## Exercise 3

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## 1 EECS 491: Probabilistic Graphical Models Assignment 4

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## 2 Exercise 3

In this notebook we will utilize PCA to compute the principle components of the MNIST dataset. First let us load in the MNIST dataset:

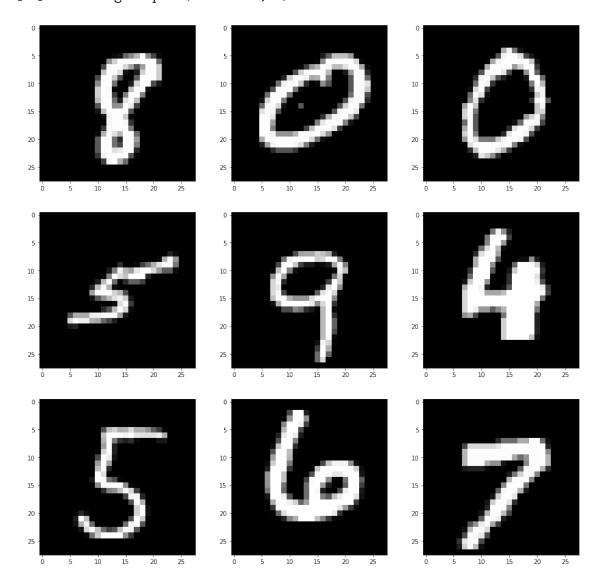
```
In [16]: import gzip, pickle
    import numpy as np
    import matplotlib.pyplot as plt
    from sklearn.decomposition import PCA
    %matplotlib inline

In [17]: with gzip.open('data/mnist.pkl.gz', 'rb') as f:
        f = pickle._Unpickler(f)
        f.encoding = 'latin1'
        trainSet, validSet, testSet = f.load()
        mnistData = np.concatenate((trainSet[0], validSet[0], testSet[0]), axis=0)
```

The following is a function taken from the demo that displays a number of image samples from a given dataset:

We will use it to display 9 images from the MNIST dataset:

In [19]: showImageSamples(mnistData, 9)



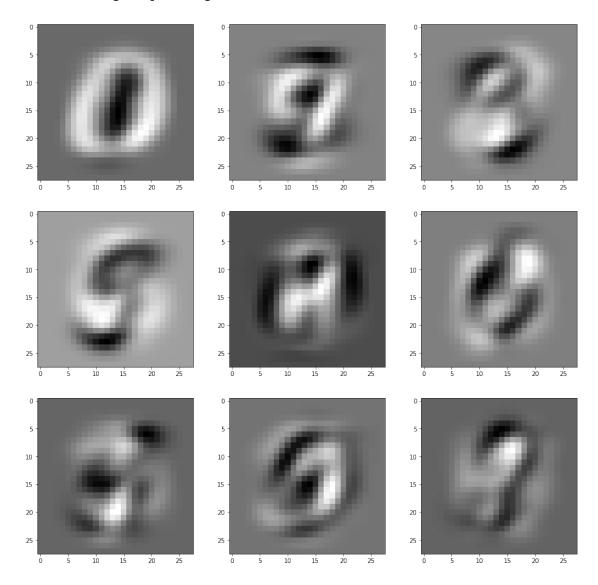
Now let us implement a function that will reduce the dimensionality of a given dataset using PCA, and return the eigenvectors and eigenvalues of the covariance matrix of the dataset:

```
In [20]: def dimReducePCA(data, dims):
    pca = PCA(n_components=dims)
    reduced_data = pca.fit_transform(data)
    pca = PCA()
    pca.fit(data)
    eigen_vectors = pca.components_.T
    eigen_values = pca.explained_variance_
    return reduced_data, eigen_vectors, eigen_values
```

Let's call that function on the MNIST dataset:

In [23]: mnistData2D, eigenVectors, eigenValues = dimReducePCA(mnistData, 2)
Now, using the eigenvectors returned, let's show the first few primary components of MNIST:

In [22]: showImageSamples(eigenVectors.T, 9, random=False)



In the next exercise, we will use this dimension reduced MNIST dataset to train a Gaussian Mixture Model.