

R-code for Chapter 8: Selection of regular vine copula models

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Required R-packages

- VineCopula
- rafalib
- TSP

Section 8.3 The Dissmann algorithm for sequential top-down.

Example 8.4 WINE7: Illustration of the Dissmann Algorithm.

Read in data and set column names

```
reddata<-read.csv(file="winequality-red.csv",sep=";")
n<-length(reddata[,1])
colnames(reddata)<-c("acf","acv","acc","sugar","clor","sf","st","den","ph","sp","alc","quality")
reddata7<-reddata[,c(1,2,3,5,7,8,9)]
```

Transform original data to copula data (udata) using ranks and then to marginal normalized data (zdata)

```

udata<-reddata
zdata<-reddata
for(i in 1:12){
udata[,i]<-rank(reddata[,i])/(n+1)
zdata[,i]<-qnorm(udata[,i])
}
udata7<-udata[,c(1,2,3,5,7,8,9)]

```

Table 8.2: WINE7: Ordered weights (pairwise absolute empirical Kendall's τ values of copula data) for all pairs of variables.

```

options(digits=3)
temp<-abs(cor(reddata7,method = "kendall"))
cor.names<-outer(names(reddata7),names(reddata7),paste)
cor.vec<-as.vector(temp)
names(cor.vec)<-as.vector(cor.names)
names.cor.vec<-as.vector(cor.names)
sort.cor.vec<-sort(cor.vec,decreasing=TRUE)
names.sort.cor.vec<-names.cor.vec[sort.list(cor.vec,decreasing=TRUE)]
names(sort.cor.vec)<-names(names.sort.cor.vec)
dup.list<-duplicated(sort.cor.vec)
unique.sort.cor<-sort.cor.vec[dup.list==FALSE]
names(unique.sort.cor)<-names.sort.cor.vec[dup.list==FALSE]
unique.sort.cor[-1]

```

```

##  ph acf  acc acf  den acf  acc acv  ph acc den clor  den acc  ph den
##  0.5278  0.4843  0.4575  0.4284  0.3898  0.2879  0.2457  0.2172
##  acv acf clor acf  ph clor  ph acv clor acv  st clor  den st clor acc
##  0.1852  0.1760  0.1627  0.1587  0.1096  0.0916  0.0877  0.0767
##  st acv  st acf  den acv  st acc  ph st
##  0.0637  0.0569  0.0159  0.0116  0.0068

```

Example 8.5 WINE7 Fitted R-vine, C- and D-vine copulas

Fitting R-vines and Gauss R-vines using sequential estimation with restricted pair copula family set

```

fit.rv=RVineStructureSelect(udata7, familyset=c(1,5,6,4),
selectioncrit="AIC",
indeptest=FALSE, level=0.05)
fit.rv.ind=RVineStructureSelect(udata7, familyset=c(1,5,6,4),
selectioncrit="AIC",
indeptest=TRUE, level=0.05)
fit.Gauss=RVineStructureSelect(udata7, familyset=1, selectioncrit="AIC",
indeptest=FALSE, level=0.05)
fit.Gauss.ind=RVineStructureSelect(udata7, familyset=1, selectioncrit="AIC",
indeptest=TRUE, level=0.05)
summary(fit.rv)

```

```

## tree          edge | family  cop   par  par2 |   tau  utd  ltd
## -----
##    1           3,2 |     5    F  -4.44 0.00 | -0.42  -   -
##           1,3 |     4    G   1.80 0.00 |  0.44 0.53  -
##           4,5 |    14   SG   1.09 0.00 |  0.09  -  0.12
##           6,4 |    16   SJ   1.64 0.00 |  0.26  -  0.48
##           1,6 |     4    G   1.82 0.00 |  0.45 0.54  -
##           7,1 |    34  G270 -1.99 0.00 | -0.50  -   -
##    2           1,2;3 |     5    F   1.15 0.00 |  0.13  -   -
##           7,3;1 |    24  G90  -1.10 0.00 | -0.09  -   -
##           6,5;4 |    16   SJ   1.08 0.00 |  0.04  -  0.10
##           1,4;6 |    36  J270 -1.06 0.00 | -0.03  -   -
##           7,6;1 |     4    G   1.24 0.00 |  0.19 0.25  -
##    3           7,2;1,3 |     6    J   1.03 0.00 |  0.01 0.03  -
##           6,3;7,1 |     5    F  -0.59 0.00 | -0.07  -   -
##           1,5;6,4 |    34  G270 -1.16 0.00 | -0.14  -   -
##           7,4;1,6 |    26  J90  -1.26 0.00 | -0.13  -   -
##    4           6,2;7,1,3 |    16   SJ   1.33 0.00 |  0.16  -  0.32
##           4,3;6,7,1 |     5    F  -0.08 0.00 | -0.01  -   -
##           7,5;1,6,4 |    24  G90  -1.14 0.00 | -0.12  -   -
##    5           4,2;6,7,1,3 |    14   SG   1.10 0.00 |  0.09  -  0.12
##           5,3;4,6,7,1 |     5    F   0.81 0.00 |  0.09  -   -
##    6           5,2;4,6,7,1,3 |     6    J   1.06 0.00 |  0.03 0.08  -
## ---
## type: R-vine    logLik: 2525    AIC: -5009    BIC: -4896
## ---
## 1 <-> acf,    2 <-> acv,    3 <-> acc,    4 <-> clor,    5 <-> st,    6 <-> den,
## 7 <-> ph

