

```

1  from google.colab import drive
2  drive.mount('/content/drive')
3  import numpy as np
4  import scipy.io
5  import pandas as pd
6  from skimage import color
7  from skimage import io
8  import math
9  import matplotlib.pyplot as plt
10 from sklearn.model_selection import train_test_split
11 from sklearn.metrics import accuracy_score, confusion_matrix
12 from sklearn.decomposition import PCA

```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mour



```

1  ctScans = scipy.io.loadmat('/content/drive/My Drive/CCE-AIMIA/ctscan_embeddings_hw2.mat')

```

```

1  print(ctScans['feat'].shape)
2  X = ctScans['feat']

```

↪ (3554, 1024)

+ Code

+ Text

```

1  pca_dims = PCA()
2  pca_dims.fit(X)
3  cumsum = np.cumsum(pca_dims.explained_variance_ratio_)
4  d = np.argmax(cumsum >= 0.95) + 1
5  print(d)

```

33

```

1  #n_components=100
2  n_components=50

```

```

1  pca = PCA(n_components)
2  X_reduced = pca.fit_transform(X)
3  X_recon = pca.inverse_transform(X_reduced)

```

```

1  print("reduced shape: " + str(X_reduced.shape))
2  print("reconstructed shape: " + str(X_recon.shape))

```

reduced shape: (3554, 50)  
reconstructed shape: (3554, 1024)

```

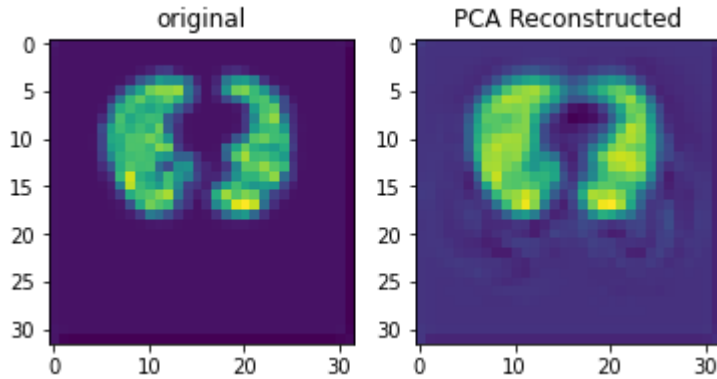
1  f = plt.figure()
2  f.add_subplot(1,2, 1)

```

```

3 plt.title("original")
4 plt.imshow(X[100].reshape((32,32)))
5 f.add_subplot(1,2, 2)
6
7 plt.title("PCA Reconstructed")
8 plt.imshow(X_recon[100].reshape((32,32)))
9 plt.show(block=True)

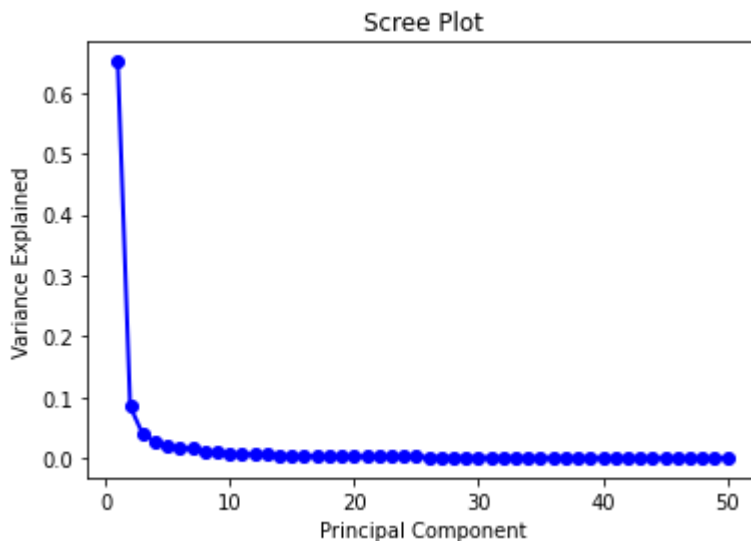
```



```

1 PC_values = np.arange(pca.n_components_) + 1
2 plt.plot(PC_values, pca.explained_variance_ratio_, 'o-', linewidth=2, color='blue')
3 plt.title('Scree Plot')
4 plt.xlabel('Principal Component')
5 plt.ylabel('Variance Explained')
6 plt.show()

