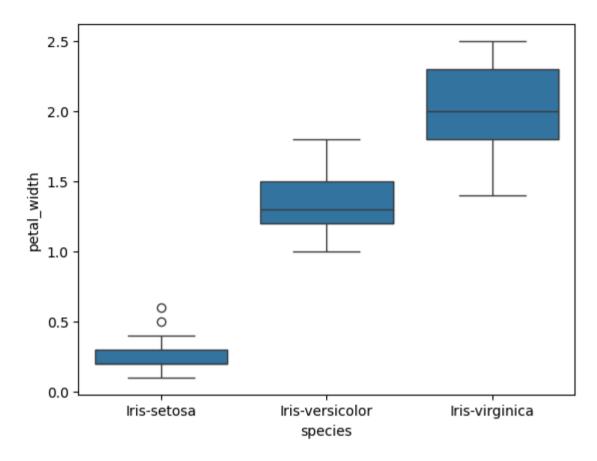
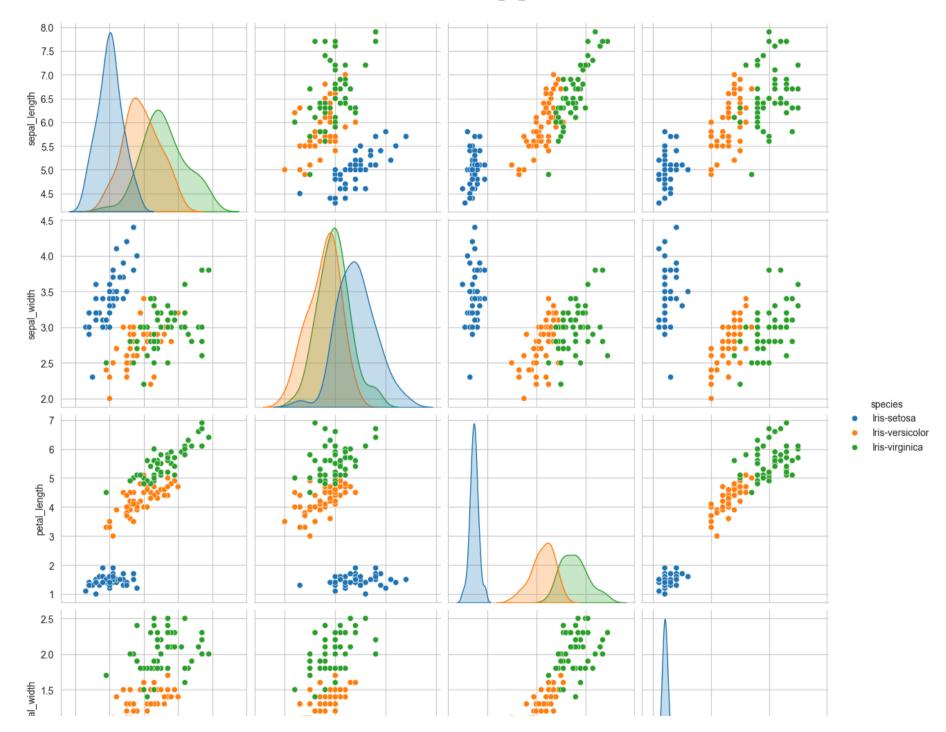
## k-means

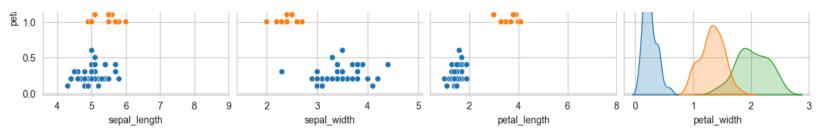
```
In [ ]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.cluster import KMeans
In [ ]: iris = pd.read csv("IRIS.csv")
In [ ]: iris.info()
       <class 'pandas.core.frame.DataFrame'>
       RangeIndex: 150 entries, 0 to 149
       Data columns (total 5 columns):
                         Non-Null Count Dtype
            Column
           sepal length 150 non-null
                                         float64
           sepal width 150 non-null
                                         float64
           petal length 150 non-null
                                         float64
           petal_width 150 non-null
                                         float64
           species
                         150 non-null
                                         object
       dtypes: float64(4), object(1)
       memory usage: 6.0+ KB
In [ ]: sns.boxplot(x = "species",y = "petal width" , data = iris)
        plt.show()
```



```
In [ ]: sns.set_style("whitegrid")
    sns.pairplot(iris, hue = "species" , size=3)
    plt.show()
```

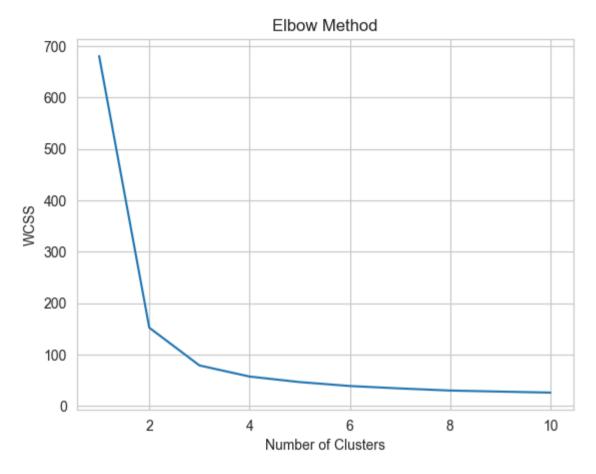
c:\Users\Deepthi\AppData\Local\Programs\Python\Python38\lib\site-packages\seaborn\axisgrid.py:2100: UserWarning: The `size` par
ameter has been renamed to `height`; please update your code.
warnings.warn(msg, UserWarning)





```
In []: data = iris.drop(columns=['species'])
In []: from sklearn.cluster import KMeans
    wcss = []
    for i in range(1,11):
        kmeans = KMeans(n_clusters= i , init='k-means++',max_iter=300, n_init= 10 , random_state=0)
        kmeans.fit(data)
        wcss.append(kmeans.inertia_)

In []: plt.plot(range(1, 11), wcss)
    plt.title('Elbow Method')
    plt.xlabel('Number of Clusters')
    plt.ylabel('WCSS')
    plt.show()
```



## PCA USING HOUUSING DATSET

```
import pandas as pd
        import numpy as np
In [ ]: data = pd.read_csv("newhousing.csv")
In [ ]: data.head()
Out[]:
             price area bedrooms bathrooms stories mainroad guestroom basement hotwaterheating airconditioning parking prefarea
        0 5250000 5500
                                3
                                                                       0
                                                                                                0
                                                                                                                               0
        1 4480000 4040
                                                  2
                                                                       0
                                                                                 0
                                                                                                                               0
        2 3570000 3640
                                2
                                                                       0
                                                                                 0
                                                                                                0
                                                                                                              0
                                                                                                                       0
                                                                                                                               0
                                                  1
        3 2870000 3040
                                2
                                                  1
                                                            0
                                                                       0
                                                                                                                       0
                                                                                                                               0
                                                                                 0
                                                                                                0
        4 3570000 4500
                                2
                                           1
                                                  1
                                                            0
                                                                       0
                                                                                 0
                                                                                                0
                                                                                                              0
                                                                                                                       0
                                                                                                                               0
In [ ]: # What type of values are stored in the columns?
