

# Analysis of eight daily European bank asset values using vine copulas between 2013-2017 (Application to Lecture 2)

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# Load packages

```
library(VineCopula) # primary software for vine copulas
library(rafalib) # allows for nicer multiple plots
library(ggplot2) # allows for nice plots
library(reshape2) # allows to merge data
library(fGarch) # fits GARCH times series models
library(TSP) # needed to fit D-Vines
```

# Setup

# Read in data

```
load("stock_data1317.Rdata")
lret<-data.frame(stock_data1317$lret)
data.org<-data.frame(stock_data1317$data_orinigal)
colnames(lret)
```

```
## [1] "ACA.PA" "BBVA.MC" "BNP.PA" "CBK.DE" "DBK.DE" "GLE.PA" "ISP.MI"
## [8] "SAN.MC" "BAS.DE" "BAYN.DE" "FME.DE" "FRE.DE" "LIN.DE" "MRK.DE"
## [15] "SDF.DE"
```

## Create bank data with create daily dates

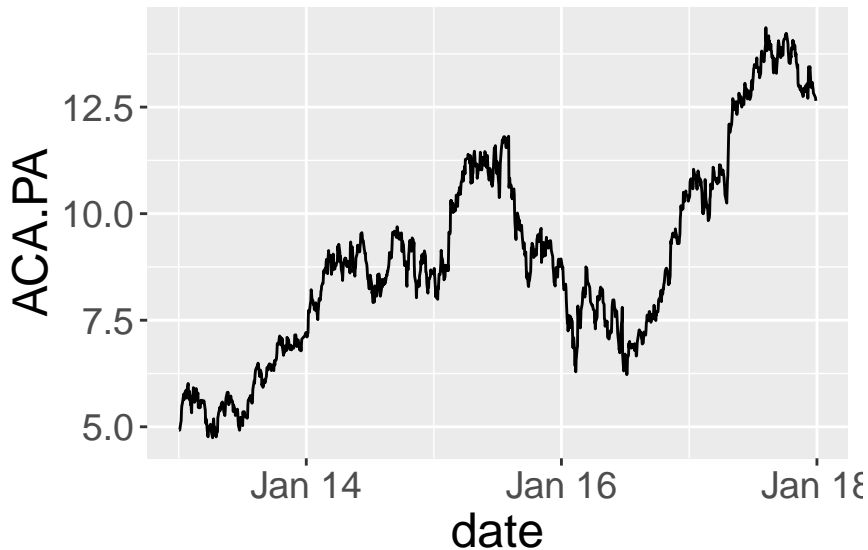
```
bank.names<-colnames(lret)[1:8]
dates<-as.Date(rownames(data.org))
data.bank<-data.frame(dates=dates, data.org[,bank.names])
dim(data.bank)
```

```
## [1] 1280    9
```

```
lret.bank<-data.frame(dates=as.Date(rownames(lret)),
                      lret[,bank.names])
dim(lret.bank)
```

```
## [1] 1279    9
```

## Show time trend of single stock "ACA.PA"

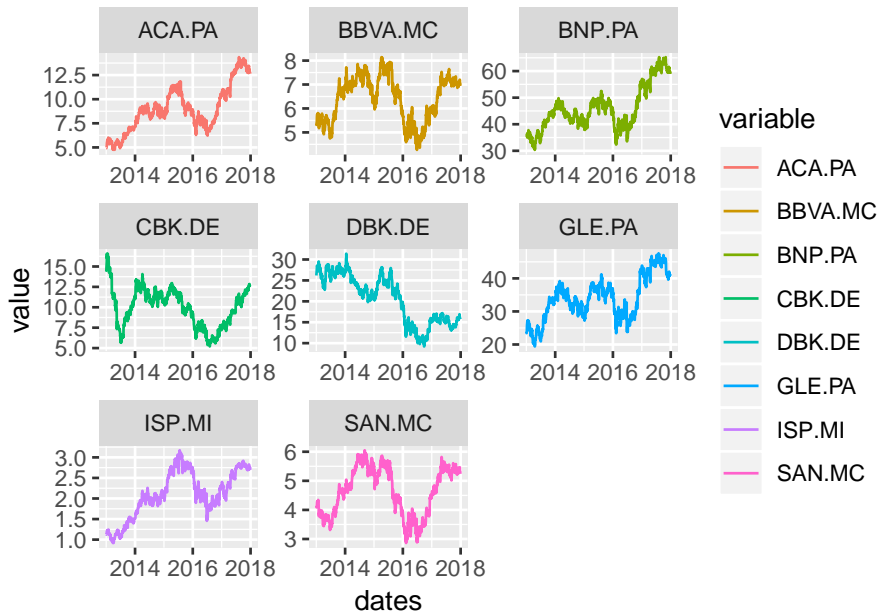


## R-code: Show time trend of single stock "ACA.PA"

```
ggplot(data.bank, aes(dates, ACA.PA)) +  
  geom_line() +  
  scale_x_date(labels = scales::date_format("%b %y")) +  
  xlab("date") +  
  ylab("ACA.PA") +  
  theme(plot.title = element_text(lineheight = 0.8,  
    face = "bold", size = 20)) +  
  theme(text = element_text(size = 18))
```



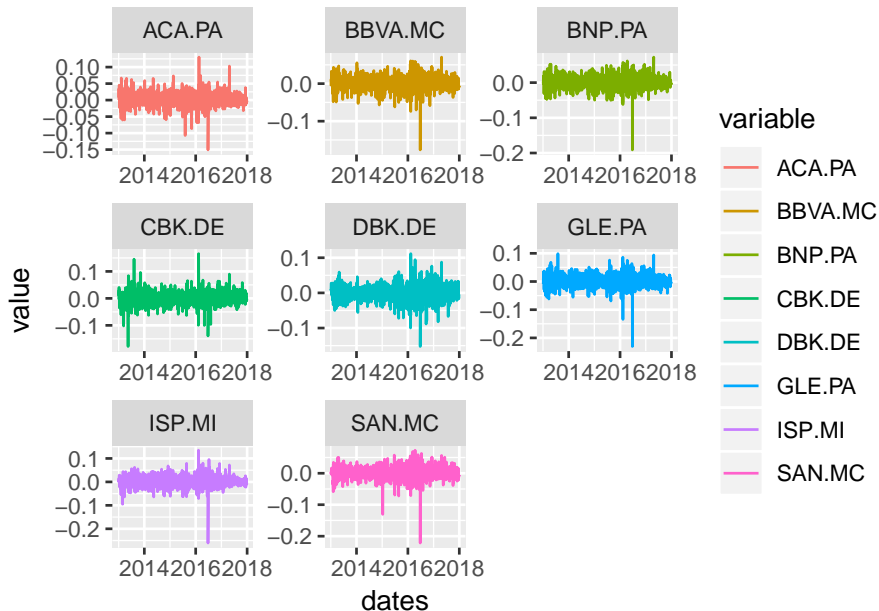
# Time plot of bank stocks



## R:code: Time plot of bank stocks

```
df <- melt(data.bank[, 1:9], id="dates")
ggplot(df) + geom_line(aes(x=dates, y=value, color=variable))
+ facet_wrap( ~ variable, scales="free")
```

# Time plot of log returns for banks



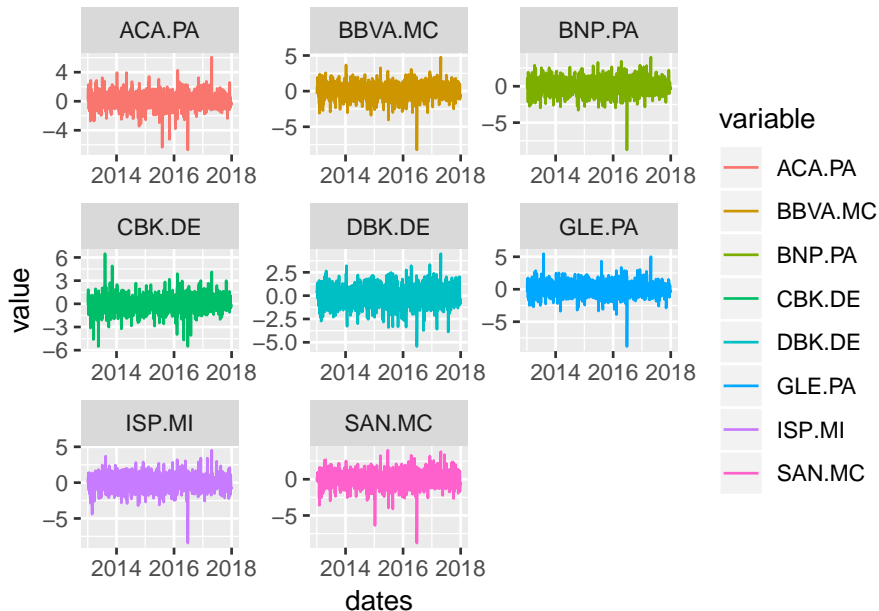
## R code: Time plot of log returns for banks

```
df <- melt(lret.bank[, 1:9], id="dates")
ggplot(df) + geom_line(aes(x=dates, y=value, color=variable))
+ facet_wrap( ~ variable, scales="free")
```

## Fit marginal GARCH models with Student t errors

```
fits = list() # Garch(1,1) fits
res = list() # standardized residuals
pvalues = rep(0,8) # p-values of KS test for t distribution
resdata= matrix(0,dim(lret.bank)[1],8)
udata = matrix(0,dim(lret.bank)[1],8)
for(i in 1:8){
fits[[i]] = garchFit(formula=~garch(1,1),
                    data=lret.bank[,i+1], trace=FALSE,
                    include.mean=TRUE, cond.dist="std")
res[[i]] = residuals(fits[[i]], standardize=TRUE)
pvalues[i] = ks.test(res[[i]], "pstd",
                    0, 1, coef(fits[[i]))[5])$p.value
#transform standardized residuals with Student t cdf
udata[,i] = pstd(res[[i]], nu=coef(fits[[i]))[5])
resdata[,i]= res[[i]]
}
colnames(udata)<-colnames(lret.bank[,2:9])
colnames(resdata)<-c("ACA", "BBVA", "BNP", "CBK", "DBK", "GLE", "ISP", "SAN")
rownames(udata)<-rownames(lret.bank)
rownames(resdata)<-rownames(lret.bank)
```

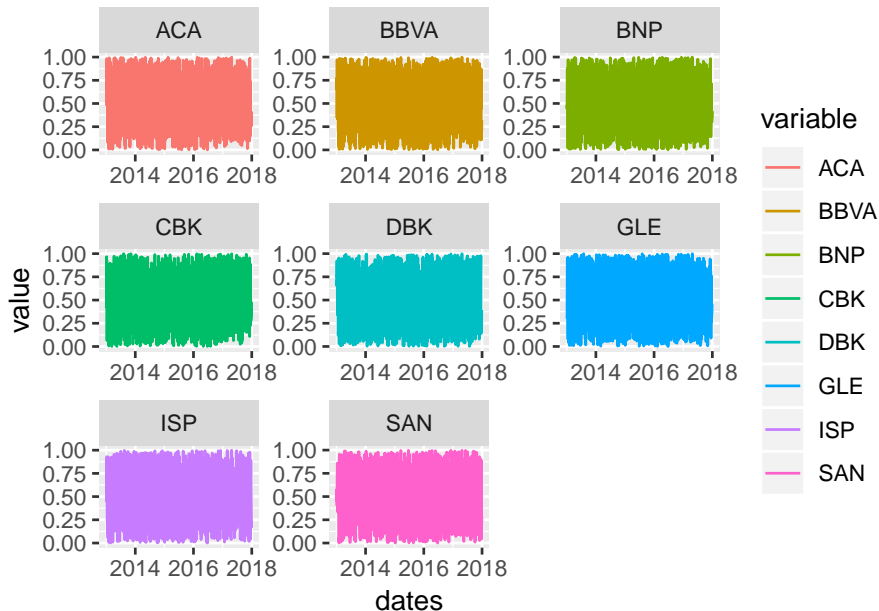
# time plot of standardized residuals for banks



## R code: time plot of standardized residuals for banks

```
res.dates<-data.frame(dates=as.Date(rownames(lret)),resdata)
df <- melt(res.dates[, 1:9], id="dates")
ggplot(df) + geom_line(aes(x=dates, y=value, color=variable))+
  facet_wrap( ~ variable, scales="free")
```

# time plot of copula data for banks





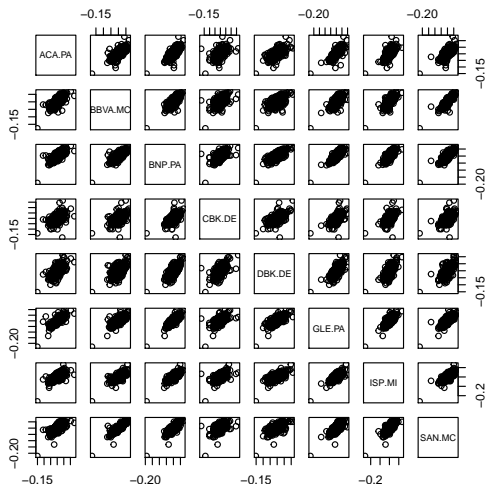
## R code: time plot of copula data for banks

```
udata.dates<-data.frame(dates=as.Date(rownames(lret)),udata)
df <- melt(udata.dates[, 1:9], id="dates")
ggplot(df) + geom_line(aes(x=dates, y=value, color=variable))
+ facet_wrap(~ variable, scales="free")
```

## Exploring pairwise dependence

# pairs plot ignoring serial dependence

```
pairs(lret.bank[,2:9])
```



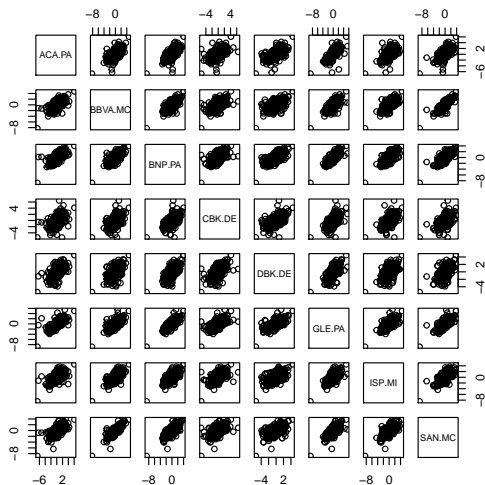
## Kendall's tau ignoring serial dependence

```
round(cor(lret.bank[,2:9],method="kendall"), digits=2)
```

##	ACA.PA	BBVA.MC	BNP.PA	CBK.DE	DBK.DE	GLE.PA	ISP.MI	SAN.MC
## ACA.PA	1.00	0.56	0.61	0.50	0.53	0.65	0.52	0.56
## BBVA.MC	0.56	1.00	0.59	0.46	0.52	0.60	0.52	0.72
## BNP.PA	0.61	0.59	1.00	0.47	0.56	0.68	0.55	0.60
## CBK.DE	0.50	0.46	0.47	1.00	0.53	0.53	0.45	0.47
## DBK.DE	0.53	0.52	0.56	0.53	1.00	0.58	0.47	0.53
## GLE.PA	0.65	0.60	0.68	0.53	0.58	1.00	0.56	0.59
## ISP.MI	0.52	0.52	0.55	0.45	0.47	0.56	1.00	0.53
## SAN.MC	0.56	0.72	0.60	0.47	0.53	0.59	0.53	1.00

# Dependence among standardized residuals (x-scale)

pairs(resdata)



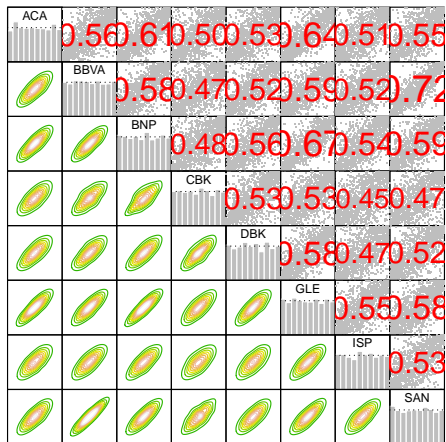
# Kendall's tau among standardized residuals

```
round(cor(resdata,method="kendall"), digits=2)
```

##	ACA.PA	BBVA.MC	BNP.PA	CBK.DE	DBK.DE	GLE.PA	ISP.MI	SAN.MC
## ACA.PA	1.00	0.56	0.61	0.50	0.53	0.64	0.51	0.55
## BBVA.MC	0.56	1.00	0.58	0.47	0.52	0.59	0.52	0.72
## BNP.PA	0.61	0.58	1.00	0.48	0.56	0.67	0.54	0.59
## CBK.DE	0.50	0.47	0.48	1.00	0.53	0.53	0.45	0.47
## DBK.DE	0.53	0.52	0.56	0.53	1.00	0.58	0.47	0.52
## GLE.PA	0.64	0.59	0.67	0.53	0.58	1.00	0.55	0.58
## ISP.MI	0.51	0.52	0.54	0.45	0.47	0.55	1.00	0.53
## SAN.MC	0.55	0.72	0.59	0.47	0.52	0.58	0.53	1.00

## Normalized contour plots (z-scale)

```
udata.cop<-as.copuladata(udata)  
pairs(udata.cop)
```



# Parametric vine copula analysis using 4 candidate copula families



## Fit R-vine using only 4 pair copula families with and without independence test

```
rv.fsub<-RVineStructureSelect(udata.cop, family=c(1,2,3,4),
                             method = "mle")
rv.fsub.ind<-RVineStructureSelect(udata.cop, family=c(1,2,3,4),
                                 method = "mle", indeptest = TRUE)
# suppress console output
dummy<-capture.output(sum.fsub<- summary(rv.fsub))
dummy<-capture.output(sum.fsub.ind<- summary(rv.fsub.ind))
```

## Results of first two trees of rv.fsub

1 <-> ACA, 2 <-> BBVA, 3 <-> BNP, 4 <-> CBK,

5 <-> DBK, 6 <-> GLE, 7 <-> ISP, 8 <-> SAN

```
print(sum.fsub[1:14, ], digits = 3)
```

##	tree	edge	family	cop	par	par2	tau	utd	ltd
## 1	1	8,3	2	t	0.797	5.51	0.587	0.4212	0.4212
## 2	1	8,2	2	t	0.902	3.85	0.715	0.6384	0.6384
## 3	1	3,6	2	t	0.864	4.37	0.665	0.5576	0.5576
## 4	1	6,1	2	t	0.839	4.40	0.634	0.5208	0.5208
## 5	1	6,7	2	t	0.752	5.05	0.541	0.3896	0.3896
## 6	1	6,5	2	t	0.785	4.08	0.575	0.4690	0.4690
## 7	1	6,4	2	t	0.733	5.04	0.524	0.3721	0.3721
## 8	2	3,2;8	2	t	0.275	6.07	0.177	0.0846	0.0846
## 9	2	8,6;3	2	t	0.326	8.61	0.212	0.0526	0.0526
## 10	2	3,1;6	2	t	0.302	7.07	0.195	0.0708	0.0708
## 11	2	3,7;6	2	t	0.290	9.53	0.187	0.0356	0.0356
## 12	2	3,5;6	2	t	0.270	10.44	0.174	0.0257	0.0257
## 13	2	5,4;6	2	t	0.387	9.81	0.253	0.0517	0.0517
## 14	3	2,6;8,3	4	G	1.125	0.00	0.111	0.1484	0.0000

## Results of trees 2 to 5 of rv.fsub

1 <-> ACA, 2 <-> BBVA, 3 <-> BNP, 4 <-> CBK,

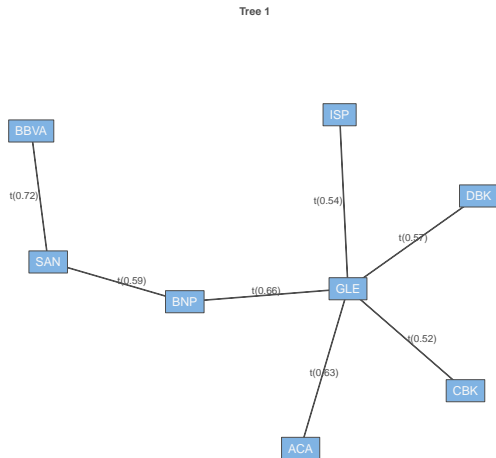
5 <-> DBK, 6 <-> GLE, 7 <-> ISP, 8 <-> SAN

```
print(sum.fsub[15:28, ], digits = 3)
```

##	tree	edge	family	cop	par	par2	tau	utd	ltd
## 15	3	8,1;3,6	1	N	0.1836	0.0	0.1175	0.00e+00	0.00e+00
## 16	3	8,7;3,6	2	t	0.2539	17.5	0.1634	3.72e-03	3.72e-03
## 17	3	8,5;3,6	2	t	0.1810	10.5	0.1159	1.59e-02	1.59e-02
## 18	3	3,4;5,6	1	N	0.0354	0.0	0.0226	0.00e+00	0.00e+00
## 19	4	2,1;8,3,6	2	t	0.1141	28.8	0.0728	3.45e-05	3.45e-05
## 20	4	1,7;8,3,6	2	t	0.1132	30.0	0.0722	2.34e-05	2.34e-05
## 21	4	1,5;8,3,6	1	N	0.0921	0.0	0.0587	0.00e+00	0.00e+00
## 22	4	8,4;3,5,6	1	N	0.1258	0.0	0.0803	0.00e+00	0.00e+00
## 23	5	2,7;1,8,3,6	1	N	0.0830	0.0	0.0529	0.00e+00	0.00e+00
## 24	5	2,5;1,8,3,6	2	t	0.0645	18.3	0.0411	5.63e-04	5.63e-04
## 25	5	1,4;8,3,5,6	2	t	0.1309	18.9	0.0836	8.63e-04	8.63e-04
## 26	6	7,5;2,1,8,3,6	1	N	0.0493	0.0	0.0314	0.00e+00	0.00e+00
## 27	6	2,4;1,8,3,5,6	1	N	0.0249	0.0	0.0158	0.00e+00	0.00e+00
## 28	7	7,4;2,1,8,3,5,6	1	N	0.1017	0.0	0.0649	0.00e+00	0.00e+00

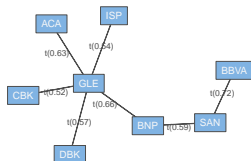
# Tree 1 of rv.fsub

```
plot(rv.fsub, tree=1, edge.labels="family-tau", type=1)
```

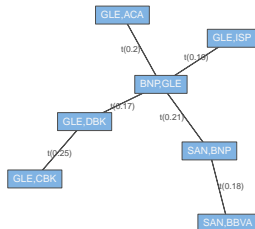


# Trees 1 to 4 of rv.sub

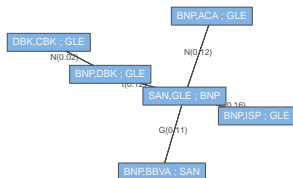
Tree 1



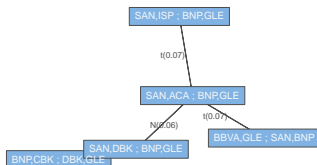
Tree 2



Tree 3

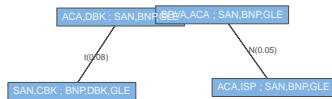


Tree 4



# Trees 5 to 7 of rv.sub

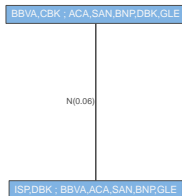
Tree 5



Tree 6



Tree 7



# Parametric vine copula analysis using all candidate copula families

## Fit R-vine using all pair copula families without and with independence test

```
rv<-RVineStructureSelect(udata.cop, family=NA,  
                        method = "mle")  
rv.ind<-RVineStructureSelect(udata.cop, family=NA,  
                            method = "mle", indeptest=TRUE)  
# suppress console output  
dummy<-capture.output(sum.rv<- summary(rv))  
dummy<-capture.output(sum.rv.ind<- summary(rv.ind))
```



## Results of first two trees of rv

1 <-> ACA, 2 <-> BBVA, 3 <-> BNP, 4 <-> CBK,

5 <-> DBK, 6 <-> GLE, 7 <-> ISP, 8 <-> SAN

```
print(sum.rv[1:14, ], digits = 3)
```

##	tree	edge	family	cop	par	par2	tau	utd	ltd
## 1	1	8,3	2	t	0.797	5.505	0.587	0.4212	0.4212
## 2	1	8,2	2	t	0.902	3.853	0.715	0.6384	0.6384
## 3	1	3,6	2	t	0.864	4.369	0.665	0.5576	0.5576
## 4	1	6,1	2	t	0.839	4.400	0.634	0.5208	0.5208
## 5	1	6,7	2	t	0.752	5.054	0.541	0.3896	0.3896
## 6	1	6,5	2	t	0.785	4.082	0.575	0.4690	0.4690
## 7	1	6,4	2	t	0.733	5.038	0.524	0.3721	0.3721
## 8	2	3,2;8	2	t	0.275	6.068	0.177	0.0846	0.0846
## 9	2	8,6;3	2	t	0.326	8.613	0.212	0.0526	0.0526
## 10	2	3,1;6	2	t	0.302	7.072	0.195	0.0708	0.0708
## 11	2	3,7;6	7	BB1	0.203	1.114	0.185	0.1367	0.0464
## 12	2	3,5;6	2	t	0.270	10.436	0.174	0.0257	0.0257
## 13	2	5,4;6	5	F	2.737	0.000	0.284	0.0000	0.0000
## 14	3	2,6;8,3	10	BB8	1.693	0.805	0.145	0.0000	0.0000

## Results of trees 2 to 5 of rv

1 <-> ACA, 2 <-> BBVA, 3 <-> BNP, 4 <-> CBK,

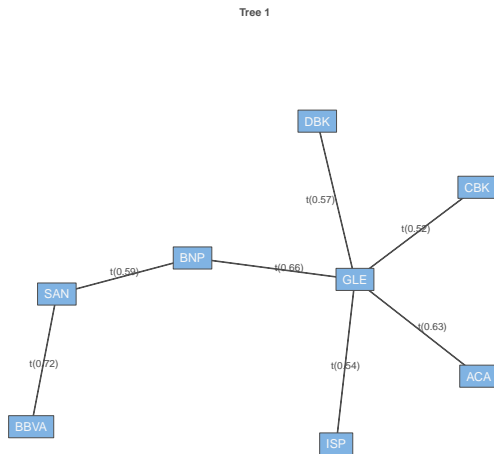
5 <-> DBK, 6 <-> GLE, 7 <-> ISP, 8 <-> SAN

```
print(sum.rv[15:28, ], digits = 2)
```

##	tree	edge	family	cop	par	par2	tau	utd	ltd
## 15	3	8,1;3,6	20	SBB8	1.950	0.6826	1.4e-01	0.0e+00	0.0e+00
## 16	3	8,7;3,6	2	t	0.256	17.8509	1.6e-01	3.5e-03	3.5e-03
## 17	3	8,5;3,6	5	F	1.290	0.0000	1.4e-01	0.0e+00	0.0e+00
## 18	3	3,4;5,6	10	BB8	1.111	0.9201	3.8e-02	0.0e+00	0.0e+00
## 19	4	2,1;8,3,6	7	BB1	0.059	1.0364	6.3e-02	4.8e-02	1.3e-05
## 20	4	1,7;8,3,6	5	F	0.783	0.0000	8.7e-02	0.0e+00	0.0e+00
## 21	4	1,5;8,3,6	5	F	0.671	0.0000	7.4e-02	0.0e+00	0.0e+00
## 22	4	8,4;3,5,6	5	F	0.949	0.0000	1.0e-01	0.0e+00	0.0e+00
## 23	5	2,7;1,8,3,6	1	N	0.085	0.0000	5.4e-02	0.0e+00	0.0e+00
## 24	5	2,5;1,8,3,6	2	t	0.061	16.0869	3.9e-02	1.2e-03	1.2e-03
## 25	5	1,4;8,3,5,6	5	F	1.030	0.0000	1.1e-01	0.0e+00	0.0e+00
## 26	6	7,5;2,1,8,3,6	5	F	0.410	0.0000	4.6e-02	0.0e+00	0.0e+00
## 27	6	2,4;1,8,3,5,6	104	Tawn	9.254	0.0001	6.2e-12	1.0e-04	0.0e+00
## 28	7	7,4;2,1,8,3,5,6	2	t	0.098	28.6734	6.2e-02	2.8e-05	2.8e-05

