

# Package ‘CopulaREMADA’

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**Title** Copula Mixed Effect Models for Multivariate Meta-Analysis of Diagnostic Test Accuracy Studies

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**Depends** R (>= 2.10), statmod, matlab, tensor

**Description** It has functions to implement the copula mixed models for multivariate meta-analysis of diagnostic test accuracy studies.

**License** GPL (>= 2.10)

**NeedsCompilation** no

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## R topics documented:

CopulaREMADA-package . . . . .	2
arthritis . . . . .	3
betaDG . . . . .	4
CopulaREMADA . . . . .	4
CT . . . . .	6
cvinesim . . . . .	7
dcop . . . . .	8
dvinesim . . . . .	9
hybridCopulaREMADA . . . . .	10
LAG . . . . .	12
mgrid . . . . .	13
MRI . . . . .	13
OGT . . . . .	14
pcondcop . . . . .	15
qcondcop . . . . .	16
quadVineCopulaREMADA . . . . .	17

rcop . . . . .	19
rCopulaREMADA . . . . .	20
rVineCopulaREMADA . . . . .	21
SROC . . . . .	23
tau2par . . . . .	24
telomerase . . . . .	25
vine.vuong . . . . .	26
VineCopulaREMADA . . . . .	28
vuong . . . . .	30

<b>Index</b>	<b>32</b>
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CopulaREMADA-package    *Copula Mixed Effect Models for Multivariate Meta-Analysis of Diagnostic Test Accuracy Studies*

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## Description

Fits copula mixed effect models for multivariate meta-analysis of diagnostic test accuracy studies proposed in Nikoloulopoulos (2015, 2016, 2017, 2018a, 2018b).

## Details

This package contains R functions to implement:

- The copula mixed model for meta-analysis of diagnostic test accuracy studies and produce SROC curves and summary operating points (a pair of average sensitivity and specificity) with a confidence region and a predictive region (Nikoloulopoulos, 2015). All the analyses presented in Section 7 of Nikoloulopoulos (2015) are given as code examples in the package;
- The vine copula mixed model for meta-analysis of diagnostic test accuracy studies accounting for disease prevalence (Nikoloulopoulos, 2017);
- The hybrid vine copula mixed model for meta-analysis of diagnostic test accuracy case-control and cohort studies (Nikoloulopoulos, 2016);
- The D-vine copula mixed model for meta-analysis and comparison of two diagnostic tests (Nikoloulopoulos, 2018).

## Author(s)

Aristidis K. Nikoloulopoulos.

## References

- Nikoloulopoulos, A.K. (2015) A mixed effect model for bivariate meta-analysis of diagnostic test accuracy studies using a copula representation of the random effects distribution. *Statistics in Medicine*, **34**, 3842–3865. doi: [10.1002/sim.6595](https://doi.org/10.1002/sim.6595).
- Nikoloulopoulos, A.K. (2017) A vine copula mixed effect model for trivariate meta-analysis of diagnostic test accuracy studies accounting for disease prevalence. *Statistical Methods in Medical Research*, **26**, 2270–2286. doi: [10.1177/0962280215596769](https://doi.org/10.1177/0962280215596769).

Nikoloulopoulos, A.K. (2016) Hybrid copula mixed models for combining case-control and cohort studies in meta-analysis of diagnostic tests. *Statistical Methods in Medical Research*, in press. doi: [10.1177/0962280216682376](https://doi.org/10.1177/0962280216682376).

Nikoloulopoulos, A.K. (2018a) On composite likelihood in bivariate meta-analysis of diagnostic test accuracy studies. *AStA Advances in Statistical Analysis*, **102**, 211–227. doi: [10.1007/s10182-0170299y](https://doi.org/10.1007/s10182-0170299y).

Nikoloulopoulos, A.K. (2018b) A D-vine copula mixed model for joint meta-analysis and comparison of diagnostic tests. *ArXiv e-prints*, arXiv:1805.09674. <https://arxiv.org/abs/1805.09674>.

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arthritis

*The rheumatoid arthritis data*

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## Description

Data obtained from a meta-analysis that aimed to determine whether anti-cyclic citrullinated peptide (anti-CCP) antibody identifies more accurately patients with rheumatoid arthritis than rheumatoid factor (RF) does. We include  $N = 22$  studies that assessed both RF and anti-CCP2 antibody for diagnosing rheumatoid arthritis.

## Format

A data frame with 22 observations on the following 8 variables.

**TP1** the number of true positives for RF

**FN1** the number of false negatives for RF

**FP1** the number of false positives for RF

**TN1** the number of true negatives for RF

**TP2** the number of true positives for anti-CCP2

**FN2** the number of false negatives for anti-CCP2

**FP2** the number of false positives for anti-CCP2

**TN2** the number of true negatives for anti-CCP2

## References

Nishimura, K., Sugiyama, D., Kogata, Y., et al. (2007) Meta-analysis: Diagnostic accuracy of anti-cyclic citrullinated peptide antibody and rheumatoid factor for rheumatoid arthritis. *Annals of Internal Medicine*, **146**(11), 797–808.

Dimou, N.L., Adam, M. and Bagos, P.G. (2016) A multivariate method for meta-analysis and comparison of diagnostic tests. *Statistics in Medicine*, **35**(20), 3509–3523.

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betaDG	<i>The beta-D-Glucan-data</i>
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### Description

Data on 8 cohort studies in the meta-analysis in Karageorgopoulos et al. (2011). The interest there is to assess *beta*-D-Glucan as a serum or plasma marker for the presence of invasive fungal infections.

### Usage

```
data(betaDG)
```

### Format

A data frame with 8 observations on the following 4 variables.

- TP** the number of true positives
- FN** the number of false negatives
- FP** the number of false positives
- TN** the number of true negatives

### References

Karageorgopoulos, D.E., Vouloumanou, E.K., Ntziora, F., et al. (2011) *beta*-D-Glucan assay for the diagnosis of invasive fungal infections: a meta-analysis. *Clinical Infectious Diseases*, **52**(6), 750–770.

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CopulaREMADA	<i>Maximum likelihood estimation for copula mixed models for diagnostic test accuracy studies</i>
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### Description

For copula mixed models for diagnostic test accuracy studies numerical evaluation of the MLE is easily done with the following steps:

1. Calculate Gauss-Legendre quadrature points `gl$nodes` and weights `gl$weights`.
2. Convert from independent uniform quadrature points to dependent uniform quadrature points that have distribution 'cop'. The inverse of the conditional distribution `qcondcop` corresponding to the copula 'cop' is used to achieve this.
3. Numerically evaluate the joint probability mass function with the bivariate integral in a double sum.

With Gauss-Legendre quadrature, the same nodes and weights are used for different functions; this helps in yielding smooth numerical derivatives for numerical optimization via quasi-Newton. Our comparisons show that  $n_q = 15$  is adequate with good precision to at least at four decimal places.

