## HUDK5124Assignment10

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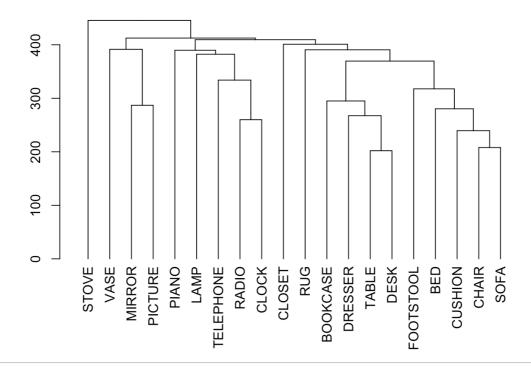
## 1

For either the sports data or the furniture data (both rated similarity, collected by Rosch et al.) posted in this session, fit an ultrametric tree to the data using the iterative projection method offered in R's "Is\_fit\_ultrametric" routine (see method below, also see posted help file "R\_fitting trees to prox data.txt"). Calculate the (squared) correlation of the model distances to the data, RSQ.

```
library(clue)
f1 <- read.table("FURN_prx_revised.txt", fill = TRUE, header = TR
UE) # add zero on diagonal before read the table
rownames(f1) <- names(f1)
f2 <- as.matrix(f1)
ultraD=ls_fit_ultrametric(f2,method=c("IP"), weights = 1, control
= list()) # fit an ultrametric tree (LS solution, by iterative pr
ojection) to a data set</pre>
```

```
## Warning in x - d: 长的对象长度不是短的对象长度的整倍数
```

```
hc <- hclust(ultraD, "ave") # method="average"
dend1 <- as.dendrogram(hc) # "print()" method
plot(dend1)</pre>
```



# calculate the "cophenetic correlation" - the corr of data & mod el distances

library(gdata)

## gdata: read.xls support for 'XLS' (Excel 97-2004) files ENABLE D.

##

## gdata: read.xls support for 'XLSX' (Excel 2007+) files ENABLED
.

##
## Attaching package: 'gdata'

## The following object is masked from 'package:stats':
##
## nobs

```
## The following object is masked from 'package:utils':
##
## object.size
```

```
## The following object is masked from 'package:base':
##
## startsWith
```

```
datvec = as.vector(lowerTriangle(f2, diag=FALSE)) # Function lowe
rTriangle() is in Package "gdata"
modvec = as.vector(ultraD)
cor(datvec, modvec)
```

```
## [1] 0.8035247
```

The correlation of data distances and model distances of the ultrametic tree is 0.7975155 for the furniture data.

## 2

Again using the iterative projection method offered in R, fit an additive tree to the same data set (using "Is\_fit\_addtree"). Calculate the (squared) correlation of the model distances to the data, RSQ.

```
\label{eq:addD} $$ addD=ls_fit_addtree(f2,method=c("IP"), weights = 1, control = list()) $$ $$ $$ fit an additive tree (LS solution, by "iterative projection" method to the data set
```

```
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑值
```

```
addD
```

```
## Dissimilarities using Additive tree distances:

## CHAIR SOFA TABLE DRESSER DESK BE

D STOVE

## SOFA 214.1667

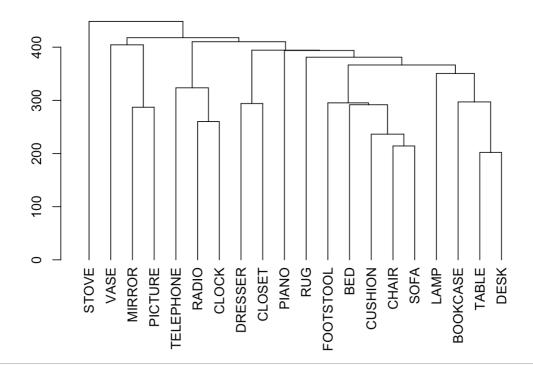
## TABLE 326.3816 341.0777

## DRESSER 356.6694 371.3655 318.4558
```

```
## DESK
           325.6594 340.3554 202.0000 317.7336
## BED
            272.6413 287.3374 332.1855 362.4734 331.4633
## STOVE
             439.7254 454.4215 401.5118 406.0981 400.7896 445.529
4
## BOOKCASE 365.8734 380.5695 297.3611 357.9477 296.6389 371.677
4 441.0037
## FOOTSTOOL 274.5898 289.2859 353.9292 384.2170 353.2070 300.188
9 467.2730
             375.7569 390.4529 337.5433 338.7913 336.8210 381.560
## LAMP
8 425.1856
## PIANO
            389.6795 404.3756 351.4659 360.3503 350.7436 395.483
4 443.4063
## MIRROR
            412.0675 426.7636 373.8539 378.4402 373.1317 417.871
4 461.1259
## RUG
             360.0412 374.7373 362.8570 393.1448 362.1348 365.845
2 476.2009
## RADIO
             404.8277 419.5237 366.6140 375.4985 365.8918 410.631
6 458.5545
             404.3277 419.0237 366.1140 374.9985 365.3918 410.131
## CLOCK
6 458.0545
## PICTURE
             414.4008 429.0969 376.1872 380.7735 375.4650 420.204
8 463.4592
## VASE
             407.1538 421.8499 368.9402 373.5265 368.2180 412.957
8 456.2122
## TELEPHONE 412.8129 427.5090 374.5993 383.4838 373.8771 418.616
9 466.5398
## CLOSET
            425.3916 440.0877 387.1780 294.0000 386.4558 431.195
6 474.8203
## CUSHION
             241.8333 231.0000 368.7443 399.0322 368.0221 315.004
1 482.0882
##
            BOOKCASE FOOTSTOOL
                                  LAMP
                                           PIANO MIRROR
UG
     RADIO
## SOFA
## TABLE
## DRESSER
## DESK
## BED
## STOVE
## BOOKCASE
## FOOTSTOOL 393.4210
## LAMP
             377.0351 403.3045
## PIANO
            390.9577 417.2271 379.4377
## MIRROR
            413.3457 439.6151 397.5276 415.7483
## RUG
             402.3488 387.5888 412.2323 426.1549 448.5429
## RADIO
            406.1059 432.3752 394.5859 391.8178 430.8965 441.30
```

```
31
## CLOCK
           405.6059 431.8752 394.0859 391.3178 430.3965 440.80
31 260.0000
## PICTURE
            415.6790 441.9484 399.8609 418.0817 287.0000 450.87
62 433.2299
## VASE
            408.4321 434.7014 392.6139 410.8347 403.1988 443.62
92 425.9829
## TELEPHONE 414.0912 440.3605 402.5712 399.8030 438.8818 449.28
84 323.7500
## CLOSET
            426.6699 452.9392 407.5136 429.0725 447.1624 461.86
71 444.2207
## CUSHION 408.2362 316.9526 418.1196 432.0422 454.4302 402.40
40 447.1904
##
               CLOCK PICTURE VASE TELEPHONE CLOSET
## SOFA
## TABLE
## DRESSER
## DESK
## BED
## STOVE
## BOOKCASE
## FOOTSTOOL
## LAMP
## PIANO
## MIRROR
## RUG
## RADIO
## CLOCK
## PICTURE 432.7299
## VASE
            425.4829 405.5322
## TELEPHONE 323.2500 441.2152 433.9682
## CLOSET 443.7207 449.4957 442.2487 452.2060
## CUSHION 446.6904 456.7636 449.5166 455.1757 467.7544
```

```
hc2 <- hclust(addD, "ave")
dend2 <- as.dendrogram(hc2)
plot(dend2)</pre>
```



