In this homework, I implemented an expectation-maximization (EM) clustering algorithm.

Part 1

I read data from hw07_data_set which contains 300 data points generated randomly from five bivariate Gaussian densities.

Part 2

To initialize your EM algorithm, I took the centroids given in the file named hw07 initial centroids.csv as the initial values for the mean vectors.

```
#initialize centroids, memberships
centroids = np.genfromtxt("hw07_initial_centroids.csv", delimiter = ",")
memberships = update_memberships(centroids, X)
print(centroids)

[[-3.0439416    0.32509753]
[-0.30377397    3.73480678]
[ 1.17587258    -2.50984601]
[ 0.91037084    0.87608158]
[ -1.98588337    2.98466811]]
```

By assigning the data points to the nearest center, I estimated the initial covariance matrices and prior probabilities before applying EM algorithm.

Part 3

After the initialization step, I implemented EM algorithm.

E Step:

$$h_{ik} = E\left[z_{ik}|X,\Phi^{(t)}\right] = \frac{p(x_i|C_k,\Phi^{(t)})P(C_k)}{\sum_{c=1}^{K} p(x_i|C_c,\Phi^{(t)})P(C_c)}$$

M Step:

$$\hat{\mu}_{k}^{(t+1)} = \frac{\sum_{i=1}^{N} h_{ik} x_{i}}{\sum_{i=1}^{N} h_{ik}}$$

$$\hat{\sigma}_{k}^{(t+1)} = \frac{\sum_{i=1}^{N} h_{ik} (x_{i} - \hat{\mu}_{k}^{(t+1)}) (x_{i} - \hat{\mu}_{k}^{(t+1)})^{T}}{\sum_{i=1}^{N} h_{ik}}$$

$$\hat{P}(C_{K}) = \frac{\sum_{i=1}^{N} h_{ik}}{N}$$

For M Step I implemented three functions to update means, covariances and priors.

After defining functions, I run EM algorithm for 100 iterations. Then centroids which is mean vectors are found as follows:

Part 4

I drew the clustering result obtained by your EM algorithm by coloring each cluster with a different color. I also drew the original Gaussian densities used to generate data points with dashed lines and the Gaussian densities EM algorithm with solid lines. I used gaussian ellipse code which is taken from github that is referenced above function.