**Usage**

```
CopulaREMADA.norm(TP, FN, FP, TN, gl, mgrid, qcond, tau2par)
CopulaREMADA.beta(TP, FN, FP, TN, gl, mgrid, qcond, tau2par)
countermonotonicCopulaREMADA.norm(TP, FN, FP, TN, gl, mgrid)
countermonotonicCopulaREMADA.beta(TP, FN, FP, TN, gl, mgrid)
```

**Arguments**

TP	the number of true positives
FN	the number of false negatives
FP	the number of false positives
TN	the number of true negatives
gl	a list containing the components of Gauss-Legendre nodes <code>gl\$nodes</code> and weights <code>gl\$weights</code>
mgrid	a list containing two matrices with the rows of the output matrix <code>x</code> are copies of the vector <code>gl\$nodes</code> ; columns of the output matrix <code>y</code> are copies of the vector <code>gl\$nodes</code> . For more details see also <a href="#">meshgrid</a>
qcond	function for the inverse of conditional copula cdf
tau2par	function for mapping Kendall's tau to copula parameter

**Value**

A list containing the following components:

minimum	the value of the estimated minimum of the negative log-likelihood
estimate	the MLE
gradient	the gradient at the estimated minimum of of the negative log-likelihood
hessian	the hessian at the estimated minimum of the negative log-likelihood
code	an integer indicating why the optimization process terminated
iterations	the number of iterations performed

For more details see [nlm](#)

**References**

Nikoloulopoulos, A.K. (2015) A mixed effect model for bivariate meta-analysis of diagnostic test accuracy studies using a copula representation of the random effects distribution. *Statistics in Medicine*, **34**, 3842–3865. doi: [10.1002/sim.6595](https://doi.org/10.1002/sim.6595).

**See Also**

[rCopulaREMADA](#)

**Examples**

```

nq=15
gl=gauss.quad.prob(nq,"uniform")
mgrid<- meshgrid(gl$n,gl$n)

data(LAG)
attach(LAG)
c270est.b=CopulaREMADA.beta(TP, FN, FP, TN, gl, mgrid, qcondln270, tau2par.cln270)
detach(LAG)

data(MRI)
attach(MRI)
c270est.n=CopulaREMADA.norm(TP, FN, FP, TN, gl, mgrid, qcondln270, tau2par.cln270)
detach(MRI)

data(CT)
attach(CT)
est.n=countermonotonicCopulaREMADA.norm(TP, FN, FP, TN, gl, mgrid)
est.b=countermonotonicCopulaREMADA.beta(TP, FN, FP, TN, gl, mgrid)
detach(CT)

```

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CT

*The computing tomography data*


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**Description**

Data on 17 studies of computed tomography (CT) for the diagnosis of lymph node metastasis in women with cervical cancer, one of three imaging techniques in the meta-analysis in Scheidler et al. (1997). Diagnosis of metastatic disease by CT relies on nodal enlargement.

**Usage**

```
data(CT)
```

**Format**

A data frame with 17 observations on the following 4 variables.

**TP** the number of true positives  
**FN** the number of false negatives  
**FP** the number of false positives  
**TN** the number of true negatives

**References**

Scheidler, J., H. Hricak, K. K. Yu, L. Subak, and M. R. Segal. (1997) Radiological evaluation of lymph node metastases in patients with cervical cancer: A meta-analysis. *Journal of the American Medical Association*, **278**, 1096–1101.

cvinesim

*Simulation from a trivariate C-vine copula***Description**

Simulation from a trivariate C-vine copula

**Usage**

```
cvinesim(N,param,qcondcop12,qcondcop13,qcondcop23,
         tau2par12,tau2par13,tau2par23)
```

**Arguments**

N	sample size
param	Kendall's tau values for margins (1,2), (1,3), (2,3 1)
qcondcop12	function for the inverse of conditional copula cdf at the (1,2) bivariate margin
qcondcop13	function for the inverse of conditional copula cdf at the (1,3) bivariate margin
qcondcop23	function for the inverse of conditional copula cdf at the (2,3 1) bivariate margin
tau2par12	function for mapping Kendall's tau at the (1,2) bivariate margin to copula parameter
tau2par13	function for mapping Kendall's tau at the (1,3) bivariate margin to copula parameter
tau2par23	function for mapping Kendall's tau at the (2,3 1) bivariate margin to the conditional copula parameter

**Details**

Choices are 'cop' in rcop are `bvn`, `frk`, `cln`, `cln90` (rotated by 90 degrees `cln`), `cln180` (rotated by 180 degrees `cln`), `cln270` (rotated by 270 degrees `cln`).

See help page for `dcop` for the abbreviations of the copula names.

**Value**

$N \times 3$  matrix with values in (0,1)

**References**

Joe H (2011) Dependence comparisons of vine copulae with four or more variables. In: Kurowicka D, Joe H, editors. *Dependence Modeling: Handbook on Vine Copulae*. Singapore: World Scientific; 2011. p. 139–164

Joe H (2014) *Dependence Modeling with Copulas*. Chapman & Hall

Joe H (2014) *CopulaModel: Dependence Modeling with Copulas*. <http://copula.stat.ubc.ca/>.

**See Also**[qcondcop](#) [dcop](#) [rcop](#)

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`dcop`*Bivariate copula densities*

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**Description**

Bivariate copula densities for parametric families.

**Usage**

```
dbvn(u, v, cpar)
dfrk(u, v, cpar)
dcln(u, v, cpar)
dcln90(u, v, cpar)
dcln270(u, v, cpar)
```

**Arguments**

<code>u</code>	value in interval 0,1; could be a vector
<code>v</code>	value in interval 0,1; could be a vector
<code>cpar</code>	copula parameter: scalar.

**Details**

Choices are 'cop' in `dcop` are `bvn`, `frk`, `cln`, `cln90` (rotated by 90 degrees `cln`), `cln180` (rotated by 180 degrees `cln`), `cln270` (rotated by 270 degrees `cln`).

The copula names are abbreviations for:

`bvn` = bivariate normal or Gaussian

`frk` = Frank

`cln` = Clayton or Mardia-Takahasi-Cook-Johnson

**Value**

pdf value(s).

**References**

Joe H (1997) *Multivariate Models and Dependence Concepts*. Chapman & Hall

Joe H (2014) *Dependence Modeling with Copulas*. Chapman & Hall

Joe H (2014) *CopulaModel: Dependence Modeling with Copulas*. <http://copula.stat.ubc.ca/>.

**See Also**[qcondcop](#) [rcop](#)



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dvinesim	Simulation from a quadrivariate D-vine copula
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### Description

Simulation from a quadrivariate D-vine copula

### Usage

```
dvinesim(nsim,param,qcond1,pcond1,tau2par1,qcond2,pcond2,tau2par2)
```

### Arguments

nsim	sample size
param	Kendall's tau values for margins (1,2), (2,3), (3,4), (1,3 2), (2,4 3), (1,4 23)
qcond1	function for the inverse conditional copula cdf at the (1,2) bivariate margin
pcond1	function for the conditional copula cdf at the (1,2) bivariate margin
tau2par1	function for mapping Kendall's tau at the (1,2) bivariate margin to copula parameter
qcond2	function for the inverse conditional copula cdf at the (3,4) bivariate margin
pcond2	function for the conditional copula cdf at the (3,4) bivariate margin
tau2par2	function for mapping Kendall's tau at the (3,4) bivariate margin to copula parameter

### Details

Choices are 'cop' in `rcop` are `bvn`, `frk`, `cln`, `cln90` (rotated by 90 degrees `cln`), `cln180` (rotated by 180 degrees `cln`), `cln270` (rotated by 270 degrees `cln`).