```

```
summary(fit.rv.ind)
```

```

## tree          edge | family  cop   par  par2 |   tau  utd  ltd
## -----
##    1           3,2 |     5    F  -4.44 0.00 | -0.42  -   -
##           1,3 |     4    G   1.80 0.00 |  0.44 0.53  -
##           4,5 |    14   SG   1.09 0.00 |  0.09  -  0.12
##           6,4 |    16   SJ   1.64 0.00 |  0.26  -  0.48
##           1,6 |     4    G   1.82 0.00 |  0.45 0.54  -
##           7,1 |    34  G270 -1.99 0.00 | -0.50  -   -
##    2           1,2;3 |     5    F   1.15 0.00 |  0.13  -   -
##           7,3;1 |    24  G90  -1.10 0.00 | -0.09  -   -
##           6,5;4 |    16   SJ   1.08 0.00 |  0.04  -  0.10
##           1,4;6 |     0    I    -    - |  0.00  -   -
##           7,6;1 |     4    G   1.24 0.00 |  0.19 0.25  -
##    3           7,2;1,3 |     0    I    -    - |  0.00  -   -
##           6,3;7,1 |     5    F  -0.59 0.00 | -0.07  -   -
##           1,5;6,4 |    34  G270 -1.16 0.00 | -0.14  -   -
##           7,4;1,6 |    26  J90  -1.26 0.00 | -0.13  -   -
##    4           6,2;7,1,3 |    16   SJ   1.33 0.00 |  0.16  -  0.32
##           4,3;6,7,1 |     0    I    -    - |  0.00  -   -
##           7,5;1,6,4 |    24  G90  -1.14 0.00 | -0.12  -   -
##    5           4,2;6,7,1,3 |    14   SG   1.10 0.00 |  0.09  -  0.12
##           5,3;4,6,7,1 |     5    F   0.82 0.00 |  0.09  -   -

```

```
## 6 5,2;4,6,7,1,3 | 6 J 1.06 0.00 | 0.03 0.08 -
## ---
## type: R-vine logLik: 2519 AIC: -5003 BIC: -4906
## ---
## 1 <-> acf, 2 <-> acv, 3 <-> acc, 4 <-> clor, 5 <-> st, 6 <-> den,
## 7 <-> ph
```

```
summary(fit.Gauss)
```

```
## tree edge | family cop par par2 | tau utd ltd
## -----
## 1 3,2 | 1 N -0.56 0.00 | -0.37 - -
## 1,3 | 1 N 0.65 0.00 | 0.45 - -
## 4,5 | 1 N 0.11 0.00 | 0.07 - -
## 6,4 | 1 N 0.44 0.00 | 0.29 - -
## 1,6 | 1 N 0.67 0.00 | 0.47 - -
## 7,1 | 1 N -0.72 0.00 | -0.51 - -
## 2 1,2;3 | 1 N 0.15 0.00 | 0.10 - -
## 7,3;1 | 1 N -0.18 0.00 | -0.11 - -
## 6,5;4 | 1 N 0.08 0.00 | 0.05 - -
## 1,4;6 | 1 N -0.03 0.00 | -0.02 - -
## 7,6;1 | 1 N 0.28 0.00 | 0.18 - -
## 3 7,2;1,3 | 1 N -0.02 0.00 | -0.02 - -
## 6,3;7,1 | 1 N -0.07 0.00 | -0.04 - -
## 1,5;6,4 | 1 N -0.24 0.00 | -0.15 - -
## 7,4;1,6 | 1 N -0.23 0.00 | -0.15 - -
## 4 6,2;7,1,3 | 1 N 0.28 0.00 | 0.18 - -
## 4,3;6,7,1 | 1 N -0.03 0.00 | -0.02 - -
## 7,5;1,6,4 | 1 N -0.21 0.00 | -0.14 - -
## 5 4,2;6,7,1,3 | 1 N 0.17 0.00 | 0.11 - -
## 5,3;4,6,7,1 | 1 N 0.12 0.00 | 0.07 - -
## 6 5,2;4,6,7,1,3 | 1 N 0.07 0.00 | 0.04 - -
## ---
## type: R-vine logLik: 2272 AIC: -4502 BIC: -4389
## ---
## 1 <-> acf, 2 <-> acv, 3 <-> acc, 4 <-> clor, 5 <-> st, 6 <-> den,
## 7 <-> ph
```

```
summary(fit.Gauss.ind)
```

```
## tree edge | family cop par par2 | tau utd ltd
## -----
## 1 3,2 | 1 N -0.56 0.00 | -0.37 - -
## 1,3 | 1 N 0.65 0.00 | 0.45 - -
## 4,5 | 1 N 0.11 0.00 | 0.07 - -
## 6,4 | 1 N 0.44 0.00 | 0.29 - -
## 1,6 | 1 N 0.67 0.00 | 0.47 - -
## 7,1 | 1 N -0.72 0.00 | -0.51 - -
## 2 1,2;3 | 1 N 0.15 0.00 | 0.10 - -
## 7,3;1 | 1 N -0.18 0.00 | -0.11 - -
## 6,5;4 | 1 N 0.08 0.00 | 0.05 - -
## 1,4;6 | 0 I - - | 0.00 - -
## 7,6;1 | 1 N 0.28 0.00 | 0.18 - -
## 3 7,2;1,3 | 0 I - - | 0.00 - -
## 6,3;7,1 | 1 N -0.07 0.00 | -0.04 - -
```

```

##          1,5;6,4 |      1  N -0.24  0.00 | -0.15  -  -
##          7,4;1,6 |      1  N -0.23  0.00 | -0.15  -  -
##    4      6,2;7,1,3 |      1  N  0.28  0.00 |  0.18  -  -
##          4,3;6,7,1 |      1  N -0.03  0.00 | -0.02  -  -
##          7,5;1,6,4 |      1  N -0.21  0.00 | -0.14  -  -
##    5      4,2;6,7,1,3 |      1  N  0.17  0.00 |  0.11  -  -
##          5,3;4,6,7,1 |      1  N  0.12  0.00 |  0.07  -  -
##    6  5,2;4,6,7,1,3 |      0  I      -      - |  0.00  -  -
## ---
## type: R-vine    logLik: 2267    AIC: -4498    BIC: -4401
## ---
## 1 <-> acf,    2 <-> acv,    3 <-> acc,    4 <-> clor,    5 <-> st,    6 <-> den,
## 7 <-> ph

```

Fitting C-vines using sequential estimation with restricted pair copula family set

```

fit.cv=RVineStructureSelect(udata7, familyset=c(1,5,6,4), selectioncrit="AIC",
                             indeptest=FALSE, level=0.05,type="CVine")
fit.cv.ind=RVineStructureSelect(udata7, familyset=c(1,5,6,4), selectioncrit="AIC",
                                indeptest=TRUE, level=0.05,type="CVine")
summary(fit.cv)

```

```

## tree          edge | family  cop   par  par2 |  tau  utd  ltd
## -----
##    1          1,4 |     16  SJ  1.34  0.00 |  0.16  -  0.32
##          1,2 |      5   F -1.69  0.00 | -0.18  -  -
##          1,5 |      1   N -0.10  0.00 | -0.06  -  -
##          1,3 |      4   G  1.80  0.00 |  0.44  0.53  -
##          1,6 |      4   G  1.82  0.00 |  0.45  0.54  -
##          7,1 |     34 G270 -1.99  0.00 | -0.50  -  -
##    2          6,4;1 |     16  SJ  1.46  0.00 |  0.21  -  0.39
##          6,2;1 |     14  SG  1.23  0.00 |  0.18  -  0.24
##          6,5;1 |     14  SG  1.19  0.00 |  0.16  -  0.21
##          6,3;1 |      5   F -0.78  0.00 | -0.09  -  -
##          7,6;1 |      4   G  1.24  0.00 |  0.19  0.25  -
##    3          3,4;6,1 |      6   J  1.11  0.00 |  0.06  0.13  -
##          3,2;6,1 |      5   F -3.84  0.00 | -0.38  -  -
##          3,5;6,1 |      1   N  0.15  0.00 |  0.10  -  -
##          7,3;6,1 |     24 G90 -1.08  0.00 | -0.08  -  -
##    4          7,4;3,6,1 |     26 J90 -1.24  0.00 | -0.12  -  -
##          7,2;3,6,1 |     36 J270 -1.06  0.00 | -0.04  -  -
##          7,5;3,6,1 |     26 J90 -1.21  0.00 | -0.11  -  -
##    5          2,4;7,3,6,1 |     14  SG  1.09  0.00 |  0.08  -  0.11
##          5,2;7,3,6,1 |      6   J  1.05  0.00 |  0.03  0.07  -
##    6  5,4;2,7,3,6,1 |     16  SJ  1.06  0.00 |  0.03  -  0.07
## ---
## type: C-vine    logLik: 2489    AIC: -4937    BIC: -4824
## ---
## 1 <-> acf,    2 <-> acv,    3 <-> acc,    4 <-> clor,    5 <-> st,    6 <-> den,
## 7 <-> ph

```

```
summary(fit.cv.ind)
```

```

## tree          edge | family  cop   par  par2 |  tau  utd  ltd

```

```

## -----
## 1      1,4 |      16  SJ  1.34  0.00 |  0.16   -  0.32
##      1,2 |      5   F  -1.69  0.00 | -0.18   -   -
##      1,5 |      1   N  -0.10  0.00 | -0.06   -   -
##      1,3 |      4   G   1.80  0.00 |  0.44  0.53  -
##      1,6 |      4   G   1.82  0.00 |  0.45  0.54  -
##      7,1 |     34  G270 -1.99  0.00 | -0.50   -   -
## 2      6,4;1 |     16  SJ   1.46  0.00 |  0.21   -  0.39
##      6,2;1 |     14  SG   1.23  0.00 |  0.18   -  0.24
##      6,5;1 |     14  SG   1.19  0.00 |  0.16   -  0.21
##      6,3;1 |      5   F  -0.78  0.00 | -0.09   -   -
##      7,6;1 |      4   G   1.24  0.00 |  0.19  0.25  -
## 3      3,4;6,1 |      0   I    -    - |  0.00   -   -
##      3,2;6,1 |      5   F  -3.84  0.00 | -0.38   -   -
##      3,5;6,1 |      1   N   0.15  0.00 |  0.10   -   -
##      7,3;6,1 |     24  G90 -1.08  0.00 | -0.08   -   -
## 4      7,4;3,6,1 |     24  G90 -1.14  0.00 | -0.12   -   -
##      7,2;3,6,1 |     36  J270 -1.06  0.00 | -0.04   -   -
##      7,5;3,6,1 |     26  J90 -1.21  0.00 | -0.11   -   -
## 5      2,4;7,3,6,1 |      4   G   1.09  0.00 |  0.09  0.11  -
##      5,2;7,3,6,1 |      6   J   1.05  0.00 |  0.03  0.07  -
## 6      5,4;2,7,3,6,1 |      0   I    -    - |  0.00   -   -
## ---
## type: C-vine      logLik: 2447      AIC: -4855      BIC: -4753
## ---
## 1 <-> acf,      2 <-> acv,      3 <-> acc,      4 <-> clor,      5 <-> st,      6 <-> den,
## 7 <-> ph

```

Fitting D-vines using sequential estimation with restricted pair copula family set

Since the traveling salesman problem is a NP hard problem, the package TSP provides different choices as solution. To be consistent with the book we fix the order to the one utilized there, which is given by

den acf ph acc acv st clor

6 1 7 3 2 5 4

```

d = dim(udata7)[2]
M = 1 - abs(TauMatrix(udata7))
hamilton = insert_dummy(TSP(M),label="cut")
sol = solve_TSP(hamilton,method="repetitive_nn")
order = cut_tour(sol,"cut")
order<-c(6,1,7,3,2,5,4)
names(order)<-c("den","acf","ph","acc","acv","st","clor")
DVM= D2RVine(order,family=rep(0,d*(d-1)/2),par=rep(0,d*(d-1)/2))
fit.dv=RVineCopSelect(data=udata7,familyset=c(1,5,6,4),indeptest=FALSE,
                      level=0.05,Matrix=DVM$Matrix,selectioncrit="AIC")
fit.dv.ind=RVineCopSelect(data=udata7,familyset=c(1,5,6,4),indeptest=TRUE,
                          level=0.05,Matrix=DVM$Matrix,selectioncrit="AIC")
summary(fit.dv)

```

```

## tree      edge | family  cop   par  par2 |  tau  utd  ltd
## -----
## 1      5,4 |     14  SG   1.09  0.00 |  0.09   -  0.12
##      2,5 |     14  SG   1.06  0.00 |  0.06   -  0.08

```

```

##          3,2 |      5      F -4.44 0.00 | -0.42  -  -
##          7,3 |      1      N -0.56 0.00 | -0.38  -  -
##          1,7 |     24     G90 -1.99 0.00 | -0.50  -  -
##          6,1 |      4      G  1.82 0.00 |  0.45 0.54  -
##  2        2,4;5 |     14     SG  1.11 0.00 |  0.10  - 0.13
##          3,5;2 |      1      N  0.10 0.00 |  0.07  -  -
##          7,2;3 |      5      F -0.80 0.00 | -0.09  -  -
##          1,3;7 |      4      G  1.40 0.00 |  0.28 0.36  -
##          6,7;1 |      4      G  1.24 0.00 |  0.19 0.25  -
##  3        3,4;2,5 |      4      G  1.19 0.00 |  0.16 0.21  -
##          7,5;3,2 |      5      F  0.16 0.00 |  0.02  -  -
##          1,2;7,3 |      4      G  1.11 0.00 |  0.10 0.13  -
##          6,3;1,7 |      5      F -0.65 0.00 | -0.07  -  -
##  4        7,4;3,2,5 |     34     G270 -1.12 0.00 | -0.11  -  -
##          1,5;7,3,2 |     26     J90 -1.15 0.00 | -0.08  -  -
##          6,2;1,7,3 |     16     SJ  1.33 0.00 |  0.16  - 0.32
##  5        1,4;7,3,2,5 |     14     SG  1.08 0.00 |  0.07  - 0.10
##          6,5;1,7,3,2 |      1      N  0.29 0.00 |  0.18  -  -
##  6        6,4;1,7,3,2,5 |     14     SG  1.25 0.00 |  0.20  - 0.26
## ---
## type: D-vine    logLik: 2451    AIC: -4860    BIC: -4747
## ---
## 1 <-> acf,    2 <-> acv,    3 <-> acc,    4 <-> clor,    5 <-> st,    6 <-> den,
## 7 <-> ph

```

```
summary(fit.dv.ind)
```

```

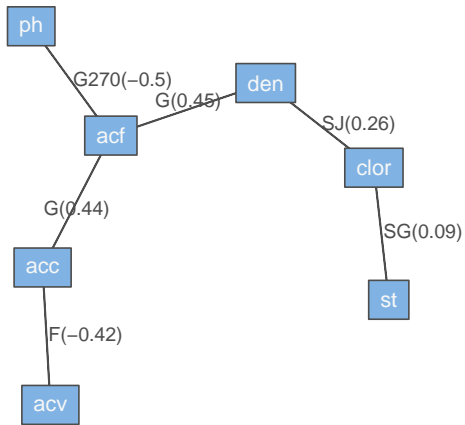
## tree          edge | family  cop   par  par2 |   tau  utd  ltd
## -----
##  1          5,4 |     14   SG  1.09 0.00 |  0.09  - 0.12
##          2,5 |     14   SG  1.06 0.00 |  0.06  - 0.08
##          3,2 |      5    F -4.44 0.00 | -0.42  -  -
##          7,3 |      1    N -0.56 0.00 | -0.38  -  -
##          1,7 |     24   G90 -1.99 0.00 | -0.50  -  -
##          6,1 |      4    G  1.82 0.00 |  0.45 0.54  -
##  2          2,4;5 |     14   SG  1.11 0.00 |  0.10  - 0.13
##          3,5;2 |      1    N  0.10 0.00 |  0.07  -  -
##          7,2;3 |      5    F -0.80 0.00 | -0.09  -  -
##          1,3;7 |      4    G  1.40 0.00 |  0.28 0.36  -
##          6,7;1 |      4    G  1.24 0.00 |  0.19 0.25  -
##  3          3,4;2,5 |      4    G  1.19 0.00 |  0.16 0.21  -
##          7,5;3,2 |      0    I  -  -  |  0.00  -  -
##          1,2;7,3 |      4    G  1.11 0.00 |  0.10 0.13  -
##          6,3;1,7 |      5    F -0.65 0.00 | -0.07  -  -
##  4          7,4;3,2,5 |     34   G270 -1.12 0.00 | -0.11  -  -
##          1,5;7,3,2 |     26   J90 -1.15 0.00 | -0.08  -  -
##          6,2;1,7,3 |     16   SJ  1.33 0.00 |  0.16  - 0.32
##  5          1,4;7,3,2,5 |     14   SG  1.08 0.00 |  0.07  - 0.10
##          6,5;1,7,3,2 |      1    N  0.28 0.00 |  0.18  -  -
##  6          6,4;1,7,3,2,5 |     14   SG  1.25 0.00 |  0.20  - 0.26
## ---
## type: D-vine    logLik: 2451    AIC: -4861    BIC: -4754
## ---
## 1 <-> acf,    2 <-> acv,    3 <-> acc,    4 <-> clor,    5 <-> st,    6 <-> den,
## 7 <-> ph

```

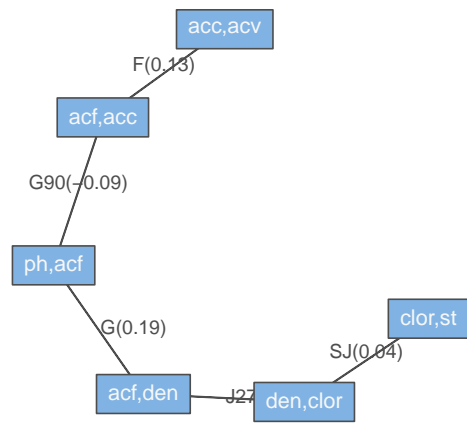
Figure 8.8: WINE7: Selected R-vine tree plots based on the Dissmann Algorithm 8.2.

```
bigpar(3,2)  
plot(fit.rv,edge.labels = "family-tau",type=1 )
```

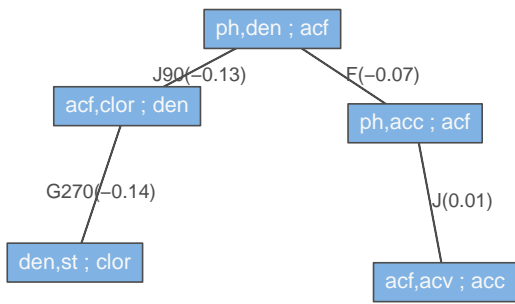

Tree 1



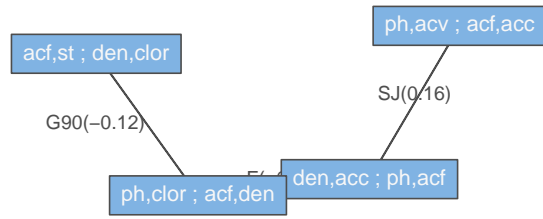
Tree 2



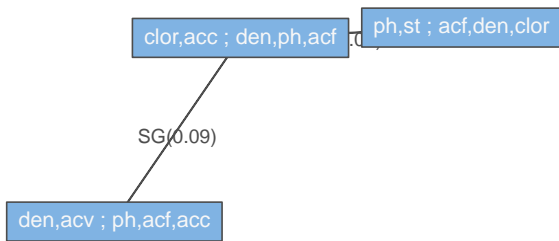
Tree 3



Tree 4



Tree 5



Tree 6

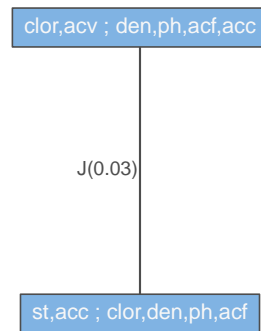
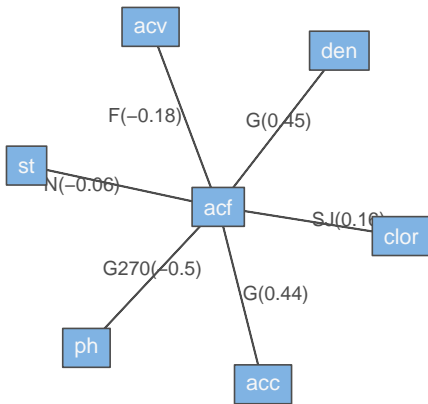


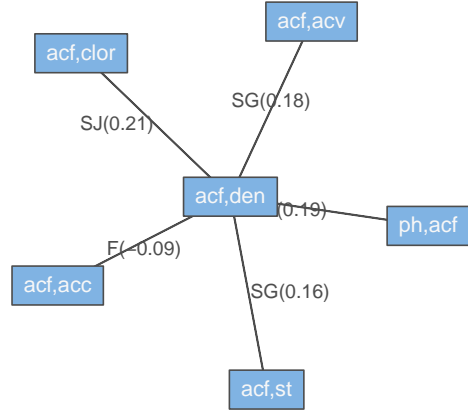
Figure 8.9: WINE7: Selected C-vine tree plots based on Dissmann Algorithm 8.2.

```
bigpar(3,2)
plot(fit.cv,edge.labels = "family-tau",type=1 )
```

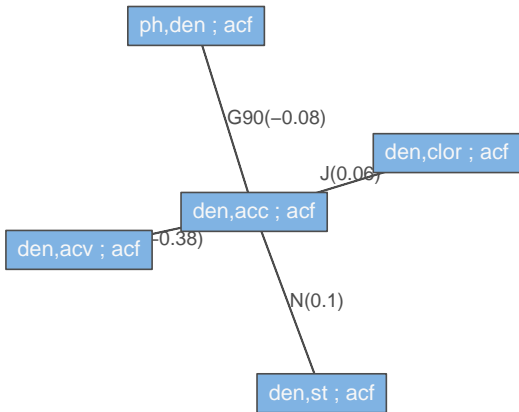
Tree 1



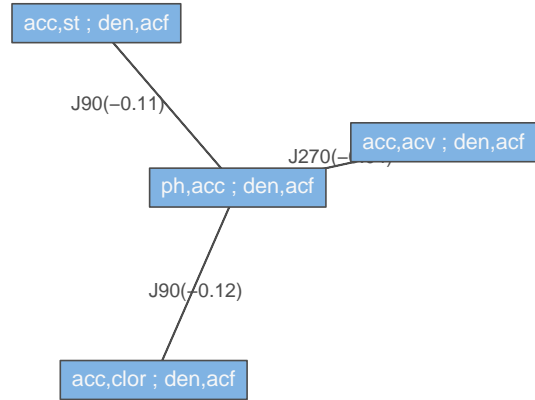
Tree 2



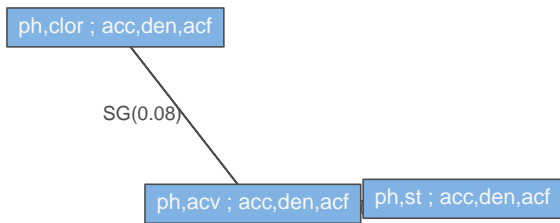
Tree 3



Tree 4



Tree 5



Tree 6

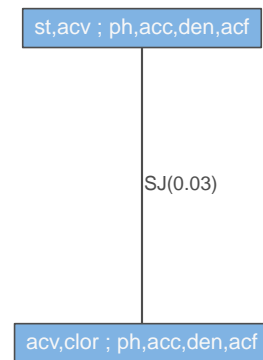
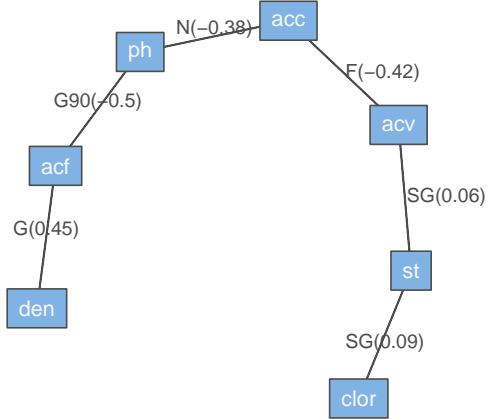


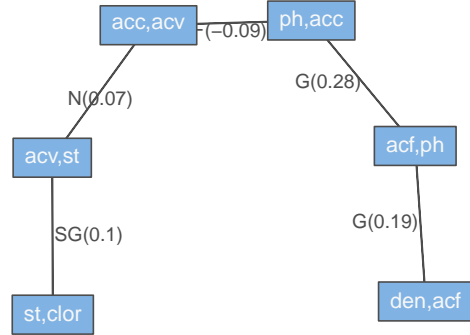
Figure 8.10: WINE7: Selected D-vine tree plots based on a solution of the traveling salesman problem

```
bigpar(3,2)
plot(fit.dv,edge.labels = "family-tau",type=1 )
```

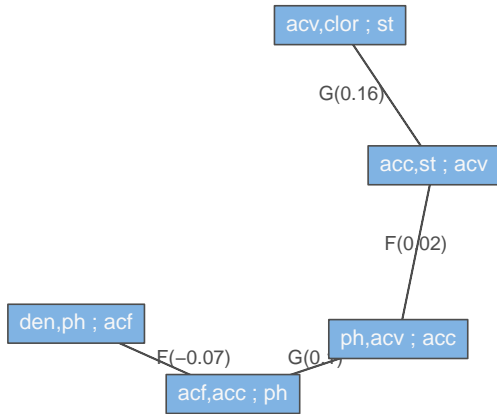
Tree 1



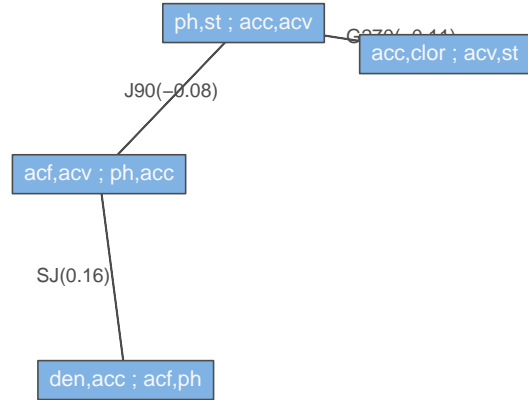
Tree 2



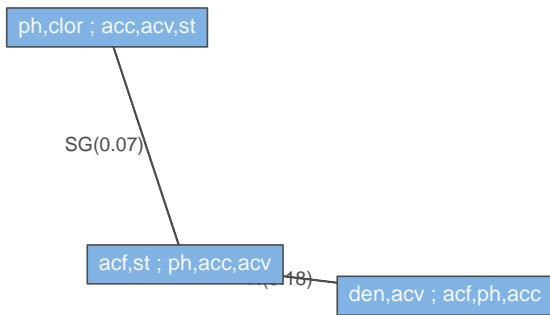
Tree 3



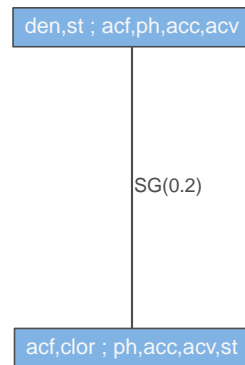
Tree 4



Tree 5



Tree 6



Analysis of chosen pair copula families for each tree

```
vine.fam<-function(fit=fit.rv.sum,familyset=c(1,2,5,6,4,14,24,34))
{
d<-ncol(fit$family)
sum.fam<-matrix(0,d-1,length(familyset))
rownames(sum.fam)<-paste("tree",2:d-1)
colnames(sum.fam)<-as.character(familyset)
for(i in d:2){
temp<-factor(as.character(fit$family[i,1:i-1]),levels=familyset)
temp
table(temp)
sum.fam[d-i+1,]<-table(temp)
}
sum.fam
}
fam.rv<-vine.fam(fit=fit.rv,familyset=c(1,5,6,4,16,26,36,14,24,34))
fam.rv.ind<-vine.fam(fit=fit.rv.ind,familyset=c(1,5,6,4,16,26,36,14,24,34,0))
fam.cv<-vine.fam(fit=fit.cv,familyset=c(1,5,6,4,16,26,36,14,24,34))
fam.cv.ind<-vine.fam(fit=fit.cv.ind,familyset=c(1,5,6,4,16,26,36,14,24,34,0))
fam.dv<-vine.fam(fit=fit.dv,familyset=c(1,5,6,4,16,26,36,14,24,34))
fam.dv.ind<-vine.fam(fit=fit.dv.ind,familyset=c(1,5,6,4,16,26,36,14,24,34,0))
```

Table 8.4: WINE7: Chosen pair copula families for different vine copula models. Models with ind at the ending denote a selection with an independence test performed for each pair copula in the Dissmann Algorithm 8.2

```
fam.out<-rbind(
c(apply(fam.rv,2,sum),0),
apply(fam.rv.ind,2,sum),
c(apply(fam.cv,2,sum),0),
apply(fam.cv.ind,2,sum),
c(apply(fam.dv,2,sum),0),
apply(fam.dv.ind,2,sum))
rownames(fam.out)<-
c("rvine","rvine-ind","cvine","cvine-ind","dvine","dvine-ind")
n.fam<-length(c(1,5,6,4,16,26,36,14,24,34,0))
familyset<-c(1,5,6,4,16,26,36,14,24,34,0)
for(i in (1:n.fam)){colnames(fam.out)[i]<-BiCopName(familyset[i])}
fam.out
```

```
##          N F J G SJ J90 J270 SG G90 G270 I
## rvine    0 5 2 3  3   1    1 2  2   2 0
## rvine-ind 0 4 1 3  3   1    0 2  2   2 3
## cvine    2 3 2 3  3   2    1 3  1   1 0
## cvine-ind 2 3 1 4  2   1    1 2  2   1 2
## dvine    3 4 0 5  1   1    0 5  1   1 0
## dvine-ind 3 3 0 5  1   1    0 5  1   1 1
```

Table 8.5: WINE7: Range of estimated (conditional) Kendall's τ values per tree for MLEestimated R-vine model (top left), Gaussian R-vine model (top right), C-vinemodel (bottom left) and D-vine model (bottom right)

Fitting R-vines, Gaussian R-vines, C-vines and D-vines using joint MLE

```
fit.mle=RVineMLE(udata7, fit.rv)
```

```
## iter 10 value -2527.262525
## iter 20 value -2527.296219
## final value -2527.297156
## converged
```

```
fit.mle.ind=RVineMLE(udata7, fit.rv.ind)
```

```
## iter 10 value -2521.146096
## final value -2521.169311
## converged
```

```
fit.mle.Gauss=RVineMLE(udata7, fit.Gauss)
```

```
## final value -2271.796370
## converged
```

```
fit.mle.Gauss.ind=RVineMLE(udata7, fit.Gauss.ind)
```

```
## final value -2267.147957
## converged
```

```
fit.cv.mle=RVineMLE(udata7, fit.cv)
```

```
## iter 10 value -2496.333280
## iter 20 value -2496.413547
## iter 30 value -2496.428978
## iter 40 value -2496.522471
## iter 50 value -2496.642744
## final value -2496.644184
## converged
```

```
fit.cv.mle.ind=RVineMLE(udata7, fit.cv.ind)
```

```
## iter 10 value -2453.953098
## iter 20 value -2454.161703
## iter 30 value -2454.169972
## iter 40 value -2454.194084
## iter 50 value -2454.391382
## iter 60 value -2454.452796
## iter 70 value -2454.457012
## final value -2454.457049
## converged
```

```
fit.dv.mle=RVineMLE(udata7, fit.dv)
```

```
## iter 10 value -2456.004582
## iter 20 value -2456.220726
## iter 30 value -2457.494929
## iter 40 value -2459.702427
## iter 50 value -2459.758492
```

```

## final value -2459.761101
## converged
fit.dv.mle.ind=RVineMLE(udata7, fit.dv.ind)

## iter 10 value -2455.790869
## iter 20 value -2456.197239
## iter 30 value -2459.090227
## iter 40 value -2459.536363
## iter 50 value -2459.541593
## final value -2459.541626
## converged

