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We will use the IRIS dataset and perform K-means on it

•	Sepal.Length [‡]	Sepal.Width [‡]	Petal.Length [‡]	Petal.Width [‡]
1	5.1	3.5	1.4	0.2
2	4.9	3.0	1.4	0.2
3	4.7	3.2	1.3	0.2
4	4.6	3.1	1.5	0.2
5	5.0	3.6	1.4	0.2
6	5.4	3.9	1.7	0.4
7	4.6	3.4	1.4	0.3
8	5.0	3.4	1.5	0.2
9	4.4	2.9	1.4	0.2
10	4.9	3.1	1.5	0.1
11	5.4	3.7	1.5	0.2
12	4.8	3.4	1.6	0.2
13	4.8	3.0	1.4	0.1
14	4.3	3.0	1.1	0.1
15	5.8	4.0	1.2	0.2
16	5.7	4.4	1.5	0.4

Code:

```
# Varun Sudhir 21BDS0040

euclidean_distance <- function(point1, point2) {
   sqrt(sum((point1 - point2)^2))
}

assign_clusters <- function(data, centroids) {
   clusters <- sapply(1:nrow(data), function(i) {
      distances <- sapply(1:nrow(centroids), function(j)
   euclidean_distance(data[i,], centroids[j,]))
      return(which.min(distances))
   })
   return(clusters)
}

update_centroids <- function(data, clusters, k) {</pre>
```

```
new_centroids <- matrix(NA, nrow = k, ncol = ncol(data))</pre>
  for (i in 1:k) {
    cluster points <- data[clusters == i, ]</pre>
    if (nrow(cluster_points) > 0) {
      new centroids[i, ] <- colMeans(cluster points)</pre>
    } else {
      new_centroids[i, ] <- data[sample(1:nrow(data), 1), ]</pre>
    }
  }
  return(new_centroids)
}
# K-means function
k_means <- function(data, k, max_iter = 100) {</pre>
  centroids <- data[sample(1:nrow(data), k), ]</pre>
  for (i in 1:max iter) {
    clusters <- assign_clusters(data, centroids)</pre>
    new_centroids <- update_centroids(data, clusters, k)</pre>
    if (all(centroids == new_centroids)) {
      cat("Convergence reached at iteration", i, "\n")
      break
    }
    centroids <- new_centroids</pre>
  list(clusters = clusters, centroids = centroids)
}
display_clusters <- function(data, clusters, centroids) {</pre>
  k <- nrow(centroids)</pre>
  for (i in 1:k) {
    cat("\nCluster", i, "\n")
    cluster_points <- data[clusters == i, ]</pre>
    cat("Centroid:", centroids[i, ], "\n")
    cat("Number of points:", nrow(cluster_points), "\n")
    cat("Variance of features:\n")
    print(apply(cluster_points, 2, var))
    cat("Minimum values of features:\n")
    print(apply(cluster_points, 2, min))
    cat("Maximum values of features:\n")
    print(apply(cluster_points, 2, max))
    distances <- apply(cluster_points, 1, function(point)</pre>
euclidean_distance(point, centroids[i, ]))
    cat("Average distance from centroid:", mean(distances), "\n")
  }
}
```

```
# Function to plot clusters and centroids
plot_clusters <- function(data, clusters, centroids) {</pre>
  data_df <- as.data.frame(data)</pre>
  data df$cluster <- as.factor(clusters)</pre>
  colnames(data_df)[1:2] <- c("Feature1", "Feature2")</pre>
  centroid_df <- as.data.frame(centroids[, 1:2])</pre>
  centroid_df$Cluster <- as.factor(1:nrow(centroid_df))</pre>
  colnames(centroid_df)[1:2] <- c("Feature1", "Feature2")</pre>
  ggplot(data_df, aes(x = Feature1, y = Feature2, color = cluster)) +
    geom_point(size = 2) +
    geom_point(data = centroid_df, aes(x = Feature1, y = Feature2), color =
"black", shape = 4, size = 5) +
    labs(title = "K-Means Clustering", x = "Feature 1", y = "Feature 2") +
    theme minimal()
}
run_k_means <- function() {</pre>
  # iris dataset (4 features: Sepal.Length, Sepal.Width, Petal.Length,
Petal.Width)
  data <- as.matrix(iris[, 1:4])</pre>
  data <- scale(data)</pre>
  cat("Enter the number of clusters (k): ")
  k <- as.integer(readline())</pre>
  result <- k_means(data, k)
  display_clusters(data, result$clusters, result$centroids)
  plot_clusters(data[, 1:2], result$clusters, result$centroids[, 1:2])
}
run_k_means()
cat("Varun Sudhir 21BDS0040")
```

Output:

```
> run_k_means()
Enter the number of clusters (k):
Convergence reached at iteration 4
Cluster 1
Centroid: 1.127627 0.07877034 0.9820815 0.9957524
Number of points: 48
Variance of features:
Sepal.Length Sepal.Width Petal.Length Petal.Width
   0.3445740
               0.3580082
                            0.1333879
                                         0.2005569
Minimum values of features:
Sepal.Length Sepal.Width Petal.Length Petal.Width
  0.06843254 -1.27867961
                           0.36367793 0.13206729
Maximum values of features:
Sepal.Length Sepal.Width Petal.Length Petal.Width
    2.483699
                 1.703886
                             1.779869
                                          1.706379
Average distance from centroid: 0.9139393
Cluster 2
Centroid: -0.06858736 -0.8904166 0.3440691 0.2834435
Number of points: 52
Variance of features:
Sepal.Length Sepal.Width Petal.Length Petal.Width
   0.2259220
               0.3417057
                            0.1024204
                                         0.1696749
Minimum values of features:
Sepal.Length Sepal.Width Petal.Length Petal.Width
  -1.1392005
              -2.4258204
                           -0.4293892 -0.2615107
Maximum values of features:
Sepal.Length Sepal.Width Petal.Length Petal.Width
   0.7930124
              -0.1315388
                            1.0434497
                                         1.5751867
Average distance from centroid: 0.8507722
Cluster 3
Centroid: -1.011191 0.8504137 -1.30063 -1.250704
Number of points: 50
Variance of features:
Sepal.Length Sepal.Width Petal.Length Petal.Width
             0.756344014 0.009677951 0.019115323
 0.181201918
Minimum values of features:
Sepal.Length Sepal.Width Petal.Length Petal.Width
   -1.863780
               -1.737536
                             -1.562342
                                          -1.442245
Maximum values of features:
Sepal.Length Sepal.Width Petal.Length Petal.Width
               3.08045544 -1.05251337 -0.78628144
 -0.05233076
Average distance from centroid: 0.8109485
```

