

K-means Clustering

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Adding cluster to a PCA plot

The usual libraries. You will probably need to install the factoextra package.

```
suppressMessages(library(tidyverse))
suppressMessages(library(MASS))
suppressMessages(library(factoextra))
```

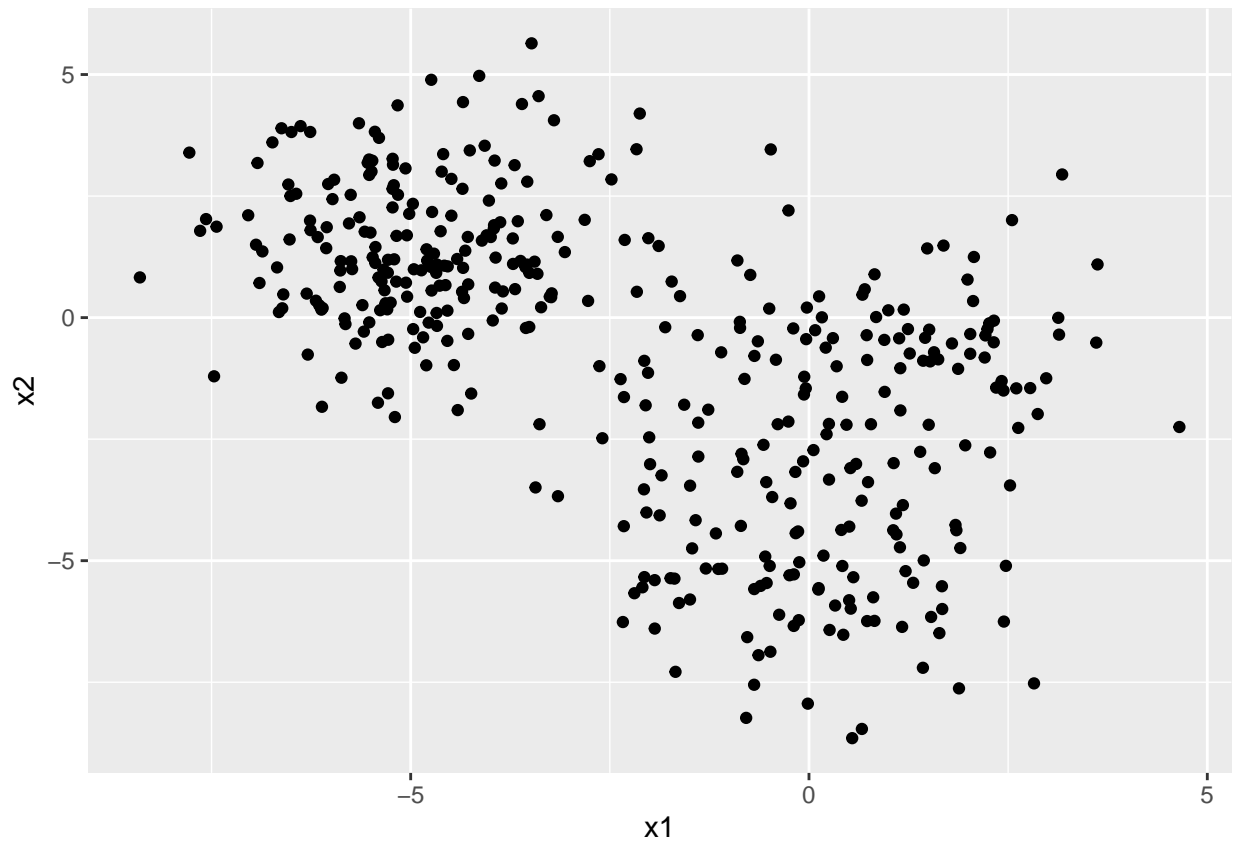
Data with more than 2 dimensions.

Create a data frame with 4 dimensions and

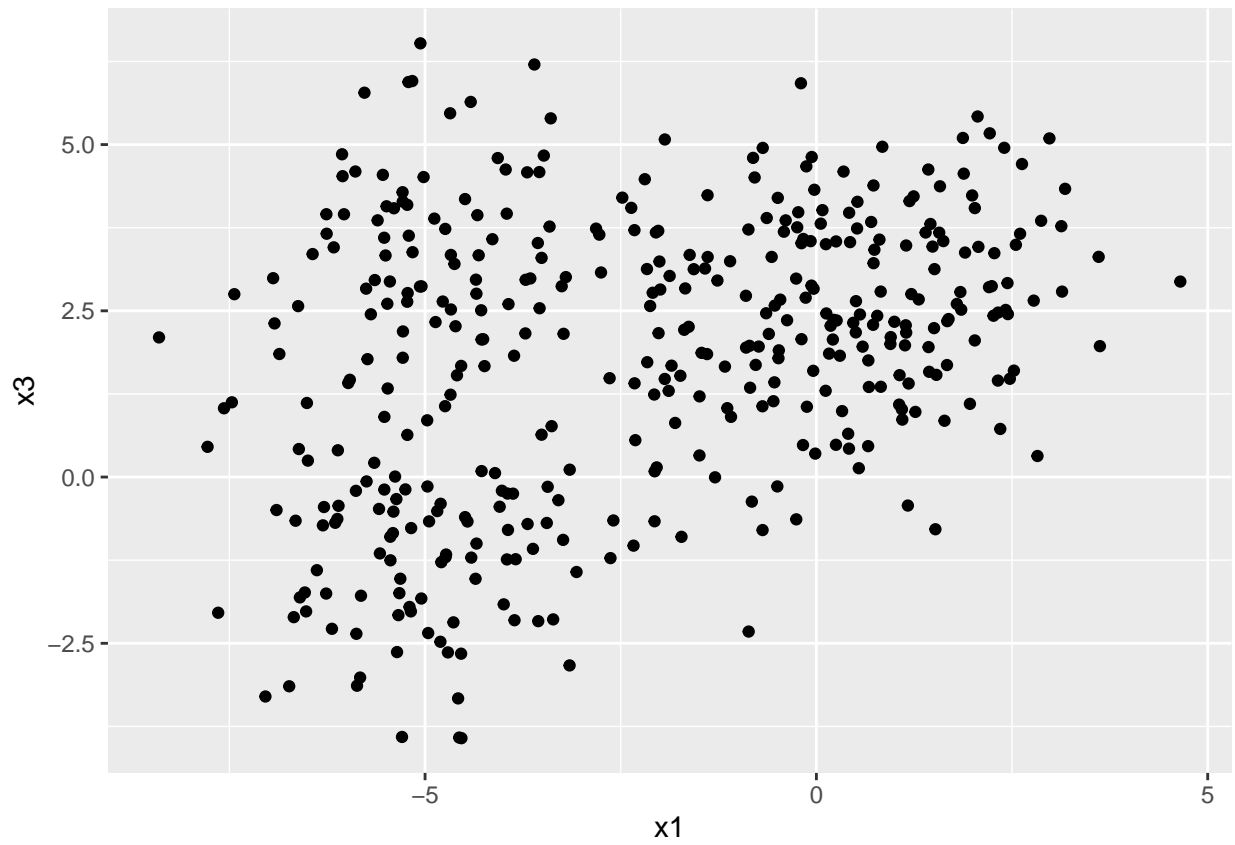
```
N <- 100
K <- 4
mu <- sample(-5:5, 5*K, rep=T)
sd0 <- 2
dat <- c()
for(k in 1:K){
  dat <- c(dat, c(mvrnorm(N, c(mu[2*k-1], mu[2*k]), diag(c(1,1)*sd0)),
                  mvrnorm(N, c(mu[2*k+1], mu[2*k+2]), diag(c(1,1)*sd0))))
}
dat <- matrix(dat, byrow=F, ncol=K)
data.df <-
  data.frame(x1=dat[,1],
             x2=dat[,2],
             x3=dat[,3],
             x4=dat[,4])
```

Since we are in four dimensions, there is no simple visualization available. You could try some pairwise plots.

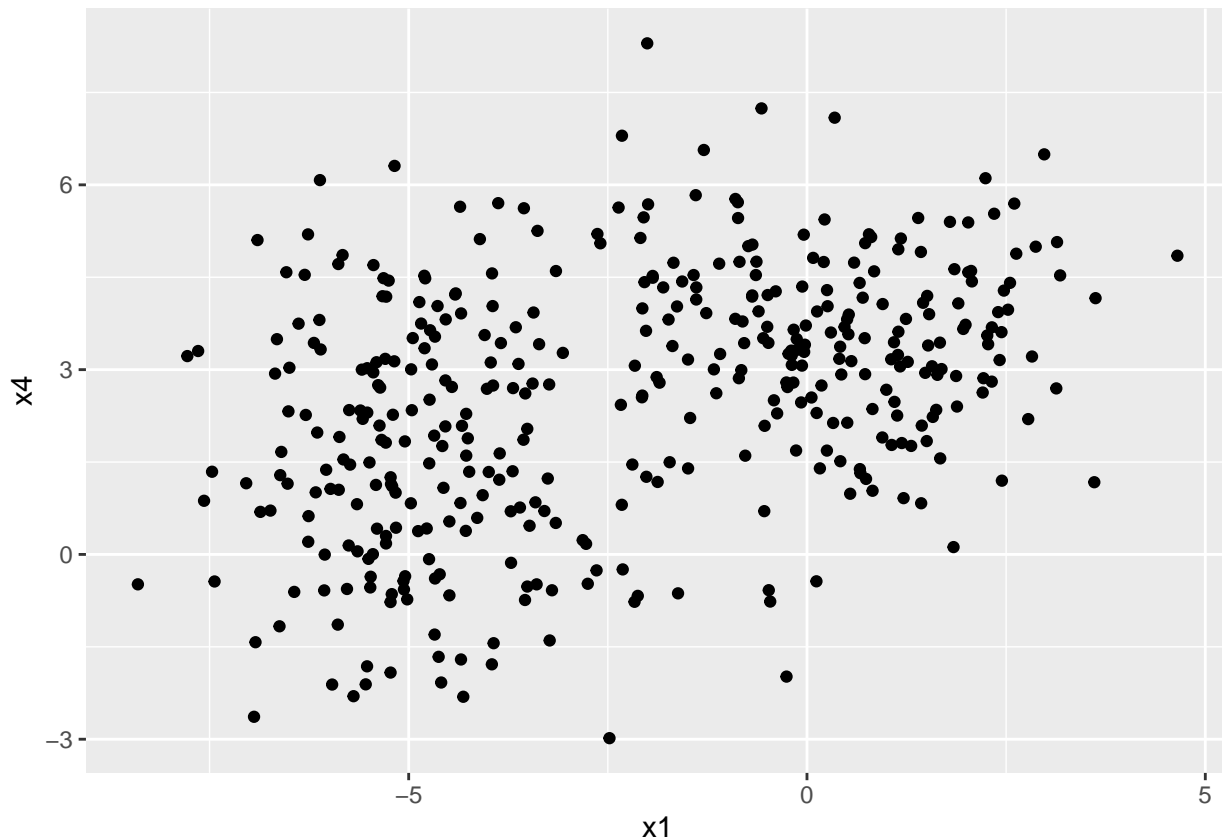
```
ggplot(data.df, aes(x1, x2)) + geom_point()
```



```
ggplot(data.df, aes(x1, x3)) + geom_point()
```



```
ggplot(data.df, aes(x1, x4)) + geom_point()
```



Not too helpful.

However, we can still cluster. Just to be safe, scale the data and repack into data frame.

```
data.df <- scale(data.df)
data.df <- data.frame(data.df)
```

Apply kmeans with, say, K=5 means.

```
K<-5
mod.km <- kmeans(data.df,K,nstart=25)
data.df$cluster <- factor(mod.km$cluster)
```

Ok..what do we do now, we have a clustering, but how does it look?

