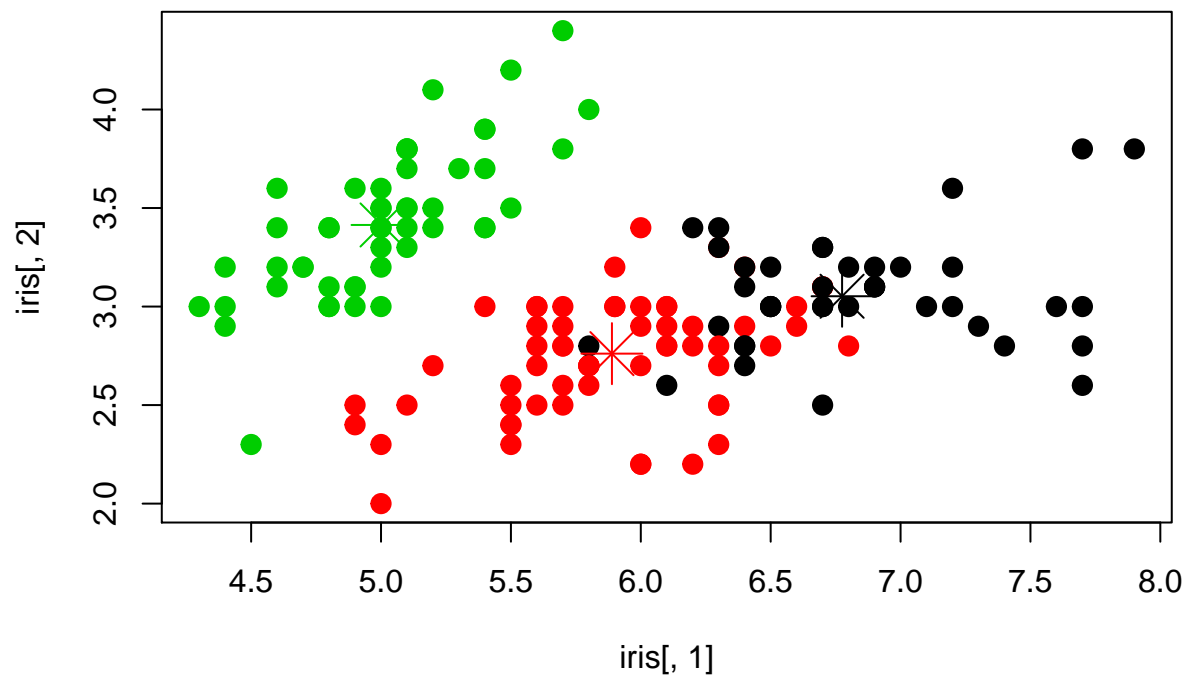


Fuzzy C-means Clustering of Iris Dataset

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```
# Unlike K-Means where each data point belongs to only one cluster,  
# in fuzzy cmeans, each data point has a fraction of membership to each cluster.  
# The goal is to figure out the membership fraction that minimize the  
# expected distance to each centroid. This method utilizes an overlapping  
# clustering algorithm that was developed by Dunn in 1973 and improved  
# by Bezdek in 1981. It is frequently used in pattern recognition.  
#  
# Advantages:  
# 1. Gives best result for overlapped data set and comparatively better than  
#   k-means algorithm.  
# 2. Unlike k-means where data point must exclusively belong to one cluster  
#   center here data point is assigned membership to each cluster center as  
#   a result of which data point may belong to more than one cluster center.  
# Disadvantages:  
# 1. Apriori specification of the number of clusters.  
# 2. With lower value of beta we get the better result but at the expense of more  
#   number of iteration.  
# 3. Euclidean distance measures can unequally weight underlying factors.  
#  
# J. C. Dunn (1973): "A Fuzzy Relative of the ISODATA Process and Its Use in  
#   Detecting Compact Well-Separated Clusters", Journal of Cybernetics 3: 32-57  
# J. C. Bezdek (1981): "Pattern Recognition with Fuzzy Objective Function Algorithms",  
#   Plenum Press, New York  
  
library(e1071)  
  
result <- cmeans(iris[,-5], 3, 100, m=2, method="cmeans")  
plot(iris[,1], iris[,2], col=result$cluster,pch=20,cex=2)  
points(result$centers[,c(1,2)], col=1:3, pch=8, cex=3)
```



```
# the visual output is very similar to K-Means
result$membership[1:3,]
```

```
##           1           2           3
## [1,] 0.001072 0.002304 0.9966
## [2,] 0.007498 0.016651 0.9759
## [3,] 0.006415 0.013760 0.9798
```

```
result$centers
```

```
## Sepal.Length Sepal.Width Petal.Length Petal.Width
## 1          6.775         3.052         5.647         2.0535
## 2          5.889         2.761         4.364         1.3973
## 3          5.004         3.414         1.483         0.2535
```