See help page for `dcop` for the abbreviations of the copula names.

### Value

$N \times 4$  matrix with values in (0,1)

### References

Joe H (2011) Dependence comparisons of vine copulae with four or more variables. In: Kurowicka D, Joe H, editors. *Dependence Modeling: Handbook on Vine Copulae*. Singapore: World Scientific; 2011. p. 139–164

Joe H (2014) *Dependence Modeling with Copulas*. Chapman & Hall

Joe H (2014) *CopulaModel: Dependence Modeling with Copulas*. <http://copula.stat.ubc.ca/>.

Nikoloulopoulos, A.K. (2018b) A D-vine copula mixed model for joint meta-analysis and comparison of diagnostic tests. *ArXiv e-prints*, arXiv:1805.09674. <https://arxiv.org/abs/1805.09674>.

**See Also**

[qcondcop](#) [dcop](#) [rcop](#)

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hybridCopulaREMADA	<i>Maximum likelihood estimation for hybrid copula mixed models for combining case-control and cohort studies in meta-analysis of diagnostic tests</i>
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**Description**

The estimated parameters can be obtained by using a quasi-Newton method applied to the logarithm of the joint likelihood. This numerical method requires only the objective function, i.e., the logarithm of the joint likelihood, while the gradients are computed numerically and the Hessian matrix of the second order derivatives is updated in each iteration. The standard errors (SE) of the ML estimates can be also obtained via the gradients and the Hessian computed numerically during the maximization process.

**Usage**

```
hybridCopulaREMADA.norm(TP, FN, FP, TN, type, perm, gl, mgrid1, mgrid2,
                        qcondcop12, qcondcop13,
                        tau2par12, tau2par13, qcond, tau2par)
hybridCopulaREMADA.beta(TP, FN, FP, TN, type, perm, gl, mgrid1, mgrid2,
                        qcondcop12, qcondcop13,
                        tau2par12, tau2par13, qcond, tau2par)
```

**Arguments**

TP	the number of true positives
FN	the number of false negatives
FP	the number of false positives
TN	the number of true negatives
type	a scalar indicating the study type: 1: Cohort study; 2: Case-control study.
perm	a scalar indicating the selected permutation of indices: 1: Pilot variable is the number of TP. The bivariate margins are 12, 13, 23 1; 2: Pilot variable is the number of TN. The bivariate margins are 23, 12, 13 2; 3: Pilot variable is the TP+FN. The bivariate margins are 13, 23, 12 3; 1:TP, 2:TN, 3:TP+FN
gl	a list containing the components of Gauss-Legendre nodes <code>gl\$nodes</code> and weights <code>gl\$weights</code>
mgrid1	a list containing three-dimensional arrays. For more details see <a href="#">meshgrid</a>

mgrid2	a list containing two matrices with the rows of the output matrix x are copies of the vector gl\$nodes; columns of the output matrix y are copies of the vector gl\$nodes. For more details see also <a href="#">meshgrid</a>
qcondcop12	function for the inverse of conditional copula cdf at the (1,2) bivariate margin of the vine
qcondcop13	function for the inverse of conditional copula cdf at the (1,3) bivariate margin of the vine
tau2par12	function for mapping Kendall's tau at the (1,2) bivariate margin of the vine to copula parameter
tau2par13	function for mapping Kendall's tau at the (1,3) bivariate margin of the vine to copula parameter
qcond	function for the inverse of conditional copula cdf
tau2par	function for mapping Kendall's tau to the bivariate copula parameter

**Value**

A list containing the following components:

minimum	the value of the estimated minimum of the negative log-likelihood
estimate	the MLE
gradient	the gradient at the estimated minimum of of the negative log-likelihood
hessian	the hessian at the estimated minimum of the negative log-likelihood
code	an integer indicating why the optimization process terminated
iterations	the number of iterations performed

For more details see [nlm](#)

**References**

Nikoloulopoulos, A.K. (2016) Hybrid copula mixed models for combining case-control and cohort studies in meta-analysis of diagnostic tests. *Statistical Methods in Medical Research*, in press. doi: [10.1177/0962280216682376](https://doi.org/10.1177/0962280216682376).

**See Also**

[VineCopulaREMADA](#), [CopulaREMADA](#)

**Examples**

```
# simulate the data from N=25 cohort studies
N=25
p=c(0.8,0.7,0.4)
g=c(0.1,0.1,0.05)
taus=c(-0.5,-0.3,-0.0001)
qcondcop12=qcondcop23=qcondcop13=qcondc1n90
tau2par12=tau2par23=tau2par13=tau2par.c1n90
simdat1=rVineCopulaREMADA.beta(N,p,g,taus,qcondcop12,qcondcop13,qcondcop23,
```

```

        tau2par12,tau2par13,tau2par23)
# simulate data from the N=25 case-control studies
tau=0.5
p=p[-3]
g=g[-3]
simdat2=rCopulaREMADA.beta(N,p,g,tau,rc1n,tau2par.c1n)
# combine the data
TP=c(simdat1$TP,simdat2$TP)
TN=c(simdat1$TN,simdat2$TN)
FP=c(simdat1$FP,simdat2$FP)
FN=c(simdat1$FN,simdat2$FN)
type=rep(c(1,2),each=N)

# fit the hybrid copula mixed model
nq=21
gl=gauss.quad.prob(nq,"uniform")
mgrid1<- meshgrid(gl$n,gl$n,gl$n,nargout=3)
mgrid2<- meshgrid(gl$n,gl$n)

perm=1
qcond=qcondc1n
tau2par=tau2par.c1n
# est=hybridCopulaREMADA.beta(TP,FN,FP,TN,type,perm,gl,mgrid1,mgrid2,
# qcondcop12,qcondcop13,tau2par12,tau2par13,qcond,tau2par)

```

---

LAG

*The lymphangiography data*


---

### Description

Data on 17 studies of lymphangiography (LAG) for the diagnosis of lymph node metastasis in women with cervical cancer, one of three imaging techniques in the meta-analysis in Scheidler et al. (1997). Diagnosis of metastatic disease by LAG is based on the presence of nodal-filling defects.

### Usage

```
data(LAG)
```

### Format

A data frame with 17 observations on the following 4 variables.

**TP** the number of true positives

**FN** the number of false negatives

**FP** the number of false positives

**TN** the number of true negatives

## References

Scheidler, J., H. Hricak, K. K. Yu, L. Subak, and M. R. Segal. (1997) Radiological evaluation of lymph node metastases in patients with cervical cancer: A meta-analysis. *Journal of the American Medical Association*, **278**, 1096–1101.

---

mgrid	<i>A list containing four-dimensional arrays</i>
-------	--

---

## Description

A list containing four-dimensional arrays. Replicates of the quadrature points that produce an 4-dimensional full grid.

