

# Coding Assignment

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```
library(mclust)
library(cluster)
library(ggplot2)
```

We firstly write a function `getA1` for simulating high-dimensional data ( $p=1000$ ) with three groups of observations where the number of observations is  $n=100$ :

```
getA1 <- function(){
  n_rows = 1000
  n_cols = 100

  k=3
  x_mus = c(0,5,5)
  x_sds = c(1,0.1,1)
  y_mus = c(5,5,0)
  y_sds = c(1,0.1,1)
  prop1 = c(0.3,0.5,0.2)

  comp1 <- sample(seq_len(k), prob=prop1, size=n_cols, replace=TRUE)
  samples1 <- cbind(rnorm(n=n_cols, mean=x_mus[comp1],sd=x_sds[comp1]),
                    rnorm(n=n_cols, mean=y_mus[comp1],sd=y_sds[comp1]))

  proj <- matrix(rnorm(n_rows* n_cols), nrow=n_rows, ncol=2)
  A1 <- samples1 %*% t(proj)
  A1 <- A1 + rnorm(n_rows* n_cols)
  return (list("data" = A1, "labels" = comp1))
}
```

We firstly take a look at a single run, and find out the optimal number of clustering; we plot total within groups sum of squares against values of  $k$ , we pick  $k$  to be the elbow point, which corresponding to  $k = 3$ .

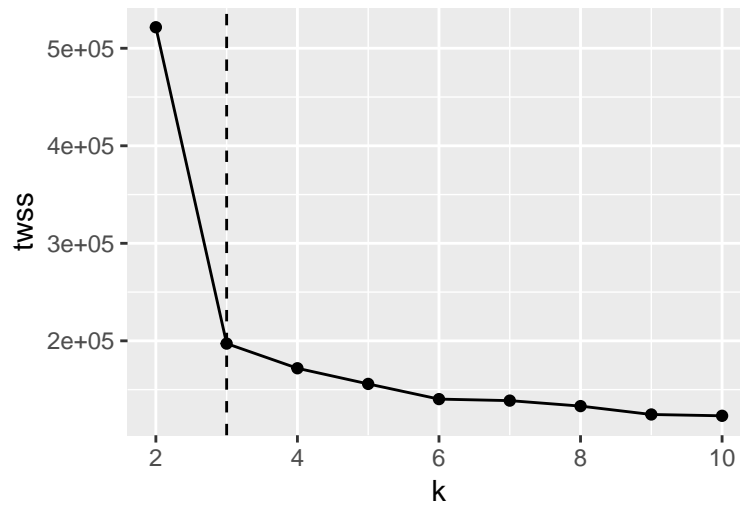
```
set.seed(1)

result = getA1()
A1=result$data

#function for calculating total within groups sum of squares
twss <- function(fit){
  return(fit$tot.withinss)
}

result = data.frame(k=c(2:10),twss=apply(2:10,function(k){twss(kmeans(A1, k,nstart = 25))}))
ggplot(data=result, aes(x=k, y=twss)) + geom_line()+geom_point()+ geom_vline(xintercept = 3,linetype = "dashed")
```

## Optimal number of clusters



k.opt=3

We generate simulated high-dimensional data and perform K-means 100 times; and we calculate the adjusted rand index and the total within clusters sum of squares for each run:

```
metrics <- data.frame(ARI=numeric(0),WSS=numeric(0))

for (i in 1:100) {
  result = getA1()
  A1 = result$data
  lbs = result$labels
  KM = kmeans(A1, k.opt, nstart = 25)
  clusters <- KM$cluster
  new <- data.frame(adjustedRandIndex(clusters, lbs), twss(KM))
  names(new)<-c("ARI", "WSS")
  metrics <- rbind(metrics, new)
}

