# K-Means and K-Medoids Clustering on Simulated Data

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Optional: Can you try to determine the shape of a data point based on its real class label and color based on its cluster membership? I thought I could kind of determine the shape of a data point based on its real class label and its color based on its cluster membership, but I am uncertain based on the two plots fro k-means and k-medoids where the smaller clusters and colors don't quite seem to follow the same pattern. I thin I'll have to observe more data.

#### Introduction

In this analysis, we simulate data to apply both **K-Means** and **K-Medoids** clustering algorithms, compare their results, and visualize the clustering outcomes.

We also measure the time taken for each clustering method.

#### Load Required Libraries

```
# Install libraries if not already installed
# install.packages("cluster")
# install.packages("factoextra")
# install.packages("ggplot2")
# install.packages("tictoc")

# Load necessary libraries
library(cluster)  # For K-Medoids (PAM)
library(factoextra)  # For visualization
library(ggplot2)  # For plotting
library(tictoc)  # For measuring execution time
```

```
# Set random seed for reproducibility
set.seed(42)

# Create 3 clusters with normal distribution
# Change n from 300 to 1500 in increments of 200
n <- 1500 # Total number of data points
cluster1 <- data.frame(x = rnorm(n/3, mean = 2, sd = 0.5), y = rnorm(n/3, mean = 2, sd = 0.5))
cluster2 <- data.frame(x = rnorm(n/3, mean = 6, sd = 0.5), y = rnorm(n/3, mean = 6, sd = 0.5))
cluster3 <- data.frame(x = rnorm(n/3, mean = 10, sd = 0.5), y = rnorm(n/3, mean = 2, sd = 0.5))

# Combine clusters into one dataset
data <- rbind(cluster1, cluster2, cluster3)
colnames(data) <- c("feature1", "feature2")</pre>
```

```
# Normalize the data
scaled_data <- scale(data)</pre>
```

# Apply K-Means Clustering

```
tic("K-Means") # Start timer
kmeans_result <- kmeans(scaled_data, centers = 6, nstart = 10)
toc() # Stop timer</pre>
```

## K-Means: 0 sec elapsed

# Apply K-Medoids Clustering

```
tic("K-Medoids") # Start timer
kmedoids_result <- pam(scaled_data, 5)
toc() # Stop timer</pre>
```

## K-Medoids: 0.65 sec elapsed

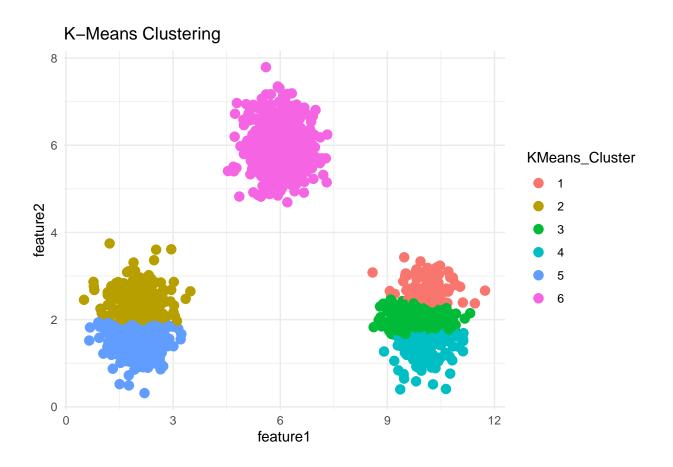
# Prep data for visualization

```
data$KMeans_Cluster <- as.factor(kmeans_result$cluster)
data$KMedoids_Cluster <- as.factor(kmedoids_result$clustering)</pre>
```

# Plotting k-means clusters

```
p1 <- ggplot(data, aes(x = feature1, y = feature2, color = KMeans_Cluster)) +
    geom_point(size = 3) +
    ggtitle("K-Means Clustering") +
    theme_minimal()

# Print the K-Means plot
print(p1)</pre>
```