## Examples

```
data(mgrid15)
dim(mgrid15$x)
dim(mgrid15$y)
dim(mgrid15$z)
dim(mgrid15$w)
```

```
data(mgrid30)
dim(mgrid30$x)
dim(mgrid30$y)
dim(mgrid30$z)
dim(mgrid30$w)
```

```
data(mgrid50)
dim(mgrid50$x)
dim(mgrid50$y)
dim(mgrid50$z)
dim(mgrid50$w)
```

---

MRI	<i>The magnetic resonance imaging data</i>
-----	--

---

## Description

Data on 10 studies of magnetic resonance imaging (MRI) for the diagnosis of lymph node metastasis in women with cervical cancer, the last imaging technique in the meta-analysis in Scheidler et al. (1997). Diagnosis of metastatic disease by MRI relies on nodal enlargement.

## Usage

```
data(MRI)
```

**Format**

A data frame with 10 observations on the following 4 variables.

**TP** the number of true positives

**FN** the number of false negatives

**FP** the number of false positives

**TN** the number of true negatives

**References**

Scheidler, J., H. Hricak, K. K. Yu, L. Subak, and M. R. Segal. (1997) Radiological evaluation of lymph node metastases in patients with cervical cancer: A meta-analysis. *Journal of the American Medical Association*, **278**, 1096–1101.

---

OGT

*The orale glucose tolerance data*

---

**Description**

Data on 10 studies of the oral glucose tolerance test for the diagnosis of diabetes mellitus in patients during acute coronary syndrome hospitalization in Ye et al. (2012).

**Usage**

data(OGT)

**Format**

A data frame with 10 observations on the following 4 variables.

**TP** the number of true positives

**FN** the number of false negatives

**FP** the number of false positives

**TN** the number of true negatives

**References**

Ye, Y., Xie, H., Zhao, X., Zhang, S. (2012) The oral glucose tolerance test for the diagnosis of diabetes mellitus in patients during acute coronary syndrome hospitalization: a meta-analysis of diagnostic test accuracy. *Cardiovascular Diabetology*, **11**(5):155.

---

pcondcop

*Bivariate copula conditional distribution functions*

---

## Description

Bivariate copula conditional distribution functions

## Usage

```
pcondbvn(v, u, cpar)
pcondfrk(v, u, cpar)
pcondcln(v, u, cpar)
pcondcln90(v, u, cpar)
pcondcln270(v, u, cpar)
```

## Arguments

v	conditioning value in interval 0,1; could be a vector
u	value in interval 0,1; could be a vector
cpar	copula parameter: scalar.

## Details

Choices appending 'cop' are bvn, frk, cln, cln90 (rotated by 90 degrees cln), cln180 (rotated by 180 degrees cln), cln270 (rotated by 270 degrees cln).

See help page for [dcop](#) for the abbreviations of the copula names.

## Value

inverse conditional cdf value(s)

## References

Joe H (1997) *Multivariate Models and Dependence Concepts*. Chapman & Hall

Joe H (2014) *Dependence Modeling with Copulas*. Chapman & Hall

Joe H (2014) *CopulaModel: Dependence Modeling with Copulas*. <http://copula.stat.ubc.ca/>.

## See Also

[dcop](#) [rcop](#)

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`qcondcop`*Bivariate copula conditional quantile functions*

---

**Description**

Bivariate copula conditional quantile functions

**Usage**

```
qcondbvn(p, u, cpar)
qcondfrk(p, u, cpar)
qcondcln(p, u, cpar)
qcondcln90(p, u, cpar)
qcondcln270(p, u, cpar)
```

**Arguments**

<code>u</code>	conditioning value in interval 0,1; could be a vector
<code>p</code>	quantile in interval 0,1; could be a vector
<code>cpar</code>	copula parameter: scalar.

**Details**

Choices appending 'cop' are bvn, frk, cln, cln90 (rotated by 90 degrees cln), cln180 (rotated by 180 degrees cln), cln270 (rotated by 270 degrees cln).

See help page for [dcop](#) for the abbreviations of the copula names.

**Value**

inverse conditional cdf value(s)

**References**

Joe H (1997) *Multivariate Models and Dependence Concepts*. Chapman & Hall

Joe H (2014) *Dependence Modeling with Copulas*. Chapman & Hall

Joe H (2014) *CopulaModel: Dependence Modeling with Copulas*. <http://copula.stat.ubc.ca/>.

**See Also**

[dcop](#) [rcop](#)



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quadVineCopulaREMADA *Maximum likelihood estimation for quadrivariate D-vine copula mixed models for joint meta-analysis and comparison of two diagnostic tests*

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## Description

The estimated parameters can be obtained by using a quasi-Newton method applied to the logarithm of the joint likelihood. This numerical method requires only the objective function, i.e., the logarithm of the joint likelihood, while the gradients are computed numerically and the Hessian matrix of the second order derivatives is updated in each iteration. The standard errors (SE) of the ML estimates can be also obtained via the gradients and the Hessian computed numerically during the maximization process.

## Usage

```
quadVineCopulaREMADA.norm(TP1, FN1, FP1, TN1, TP2, FN2, FP2, TN2,
                          gl, mgrid, qcond1, pcond1, tau2par1,
                          qcond2, pcond2, tau2par2)

quadVineCopulaREMADA.beta(TP1, FN1, FP1, TN1, TP2, FN2, FP2, TN2,
                          gl, mgrid, qcond1, pcond1, tau2par1,
                          qcond2, pcond2, tau2par2)

quadVineCopulaREMADA.norm.beta(TP1, FN1, FP1, TN1, TP2, FN2, FP2, TN2,
                                gl, mgrid, qcond1, pcond1, tau2par1,
                                qcond2, pcond2, tau2par2)
```

## Arguments

TP1	the number of true positives for test 1
FN1	the number of false negatives for test 1
FP1	the number of false positives for test 1
TN1	the number of true negatives for test 1
TP2	the number of true positives for test 2
FN2	the number of false negatives for test 2
FP2	the number of false positives for test 2
TN2	the number of true negatives for test 2
gl	a list containing the components of Gauss-Legendre nodes <code>gl\$nodes</code> and weights <code>gl\$weights</code>
mgrid	a list containing 4-dimensional arrays.
qcond1	function for the inverse conditional copula cdf at the (1,2) bivariate margin
pcond1	function for the conditional copula cdf at the (1,2) bivariate margin
tau2par1	function for mapping Kendall's tau at the (1,2) bivariate margin to copula parameter

qcond2	function for the inverse conditional copula cdf at the (3,4) bivariate margin
pcond2	function for the conditional copula cdf at the (3,4) bivariate margin
tau2par2	function for mapping Kendall's tau at the (3,4) bivariate margin to copula parameter

### Value

A list containing the following components:

minimum	the value of the estimated minimum of the negative log-likelihood
estimate	the MLE
gradient	the gradient at the estimated minimum of of the negative log-likelihood
hessian	the hessian at the estimated minimum of the negative log-likelihood
code	an integer indicating why the optimization process terminated
iterations	the number of iterations performed

For more details see [nlm](#)

### References

Nikoloulopoulos, A.K. (2018b) A D-vine copula mixed model for joint meta-analysis and comparison of diagnostic tests. *ArXiv e-prints*, arXiv:1805.09674. <https://arxiv.org/abs/1805.09674>.