        data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
       RangeIndex: 545 entries, 0 to 544
       Data columns (total 16 columns):
            Column
                            Non-Null Count Dtype
                             -----
            price
                             545 non-null
                                            int64
        1
            area
                             545 non-null
                                            int64
            bedrooms
                            545 non-null
                                            int64
        3
            bathrooms
                            545 non-null
                                            int64
                            545 non-null
            stories
                                            int64
            mainroad
                            545 non-null
                                            int64
                            545 non-null
            guestroom
                                            int64
        7
                            545 non-null
            basement
                                            int64
            hotwaterheating 545 non-null
                                            int64
            airconditioning 545 non-null
                                            int64
            parking
                             545 non-null
                                            int64
        11 prefarea
                            545 non-null
                                            int64
        12 semi-furnished 545 non-null
                                            int64
        13 unfurnished
                            545 non-null
                                            int64
        14 areaperbedroom
                            545 non-null
                                            float64
        15 bbratio
                             545 non-null
                                            float64
       dtypes: float64(2), int64(14)
       memory usage: 68.2 KB
In [ ]: X = data[['area', 'bedrooms', 'bathrooms', 'stories', 'mainroad',
               'guestroom', 'basement', 'hotwaterheating', 'airconditioning',
               'parking', 'prefarea', 'semi-furnished', 'unfurnished',
               'areaperbedroom', 'bbratio']]
        # Putting response variable to y
        y = data['price']
In [ ]: from sklearn.preprocessing import MinMaxScaler
        scaler = MinMaxScaler()
        X=scaler.fit transform(X)
In [ ]: print(X)
```

```
[[0.26460481 0.4
                              0.33333333 ... 0.
                                                        0.23353165 0.6
        [0.16426117 0.4
                                                        0.15527684 0.2
                                         ... 0.
        [0.13676976 0.2
                              0.
                                         ... 1.
                                                        0.23138768 0.4
                                         ... 0.
        [0.45360825 0.4
                               0.
                                                        0.38092941 0.2
                                         ... 0.
                                                        0.85608619 0.4
        [0.67079038 0.2
                               0.
                                          ... 0.
        [0.15931271 0.4
                              0.
                                                        0.1514177 0.2
                                                                             ]]
In [ ]: from sklearn.model selection import train test split
        X train, X test, y train, y test = train test split(X, y, test size = 0.3, random state=20)
In [ ]: from sklearn.decomposition import PCA
        pca = PCA()
        X_new = pca.fit_transform(X)
In [ ]: pca.get covariance()
```

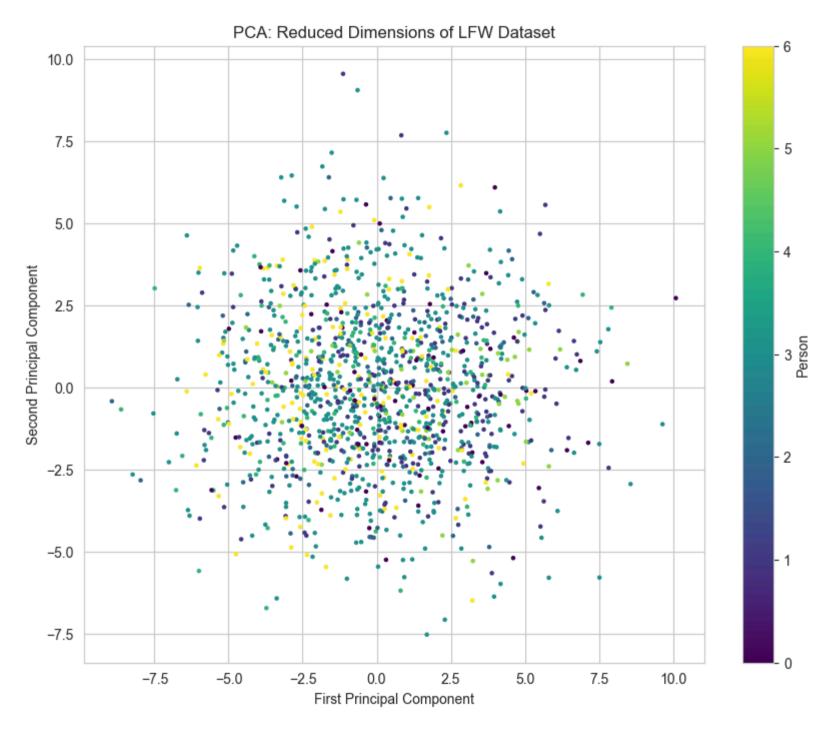
```
Out[]: array([[ 0.0222459 ,
                             0.0033434 , 0.00484185 , 0.00362267 ,
                                                                   0.01502118,
                 0.00801123, 0.00337738, -0.00028825, 0.01543005, 0.01512004,
                 0.01485796, 0.00045307, -0.00996113, 0.01621209, 0.00163042]
                             0.02178953, 0.00924492, 0.01743929, -0.00061927,
               [ 0.0033434 ,
                 0.00455208,
                             0.00685982, 0.00142337, 0.01102806,
                                                                   0.00590416,
                 0.00494941. 0.00364477. -0.00874798. -0.007912 . -0.008964331.