```



```

1 #function for mean square error
2 def mse(predict, actual):
3     return np.square(predict - actual).sum(axis = 1).mean()

```

```

1 #calculating loss and reconstructing images
2 loss = []
3 max_components = 50

```

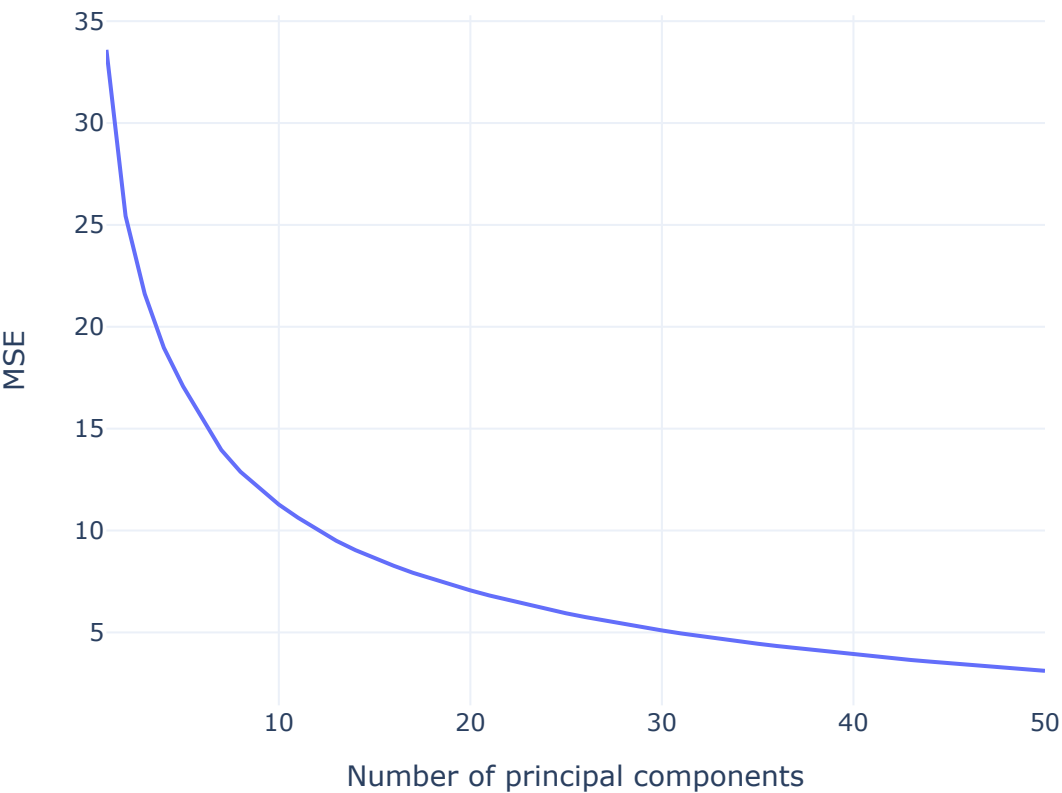
```
4 print("Processing...")
5 for num_component in range(1, max_components + 1):
6     pca = PCA(num_component)
7     X_reduced = pca.fit_transform(X)
8     X_recovered = pca.inverse_transform(X_reduced)
9     error = mse(X_recovered, X)
10    loss.append((num_component, error))
11 print()
12 print("Done!")
```

Processing...

Done!

```
1 import plotly.graph_objs as go
2 #visualizing mse vs number of principal components
3 result = list(map(list, zip(*loss)))
4 x, y = result
5 trace = go.Scatter(x = x[:,], y = y[:,])
6 data = [trace]
7 fig = go.Figure(data)
8 fig.update_layout(title = "MSE vs number of principal components",
9                    xaxis_title = "Number of principal components",
10                   yaxis_title = "MSE", template = "plotly_white")
11 fig.show()
```

MSE vs number of principal components



✓ 1s completed at 1:10 PM ● ✕