### Examples

```
nq=15
gl=gauss.quad.prob(nq,"uniform")
data(mgrid15)

data(arthritis)
attach(arthritis)

qcond1=qcondc1n270
pcond1=pcondc1n270
tau2par1=tau2par.c1n270

qcond2=qcondfrk
pcond2=pcondfrk
tau2par2=tau2par.frk

# out<-quadVineCopulaREMADA.norm(TP1, FN1, FP1, TN1, TP2, FN2, FP2, TN2,
# gl,mgrid15,qcond1,pcond1,tau2par1,qcond2,pcond2,tau2par2)

detach(arthritis)
```

---

`rcop`*Simulation from parametric bivariate copula families*

---

**Description**

Simulation from parametric bivariate copula families

**Usage**

```
rbvn(N, cpar)
rfrk(N, cpar)
rcln(N, cpar)
rcln90(N, cpar)
rcln270(N, cpar)
```

**Arguments**

N	sample size
cpar	copula parameter: scalar

**Details**

Choices are 'cop' in rcop are bvn, frk, cln, cln90 (rotated by 90 degrees cln), cln180 (rotated by 180 degrees cln), cln270 (rotated by 270 degrees cln).

See help page for [dcop](#) for the abbreviations of the copula names.

**Value**

nx2 matrix with values in (0,1)

**References**

Joe H (1997) *Multivariate Models and Dependence Concepts*. Chapman & Hall

Joe H (2014) *Dependence Modeling with Copulas*. Chapman & Hall

Joe H (2014) *CopulaModel: Dependence Modeling with Copulas*. <http://copula.stat.ubc.ca/>.

**See Also**

[qcondcop](#) [dcop](#)

rCopulaREMADA

*Simulation from copula mixed models for diagnostic test accuracy studies***Description**

To simulate the data we have used the following steps:

1. Simulate the study size  $n$  from a shifted gamma distribution with parameters  $\alpha = 1.2, \beta = 0.01, lag = 30$  and round off to the nearest integer.
2. Simulate  $(u_1, u_2)$  from a parametric family of copulas 'cop'.
3. Convert to beta realizations or normal realizations.
4. Draw the number of diseased  $n_1$  from a  $B(n, 0.43)$  distribution.
5. Set  $n_2 = n - n_1, y_j = n_j x_j$  and then round  $y_j$  for  $j = 1, 2$ .

**Usage**

```
rCopulaREMADA.norm(N,p,si,tau,rcop,tau2par)
rCopulaREMADA.beta(N,p,g,tau,rcop,tau2par)
```

**Arguments**

N	sample size
p	Pair $(\pi_1, \pi_2)$ of sensitivity/specificity
si	Pair $(\sigma_1, \sigma_2)$ of variability; normal margins
g	Pair $(\gamma_1, \gamma_2)$ of variability; beta margins
tau	Kendall's tau value
rcop	function for copula generation
tau2par	function for mapping from Kendall's tau to copula parameter

**Value**

A list containing the following simulated components:

TP	the number of true positives
FN	the number of false negatives
FP	the number of false positives
TN	the number of true negatives

**References**

Nikoloulopoulos, A.K. (2015) A mixed effect model for bivariate meta-analysis of diagnostic test accuracy studies using a copula representation of the random effects distribution. *Statistics in Medicine*, **34**, 3842–3865. doi: [10.1002/sim.6595](https://doi.org/10.1002/sim.6595).

**See Also**

[CopulaREMADA rcop](#)

**Examples**

```
nq=15
gl=gauss.quad.prob(nq,"uniform")
mgrid<- meshgrid(gl$n,gl$n)

N=20
tau=-0.5
p=c(0.7,0.9)
g=c(0.2,0.1)
simDat=rCopulaREMADA.beta(N,p,g,tau,rcln270,tau2par.cln270)
TP=simDat$TP
TN=simDat$TN
FP=simDat$FP
FN=simDat$FN
c270est.b=CopulaREMADA.beta(TP,FN,FP,TN,gl,mgrid,qcondcln270,tau2par.cln270)

si=c(2,1)
tau=0.5
simDat=rCopulaREMADA.norm(N,p,si,tau,rcln,tau2par.cln)
TP=simDat$TP
TN=simDat$TN
FP=simDat$FP
FN=simDat$FN
cest.n=CopulaREMADA.norm(TP,FN,FP,TN,gl,mgrid,qcondcln,tau2par.cln)
```

---

rVineCopulaREMADA

*Simulation from vine copula mixed models for diagnostic test accuracy studies accounting for disease prevalence*

---

**Description**

To simulate the data we have used the following steps:

1. Simulate the study size  $n$  from a shifted gamma distribution with parameters  $\alpha = 1.2, \beta = 0.01, lag = 30$  and round off to the nearest integer.
2. Simulate  $(u_1, u_2, u_3)$  from a trivariate C-vine copula. See also [cvinesim](#).
3. Convert to beta realizations or normal realizations.
4. Set the number of diseased  $n_1 = nx_1$ .
5. Set  $n_2 = n - n_1, y_j = n_j x_j$  and then round  $y_j$  for  $j = 1, 2$ .

**Usage**

```
rVineCopulaREMADA.beta(N,p,g,taus,qcondcop12,qcondcop13,qcondcop23,
                        tau2par12,tau2par13,tau2par23)
rVineCopulaREMADA.norm(N,p,si,taus,qcondcop12,qcondcop13,qcondcop23,
                       tau2par12,tau2par13,tau2par23)
```

**Arguments**

N	sample size
p	Vector $(\pi_1, \pi_2, \pi_3)$ of sensitivity/specificity/prevalence
si	Vector $(\sigma_1, \sigma_2, \sigma_3)$ of variability; normal margins
g	Vector $(\gamma_1, \gamma_2, \gamma_3)$ of variability; beta margins
taus	Kendall's tau values
qcondcop12	function for the inverse of conditional copula cdf at the (1,2) bivariate margin
qcondcop13	function for the inverse of conditional copula cdf at the (1,3) bivariate margin
qcondcop23	function for the inverse of conditional copula cdf at the (2,3 1) bivariate margin
tau2par12	function for mapping Kendall's tau at the (1,2) bivariate margin to copula parameter
tau2par13	function for mapping Kendall's tau at the (1,3) bivariate margin to copula parameter
tau2par23	function for mapping Kendall's tau at the (2,3 1) bivariate margin to the conditional copula parameter

**Value**

A list containing the following simulated components:

TP	the number of true positives
FN	the number of false negatives
FP	the number of false positives
TN	the number of true negatives

**References**

Nikoloulopoulos, A.K. (2017) A vine copula mixed effect model for trivariate meta-analysis of diagnostic test accuracy studies accounting for disease prevalence. *Statistical Methods in Medical Research*, **26**, 2270–2286. doi: [10.1177/0962280215596769](https://doi.org/10.1177/0962280215596769).

