

## PCA ANALYSIS ON MNIST DATASET

```
In [17]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings("ignore")
```

## LOAD MINSNT DATASET

```
In [18]: d0 = pd.read_csv('./mnist_train.csv')
```

```
In [19]: print (d0.head(5))
```

	label	pixel0	pixel1	pixel2	pixel3	pixel4	pixel5	pixel6	pixel
7	\								
0	1	0	0	0	0	0	0	0	
0									
1	0	0	0	0	0	0	0	0	
0									
2	1	0	0	0	0	0	0	0	
0									
3	4	0	0	0	0	0	0	0	
0									
4	0	0	0	0	0	0	0	0	
0									
	pixel8	...	pixel774	pixel775	pixel776	pixel777	pixel778		
\									
0	0	...	0	0	0	0	0		
1	0	...	0	0	0	0	0		

2	0	...	0	0	0	0	0
3	0	...	0	0	0	0	0
4	0	...	0	0	0	0	0

	pixel779	pixel780	pixel781	pixel782	pixel783
0	0	0	0	0	0
1	0	0	0	0	0
2	0	0	0	0	0
3	0	0	0	0	0
4	0	0	0	0	0

[5 rows x 785 columns]

```
In [20]: l = d0['label']
```

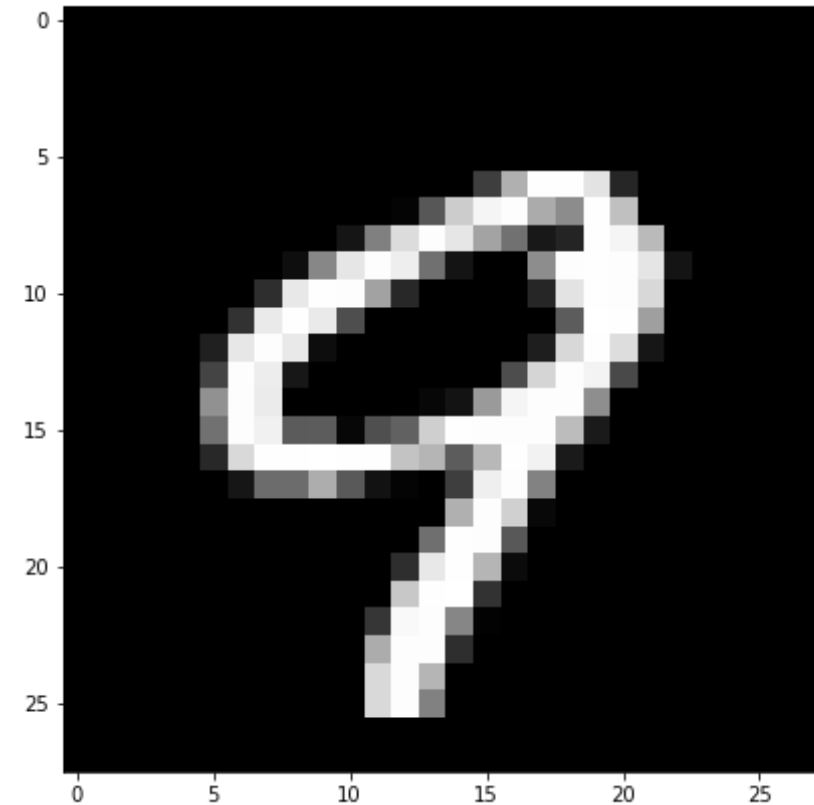
```
In [21]: d = d0.drop ("label" , axis = 1)
```

```
In [22]: print (d.shape)
          print (l.shape)
```

```
(42000, 784)
(42000,)
```

```
In [23]: #display plot number
```

```
plt.figure (figsize = (7,7))
idx = 100
grid_data = d.iloc[idx].as_matrix().reshape(28,28)
plt.imshow (grid_data , interpolation = "none" , cmap = "gray" )
plt.show()
print (l[idx])
```



9

## 2D VISUALIZATION USING PCA

```
In [24]: labels = l.head(15000)
data = d.head(15000)

print ("The shape of sample data = " , data.shape)
```

