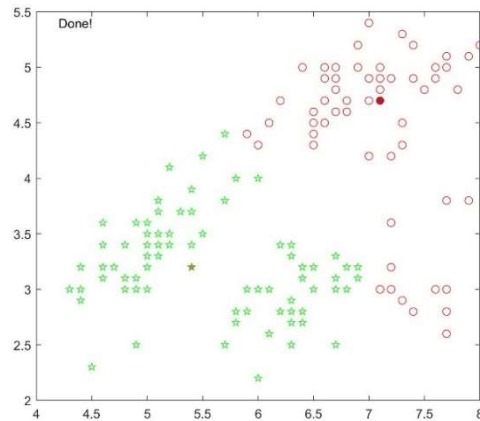


## ELEC 425 – Assignment 2

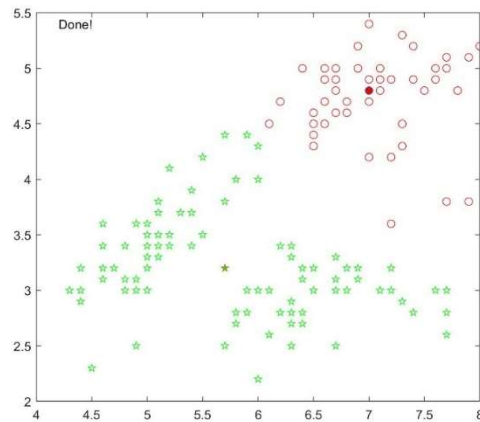
### 1 – Implement K-medians

2 clusters

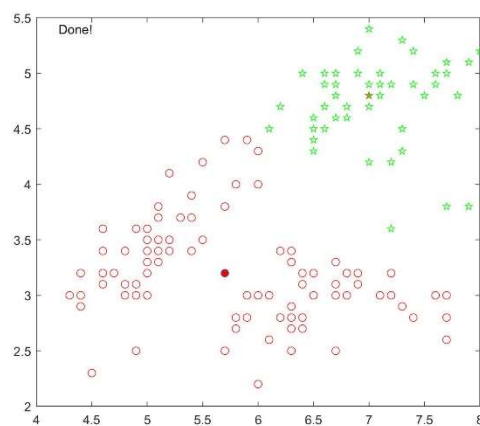
Cluster centres = [5.5, 4; 4.5, 3.2]



Cluster centres = [3.1, 5.2; 3.8, 4.2;]

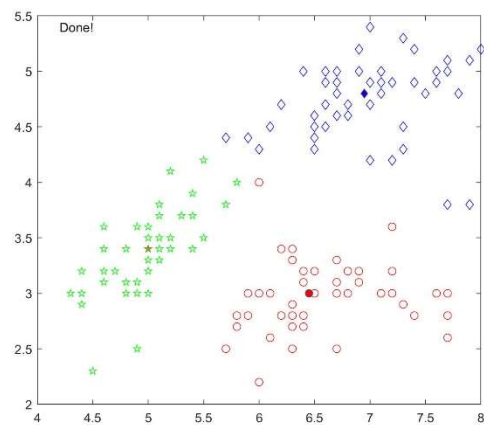


Cluster centres = [3, 3; 6, 6]