**See Also**

[rCopulaREMADA](#) [rcop](#) [cvinesim](#)

**Examples**

```

p=c(0.8,0.7,0.4)
g=c(0.1,0.1,0.05)
taus=c(-0.5,-0.3,-0.0001)
qcondcop12=qcondcop23=qcondcop13=qcondc1n90
tau2par12=tau2par23=tau2par13=tau2par.c1n90
rVineCopulaREMADA.beta(50,p,g,taus,qcondcop12,qcondcop13,qcondcop23,
tau2par12,tau2par13,tau2par23)

```

---

SROC	<i>Summary receiver operating characteristic curves for copula mixed effect models for bivariate meta-analysis of diagnostic test accuracy studies</i>
------	--

---

**Description**

Summary receiver operating characteristic (SROC) curves are demonstrated for the proposed models through quantile regression techniques and different characterizations of the estimated bivariate random effects distribution

**Usage**

```

SROC.norm(param,dcop,qcondcop,tau2par,TP,FN,FP,TN,points=TRUE,curves=TRUE)
SROC.beta(param,dcop,qcondcop,tau2par,TP,FN,FP,TN,points=TRUE,curves=TRUE)
SROC(param.beta,param.normal,TP,FN,FP,TN)

```

**Arguments**

param	A vector with the sensitivities, specificities, variabilities and Kendall's tau value (the latter only for <a href="#">SROC.norm</a> and <a href="#">SROC.beta</a> )
param.beta	A vector with the sensitivity, specificity and variabilities of the countermonotonic CopulaREMADA with beta margins
param.normal	A vector with the sensitivity, specificity and variabilities of the countermonotonic CopulaREMADA with normal margins
dcop	function for copula density
qcondcop	function for the inverse of conditional copula cdf
tau2par	function for mapping Kendall's tau to copula parameter
TP	the number of true positives
FN	the number of false negatives
FP	the number of false positives
TN	the number of true negatives
points	logical: print individual studies
curves	logical: print quantile regression curves

**Value**

Summary receiver operating characteristic curves

**References**

Nikoloulopoulos, A.K. (2015) A mixed effect model for bivariate meta-analysis of diagnostic test accuracy studies using a copula representation of the random effects distribution. *Statistics in Medicine*, **34**, 3842–3865. doi: [10.1002/sim.6595](https://doi.org/10.1002/sim.6595).

**See Also**

[CopulaREMADA](#) [rCopulaREMADA](#)

**Examples**

```
nq=15
gl=gauss.quad.prob(nq,"uniform")
mgrid<- meshgrid(gl$n,gl$n)

data(telomerase)
attach(telomerase)
est.n=countermonotonicCopulaREMADA.norm(TP, FN, FP, TN, gl, mgrid)
est.b=countermonotonicCopulaREMADA.beta(TP, FN, FP, TN, gl, mgrid)
SROC(est.b$e, est.n$e, TP, FN, FP, TN)
detach(telomerase)

data(LAG)
attach(LAG)
c180est.b=CopulaREMADA.beta(TP, FN, FP, TN, gl, mgrid, qcondc1n180, tau2par.c1n180)
SROC.beta(c180est.b$e, d1n180, qcondc1n180, tau2par.c1n180, TP, FN, FP, TN)
detach(LAG)

data(MRI)
attach(MRI)
c270est.n=CopulaREMADA.norm(TP, FN, FP, TN, gl, mgrid, qcondc1n270, tau2par.c1n270)
SROC.norm(c270est.n$e, d1n270, qcondc1n270, tau2par.c1n270, TP, FN, FP, TN)
detach(MRI)
```

---

tau2par

---

*Mapping of Kendall's tau and copula parameter*


---

**Description**

Bivariate copulas: mapping of Kendall's tau and copula parameter.



**Usage**

```
tau2par.bvn(tau)
tau2par.frk(tau)
tau2par.cln(tau)
tau2par.cln90(tau)
tau2par.cln180(tau)
tau2par.cln270(tau)
```

**Arguments**

tau                    Kendall's tau for the copula

**Details**

For abbreviations of names of copula families (after the dot in function names), see [dcop](#) help page.

**Value**

copula parameter

**References**

Joe H (1997) *Multivariate Models and Dependence Concepts*. Chapman & Hall

Joe H (2014) *Dependence Modeling with Copulas*. Chapman & Hall

Joe H (2014) *CopulaModel: Dependence Modeling with Copulas*. <http://copula.stat.ubc.ca/>.

**See Also**

[dcop](#)

---

telomerase

*The telomerase data*

---

**Description**

In Glas et al. (2003) the telomerase marker for the diagnosis of bladder cancer is evaluated using 10 studies. The interest was to define if this non-invasive and cheap marker could replace the standard of cystoscopy or histopathology.

**Usage**

```
data(telomerase)
```

**Format**

A data frame with 10 observations on the following 4 variables.

**TP** the number of true positives

**FN** the number of false negatives

**FP** the number of false positives

**TN** the number of true negatives

**References**

Glas A.S., Roos D., Deutekom M., Zwinderman A.H., Bossuyt P.M., Kurth K.H. (2003) Tumor markers in the diagnosis of primary bladder cancer. A systematic review. *Journal of Urology*, **169**(6), 1975–82.

---

vine.vuong	<i>Vuong's test for the comparison of non-nested vine copula mixed models for diagnostic test accuracy studies</i>
------------	--

---

**Description**

Vuong (1989)'s test for the comparison of non-nested vine copula mixed models for diagnostic test accuracy studies. It shows if a vine copula mixed model provides better fit than the standard GLMM. We compute the Vuong's test with Model 1 being the vine copula mixed model with BVN copula and normal margins, i.e., the standard GLMM.

**Usage**

```
vine.vuong.beta(qcondcop12, qcondcop13, qcondcop23,
tau2par12, tau2par13, tau2par23, param1, param2, TP, FN, FP, TN, perm, gl, mgrid)
vine.vuong.norm(qcondcop12, qcondcop13, qcondcop23,
tau2par12, tau2par13, tau2par23, param1, param2, TP, FN, FP, TN, perm, gl, mgrid)
tvine.vuong.beta(qcondcop12, qcondcop13,
tau2par12, tau2par13, param1, param2, TP, FN, FP, TN, perm, gl, mgrid)
tvine.vuong.norm(qcondcop12, qcondcop13,
tau2par12, tau2par13, param1, param2, TP, FN, FP, TN, perm, gl, mgrid)
tvine2.vuong.beta(qcondcop12, qcondcop13,
tau2par12, tau2par13, param1, param2, TP, FN, FP, TN, perm, gl, mgrid)
tvine2.vuong.norm(qcondcop12, qcondcop13,
tau2par12, tau2par13, param1, param2, TP, FN, FP, TN, perm, gl, mgrid)
```