The shape of sample data = (15000, 784)

```
In [25]: from sklearn.preprocessing import StandardScaler
```

```
In [26]: standardized_data = StandardScaler().fit_transform(data)
print (standardized_data.shape)

(15000, 784)
```

```
In [27]: #Find covariance matrix
sample_data = standardized_data

#matrix multiplication using numpy
covar_matrix = np.matmul (sample_data.T , sample_data)
print ("The shape of variance matrix is = " , covar_matrix.shape)

The shape of variance matrix is = (784, 784)
```

```
In [28]: #finding top two eigen values and eigen vectors for projecting into 2 d
dimension space

from scipy.linalg import eig

values , vectors = eig (covar_matrix , eigvals = (782, 783))

print ("Shape of eigen vectors" , vectors.shape)

vectors = vectors.T

print ("Updated shape of eigen vectors" , vectors.shape)

Shape of eigen vectors (784, 2)
Updated shape of eigen vectors (2, 784)
```

```
In [29]: #projecting sample data on sample plane

import matplotlib.pyplot as plt

new_coordinates = np.matmul (vectors , sample_data.T)
```

```
print ("Resultant new data points " , vectors.shape , "X" , sample_data.T.shape, " = " , new_coordinates.shape)
```

Resultant new data points (2, 784) X (784, 15000) = (2, 15000)

```
In [30]: import pandas as pd

new_coordinates = np.vstack ((new_coordinates, labels)).T
```

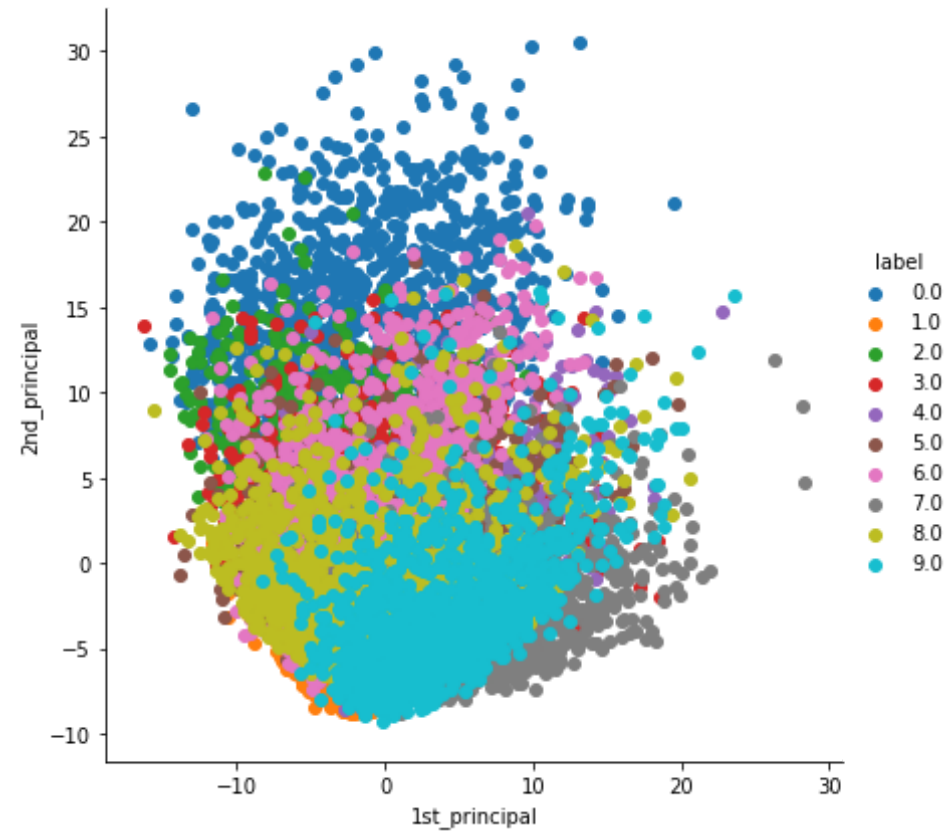
vstack is used to stack arrays vertically

```
In [35]: # creating a new data frame for plotting the labeled points.
dataframe = pd.DataFrame(data=new_coordinates, columns=("1st_principal"
, "2nd_principal", "label"))
print(dataframe.head())
```