               [ 0.00484185, 0.00924492, 0.02805286, 0.01579683, 0.00247572,
                 0.0081096, 0.00816694, 0.00235541, 0.0145631, 0.00853796,
                 0.0045107, 0.0024656, -0.01038631, -0.00118307, 0.02330959]
               [ 0.00362267, 0.01743929, 0.01579683, 0.08361591,
                                                                   0.01226952,
                 0.00481989, -0.02380599, 0.00114117, 0.03949339,
                                                                   0.00378253,
                 0.00545062, -0.00052055, -0.01126214, -0.00698145, 0.0018242],
               [0.01502118, -0.00061927, 0.00247572, 0.01226952, 0.12154614,
                 0.01232461, 0.00732596, -0.00086009, 0.01709727, 0.02046906,
                             0.00196978, -0.02178562, 0.01175176,
                 0.02956692,
                                                                   0.00202894],
               [ 0.00801123, 0.00455208, 0.0081096, 0.00481989,
                                                                   0.01232461,
                 0.14657312, 0.06802482, -0.00082636, 0.02460874, 0.00411945,
                 0.02613667, 0.00109957, -0.01779547, 0.0028055,
                                                                   0.00555808],
               [ 0.00337738, 0.00685982, 0.00816694, -0.02380599,
                                                                   0.00732596,
                 0.06802482,
                             0.22805586, 0.00043848,
                                                      0.01051673,
                                                                   0.00706287,
                 0.0462156 , 0.01184903 , -0.02643686 , -0.00109756 ,
                                                                   0.00432737],
               [-0.00028825, 0.00142337, 0.00235541, 0.00114117, -0.00086009,
                -0.00082636.
                             0.00043848, 0.04384781, -0.01266527, 0.00408122,
                -0.0052786 ,
                             0.00659404, -0.00581827, -0.00061124,
                                                                   0.00086245],
               [ 0.01543005, 0.01102806, 0.0145631, 0.03949339,
                                                                   0.01709727,
                 0.02460874, 0.01051673, -0.01266527, 0.21639234, 0.02126507,
                 0.02316851, -0.01220656, -0.02054439, 0.0060774, 0.00575573]
               [ 0.01512004, 0.00590416, 0.00853796,
                                                      0.00378253, 0.02046906,
                 0.00411945, 0.00706287, 0.00408122, 0.02126507,
                                                                   0.08248111,
                 0.01116545, 0.00585649, -0.02233877, 0.0086949,
                                                                   0.00352372],
               [ 0.01485796, 0.00494941, 0.0045107, 0.00545062,