metrics
```

```
##      ARI    WSS
## 1  1.0000 195850
## 2  1.0000 168021
## 3  0.9356 196977
## 4  1.0000 193370
## 5  1.0000 183157
## 6  1.0000 175918
## 7  1.0000 171186
## 8  1.0000 186844
## 9  1.0000 184879
## 10 1.0000 181435
## 11 1.0000 172146
## 12 1.0000 191679
## 13 1.0000 185770
## 14 1.0000 162601
## 15 1.0000 196046
## 16 0.9722 206972
## 17 0.9644 213957
```

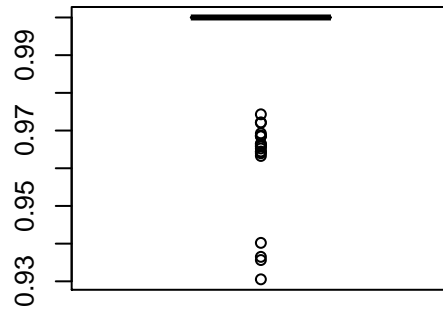
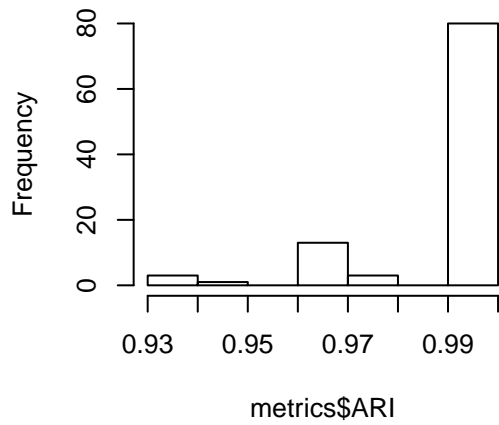
##	18	1.0000	210627
##	19	1.0000	174344
##	20	1.0000	236300
##	21	0.9365	219279
##	22	1.0000	178922
##	23	1.0000	199970
##	24	1.0000	184080
##	25	1.0000	165469
##	26	1.0000	173994
##	27	1.0000	192761
##	28	1.0000	200808
##	29	1.0000	185874
##	30	1.0000	228024
##	31	1.0000	209945
##	32	1.0000	212046
##	33	0.9686	172977
##	34	0.9658	204462
##	35	1.0000	189869
##	36	1.0000	183481
##	37	1.0000	212827
##	38	1.0000	182801
##	39	1.0000	192912
##	40	1.0000	169618
##	41	0.9685	174324
##	42	1.0000	208254
##	43	0.9402	185029
##	44	1.0000	199699
##	45	1.0000	193614
##	46	0.9652	194784
##	47	1.0000	200334
##	48	1.0000	210419
##	49	0.9664	205230
##	50	1.0000	197194
##	51	0.9644	186062
##	52	1.0000	183654
##	53	1.0000	165188
##	54	0.9659	222213
##	55	1.0000	188083
##	56	1.0000	182992
##	57	1.0000	207603
##	58	1.0000	194339
##	59	0.9665	209267
##	60	0.9743	208061
##	61	1.0000	192761
##	62	1.0000	186339
##	63	1.0000	205345
##	64	1.0000	184401
##	65	1.0000	205249
##	66	1.0000	193820
##	67	1.0000	185795
##	68	1.0000	197389
##	69	1.0000	202249
##	70	1.0000	205306
##	71	1.0000	197607

```
## 72 1.0000 185899
## 73 1.0000 178066
## 74 1.0000 190200
## 75 1.0000 195433
## 76 1.0000 179916
## 77 0.9305 207111
## 78 1.0000 196342
## 79 1.0000 180300
## 80 1.0000 200982
## 81 1.0000 183801
## 82 0.9721 180053
## 83 1.0000 204807
## 84 0.9640 197556
## 85 1.0000 205202
## 86 1.0000 210403
## 87 0.9633 227681
## 88 1.0000 194895
## 89 0.9693 227222
## 90 1.0000 187314
## 91 1.0000 187858
## 92 1.0000 184723
## 93 1.0000 217130
## 94 1.0000 167672
## 95 0.9687 158808
## 96 1.0000 193799
## 97 1.0000 205198
## 98 1.0000 166794
## 99 1.0000 228617
## 100 1.0000 172329
```

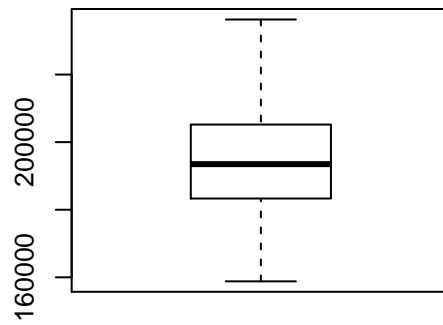
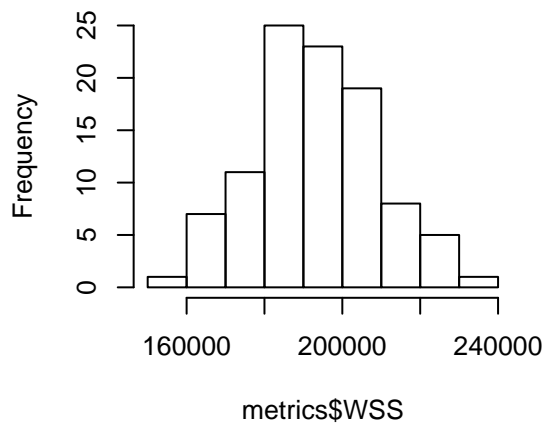
Now we use boxplot and histogram to view the result of adjusted rand index and the total within clusters sum of squares:

```
par(mfrow=c(2,2))
hist(metrics$ARI)
boxplot(metrics$ARI)
hist(metrics$WSS)
boxplot(metrics$WSS)
```

**Histogram of metrics\$ARI**



**Histogram of metrics\$WSS**



By the result of adjusted rand index, we know our K-means model has great accuracy.