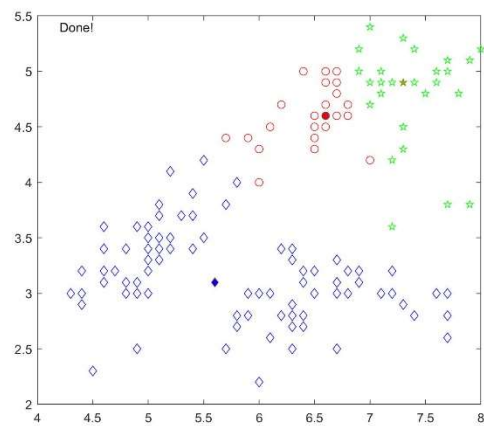


3 clusters

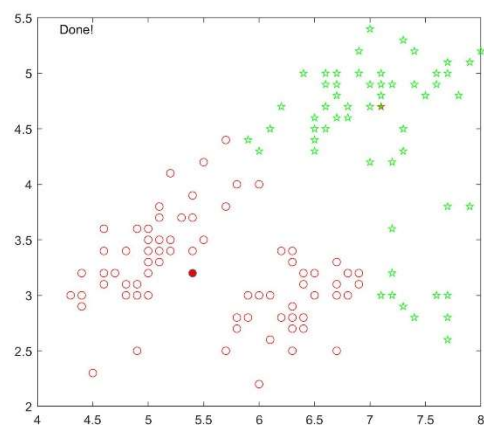
Cluster centres =  $[3, 3; 4, 4; 6, 6]$



Cluster centres =  $[3.5, 4.3; 3.9, 5.6; 3.4, 3.8]$

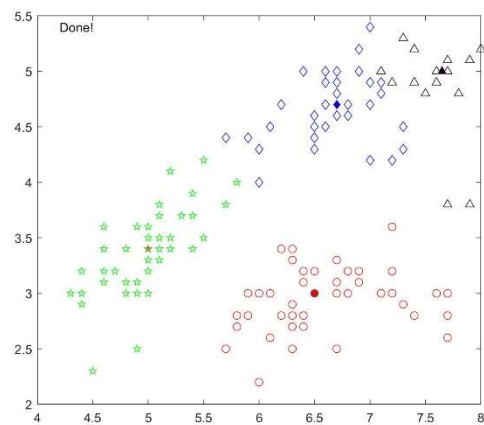


Cluster centres =  $[4, 5; 4.8, 5.2; 3.9, 6]$

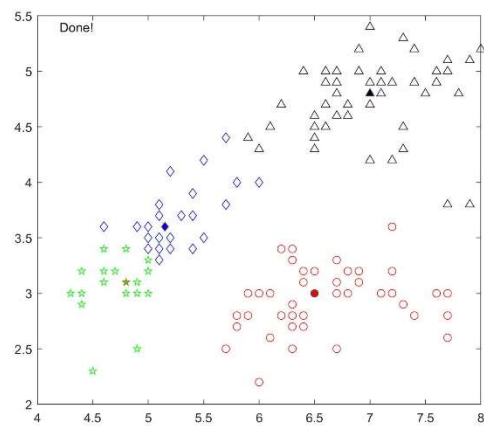


4 clusters

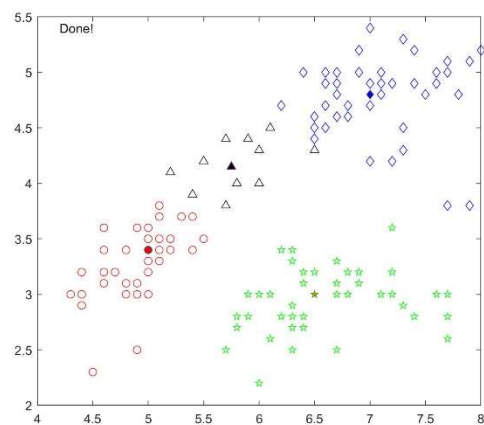
Cluster centres =  $[3, 3; 4, 4; 5, 5; 6, 6]$



Cluster centres =  $[6, 4; 4, 3; 5, 4; 6, 5]$



Cluster centres =  $[4, 5; 8, 3; 7.6, 5; 8, 3.5]$



The relevant code can be found in `k_medians.m` (code was too large to include in this document).

## 2 – Prove the EM Updating Algorithm Used in K-medians

Similar to the proof that K-means minimizes the following loss function:

$$J = \sum_{n=1}^N \sum_{k=1}^K r_{nk} \|x_n - \mu_k\|^2$$

The K-median algorithm minimizes the following loss function:

$$J = \sum_{n=1}^N \sum_{k=1}^K r_{nk} |x_n - \mu_k|$$

The terms involving  $n$  are independent, so each  $n$  can be optimized for separately by choosing  $r_{nk}$  to be 1 for whichever value of  $k$  gives the minimum value of  $|x_n - \mu_k|$ :

$$r_{nk} = \begin{cases} 1 & \text{if } k = \operatorname{argmin}_j |x_n - \mu_j| \\ 0 & \text{otherwise} \end{cases}$$

To optimize  $\mu_k$  with  $r_{nk}$  fixed, take the derivative with respect to  $\mu_k$  and set it to 0:

