

# Exp 6

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Implementation and analysis of clustering algorithms like

1. K-Means

2. Agglomerative

**\*\* K Means Clustering\*\*** setwd("E:/R Orientation")

**head**(iris)

## Sepal.Length Sepal.Width Petal.Length Petal.Width Species ## 1 5.1 3.5 1.4 0.2 setosa ## 2 4.9 3.0 1.4 0.2 setosa  
## 3 4.7 3.2 1.3 0.2 setosa ## 4 4.6 3.1 1.5 0.2 setosa ## 5 5.0 3.6 1.4 0.2 setosa ## 6 5.4 3.9 1.7 0.4 setosa

**library**(ggplot2)

**ggplot**(iris, **aes**(Petal.Length, Petal.Width, **color** = Species)) + **geom\_point**() 1

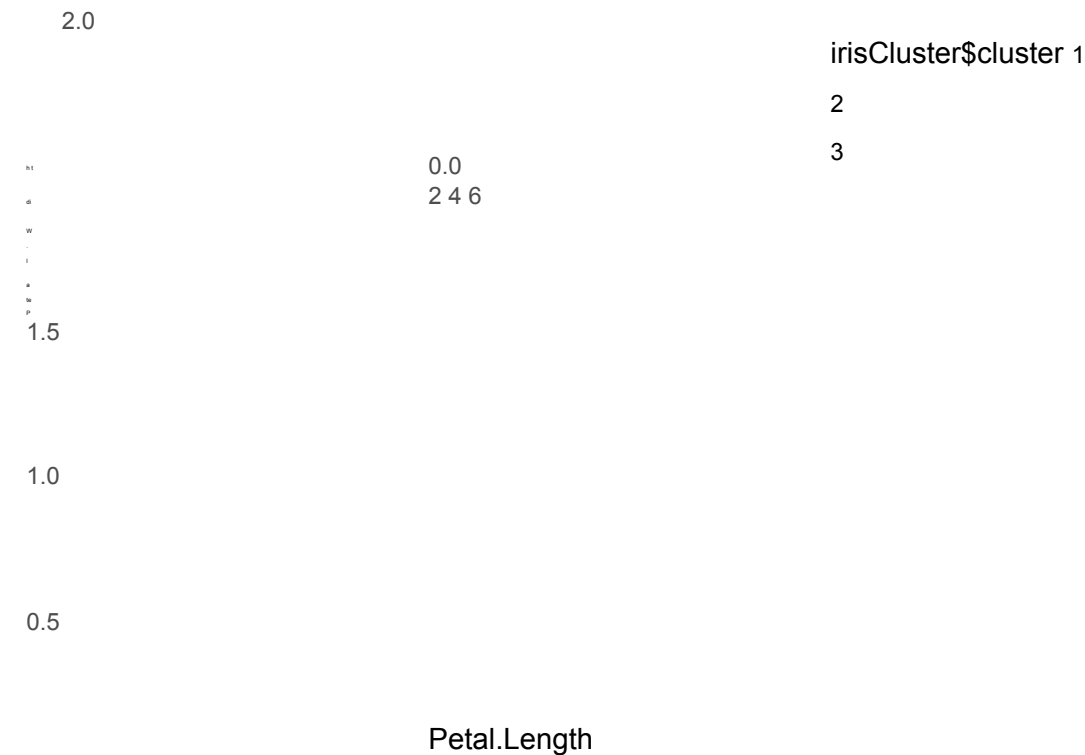


```
set.seed(20)
irisCluster <- kmeans(iris[, 3:4], 3, nstart = 20)
irisCluster

## K-means clustering with 3 clusters of sizes 52, 48, 50
##
## Cluster means:
## Petal.Length Petal.Width
## 1 4.269231 1.342308
## 2 5.595833 2.037500
```

2

```
##  
## setosa versicolor virginica  
## 1 0 48 4  
## 2 0 2 46  
## 3 50 0 0  
  
irisCluster$cluster <- as.factor(irisCluster$cluster)  
ggplot(iris, aes(Petal.Length, Petal.Width, color = irisCluster$cluster)) + geom_point()
```



### **#Agglomerative Clustering**

```
{ r Agglomerative} head(iris) clusters <- hclust(dist(iris[, 3:4])) plot(clusters) clusterCut <- cutree(clusters, 3)
table(clusterCut, iris$Species) clusters <- hclust(dist(iris[, 3:4]), method = 'average') plot(clusters) clusterCut <-
cutree(clusters, 3) table(clusterCut, iris$Species) ggplot(iris, aes(Petal.Length, Petal.Width, color = iris$Species)) +
geom_point(alpha = 0.4, size = 3.5) + geom_point(col = clusterCut) + scale_color_manual(values = c('black', 'red',
'green'))
```