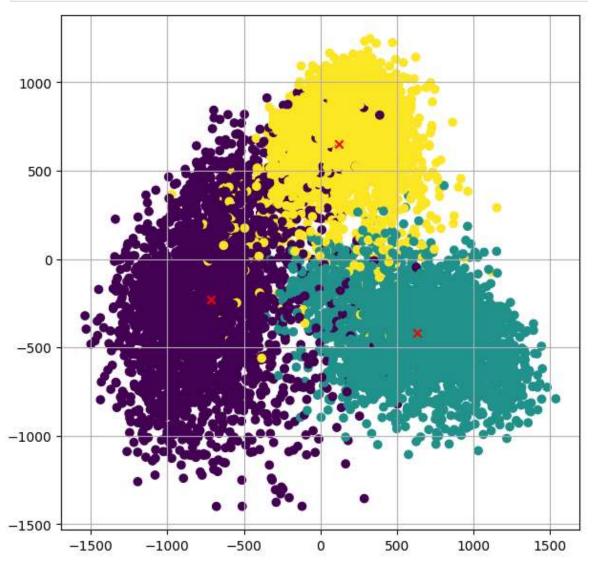
```
In [ ]: import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        data = pd.read_csv('data.csv')
        print(data.shape)
        print(data.head())
       (42000, 785)
          label pixel0 pixel1 pixel2 pixel3 pixel4 pixel5 pixel6 pixel7
              1
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       1
              0
                      0
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       2
              1
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                                                                                 0
       3
              4
                      0
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          pixel8 ... pixel774 pixel775 pixel776 pixel777 pixel778 pixel779
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                  . . .
       1
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       4
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          pixel780 pixel781 pixel782 pixel783
       0
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       1
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       2
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                                                0
                 0
       3
                 0
                            0
                                      0
                                                0
       4
                 0
       [5 rows x 785 columns]
In [ ]: data_array = np.array(data)
        print(data_array.shape)
        X = data_array[:, 1:].T
        Y = data_array[:, 0]
        print(X.shape, Y.shape)
       (42000, 785)
       (784, 42000) (42000,)
In [ ]: num1 = 3
        num2 = 6
        num3 = 9
        idx = (Y[:] == num1) | (Y[:] == num2) | (Y[:] == num3)
        X = X[:, idx]
        Y = Y[idx]
        print(X.shape, Y.shape)
       (784, 12676) (12676,)
In [ ]: def train_test_split(X,Y,N_train):
             d, N = X.shape
             #define the array of indices
```

```
idx = np.arange(0,N)
             #shuffle the indices
             np.random.shuffle(idx)
            #extract train and test indices
            train_idx = idx[:N_train]
            test_idx = idx[N_train:]
            #extract data
            X_train = X[:, train_idx]
            Y_train = Y[train_idx]
            X_test = X[:, test_idx]
            Y_{\text{test}} = Y[\text{test_idx}]
             return (X_train, Y_train), (X_test, Y_test)
In [ ]: N_train = 8000
        (X_train, Y_train), (X_test, Y_test) = train_test_split(X, Y, N_train)
        print(X train.shape, X test.shape)
        print(Y_train.shape, Y_test.shape)
       (784, 8000) (784, 4676)
       (8000,) (4676,)
In [ ]: def centroid(X):
             return np.mean(X, axis=1, keepdims=True)
        CX = centroid(X train)
        print(CX.shape)
        # Center the dataset
        Xc = X train - CX
        print(np.linalg.norm(centroid(Xc),2))
        # Compute PCA of X_{train} with k = 2
        U,s,VT = np.linalg.svd(Xc, full_matrices = False)
        k = 2
        U_k = U[:, :k]
        Z_k = U_k.T @ Xc
        print(Z_k.shape)
        idx1 = Y_train[:] == num1
        idx2 = Y_train[:] == num2
        idx3 = Y_train[:] == num3
        Xc1 = Xc[:, idx1]
        Xc2 = Xc[:, idx2]
        Xc3 = Xc[:, idx3]
        Z_k1 = Z_k[:, idx1]
        Z_k2 = Z_k[:, idx2]
        Z_k3 = Z_k[:, idx3]
        C_k1 = np.reshape(np.mean(Z_k1, axis=1),(2,1))
        C_k2 = np.reshape(np.mean(Z_k2, axis=1),(2,1))
        C k3 = np.reshape((np.mean(Z k3, axis=1)),(2,1))
```

```
(784, 1)
1.4817588995350117e-12
(2, 8000)
```

```
In [ ]: #visualize the points and the centroids
    plt.figure(figsize=(7, 7))
    plt.scatter(Z_k[0,:], Z_k[1,:], c = Y_train[:])
    plt.scatter(C_k1[0,:], C_k1[1,:], c = "r", marker='x')
    plt.scatter(C_k2[0,:], C_k2[1,:], c = "r", marker='x')
    plt.scatter(C_k3[0,:], C_k3[1,:], c = "r", marker='x')
    plt.grid()
    plt.show()
```



```
In []: ##### AVERAGE DISTANCE OF CLUSTERS TO CENTROID #####

def average_distance_to_centroid(Z_k, C_k):
    d, N = Z_k.shape
    dist = 0
    for i in range(N):
        dist += np.linalg.norm(Z_k[:, i] - C_k, 2)
    return dist/N

d1 = average_distance_to_centroid(Z_k1, C_k1)
    d2 = average_distance_to_centroid(Z_k2, C_k2)
    d3 = average_distance_to_centroid(Z_k3, C_k3)
    print('Average distance of cluster 1 to centroid 1: ', d1)
```

```
Average distance of cluster 1 to centroid 1: 910.3463987544986
       Average distance of cluster 2 to centroid 2: 1333.8120349290812
       Average distance of cluster 3 to centroid 3: 790.3925517798625
In [ ]: | ##### AVERAGE DISTANCE OF CLUSTERS TO CENTROID IN TEST SET #####
        def average_distance_to_centroid2(X, Y, C1, C2, C3):
            idx1 = (Y[:] == num1)
            idx2 = (Y[:] == num2)
            idx3 = (Y[:] == num3)
            X1 = X[:, idx1]
            X2 = X[:, idx2]
            X3 = X[:, idx3]
            avg_dist1 = average_distance_to_centroid(X1, C1)
            avg dist2 = average distance to centroid(X2, C2)
            avg_dist3 = average_distance_to_centroid(X3, C3)
            return avg_dist1, avg_dist2, avg_dist3
        U, s, VT = np.linalg.svd(X_test, full_matrices = False)
        k = 2
        U_k_{\text{test}} = U[:, :k]
        Z_k_test = U_k_test.T @ X_test
        avg_dist1, avg_dist2, avg_dist3 = average_distance_to_centroid2(Z_k_test, Y_test
        print("Average distance of test set to centroid 1:", avg_dist1)
        print("Average distance of test set to centroid 2:", avg_dist2)
        print("Average distance of test set to centroid 3:", avg_dist3)
       Average distance of test set to centroid 1: 1817.0260463719023
       Average distance of test set to centroid 2: 2665.400148489955
       Average distance of test set to centroid 3: 2816.15894485272
In [ ]: def centroid(X):
            return np.mean(X, axis=1)
        def classificationAlgo(C1, C2, C3, x):
            dist1 = np.linalg.norm(x - C1)
            dist2 = np.linalg.norm(x - C2)
            dist3 = np.linalg.norm(x - C3)
            minDist = min(dist1, dist2, dist3)
            if minDist == dist1:
                return num1
            if minDist == dist2:
                return num2
            if minDist == dist3:
                return num3
        counter = 0
        C1 = centroid(Xc1)
        C2 = centroid(Xc2)
        C3 = centroid(Xc3)
        for i in range(Y_test.shape[0]):
            c = classificationAlgo(C1, C2, C3, X_test[:, i])
            if c == Y test[i]:
                counter += 1
```