                                                                   0.02956692,
                 0.02613667, 0.0462156, -0.0052786, 0.02316851, 0.01116545,
                 0.18003238, -0.002415 , -0.01618659, 0.00810504, -0.00150924]
               [ 0.00045307, 0.00364477, 0.0024656, -0.00052055,
                                                                   0.00196978,
                 0.00109957, 0.01184903, 0.00659404, -0.01220656, 0.00585649,
                -0.002415 , 0.24347679 , -0.13628575 , -0.00192466 , -0.00122922 ]
               [-0.00996113, -0.00874798, -0.01038631, -0.01126214, -0.02178562,
                -0.01779547, -0.02643686, -0.00581827, -0.02054439, -0.02233877,
                -0.01618659, -0.13628575, 0.22033864, -0.00394716, -0.00371816],
               [0.01621209, -0.007912, -0.00118307, -0.00698145, 0.01175176,
```

```
0.0028055 , -0.00109756, -0.00061124, 0.0060774 , 0.0086949 ,
                 0.00810504, -0.00192466, -0.00394716, 0.01820446, 0.00581644],
                [ 0.00163042, -0.00896433, 0.02330959, 0.0018242, 0.00202894,
                 0.00555808, 0.00432737, 0.00086245, 0.00575573, 0.00352372,
                -0.00150924, -0.00122922, -0.00371816, 0.00581644, 0.03663016]])
In [ ]: explained variance=pca.explained variance ratio
        explained variance
Out[]: array([0.2295475 , 0.17794124, 0.13925925, 0.09592225, 0.07018757,
               0.06282636, 0.05155462, 0.04543956, 0.04320347, 0.03114886,
               0.02425278, 0.01574126, 0.01217287, 0.00052413, 0.00027827
In [ ]: pca=PCA(n components=3)
        X new=pca.fit transform(X)
        X train new, X test new, y train, y test = train test split(X new, y, test size=0.2, random state=42)
In [ ]: from sklearn.neighbors import KNeighborsClassifier
        knn pca = KNeighborsClassifier(7)
        knn pca.fit(X train new, y train)
        print("Train score after PCA",knn pca.score(X train new,y train),"%")
        print("Test score after PCA",knn pca.score(X test new,y test),"%")
       Train score after PCA 0.15825688073394495 %
       Test score after PCA 0.009174311926605505 %
```

## **PCA** with images

```
In [ ]: pca = PCA(n_components=2)
    X_pca = pca.fit_transform(X)

In [ ]: plt.figure(figsize=(10, 8))
    scatter = plt.scatter(X_pca[:, 0], X_pca[:, 1], c=y, cmap='viridis', s=5)
    plt.xlabel('First Principal Component')
    plt.ylabel('Second Principal Component')
    plt.title('PCA: Reduced Dimensions of LFW Dataset')
    plt.colorbar(scatter, label='Person')
    plt.show()
```

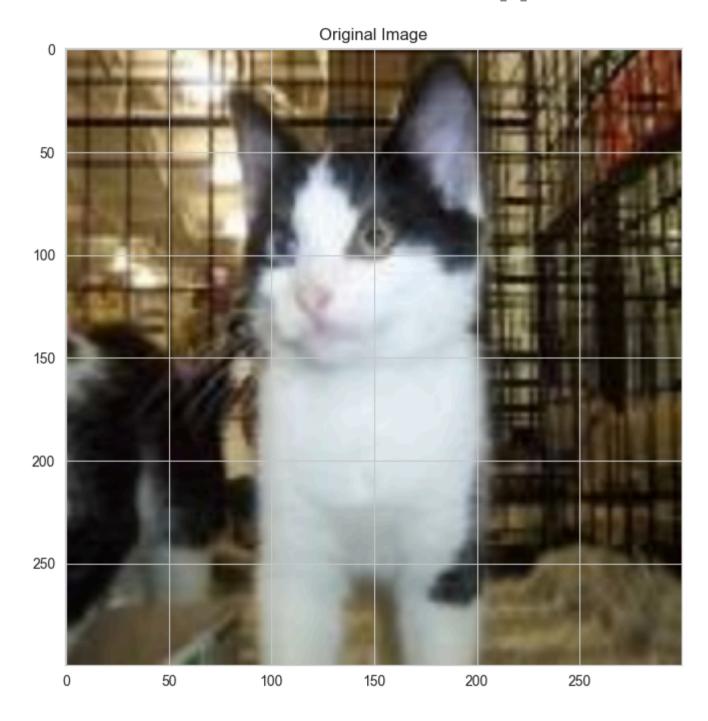


```
In []:
import os
import cv2
image_folder=r'C:\Users\Deepthi\OneDrive\Desktop\programs\mlt lab works\mlt lab 8\Cat_Pic'
def load_images(folder, target_size=(300, 300)):
    image_files = sorted([os.path.join(folder, file) for file in os.listdir(folder) if file.endswith(('.jpg', '.png'))])
    images = [cv2.resize(cv2.imread(file), target_size) for file in image_files]
    return np.array(images)

# