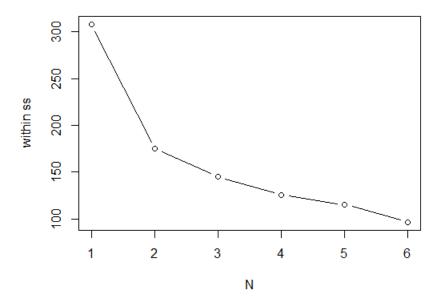
```
Problem(1)
Open the data:
crime <- as.data.frame(crime)
#rm the outlier#
pairs(crime)
which(crime$Murder>15)
rm.crime=crime[-24,]
sd=sapply(rm.crime,sd)
crime.s=sweep(rm.crime,2,sd,FUN="/")
#find the proper K#
n=nrow(crime.s)
wss=rep(0,6)
wss[1]=(n-1)*sum(sapply(crime.s,var))
for(i in 2:6){
wss[i]=sum(kmeans(crime.s,centers=i)$withinss)
}
plot(1:6,wss,type="b",xlab="N",ylab="within ss")
final=kmeans(crime.s,centers=3)
names(final)
result=list(names(final$cluster[final$cluster==1]),names(final$cluster[final$cluster==2]),names(final$clu
ster[final$cluster==3]))
#show the result:
names(result)=c("group1","group2","group3")
result
```



3 centers would be much better. And let K=3, we get:

```
> result
$group1
  [1] "ME" "NH" "VT" "PA" "IN" "WI" "MN" "IA" "ND" "SD" "NE" "KS" "VA" "WV" "K
Y" "MS" "AR" "MT" "ID"
[20] "WY" "UT" "HI"

$group2
  [1] "MA" "RI" "CT" "NY" "NJ" "OH" "IL" "MO" "DE" "MD" "NC" "SC" "GA" "TN" "A
L" "OK"

$group3
  [1] "MI" "FL" "LA" "TX" "CO" "NM" "AZ" "NV" "WA" "OR" "CA" "AK"
```

The result is different from the textbook, where it uses ranges to standardize variables.

Problem (2)

library(MASS)

library("KernSmooth")

data(pottery)

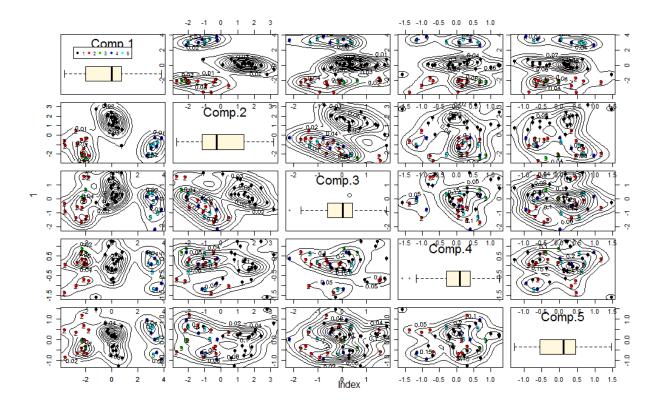
fix(pottery)

names(pottery)

pr.out=princomp(pottery[,1:9],cor=T)

```
summary(pr.out,loadings=T)
names(pr.out)
score=as.data.frame(pr.out$scores[,1:5])
score$kiln=pottery$kiln
library(SciViews)# use the function "panel.boxplot"
#draw the scatterplot matrix
pairs(score[,1:5],
   diag.panel =panel.boxplot,
   panel = function (x,y) {
    data <- data.frame(cbind(x,y))
    par(new = TRUE)
    plot(x,y,pch=16,col=score$kiln)
    text(x,y,labels=score$kiln)
    den <- bkde2D(data, bandwidth = sapply(data, dpik))</pre>
    contour(x = den$x1, y = den$x2,
        z = den$fhat, axes = FALSE,add=T)
#to add a legend#
par(new=T)
plot(1,axes=F)
```

legend("topleft",pch=16,col=c(1,2,3,4,5),legend=c(1,2,3,4,5),ncol=5,cex=0.5,inset=0)



Problem(3)

data("USairpollution")

data=USairpollution

r.data=data[,-1]

library(mclust)

#do model based cluster analysis by using Mclust() function:

r.data.m=Mclust(r.data)

#show the result of cluster

summary(r.data.m,parameters = T)

# BIC for model selection and boxplot to show the distribution of so2 in each cluster

plot(r.data.m)

names(r.data.m)

result=as.data.frame(r.data.m\$classification)

data\$group=result[,1]

## names(data)

boxplot(data\$\$O2~data\$group,ylab="level of so2",xlab="group")

the result of cluster analysis by finite mixtures method:

> summary(r.data.m,parameters = F)

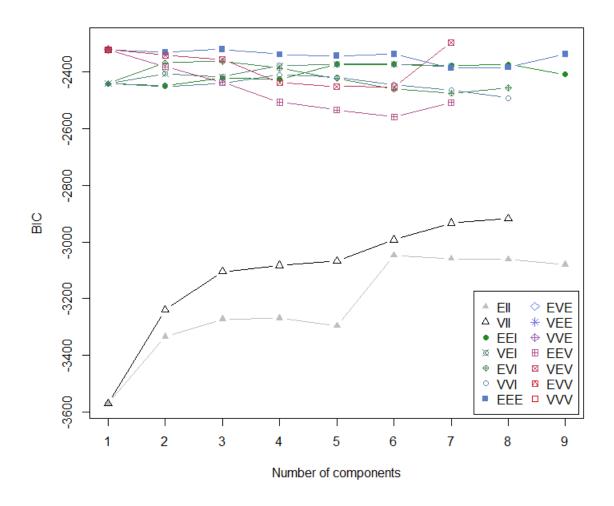
Gaussian finite mixture model fitted by EM algorithm

Mclust VEV (ellipsoidal, equal shape) model with 7 components:

log.likelihood n df -841.0466 41 165 -2294.833 -2294.833

Clustering table:

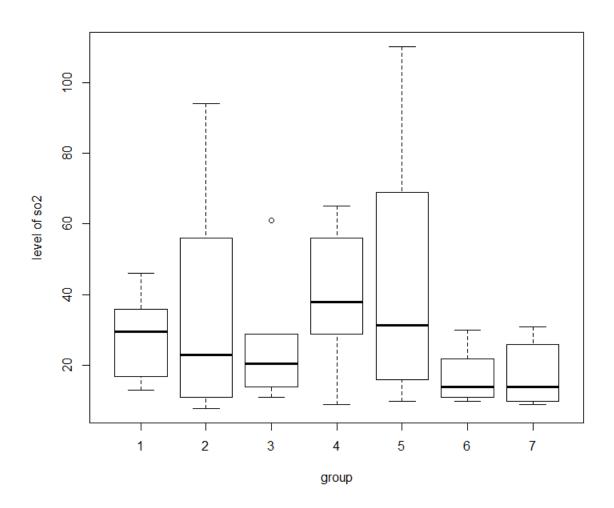
1 2 3 4 5 6 7 6 5 6 6 6 7 5



Model selection process induced by BIC criterion:

The BIC criterion selects model VEV and 7 clusters as the optimal solution.

How Sulphur dioxide concentration varies in the clusters?



## The variation of so2 in each cluster:

plot(1,ylim=c(8,110),xlim=c(1,10),main="so2 variation",xlab="NA",ylab="SO2 level",axes=F)

for(i in unique(data\$group)){

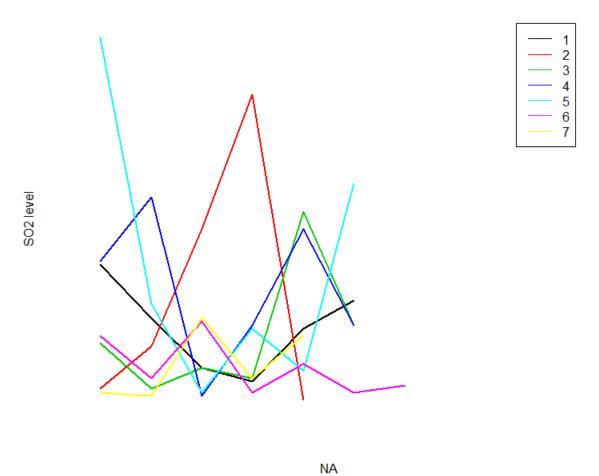
ind=data[data\$group==i,]

lines(seq(length(ind\$SO2)),ind\$SO2,col=i,lwd=2,type="I")

}

legend("topright",col=unique(data\$group),lty=1,legend=unique(data\$group))

## so2 variation



I don't think the so2 variation in each cluster is a good indication of how well the cluster analysis goes.