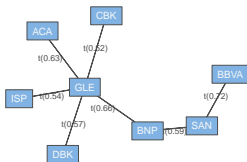
# Tree 1 of rv

```
plot(rv, tree=1, edge.labels="family-tau", type=1)
```

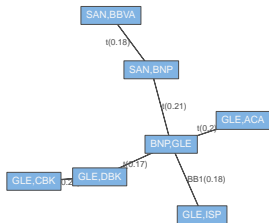


# Trees 1 to 4 of rv

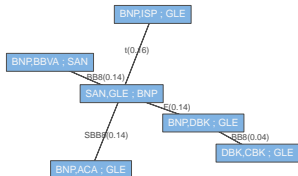
Tree 1



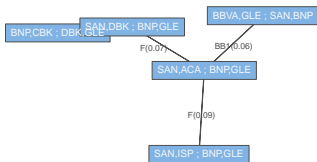
Tree 2



Tree 3

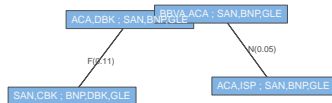


Tree 4

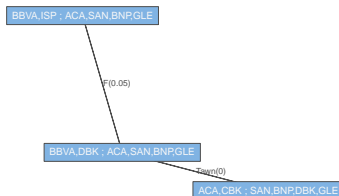


# Trees 5 to 7 of rv

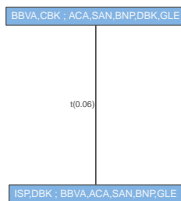
Tree 5



Tree 6

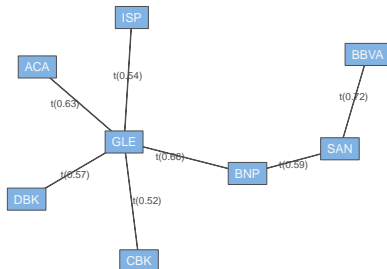


Tree 7

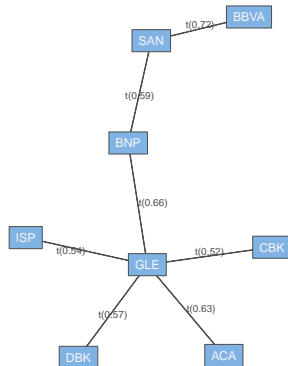


# Comparing first tree of R-vine with 4 and unrestricted candidate pair copulas

Tree 1



Tree 1



## Fitting C-vines

## Fitting C-vines using RVineStructureSelect

```
cv.fsub<-RVineStructureSelect(udata.cop, family=1:4,  
                              method = "mle",type="CVine")  
cv.fsub.ind<-RVineStructureSelect(udata.cop, family=1:4,  
                                  method = "mle",type="CVine",indeptest = TRUE)  
cv<-RVineStructureSelect(udata.cop, family=NA,  
                          method = "mle",type="CVine")  
cv.ind<-RVineStructureSelect(udata.cop, family=NA,  
                              method = "mle",type="CVine",indeptest = TRUE)
```



## Fitting D-vines

## Fitting D-vines using the TSP package

```
d = dim(udata.cop)[2]
M = 1 - abs(TauMatrix(udata.cop))
hamilton = insert_dummy(TSP(M), label="cut")
sol = solve_TSP(hamilton, method="repetitive_nn")
order = cut_tour(sol, "cut")
DVM= D2RVine(order, family=rep(0, d*(d-1)/2), par=rep(0, d*(d-1)/2))
dv.fsub=RVineCopSelect(data=udata.cop,
                       familyset=1:4, Matrix=DVM$Matrix)
dv.fsub.ind=RVineCopSelect(data=udata.cop, familyset=1:4,
                           indeptest=TRUE, Matrix=DVM$Matrix)
dv=RVineCopSelect(data=udata.cop, familyset=NA, Matrix=DVM$Matrix)
dv.ind=RVineCopSelect(data=udata.cop, familyset=NA,
                      indeptest=TRUE, Matrix=DVM$Matrix)
```

## Comparison of R-vines, C-vines and D-vines

# Output function

```
vine.out<-function(fit=out12.3fixed,data=udata.f3,digits=2){  
df<-sum(abs(fit$par)>0)+sum(fit$par2>0)  
out<-round(c(RVineLogLik(data, fit)$loglik,df,  
            RVineAIC(data,fit)$AIC,  
            RVineBIC(data,fit)$BIC),digits)  
names(out)<-c("loglik", "par", "AIC", "BIC")  
out  
}
```