print('Average distance of cluster 2 to centroid 2: ', d2)
print('Average distance of cluster 3 to centroid 3: ', d3)

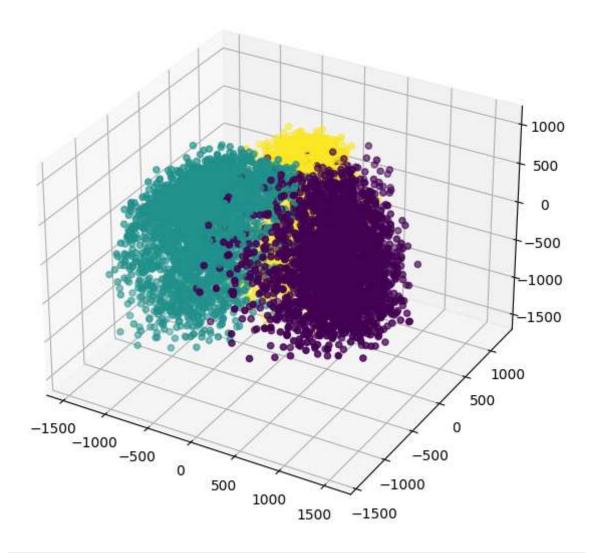
```
print(f"Accuracy: {counter*100/len(Y_test)} %")
```