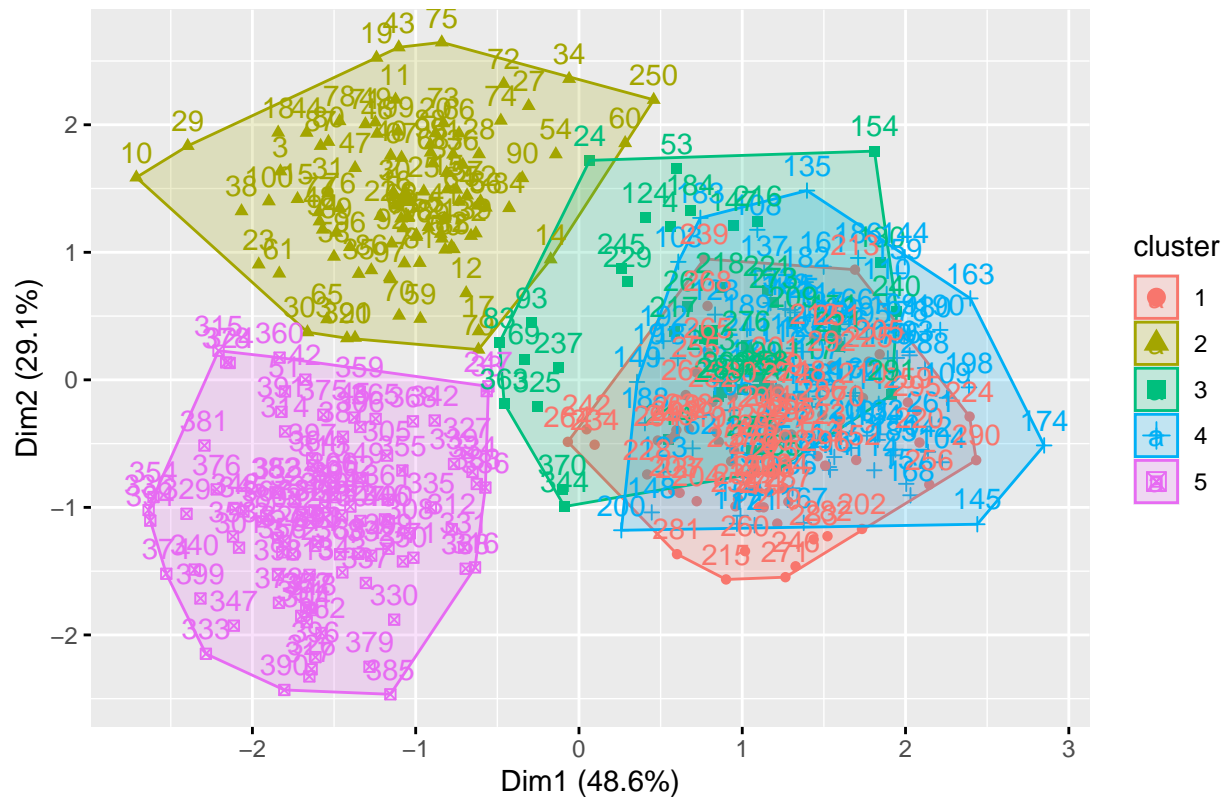
Here's the plan: Perform a Principal Component Analysis and project into 2-dimensional space. Carry the clusters along with the projection and see what we have.

The `fviz_cluster` function will do this.

I.e, `fviz_cluster` will project onto the “best” two dimensions. This is essentially the biplot with clustering information included.

```
## make sure we only use the original data!
fviz_cluster(mod.km,data=data.df[,1:4])
```

Cluster plot



The boundaries are the “convex hulls” around each cluster. These are added to help visualize the clustering.

Note: We can build this ourselves (except the convex hulls).