modvec2 = as.vector(addD)

cor(datvec, modvec2) # The correlation of data distances and model distances of the additive tree is 0.9067304 for the furniture dat a at this run.

## ## [1] 0.9012762

```
# However, IP uses a semi-randomized start. It is susceptible to
local minima, so run multiple starts.
corvector <- c(rep(0,25))
for (i in 1:25)
{addD3=ls_fit_addtree(f2,method=c("IP"), weights = 1, control = 1
ist())
modvec3=as.vector(addD3)
corvector[i]=cor(datvec,modvec3) # calculate linear fit = corr of
data & model distances
}</pre>
```

## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑值

```
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑
值
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑
值
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑
值
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑
佰
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑
佰
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑
值
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑
值
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑
佰
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑
值
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑
值
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑
值
```

```
## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑值 ## Warning in any(diff(weights)): 将种类为'double'的参数强迫转化为逻辑值
```

corvector # the vector of fits shows that local minima are very c ommon (cf. Smith, 1998)

```
## [1] 0.9086624 0.9030464 0.9105745 0.9065246 0.9040460 0.90290 98 0.9004377
## [8] 0.9092728 0.9063303 0.9093261 0.9091252 0.8974086 0.90649 95 0.9034255
## [15] 0.9052995 0.9041801 0.9001116 0.9072837 0.9011703 0.89757 60 0.9047619
## [22] 0.9054518 0.9087903 0.9036018 0.8991714
```

Also fit an additive tree to the data using GTREE. Compare the fit of this solution to the results using the R methods. Comment. Although I cannot run the executable file "gtree.exe" anyhow, I will set down the furthest step I am currently able to get. 1) Download Pascal Compiler for Mac: go to Website free pascal (http://www.freepascal.org/down/i386/macosx-hungary.var) and download fpc-3.0.2.intel-macosx.dmg (ftp://ftp.hu.freepascal.org/pub/fpc/dist/3.0.2/i386-macosx/fpc-3.0.2.intel-macosx.dmg) 2) Install the Mac OS X installer package named "fpc-3.0.2.intel-macosx.pkg" 3) Download the file gtree.pas (http://netlib.sandia.gov/mds/gtree.pas) 4) Open "Terminal" on Mac and Type the following lines after the dollar sign: \$ cd address\_of\_the\_repository\_where\_I\_saved\_gtree.pas \$ fpc gtree.pas 5) Open the folder where I saved "gtree.pas", I saw a Unix executable file named "gtree" 6) Double click that file and start fitting the additive tree. (But it always shows error 2 (fild not found). That's where I cannot get anything further.)

**References** Smith, T. J. (1998). A comparison of three additive tree algorithms that rely on a leastsquares