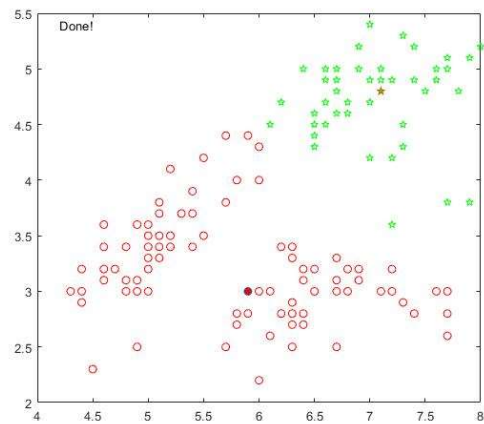
$$\begin{aligned} \frac{dJ}{d\mu_k} &= \sum_{n=1}^N \sum_{k=1}^K r_{nk} * \operatorname{sign}[\mu_k - x_n] \\ 0 &= \sum_{n=1}^N \sum_{k=1}^K r_{nk} * \operatorname{sign}[\mu_k - x_n] \\ &= (\#x_n \text{ smaller than } \mu_k) - (\#x_n \text{ bigger than } \mu_k) \\ \hat{\mu}_k &= \operatorname{median}[x_1, \dots, x_n] \end{aligned}$$

Thus, we can see that to optimize the K-median loss function, we reach the conclusion that setting  $\mu_k$  to the median of all points assigned to cluster  $k$  optimizes this loss function. Therefore, it is proven that the K-medians algorithm minimizes that error function.

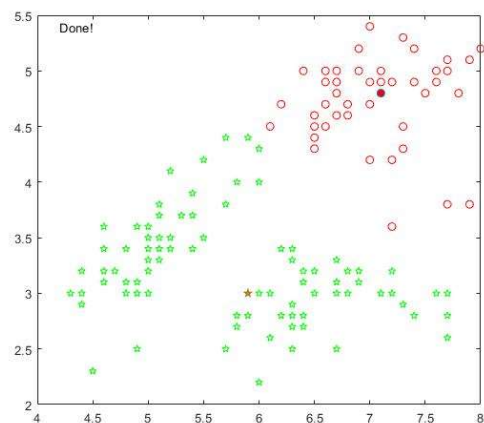
### 3 - K-medoids

2 clusters

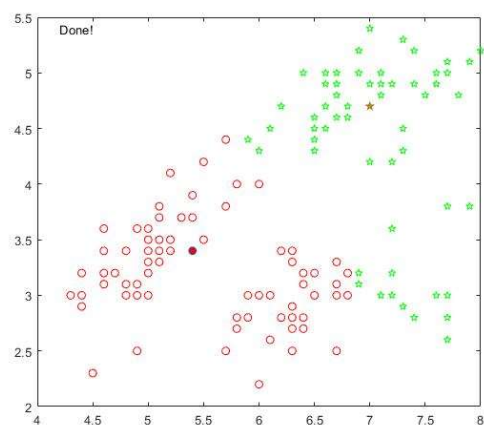
*Cluster Centres = [5.9, 3.0; 7.1, 4.8]*