Solution: Build it yourself

Cluster with $k=4$

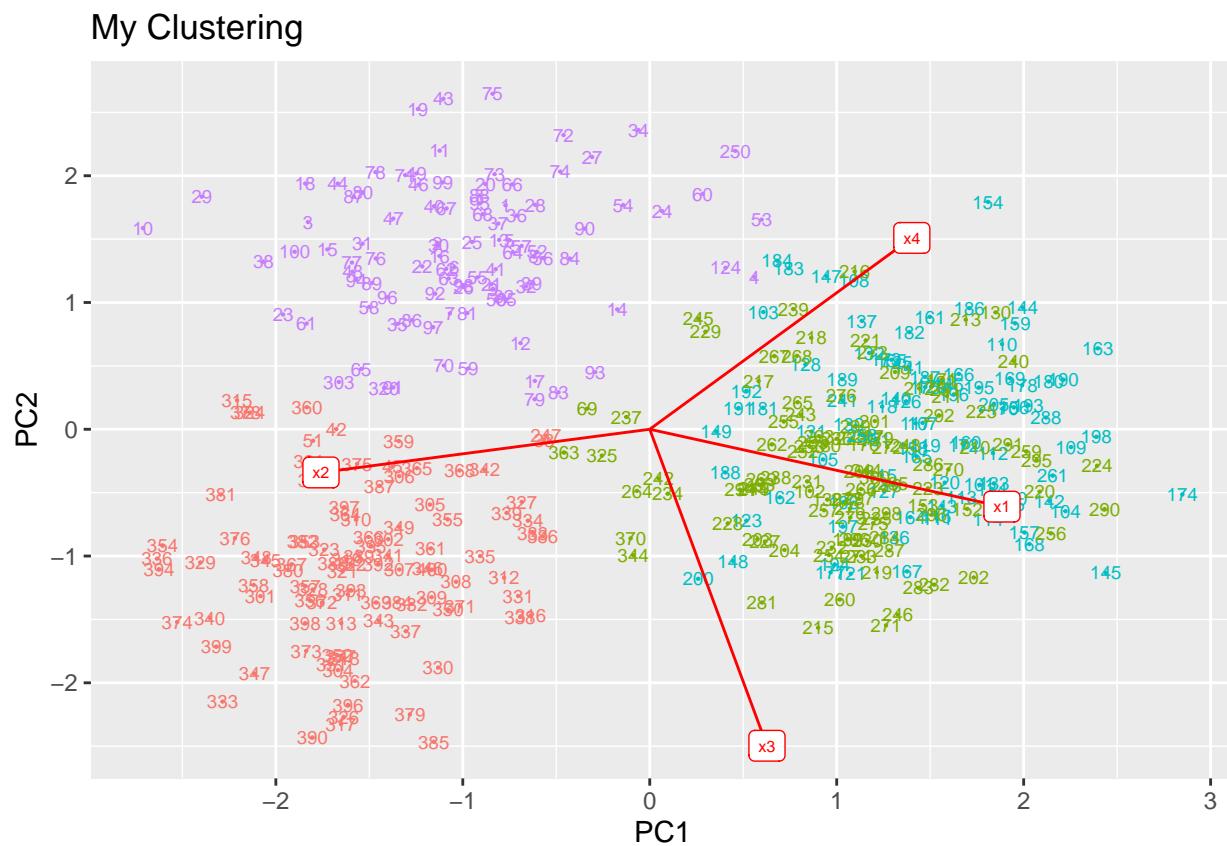
```
data.mat <- data.matrix(data.df[,1:4])
mod.km <- kmeans(data.mat,centers=4,nstart=10)
table(mod.km$cluster)
```

```
##
##  1  2  3  4
## 99 108 94 99
```

```
mod.pr <- prcomp(data.mat)
## Pull off the rotation matrix.
rot.mat <- mod.pr$rotation
## Rotate the data
dataRotate.mat <- data.mat %*% rot.mat
dataRotate.df <- data.frame(dataRotate.mat)
dataRotate.df$cluster <- factor(mod.km$cluster)
dataRotate.df$id <- 1:nrow(data.df)
loading.df <- data.frame(rot.mat)
loading.df$pred <- rownames(rot.mat)
```

Now create the plot, include the loading vectors as well.

```
load.scale <- 3
dataRotate.df%>%
  ggplot()+
  geom_point(aes(PC1,PC2,color=cluster),size=.1)+
  geom_text(aes(PC1,PC2,color=cluster,label=id),
            size=2.5)+
  geom_segment(data=loading.df,
              aes(x=0,xend=load.scale*PC1,
                  y=0,yend=load.scale*PC2),
              color="red")+
  geom_label(data=loading.df,
            aes(load.scale*PC1,
                load.scale*PC2,
                label=pred),
            color="red", size=2)+
  guides(color=F)+
  ggtitle("My Clustering")
```



The data are complete random so there isn't much to say about the loading directions.