	1st_principal	2nd_principal	label
0	-5.558661	-5.043558	1.0
1	6.193635	19.305278	0.0
2	-1.909878	-7.678775	1.0
3	5.525748	-0.464845	4.0
4	6.366527	26.644289	0.0

## PLOTTING 2D DataPoints with seaborn

```
In [36]: import seaborn as sn
sn.FacetGrid(dataframe, hue="label", size=6).map(plt.scatter, '1st_principal', '2nd_principal').add_legend()
plt.show()
```



## PCA using SCIKIT LEARN

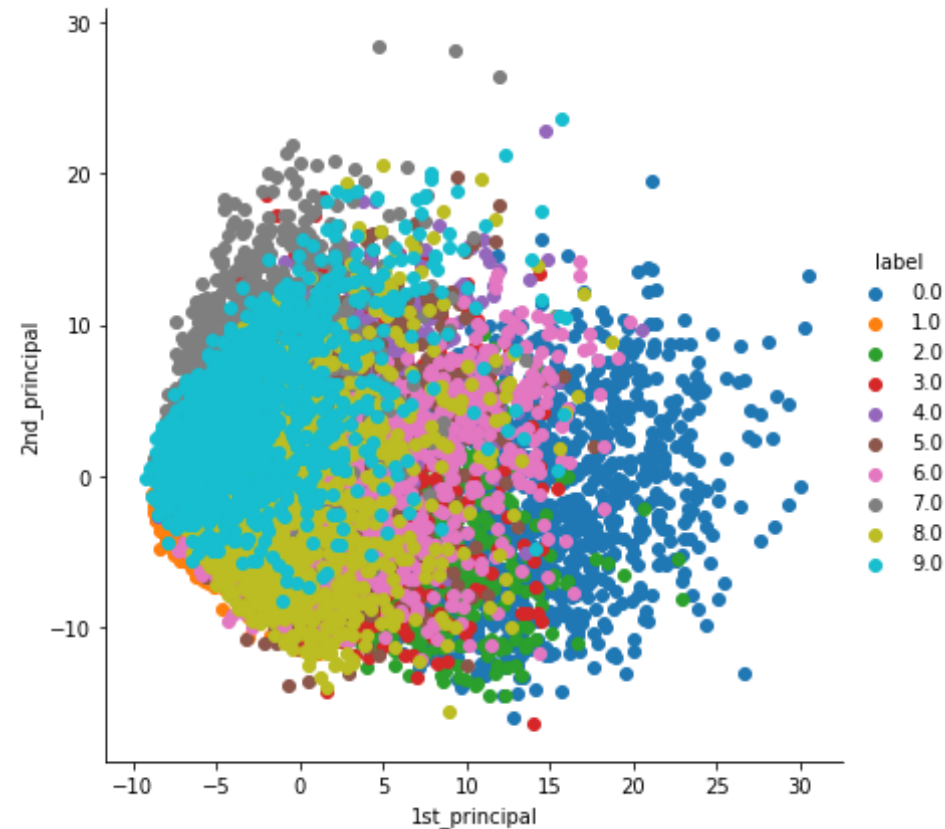
```
In [37]: from sklearn import decomposition  
pca = decomposition.PCA()
```

```
In [38]: pca.n_components = 2  
pca_data = pca.fit_transform(sample_data)  
  
# pca_reduced will contain the 2-d projects of simple data  
print("shape of pca_reduced.shape = ", pca_data.shape)
```

```
shape of pca_reduced.shape = (15000, 2)
```

```
In [39]: pca_data = np.vstack((pca_data.T, labels)).T

# creating a new data fram which help us in plotting the result data
pca_df = pd.DataFrame(data=pca_data, columns=("1st_principal", "2nd_principal", "label"))
sns.FacetGrid(pca_df, hue="label", size=6).map(plt.scatter, '1st_principal', '2nd_principal').add_legend()
plt.show()
```



```
In [ ]:
```