Accuracy: 94.803250641574 %

```
In [ ]: k = 3
        def centroid(X):
            return np.mean(X, axis=1, keepdims=True)
        (X_train, Y_train), (X_test, Y_test) = train_test_split(X, Y, N_train)
        CX = centroid(X_train)
        Xc = X_{train} - CX
        U, s, VT = np.linalg.svd(Xc, full_matrices=False)
        idx1 = Y_train[:] == num1
        idx2 = Y_train[:] == num2
        idx3 = Y_train[:] == num3
        Xc1 = Xc[:, idx1]
        Xc2 = Xc[:, idx2]
        Xc3 = Xc[:, idx3]
        C1 = centroid(Xc1)
        C2 = centroid(Xc2)
        C3 = centroid(Xc3)
        U_k = U[:, :k]
        Z_k = U_k.T @ Xc
        Z_k1 = Z_k[:, idx1]
        Z_k2 = Z_k[:, idx2]
        Z_k3 = Z_k[:, idx3]
        C_k1 = np.reshape(np.mean(Z_k1, axis=1), (k,1))
        C_k2 = np.reshape(np.mean(Z_k2, axis=1), (k,1))
        C_k3 = np.reshape(np.mean(Z_k3, axis=1), (k,1))
        counter = 0
        for i in range(Y_test.shape[0]):
            c = classificationAlgo(C1, C2, C3, X_test[:, i])
            if c == Y_test[i]:
                counter += 1
        print(f"Accuracy: {counter*100/len(Y_test)} %")
```

Accuracy: 34.40975192472199 %

```
In [ ]: fig = plt.figure(figsize=(7, 7))
    ax = fig.add_subplot(projection='3d')
    ax.scatter(Z_k[0,:], Z_k[1,:], Z_k[2,:], c = Y_train[:])
    plt.show()
```



```
In [ ]: k = 5
        def centroid(X):
             return np.mean(X, axis=1, keepdims=True)
         (X_train, Y_train), (X_test, Y_test) = train_test_split(X, Y, N_train)
        CX = centroid(X_train)
        Xc = X_{train} - CX
        U,s,VT = np.linalg.svd(Xc, full_matrices=False)
        idx1 = Y_train[:] == num1
        idx2 = Y_train[:] == num2
        idx3 = Y_train[:] == num3
        Xc1 = Xc[:, idx1]
        Xc2 = Xc[:, idx2]
        Xc3 = Xc[:, idx3]
        C1 = centroid(Xc1)
        C2 = centroid(Xc2)
        C3 = centroid(Xc3)
        U_k = U[:, :k]
        Z_k = U_k.T @ Xc
        Z_k1 = Z_k[:, idx1]
        Z_k2 = Z_k[:, idx2]
        Z_k3 = Z_k[:, idx3]
```

```
C_k1 = np.reshape(np.mean(Z_k1, axis=1), (k,1))
C_k2 = np.reshape(np.mean(Z_k2, axis=1), (k,1))
C_k3 = np.reshape(np.mean(Z_k3, axis=1), (k,1))

counter = 0

for i in range(Y_test.shape[0]):
    c = classificationAlgo(C1, C2, C3, X_test[:, i])
    if c == Y_test[i]:
        counter += 1

print(f"Accuracy: {counter*100/len(Y_test)} %")
```

Accuracy: 34.302822925577416 %

```
In [ ]: k = 10
        def centroid(X):
            return np.mean(X, axis=1, keepdims=True)
         (X_train, Y_train), (X_test, Y_test) = train_test_split(X, Y, N_train)
        CX = centroid(X train)
        Xc = X_{train} - CX
        U,s,VT = np.linalg.svd(Xc, full_matrices=False)
        idx1 = Y_train[:] == num1
        idx2 = Y train[:] == num2
        idx3 = Y_train[:] == num3
        Xc1 = Xc[:, idx1]
        Xc2 = Xc[:, idx2]
        Xc3 = Xc[:, idx3]
        C1 = centroid(Xc1)
        C2 = centroid(Xc2)
        C3 = centroid(Xc3)
        U_k = U[:, :k]
        Z_k = U_k.T @ Xc
        Z_k1 = Z_k[:, idx1]
        Z_k2 = Z_k[:, idx2]
        Z_k3 = Z_k[:, idx3]
        C_k1 = np.reshape(np.mean(Z_k1, axis=1), (k,1))
        C_k2 = np.reshape(np.mean(Z_k2, axis=1), (k,1))
        C_k3 = np.reshape(np.mean(Z_k3, axis=1), (k,1))
        counter = 0
        for i in range(Y_test.shape[0]):
            c = classificationAlgo(C1, C2, C3, X_test[:, i])
            if c == Y test[i]:
                counter += 1
        print(f"Accuracy: {counter*100/len(Y_test)} %")
```

Accuracy: 33.960650128314796 %