## Apply output function to all fitted vine models

```
out.table<-rbind(  
  vine.out(fit=rv.fsub,data=udata.cop),  
  vine.out(fit=rv.fsub.ind,data=udata.cop),  
  vine.out(fit=rv,data=udata.cop),  
  vine.out(fit=rv.ind,data=udata.cop),  
  vine.out(fit=cv.fsub,data=udata.cop),  
  vine.out(fit=cv.fsub.ind,data=udata.cop),  
  vine.out(fit=cv,data=udata.cop),  
  vine.out(fit=cv.ind,data=udata.cop),  
  vine.out(fit=dv.fsub,data=udata.cop),  
  vine.out(fit=dv.fsub.ind,data=udata.cop),  
  vine.out(fit=dv,data=udata.cop),  
  vine.out(fit=dv.ind,data=udata.cop)  
)  
row.names(out.table)<-c("rv.fsub", "rv.fsub.ind", "rv", "rv.ind",  
  "cv.fsub", "cv.fsub.ind", "cv", "cv.ind",  
  "dv.fsub", "dv.fsub.ind", "dv", "dv.ind")
```

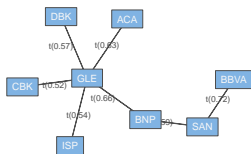
# Comparison of all fitted vine models

```
round(out.table,digits=2)
```

##		loglik	par	AIC	BIC
##	rv.fsub	5688.25	47	-11282.51	-11040.28
##	rv.fsub.ind	5686.90	45	-11283.80	-11051.88
##	rv	5729.65	48	-11363.29	-11115.91
##	rv.ind	5724.21	44	-11360.41	-11133.64
##	cv.fsub	5684.91	47	-11275.82	-11033.59
##	cv.fsub.ind	5677.10	43	-11268.19	-11046.58
##	cv	5710.42	47	-11326.83	-11084.60
##	cv.ind	5703.89	43	-11321.79	-11100.17
##	dv.fsub	5675.02	49	-11252.03	-10999.49
##	dv.fsub.ind	5675.02	49	-11252.03	-10999.49
##	dv	5712.61	47	-11331.21	-11088.98
##	dv.ind	5712.61	47	-11331.21	-11088.98

# Comparing first tree of R-vine, C-vine and D-vine pair

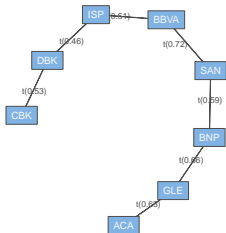
Tree 1



Tree 1



Tree 1



## Storing vine specifications



# R-vine matrix of rv

1 <-> ACA, 2 <-> BBVA, 3 <-> BNP, 4 <-> CBK,

5 <-> DBK, 6 <-> GLE, 7 <-> ISP, 8 <-> SAN

Column 1 identifies the following edges:

47;2183556, 42;18356, 41;8365; 48;356, 43;56 45;6 46

```
rv$Matrix
```

```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
## [1,]    4    0    0    0    0    0    0    0
## [2,]    7    5    0    0    0    0    0    0
## [3,]    2    7    7    0    0    0    0    0
## [4,]    1    2    2    1    0    0    0    0
## [5,]    8    1    1    2    6    0    0    0
## [6,]    3    8    8    8    2    2    0    0
## [7,]    5    3    3    3    8    3    3    0
## [8,]    6    6    6    6    3    8    8    8
```

# C-vine matrix of rv

1 <-> ACA, 2 <-> BBVA, 3 <-> BNP, 4 <-> CBK,

5 <-> DBK, 6 <-> GLE, 7 <-> ISP, 8 <-> SAN

```
cv$Matrix
```

```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
## [1,]  2    0    0    0    0    0    0    0
## [2,]  4    3    0    0    0    0    0    0
## [3,]  3    4    4    0    0    0    0    0
## [4,]  7    7    7    1    0    0    0    0
## [5,]  1    1    1    7    5    0    0    0
## [6,]  5    5    5    5    7    7    0    0
## [7,]  8    8    8    8    8    8    6    0
## [8,]  6    6    6    6    6    6    8    8
```

# D-vine matrix of rv

1 <-> ACA, 2 <-> BBVA, 3 <-> BNP, 4 <-> CBK,

5 <-> DBK, 6 <-> GLE, 7 <-> ISP, 8 <-> SAN

```
dv$Matrix
```

```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
## [1,]    4    0    0    0    0    0    0    0
## [2,]    1    5    0    0    0    0    0    0
## [3,]    6    1    7    0    0    0    0    0
## [4,]    3    6    1    2    0    0    0    0
## [5,]    8    3    6    1    8    0    0    0
## [6,]    2    8    3    6    1    3    0    0
## [7,]    7    2    8    3    6    1    6    0
## [8,]    5    7    2    8    3    6    1    1
```