examples

```
## Not run:
corr.nn4pbo(0.5, c(0.2, 0.5), 0.235)

## End(Not run)
```

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corr.nn4pn	<i>Finds the underlying bivariate normal correlation given the correlation for a count-normal pair.</i>
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## Description

This function computes the underlying bivariate normal correlation given the specified correlation for a pair of count and normal variables.

## Usage

```
corr.nn4pn(lam, PN.cor)
```

## Arguments

lam	Rate parameter for the count variable.
PN.cor	Pre-specified correlation for a pair of count and normal variables.

## Value

Correlation of underlying bivariate normal data.

## Examples

```
## Not run:
corr.nn4pn(0.5, 0.32)

## End(Not run)
```

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corr.nn4pp	<i>Finds the underlying bivariate normal correlation given the correlation for a pair of count variables.</i>
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### Description

This function computes the underlying bivariate normal correlation given the specified correlation for a pair of count variables.

### Usage

```
corr.nn4pp(lambda1, lambda2, PP.cor)
```

### Arguments

lambda1	Rate parameter for the first count variable.
lambda2	Rate parameter for the second count variable.
PP.cor	Pre-specified correlation for a pair of count variables.

### Value

Correlation of underlying bivariate normal data.

### References

Amatya, A. & Demirtas, H. (2015). Simultaneous generation of multivariate mixed data with Poisson and normal marginals. *Journal of Statistical Computation and Simulation*, 85(15), 3129-3139.

### Examples

```
## Not run:  
corr.nn4pp(0.5, 2, 0.4)  
  
## End(Not run)
```

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genPBONdata	<i>Generates correlated data with multiple count, binary, ordinal and normal variables</i>
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### Description

This function simulates a multivariate data set that is composed of count, binary, ordinal and normal variables with specified marginals and a correlation matrix.

**Usage**

```
genPBONdata(n, no_pois, no_bin, no_ord, no_norm, inter.mat, lamvec, prop_vec_bin,
  prop_vec_ord, nor.mean, nor.var)
```

**Arguments**

n	Number of rows
no_pois	Number of count variables
no_bin	Number of binary variables
no_ord	Number of ordinal variables
no_norm	Number of normal variables
inter.mat	The intermediate correlation matrix obtained from function intermat
lamvec	A vector of marginal rates for the count variables
prop_vec_bin	A vector of probabilities for the binary variables
prop_vec_ord	A vector of probabilities for the ordinal variables. For each of the variable, the i-th element of the pvec is the cumulative probability defining the marginal distribution of the ordinal variable. If the variable has k categories, the i-th element of p will contain k-1 probabilities. The k-th element is implicitly 1.
nor.mean	A vector of means for the normal variables
nor.var	A vector of variances for the normal variables

**Value**

data	A simulated data matrix of size $n \times (\text{no\_pois} + \text{no\_bin} + \text{no\_ord} + \text{no\_norm})$ , of which the first no_pois are count variables, followed by no_bin binary variables, no_ord ordinal variables, and lastly no_norm normal variables.
n.rows	Number of rows in the simulated data
prob.bin	A vector of probabilities for the binary variables
prob.ord	A vector of probabilities for the ordinal variables
nor.mean	A vector of means for the normal variables
nor.var	A vector of variances for the normal variables
lamvec	A vector of rate parameters for the count variables
n.pois	Number of count variables
n.bin	Number of binary variables
n.ord	Number of ordinal variables
n.norm	Number of normal variables
final.corr	The final correlation matrix for the simulated data



**Examples**

```
## Not run:

ss=10000
num_pois<-2
num_bin<-1
num_ord<-2
num_norm<-1

lamvec=sample(10,2)
pbin=runif(1)
pord=list(c(0.1, 0.9), c(0.2, 0.3, 0.5))
nor.mean=3.1
nor.var=0.85
M=c(-0.05, 0.26, 0.14, 0.09, 0.14, 0.12, 0.13, -0.02, 0.17, 0.29,
-0.04, 0.19, 0.10, 0.35, 0.39)
N=diag(6)
N[lower.tri(N)]=M
TV=N+t(N)
diag(TV)<-1
intmat<-intermat(num_pois,num_bin,num_ord,num_norm,corr_mat=TV,pbin,pord,lamvec,
nor.mean,nor.var)

genPBONdata(ss,num_pois,num_bin,num_ord,num_norm,intmat,lamvec,pbin,pord,nor.mean,nor.var)

## End(Not run)
```

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intermat	<i>Calculates and assembles the intermediate correlation matrix entries for the multivariate normal data.</i>
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**Description**

This function computes and assembles the correlation entries for the intermediate multivariate normal data.

**Usage**

```
intermat(no_pois, no_bin, no_ord, no_norm, corr_mat, prop_vec_bin, prop_vec_ord,
lam_vec, nor_mean, nor_var)
```

**Arguments**

no_pois	Number of the count variables.
no_bin	Number of the binary variables.
no_ord	Number of the ordinal variables.
no_norm	Number of the normal variables.
corr_mat	Pre-specified correlation matrix for the multivariate data.

prop_vec_bin	Vector of probabilities for the binary variables.
prop_vec_ord	Vector of probabilities for the ordinal variables.
lam_vec	Vector of rate parameters for the count variables.
nor_mean	Vector of means for the normal variables.
nor_var	Vector of variances for the normal variables.

**Value**

The intermediate correlation matrix that will be used later for multivariate normal data simulation.