```

Table 8.5: WINE7: Range of estimated (conditional) Kendall's τ values per tree for MLE estimated R-vine model (top left), Gaussian R-vine model (top right), C-vine model (bottom left) and D-vine model (bottom right)

```

round(RVinePar2Tau(fit.mle$RVM),digits=2)

##      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
## [1,] 0.00 0.00 0.00 0.00 0.00 0.0  0
## [2,] 0.03 0.00 0.00 0.00 0.00 0.0  0
## [3,] 0.09 0.09 0.00 0.00 0.00 0.0  0
## [4,] 0.16 -0.01 -0.13 0.00 0.00 0.0  0
## [5,] 0.01 -0.06 -0.13 -0.13 0.00 0.0  0
## [6,] 0.11 -0.09 0.04 -0.03 0.20 0.0  0
## [7,] -0.41 0.44 0.09 0.26 0.46 -0.5  0

```

```

round(RVinePar2Tau(fit.mle.Gauss$RVM),digits=2)

##      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
## [1,] 0.00 0.00 0.00 0.00 0.00 0.00  0
## [2,] 0.04 0.00 0.00 0.00 0.00 0.00  0
## [3,] 0.11 0.07 0.00 0.00 0.00 0.00  0
## [4,] 0.18 -0.02 -0.14 0.00 0.00 0.00  0
## [5,] -0.02 -0.04 -0.15 -0.15 0.00 0.00  0
## [6,] 0.10 -0.11 0.05 -0.02 0.18 0.00  0
## [7,] -0.37 0.45 0.07 0.29 0.46 -0.51  0

```

```

round(RVinePar2Tau(fit.cv.mle$RVM),digits=2)

##      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
## [1,] 0.00 0.00 0.00 0.00 0.00 0.00  0
## [2,] 0.04 0.00 0.00 0.00 0.00 0.00  0
## [3,] 0.08 0.03 0.00 0.00 0.00 0.00  0
## [4,] -0.12 -0.03 -0.11 0.00 0.00 0.00  0
## [5,] 0.06 -0.38 0.10 -0.07 0.00 0.00  0
## [6,] 0.20 0.17 0.15 -0.08 0.20 0.00  0
## [7,] 0.14 -0.16 -0.05 0.42 0.46 -0.49  0

```

```

round(RVinePar2Tau(fit.dv.mle$RVM),digits=2)

##      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
## [1,] 0.00 0.00 0.00 0.00 0.00 0.00  0
## [2,] 0.20 0.00 0.00 0.00 0.00 0.00  0

```



```
## [3,] 0.08 0.19 0.00 0.00 0.00 0.00 0
## [4,] -0.11 -0.07 0.15 0.00 0.00 0.00 0
## [5,] 0.14 0.01 0.10 -0.06 0.00 0.00 0
## [6,] 0.09 0.06 -0.09 0.29 0.19 0.00 0
## [7,] 0.10 0.05 -0.40 -0.35 -0.49 0.46 0
```

```
tree.kendall.range<-function(RVM=fit.mle$RVM){
temp<-round(RVinePar2Tau(RVM),digits=2)
d<-ncol(temp)
treerange<-matrix(0,d-1,2)
rownames(treerange)<-paste("tree",1:(d-1))
colnames(treerange)<-c("min","max")
for(i in 1:(d-1)){
treerange[i,]<-range(temp[d-i+1,1:(d-i)])
}
treerange
}
tree.kendall.range(fit.mle$RVM)
```

```
##          min  max
## tree 1 -0.50 0.46
## tree 2 -0.09 0.20
## tree 3 -0.13 0.01
## tree 4 -0.13 0.16
## tree 5  0.09 0.09
## tree 6  0.03 0.03
```

```
tree.kendall.range(fit.mle.Gauss$RVM)
```

```
##          min  max
## tree 1 -0.51 0.46
## tree 2 -0.11 0.18
## tree 3 -0.15 -0.02
## tree 4 -0.14 0.18
## tree 5  0.07 0.11
## tree 6  0.04 0.04
```

```
tree.kendall.range(fit.cv.mle$RVM)
```

```
##          min  max
## tree 1 -0.49 0.46
## tree 2 -0.08 0.20
## tree 3 -0.38 0.10
## tree 4 -0.12 -0.03
## tree 5  0.03 0.08
## tree 6  0.04 0.04
```

```
tree.kendall.range(fit.dv.mle$RVM)
```

```
##          min  max
## tree 1 -0.49 0.46
## tree 2 -0.09 0.29
## tree 3 -0.06 0.14
## tree 4 -0.11 0.15
## tree 5  0.08 0.19
## tree 6  0.20 0.20
```