*Cluster Centres = [7.1, 4.8; 5.9, 3.0]*

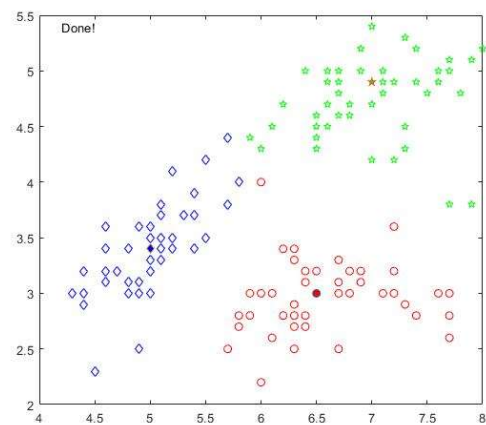


*Cluster Centres = [5.4, 3.4; 7.0, 4.7]*

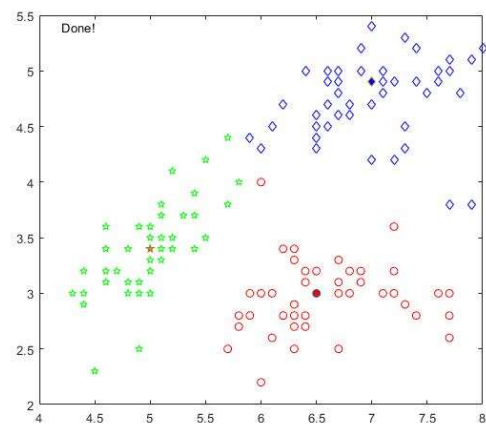


3 clusters

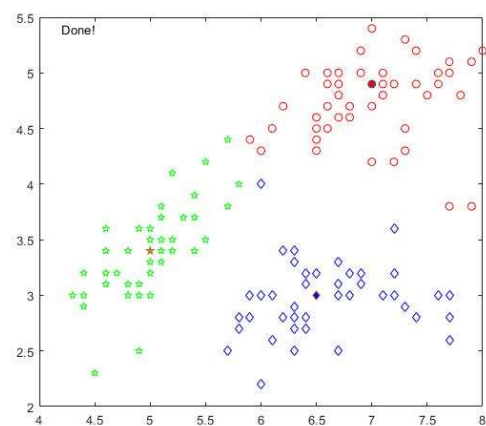
*Cluster Centres = [6.5, 3.0; 7.0, 4.9; 5.0, 3.4]*



*Cluster Centres = [6.5, 3.0; 5.0, 3.4; 7.0, 4.9]*

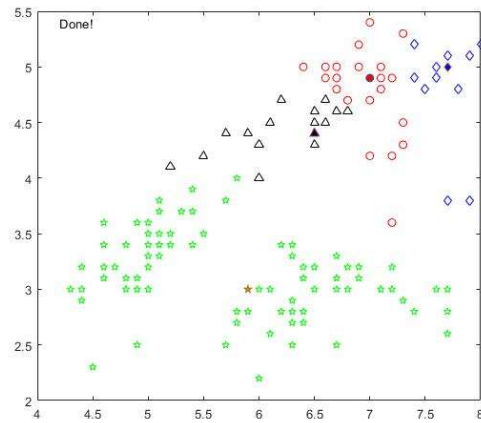


*Cluster Centres = [7.0, 4.9; 5.0, 3.4; 6.5, 3.0]*

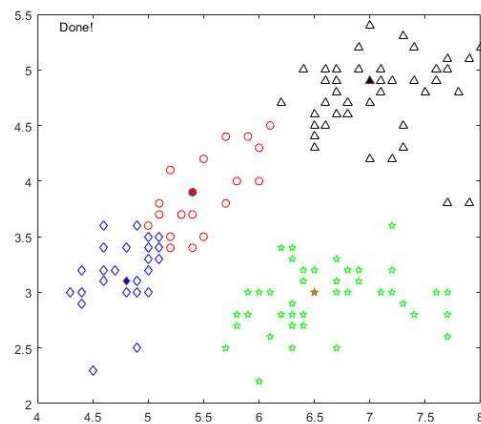


4 clusters

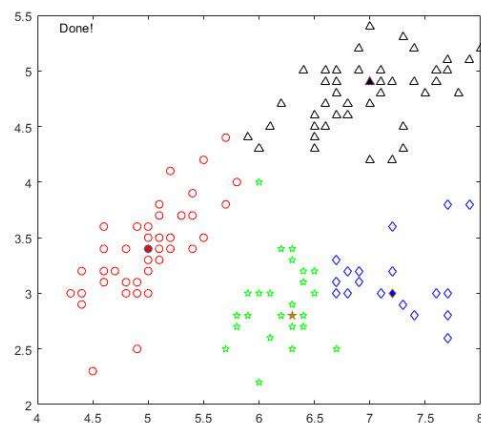
*Cluster Centres = [7.0, 4.9; 5.9, 3.0; 7.7, 5.0; 6.5, 4.4]*



*Cluster Centres = [5.4, 3.9; 6.5, 3.0; 4.8, 3.1; 7.0, 4.9]*



*Cluster Centres = [5.0, 3.4; 6.3, 2.8; 7.2, 3.0; 7.0, 4.9]*



The relevant code can be found in `k_medoids.m` (code was too large to include in this document).