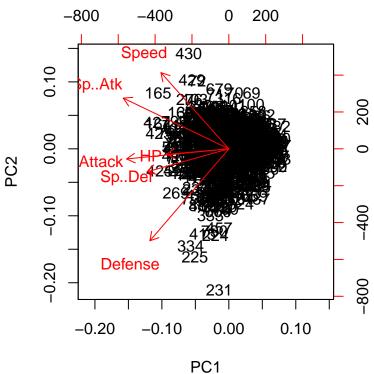
# Pokemon

```
library(readr)
Pokemon <- read.csv("Pokemon.csv")</pre>
View(Pokemon)
names (Pokemon)
                                    "Type.1"
    [1] "X."
                                                 "Type.2"
                                                               "Total"
##
                      "Name"
    [6] "HP"
                      "Attack"
                                   "Defense"
                                                 "Sp..Atk"
                                                               "Sp..Def"
## [11] "Speed"
                      "Generation" "Legendary"
Pokemon_data <- as.matrix(Pokemon[,c(6:11)])</pre>
##Peform PCA
colMeans(Pokemon_data)
              Attack Defense Sp..Atk Sp..Def
                                                     Speed
## 69.25875 79.00125 73.84250 72.82000 71.90250 68.27750
pok.pr <- prcomp(Pokemon_data,scale=TRUE)</pre>
summary(pok.pr)
## Importance of components:
                              PC1
                                     PC2
                                             PC3
                                                    PC4
                                                             PC5
                                                                     PC6
## Standard deviation
                           1.6466 1.0457 0.8825 0.8489 0.65463 0.51681
## Proportion of Variance 0.4519 0.1822 0.1298 0.1201 0.07142 0.04451
## Cumulative Proportion 0.4519 0.6342 0.7640 0.8841 0.95549 1.00000
# Interpreting PCA results with scale=TRUE
biplot(pok.pr)
                 -10
                                        20
          -20
                          0
                                 10
                                                30
                                   Defense
     0.1
                                                     10
     0.0
                                                     0
                                        Attack
     0.1
                                   Speed
                             430
               -0.1
                                   0.1
                                             0.2
                         0.0
```

PC<sub>1</sub>

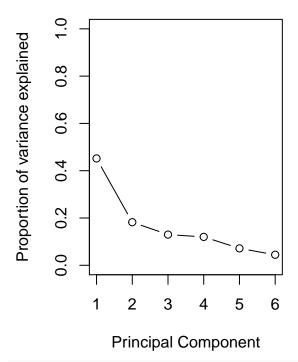
```
#Interpreting PCA results with scale=FALSE
pok.pr.ns <- prcomp(Pokemon_data,scale=FALSE)
biplot(pok.pr.ns)</pre>
```



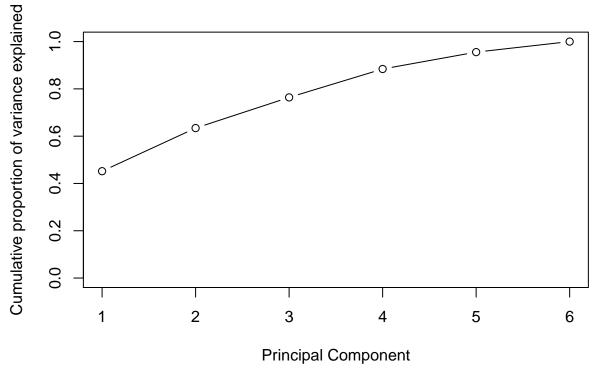
```
##Calculating variability of each component
pr.var <- pok.pr$sdev ^2

#Variance explained by each PC
pve <- pr.var/sum(pr.var)

# Plotting variance explained by each component
par(mfrow=c(1,2))
plot(pve,xlab="Principal Component",ylab="Proportion of variance explained",ylim=c(0,1),type="b")</pre>
```



plot(cumsum(pve),xlab="Principal Component" ,ylab="Cumulative proportion of variance explained",ylim=c(