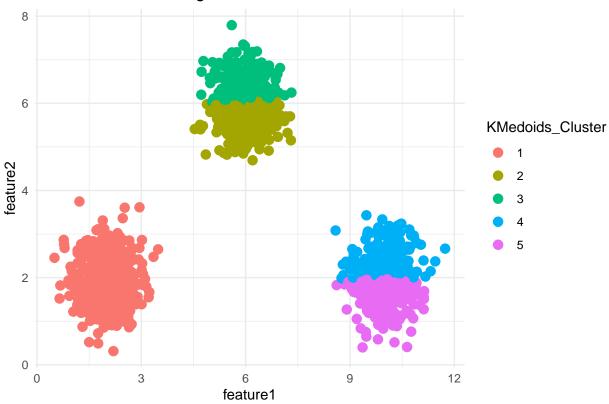


# Plotting k-medoids clusters

```
p2 <- ggplot(data, aes(x = feature1, y = feature2, color = KMedoids_Cluster)) +
   geom_point(size = 3) +
   ggtitle("K-Medoids Clustering") + theme_minimal()

# Print the K-Means plot
print(p2)</pre>
```

### K-Medoids Clustering



# # Load the Iris dataset library(datasets) data(iris) summary(iris)

```
Sepal.Width
    Sepal.Length
                                   Petal.Length
                                                   Petal.Width
##
##
   Min.
          :4.300
                   Min.
                        :2.000
                                   Min. :1.000
                                                         :0.100
                                                  Min.
   1st Qu.:5.100
                   1st Qu.:2.800
                                   1st Qu.:1.600
                                                  1st Qu.:0.300
  Median :5.800
                   Median :3.000
                                   Median :4.350
                                                  Median :1.300
##
   Mean
         :5.843
                   Mean :3.057
                                   Mean :3.758
                                                  Mean :1.199
##
   3rd Qu.:6.400
                   3rd Qu.:3.300
                                   3rd Qu.:5.100
                                                  3rd Qu.:1.800
          :7.900
                   Max. :4.400
##
   Max.
                                   Max. :6.900
                                                  Max. :2.500
##
         Species
##
   setosa
             :50
##
   versicolor:50
   virginica:50
##
##
##
```

```
df <- iris
scaled_data <-df</pre>
```

# I'm afraid from this point on I did something wrong and didn't know how to incorporate the Iris data

## Apply K-Means Clustering for Iris

```
# Apply k-means clustering with k = 3 (for the three species of Iris)
set.seed(20) # for reproducibility
iris_cluster <- kmeans(iris[, 1:4], centers = 3)

# Add the cluster assignments to the iris dataset
iris$Cluster <- as.factor(iris_cluster$cluster)</pre>
```

## Apply K-Medoids Clustering for Iris

```
tic("K-Medoids") # Start timer
kmedoids_result <- pam(scaled_data, 3)
toc() # Stop timer</pre>
```

## K-Medoids: 0 sec elapsed

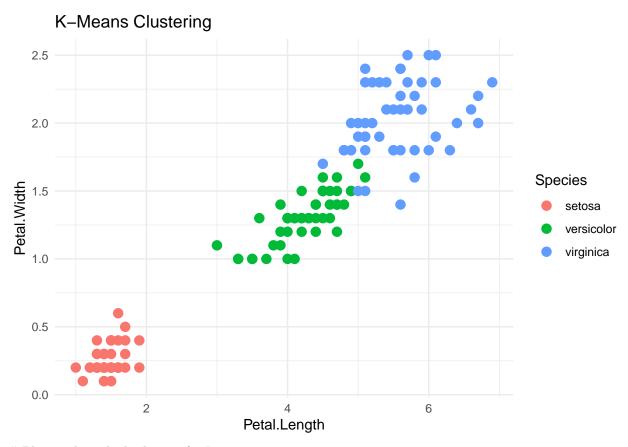
# Prep data for visualization for Iris

```
data$KMeans_Cluster <- as.factor(kmeans_result$cluster)
data$KMedoids_Cluster <- as.factor(kmedoids_result$clustering)</pre>
```

## Plotting k-means clusters fro Iris

```
p1 <- ggplot(scaled_data, aes(x = Petal.Length, y = Petal.Width, color = Species)) +
    geom_point(size = 3) +
    ggtitle("K-Means Clustering") +
    theme_minimal()

# Print the K-Means plot
print(p1)</pre>
```



# Plotting k-medoids clusters for Iris

```
p2 <- ggplot(scaled_data, aes(x = Petal.Length, y = Petal.Width, color = Species)) +
    geom_point(size = 3) +
    ggtitle("K-Medoids Clustering") +
    theme_minimal()

# Print the K-Medoids plot
print(p2)</pre>
```