**Arguments**

qcondcop12	function for the inverse of conditional copula cdf at the (1,2) bivariate margin for Model 2
qcondcop13	function for the inverse of conditional copula cdf at the (1,3) bivariate margin for Model 2

qcondcop23	function for the inverse of conditional copula cdf at the (2,3 1) bivariate margin for Model 2
tau2par12	function for mapping Kendall's tau at the (1,2) bivariate margin to copula parameter for Model 2
tau2par13	function for mapping Kendall's tau at the (1,3) bivariate margin to copula parameter for Model 2
tau2par23	function for mapping Kendall's tau at the (2,3 1) bivariate margin to the conditional copula parameter for Model 2
param1	parameters for the Model 1. i.e., the GLMM
param2	parameters for the Model 2
TP	the number of true positives
FN	the number of false negatives
FP	the number of false positives
TN	the number of true negatives
perm	a scalar indicating the selected permutation of indices: 1: Pilot variable is the number of TP. The bivariate margins are 12, 13, 23 1; 2: Pilot variable is the number of TN. The bivariate margins are 23, 12, 13 2; 3: Pilot variable is the TP+FN. The bivariate margins are 13, 23, 12 3; 1:TP, 2:TN, 3:TP+FN
gl	a list containing the components of Gauss-Legendre nodes <code>gl\$nodes</code> and weights <code>gl\$weights</code>
mgrid	a list containing three-dimensional arrays. For more details see <a href="#">meshgrid</a>

### Value

A list containing the following components:

z	the test statistic
p-value	the $p$ -value

### References

Nikoloulopoulos, A.K. (2017) A vine copula mixed effect model for trivariate meta-analysis of diagnostic test accuracy studies accounting for disease prevalence. *Statistical Methods in Medical Research*, **26**, 2270–2286. doi: [10.1177/0962280215596769](https://doi.org/10.1177/0962280215596769).

Vuong Q.H. (1989) Likelihood ratio tests for model selection and non-nested hypotheses. *Econometrica*, **57**, 307–333.

### See Also

[CopulaREMADA](#)

**Examples**

```

nq=15
gl=gauss.quad.prob(nq,"uniform")
mgrid=meshgrid(gl$n,gl$n,gl$n,nargout=3)

data(betaDG)
attach(betaDG)
#nest.n2=VineCopulaREMADA.norm(TP, FN, FP, TN, 2, gl, mgrid,
#qcondbvn, qcondbvn, qcondbvn,
#tau2par.bvn, tau2par.bvn, tau2par.bvn)
nest.n2.est= #nest.n2$e
c(0.87186926, 0.13696066, 0.70614956, 0.69152133,
0.51780203, 0.70883558, -0.41354870, 0.07701287, -0.12111253)
#c090est.b2=VineCopulaREMADA.beta(TP, FN, FP, TN, 2, gl, mgrid,
#qcondcln90, qcondcln, qcondcln90, tau2par.cln90, tau2par.cln, tau2par.cln90)
c090est.b2.est= #c090est.b2$e
c(0.85528463, 0.14667571, 0.68321231, 0.04897466,
0.02776290, 0.08561436, -0.34639172, 0.04621924, -0.21627977)
c090vuong.b2=vine.vuong.beta(qcondcln90, qcondcln, qcondcln90,
tau2par.cln90, tau2par.cln, tau2par.cln90,
nest.n2.est, c090est.b2.est, TP, FN, FP, TN, 2, gl, mgrid)
c090vuong.b2
detach(betaDG)

```

---

VineCopulaREMADA	<i>Maximum likelihood estimation for (truncated) vine copula mixed models for diagnostic test accuracy studies accounting for disease prevalence</i>
------------------	--

---

**Description**

The estimated parameters can be obtained by using a quasi-Newton method applied to the logarithm of the joint likelihood. This numerical method requires only the objective function, i.e., the logarithm of the joint likelihood, while the gradients are computed numerically and the Hessian matrix of the second order derivatives is updated in each iteration. The standard errors (SE) of the ML estimates can be also obtained via the gradients and the Hessian computed numerically during the maximization process.

**Usage**

```

VineCopulaREMADA.norm(TP, FN, FP, TN, perm, gl, mgrid,
                      qcondcop12, qcondcop13, qcondcop23,
                      tau2par12, tau2par13, tau2par23)
VineCopulaREMADA.beta(TP, FN, FP, TN, perm, gl, mgrid,
                      qcondcop12, qcondcop13, qcondcop23,
                      tau2par12, tau2par13, tau2par23)
tVineCopulaREMADA.norm(TP, FN, FP, TN, perm, gl, mgrid,
                      qcondcop12, qcondcop13,

```

```

                                tau2par12, tau2par13)
tVineCopulaREMADA.beta(TP, FN, FP, TN, perm, gl, mgrid,
                        qcondcop12, qcondcop13,
                        tau2par12, tau2par13)

```

### Arguments

TP	the number of true positives
FN	the number of false negatives
FP	the number of false positives
TN	the number of true negatives
perm	a scalar indicating the selected permutation of indices: 1: Pilot variable is the number of TP. The bivariate margins are 12, 13, 23 1; 2: Pilot variable is the number of TN. The bivariate margins are 23, 12, 13 2; 3: Pilot variable is the TP+FN. The bivariate margins are 13, 23, 12 3; 1:TP, 2:TN, 3:TP+FN
gl	a list containing the components of Gauss-Legendre nodes <code>gl\$nodes</code> and weights <code>gl\$weights</code>
mgrid	a list containing three-dimensional arrays. For more details see <a href="#">meshgrid</a>
qcondcop12	function for the inverse of conditional copula cdf at the (1,2) bivariate margin
qcondcop13	function for the inverse of conditional copula cdf at the (1,3) bivariate margin
qcondcop23	function for the inverse of conditional copula cdf at the (2,3 1) bivariate margin
tau2par12	function for mapping Kendall's tau at the (1,2) bivariate margin to copula parameter
tau2par13	function for mapping Kendall's tau at the (1,3) bivariate margin to copula parameter
tau2par23	function for mapping Kendall's tau at the (2,3 1) bivariate margin to the conditional copula parameter

### Value

A list containing the following components:

minimum	the value of the estimated minimum of the negative log-likelihood
estimate	the MLE
gradient	the gradient at the estimated minimum of of the negative log-likelihood
hessian	the hessian at the estimated minimum of the negative log-likelihood
code	an integer indicating why the optimization process terminated
iterations	the number of iterations performed

For more details see [nlm](#)

## References

Nikoloulopoulos, A.K. (2017) A vine copula mixed effect model for trivariate meta-analysis of diagnostic test accuracy studies accounting for disease prevalence. *Statistical Methods in Medical Research*, **26**, 2270–2286. doi: [10.1177/0962280215596769](https://doi.org/10.1177/0962280215596769).

## See Also

[rVineCopulaREMADA](#)

## Examples

```
nq=15
gl=gauss.quad.prob(nq,"uniform")
mgrid=meshgrid(gl$n,gl$n,gl$n,nargout=3)

data(OGT)
attach(OGT)
#out=tVineCopulaREMADA.norm(TP, FN, FP, TN, 1, gl, mgrid,
#qcondbvn, qcondbvn, tau2par.bvn, tau2par.bvn)
detach(OGT)
```

---

vuong

*Vuong's test for the comparison of non-nested copula mixed models for diagnostic test accuracy studies*

---

## Description

Vuong (1989)'s test for the comparison of non-nested copula mixed models for diagnostic test accuracy studies. It shows if a copula mixed model provides better fit than the standard GLMM. We compute the Vuong's test with Model 1 being the copula mixed model with BVN copula and normal margins, i.e., the standard GLMM.