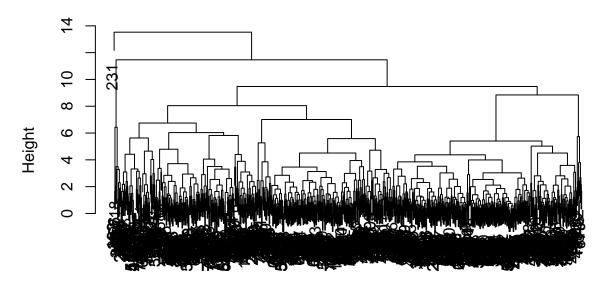


```
#Performing hierarchical clustering using euclidean distance
scaled_data <- scale(Pokemon_data)

Pokemon_hclust<- hclust(dist(scaled_data),method="complete")

plot(Pokemon_hclust)</pre>
```

## **Cluster Dendrogram**

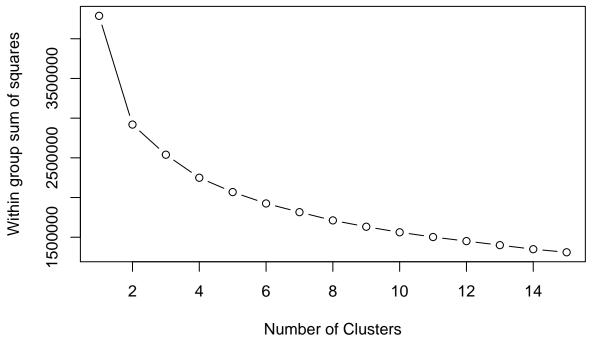


dist(scaled\_data)
hclust (\*, "complete")

```
# selecting number of clusters
Pokemon_hclust_clusters <- cutree(Pokemon_hclust,k=4)</pre>
```

```
#intialize wss
wss <- 0
for(i in 1:15){

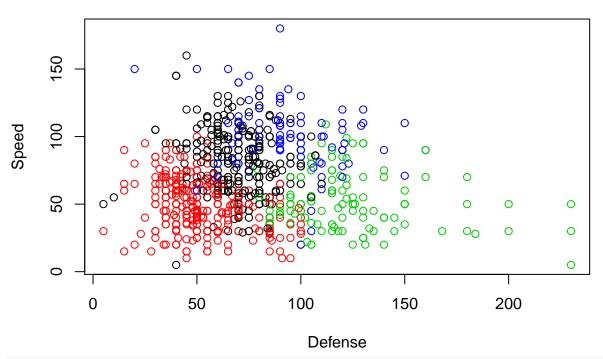
km.out<- kmeans(Pokemon_data,centers = i,nstart = 20,iter.max=50)
    wss[i] <- km.out$tot.withinss
}
plot(1:15,wss,type="b",xlab="Number of Clusters",ylab="Within group sum of squares")</pre>
```



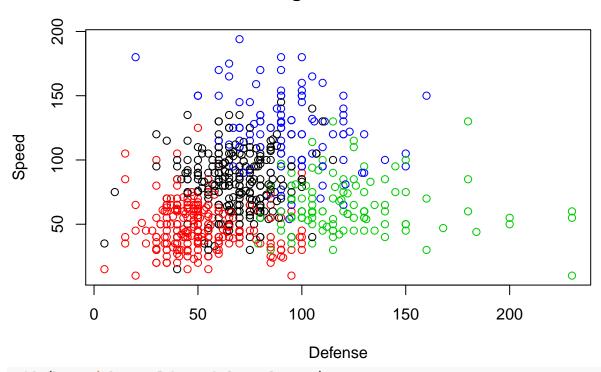
```
k <- 4
# Build model with k clusters: km.out
km.out <- kmeans(Pokemon_data, centers = k, nstart = 20, iter.max = 50)
km.out
## K-means clustering with 4 clusters of sizes 288, 283, 115, 114
##
## Cluster means:
##
                   ΗP
                              Attack
                                              Defense
                                                                Sp..Atk Sp..Def
                                                                                                      Speed
## 1 79.18056 81.31944 69.19097 82.01042 77.53125
                                                                                                80.10417
## 2 50.29682 54.03180
                                            51.62898
                                                             47.90459 49.15548
                                                                                                49.74912
## 3 71.30435 92.91304 121.42609
                                                              63.89565 88.23478 52.36522
## 4 89.20175 121.09649 92.73684 120.45614 97.67544 100.44737
##
## Clustering vector:
##
          [36] \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 
      [71] \ 1 \ 4 \ 2 \ 1 \ 3 \ 2 \ 1 \ 1 \ 2 \ 1 \ 2 \ 3 \ 3 \ 1 \ 1 \ 2 \ 3 \ 3 \ 2 \ 1 \ 2 \ 2 \ 1 \ 2 \ 1 \ 2 \ 3 \ 2 \ 1 \ 1 \ 4 \ 3 \ 2
## [106] 1 2 3 2 1 2 1 2 3 1 3 2 2 3 2 3 1 1 1 4 2 1 2 1 2 1 1 1 1 1 1 1 3 4 1 2
## [141] 1 4 1 2 2 1 1 4 1 2 3 2 3 1 4 1 4 4 4 2 1 4 4 4 4 2 1 1 2 1 1 2 1 1
## [246] 3 2 1 4 1 2 3 1 1 2 2 3 2 2 2 1 1 4 4 3 2 1 4 4 4 4 4 2 1 1 4 2 1 4 4
## [316] 1 2 2 2 1 2 1 2 3 2 2 2 3 2 3 2 3 3 3 2 1 1 2 1 4 1 1 1 1 1 1 2 1 2 1 4
## [386] 2 1 4 2 3 1 1 1 4 2 2 1 4 2 1 1 2 3 3 3 2 2 3 4 4 2 3 3 4 3 3 3 4 4 4
## [491] 3 2 1 4 4 1 2 4 4 2 3 2 3 2 1 1 2 1 2 2 1 4 1 3 3 3 3 4 4 1 1 3 1 3 1
 \#\# \ [561] \ 2 \ 1 \ 1 \ 2 \ 1 \ 2 \ 3 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 1 \ 2 \ 3 \ 3 \ 2 \ 1 \ 2 \ 1 \ 3 \ 2 \ 3 \ 3
```

```
## [666] 1 2 1 2 2 1 2 1 4 2 1 1 2 1 1 2 1 3 2 3 2 3 3 2 1 2 3 1 3 2 1 4 2 1 4
## [736] 2 1 2 2 1 2 1 2 1 1 2 1 1 2 3 4 3 2 1 2 1 2 1 2 3 2 3 2 1 2 1 2 3 2 3
## [771] 1 1 1 3 2 1 4 1 2 1 2 2 2 2 3 3 3 3 2 3 2 1 4 4 4 3 4 4 4 4
## Within cluster sum of squares by cluster:
## [1] 871531.3 513476.6 455965.7 408271.9
  (between_SS / total_SS = 47.6 %)
##
## Available components:
## [1] "cluster"
                "centers"
                            "totss"
                                       "withinss"
## [5] "tot.withinss" "betweenss"
                            "size"
                                       "iter"
## [9] "ifault"
# Plot of Defense vs. Speed by cluster membership
plot(Pokemon_data[, c("Defense", "Speed")],
   col = km.out$cluster,
   main = paste("k-means clustering of Pokemon with", k, "clusters"),
   xlab = "Defense", ylab = "Speed")
```

### k-means clustering of Pokemon with 4 clusters



#### k-means clustering of Pokemon with 4 clusters



#### table(km.out\$cluster,Pokemon\_hclust\_clusters)

```
##
      Pokemon_hclust_clusters
##
##
        32 245
                 11
                      0
##
                      0
        10 104
                  0
##
                      1
##
         5 109
# Create a hierarchical clustering model: wisc.pr.hclust
pokemon.pr.hclust <- hclust(dist(pok.pr$x[, 1:6]), method = "complete")</pre>
# Cut model into 4 clusters: wisc.pr.hclust.clusters
pokemon.pr.hclust.clusters <- cutree(pokemon.pr.hclust, k = 4)</pre>
```