**References**

- Barberio, A. & Ferrari, P.A. (2015). GenOrd: Simulation of discrete random variables with given correlation matrix and marginal distributions. <https://cran.r-project.org/web/packages/GenOrd/index.html>.
- Demirtas, H. & Hedeker, D. (2011). A practical way for computing approximate lower and upper correlation bounds. *American Statistician*, 65(2), 104-109.
- Demirtas, H. & Hedeker, D. (2016). Computing the point-biserial correlation under any underlying continuous distribution. *Communications in Statistics–Simulation and Computation*, 45(8), 2744-2751.
- Ferrari, P.A. and Barberio, A. (2012). Simulating ordinal data. *Multivariate Behavioral Research*, 47(4), 566-589.

**See Also**

[corr.nn4bb](#), [corr.nn4bn](#), [corr.nn4on](#), [corr.nn4pbo](#), [corr.nn4pn](#), [corr.nn4pp](#), and [validation.specs](#).

**Examples**

```
## Not run:
num_pois<-2
num_bin<-1
num_ord<-2
num_norm<-1
lamvec=sample(10,2)
pbin=runif(1)
pord=list(c(0.3, 0.7), c(0.2, 0.3, 0.5))
nor.mean=3.1
nor.var=0.85
M=
c(-0.05, 0.26, 0.14, 0.09, 0.14, 0.12, 0.13, -0.02, 0.17, 0.29, -0.04, 0.19, 0.10, 0.35, 0.39)
N=diag(6)
N[lower.tri(N)]=M
TV=N+t(N)
diag(TV)<-1
intmat<-
intermat(num_pois,num_bin,num_ord,num_norm,corr_mat=TV,pbin,pord,lamvec,nor.mean,nor.var)

## End(Not run)
```

---

validation.specs	<i>Validates user-specified parameters</i>
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### Description

This function checks the validity of user specified parameters including rate parameters for count variables, proportion parameters for binary and ordinary variables, mean and variance parameters for normal data, as well as the validity of entries in the correlation matrix. This function also computes the lower and upper limits for each pairwise correlation based on the marginal probabilities for range violation checks.

### Usage

```
validation.specs(no.pois, no.bin, no.ord, no.norm, corr.mat, prop.vec.bin,
prop.vec.ord, lamvec, nor.mean, nor.var)
```

```
validation_specs(no.pois, no.bin, no.ord, no.norm, corr.mat, prop.vec.bin,
prop.vec.ord, lamvec, nor.mean, nor.var) #deprecated
```

### Arguments

no.pois	Number of count variables.
no.bin	Number of binary variables.
no.ord	Number of ordinal variables.
no.norm	Number of normal variables.
corr.mat	User specified correlation matrix for the multivariate data.
prop.vec.bin	Vector of probabilities corresponding to each of the binary variables.
prop.vec.ord	Vector of probabilities corresponding to each of the ordinal variables. For each of the ordinal variable, the i-th element of the probability vector is the cumulative probability defining the marginal distribution of the ordinal variable. If the variable has k categories, the i-th element of p will contain k-1 probabilities. The k-th element is implicitly 1.
lamvec	Vector of rate parameters for the count variables.
nor.mean	Vector of means for the normal variables.
nor.var	Vector of variances for the normal variables.

### Details

This function computes the lower and upper bounds for all possible pairs that involve count, binary, ordinal and normal variables.

### Value

The function returns TRUE if no specification problem is encountered. Otherwise, it returns an error message.

## References

Demirtas, H. and Hedeker, D. (2011). A practical way for computing approximate lower and upper correlation bounds. *The American Statistician*, 65(2), 104-109.

Demirtas, H., Hedeker, D., and Mermelstein, R.J. (2012). Simulation of massive public health data by power polynomials. *Statistics in Medicine*, 31(27), 3337-3346.

## Examples

```
## Not run:

num_pois<-1
num_bin<-1
num_ord<-1
num_norm<-1
lambda<-c(1)
pbin<-c(0.3)
pord<-list(c(0.3,0.6))
normean<-15
norvar<-7
corr.mat=matrix(c(1,0.2,0.1,0.3, 0.2,1,0.5,0.4, 0.1,0.5,1, 0.7, 0.3, 0.4, 0.7, 1),4,4)
validation.specs(num_pois, num_bin, num_ord, num_norm,
corr.mat, pbin, pord, lambda, normean,norvar)

num_pois<-2
num_bin<-2
num_ord<-2
num_norm<-0
lambda<-c(1,2)
pbin<-c(0.3,0.5)
pord<-list(c(0.3,0.6),c(0.5,0.6))
corr.mat=matrix(0.64,6,6)
diag(corr.mat)=1
validation.specs(num_pois, num_bin, num_ord, num_norm,
corr.mat, pbin, pord, lambda, nor.mean=NULL, nor.var=NULL)

# An example with an invalid target correlation matrix (bound violation).
num_pois<-1
num_bin<-2
num_ord<-2
num_norm<-1
lamvec=c(1)
pbin=c(0.3, 0.7)
pord=list(c(0.2, 0.5), c(0.4, 0.7, 0.8))
nor.mean=2.1
nor.var=0.75
M=c(-0.35, 0.26, 0.34, 0.09, 0.14, 0.12, 0.30, -0.02, 0.17, 0.29, -0.04, 0.19,
0.10, 0.35, 0.39)
N=diag(6)
N[lower.tri(N)]=M
TV=N+t(N)
```

```
diag(TV)<-1
validation.specs(num_pois, num_bin, num_ord, num_norm, corr.mat=TV, pbin, pord,
lamvec, normean, norvar)

# An example with a non-positive definite correlation matrix.
pbin=c(0.3, 0.7)
TV1=TV
TV1[3,2]=TV[2,3]=5
validation.specs(num_pois, num_bin, num_ord, num_norm, corr.mat=TV1, pbin, pord,
lamvec, normean, norvar)

## End(Not run)
```

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