## Usage

```
vuong.norm(qcond, tau2par, param1, param2, TP, FN, FP, TN, gl, mgrid)
vuong.beta(qcond, tau2par, param1, param2, TP, FN, FP, TN, gl, mgrid)
countermonotonicity.vuong(param1, param2, TP, FN, FP, TN, gl, mgrid)
```

## Arguments

qcond	function for conditional copula cdf for Model 2
tau2par	function for mapping Kendall's tau to copula parameter for Model 2
param1	parameters for the Model 1. i.e., the GLMM
param2	parameters for the Model 2
TP	the number of true positives
FN	the number of false negatives

FP	the number of false positives
TN	the number of true negatives
gl	a list containing the components of Gauss-Legendre nodes <code>gl\$nodes</code> and weights <code>gl\$weights</code>
mgrid	a list containing two matrices with the rows of the output matrix <code>X</code> are copies of the vector <code>gl\$nodes</code> ; columns of the output matrix <code>Y</code> are copies of the vector <code>gl\$nodes</code> . For more details see <a href="#">meshgrid</a>

### Value

A list containing the following components:

z	the test statistic
p-value	the $p$ -value

### References

Nikoloulopoulos, A.K. (2015) A mixed effect model for bivariate meta-analysis of diagnostic test accuracy studies using a copula representation of the random effects distribution. *Statistics in Medicine*, **34**, 3842–3865. doi: [10.1002/sim.6595](https://doi.org/10.1002/sim.6595).

Vuong Q.H. (1989) Likelihood ratio tests for model selection and non-nested hypotheses. *Econometrica*, **57**:307–333.

### See Also

[CopulaREMADA](#)

### Examples

```
nq=15
gl=gauss.quad.prob(nq,"uniform")
mgrid<- meshgrid(gl$n,gl$n)

data(MRI)
attach(MRI)
c270est.b=CopulaREMADA.beta(TP, FN, FP, TN, gl, mgrid, qcondln270, tau2par.cln270)
nest.n=CopulaREMADA.norm(TP, FN, FP, TN, gl, mgrid, qcondbvn, tau2par.bvn)
c90est.n=CopulaREMADA.norm(TP, FN, FP, TN, gl, mgrid, qcondln90, tau2par.cln90)
vuong.beta(qcondln270, tau2par.cln270, nest.n$e, c270est.b$e, TP, FN, FP, TN, gl, mgrid)
vuong.norm(qcondln90, tau2par.cln90, nest.n$e, c90est.n$e, TP, FN, FP, TN, gl, mgrid)
detach(MRI)

data(CT)
attach(CT)
est.n=countermonotonicCopulaREMADA.norm(TP, FN, FP, TN, gl, mgrid)
est.b=countermonotonicCopulaREMADA.beta(TP, FN, FP, TN, gl, mgrid)
countermonotonicity.vuong(est.n$e, est.b$e, TP, FN, FP, TN, gl, mgrid)
detach(CT)
```

# Index

## \*Topic **copula**

- CopulaREMADA, 4
- cvinesim, 7
- dcop, 8
- dvinesim, 9
- hybridCopulaREMADA, 10
- pcondcop, 15
- qcondcop, 16
- quadVineCopulaREMADA, 17
- rcop, 19
- tau2par, 24
- VineCopulaREMADA, 28

## \*Topic **datasets**

- arthritis, 3
- betaDG, 4
- CT, 6
- LAG, 12
- mgrid, 13
- MRI, 13
- OGT, 14
- telomerase, 25

## \*Topic **dependence**

- tau2par, 24

## \*Topic **distribution**

- CopulaREMADA, 4
- dcop, 8
- hybridCopulaREMADA, 10
- pcondcop, 15
- qcondcop, 16
- quadVineCopulaREMADA, 17
- VineCopulaREMADA, 28

## \*Topic **graphics**

- SROC, 23

## \*Topic **maximum likelihood**

- CopulaREMADA, 4
- hybridCopulaREMADA, 10
- quadVineCopulaREMADA, 17
- vine.vuong, 26
- VineCopulaREMADA, 28

- vuong, 30

## \*Topic **package**

- CopulaREMADA-package, 2

## \*Topic **simulation**

- cvinesim, 7
- dvinesim, 9
- rcop, 19
- rCopulaREMADA, 20
- rVineCopulaREMADA, 21

- arthritis, 3

- betaDG, 4

- CopulaREMADA, 4, 11, 21, 24, 27, 31

- CopulaREMADA-package, 2

- countermonotonicCopulaREMADA  
(CopulaREMADA), 4

- countermonotonicity.vuong (vuong), 30

- CT, 6

- cvinesim, 7, 21, 22

- dbvn (dcop), 8

- dcln (dcop), 8

- dcln180 (dcop), 8

- dcln270 (dcop), 8

- dcln90 (dcop), 8

- dcop, 7, 8, 8, 9, 10, 15, 16, 19, 25

- dfrk (dcop), 8

- dvinesim, 9

- hybridCopulaREMADA, 10

- LAG, 12

- meshgrid, 5, 10, 11, 27, 29, 31

- mgrid, 13

- mgrid15 (mgrid), 13

- mgrid30 (mgrid), 13

- mgrid50 (mgrid), 13

- MRI, 13



nlm, [5](#), [11](#), [18](#), [29](#)

OGT, [14](#)

pcondbvn (pcondcop), [15](#)  
pcondc1n (pcondcop), [15](#)  
pcondc1n180 (pcondcop), [15](#)  
pcondc1n270 (pcondcop), [15](#)  
pcondc1n90 (pcondcop), [15](#)  
pcondcop, [15](#)  
pcondfrk (pcondcop), [15](#)

qcondbvn (qcondcop), [16](#)  
qcondc1n (qcondcop), [16](#)  
qcondc1n180 (qcondcop), [16](#)  
qcondc1n270 (qcondcop), [16](#)  
qcondc1n90 (qcondcop), [16](#)  
qcondcop, [8](#), [10](#), [16](#), [19](#)  
qcondfrk (qcondcop), [16](#)  
quadVineCopulaREMADA, [17](#)

rbvn (rcop), [19](#)  
rc1n (rcop), [19](#)  
rc1n180 (rcop), [19](#)  
rc1n270 (rcop), [19](#)  
rc1n90 (rcop), [19](#)  
rcop, [8](#), [10](#), [15](#), [16](#), [19](#), [21](#), [22](#)  
rCopulaREMADA, [5](#), [20](#), [22](#), [24](#)  
rfrk (rcop), [19](#)  
rVineCopulaREMADA, [21](#), [30](#)

SROC, [23](#)

SROC.beta, [23](#)

SROC.norm, [23](#)

tau2par, [24](#)

telomerase, [25](#)

tvine.vuong.beta (vine.vuong), [26](#)  
tvine.vuong.norm (vine.vuong), [26](#)  
tvine2.vuong.beta (vine.vuong), [26](#)  
tvine2.vuong.norm (vine.vuong), [26](#)  
tVineCopulaREMADA (VineCopulaREMADA), [28](#)

vine.vuong, [26](#)

VineCopulaREMADA, [11](#), [28](#)

vuong, [30](#)