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In [124]: import numpy as np
import matplotlib
import matplotlib.pyplot as plt
import numpy.matlib
import scipy
%matplotlib inline
```

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In [2]: def plotCurrent(X, Rnk, Kmus):
    N, D = np.shape(X)
    K = np.shape(Kmus)[0]

    InitColorMat = np.matrix([[1, 0, 0],
                               [0, 1, 0],
                               [0, 0, 1],
                               [0, 0, 0],
                               [1, 1, 0],
                               [1, 0, 1],
                               [0, 1, 1]])

    KColorMat = InitColorMat[0:K]
    colorVec = Rnk.dot(KColorMat)
    muColorVec = np.eye(K).dot(KColorMat)

    plt.scatter(X[:,0], X[:,1], edgecolors=colorVec, marker='o', facecolors='none')
    plt.scatter(Kmus[:,0], Kmus[:,1], c=muColorVec, marker='D', s=50);
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In [53]: def calcSqDistances(X, Kmus):
    return scipy.spatial.distance.cdist(X, Kmus, 'sqeuclidean')
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In [74]: def determineRnk(sqDmat):
    return (sqDmat == np.min(sqDmat,1).reshape(-1, 1)).astype(int)
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In [106]: def recalcMus(X, Rnk):
    return Rnk.T.dot(X)/np.sum(Rnk, 0).T.reshape(-1,1)
```

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In [111]: def runKMeans(K, fileString):
            # Load data file specified by fileString from Bishop book
            X = np.loadtxt(fileString)

            # Determine and store data set information
            N = np.shape(X)[0]
            D = np.shape(X)[1]

            # Allocate space for the K mu vectors
            Kmus = np.zeros((K, D))

            # Initialize cluster centers by randomly picking points from the data
            rndinds = np.random.permutation(N)
            Kmus = X[rndinds[:K]];

            # Specify the maximum number of iterations to allow
            maxiters = 1000;

            for iter in range(maxiters):
                # Assign each data vector to closest mu vector as per Bishop (9.2)
                # Do this by first calculating a squared distance matrix where the
                # contains the squared distance from the nth data vector to the kth

                # sqDmat will be an N-by-K matrix with the n,k entry as specified at
                sqDmat = calcSqDistances(X, Kmus);

                # given the matrix of squared distances, determine the closest cluster
                # center for each data vector

                # R is the "responsibility" matrix
                # R will be an N-by-K matrix of binary values whose n,k entry is set
                # per Bishop (9.2)
                # Specifically, the n,k entry is 1 if point n is closest to cluster
                # and is 0 otherwise
                Rnk = determineRnk(sqDmat)

                KmusOld = Kmus
                #plotCurrent(X, Rnk, Kmus)
                #plt.show()

                # Recalculate mu values based on cluster assignments as per Bishop
                Kmus = recalcMus(X, Rnk)

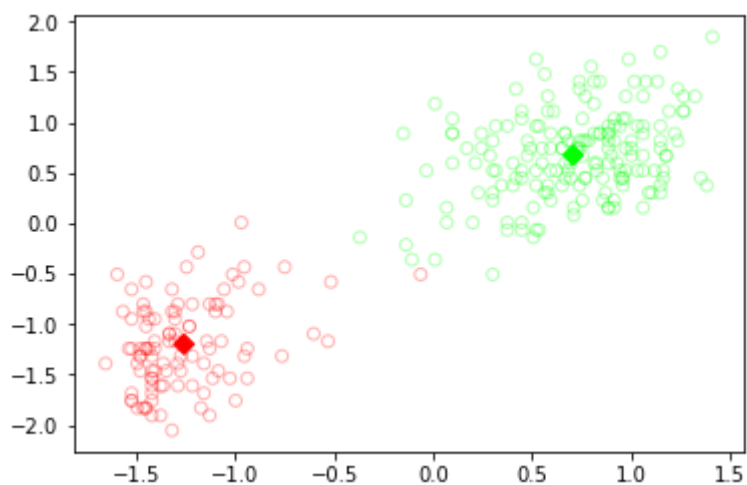
                # Check to see if the cluster centers have converged. If so, break
                if sum(abs(KmusOld.flatten() - Kmus.flatten())) < 1e-6:
                    break

            plotCurrent(X, Rnk, Kmus)

```

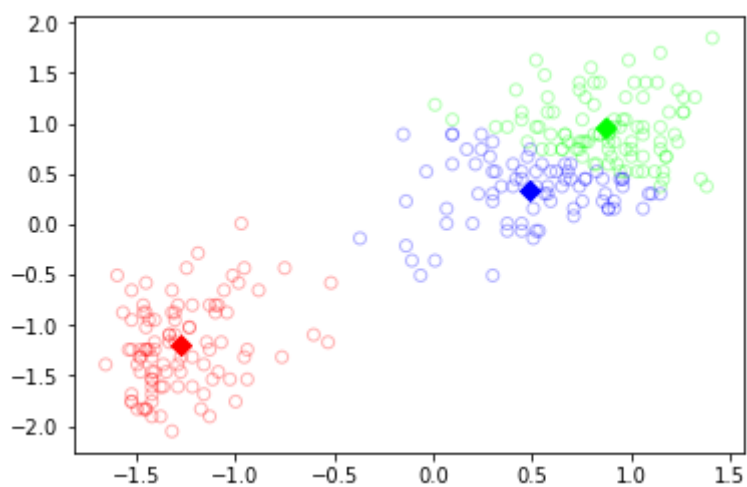
K = 2

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In [112]: runKMeans(2, 'scaledfaithful.txt')
```



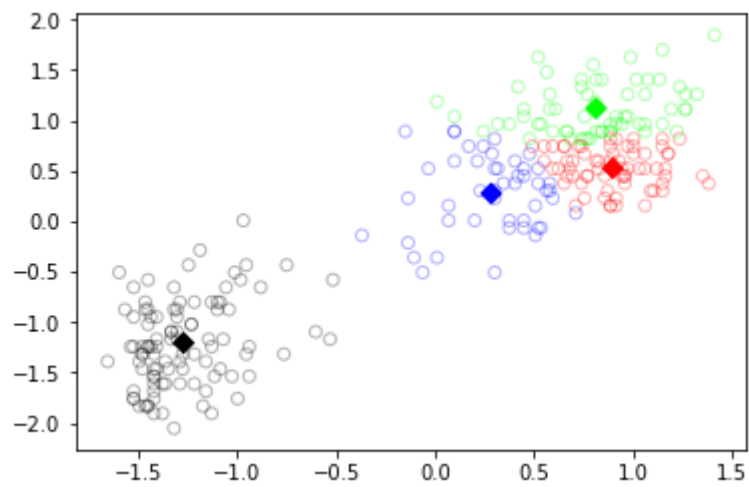
K = 3

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In [114]: runKMeans(3, 'scaledfaithful.txt')
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K = 4

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In [123]: runKMeans(4, 'scaledfaithful.txt')
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In [ ]:
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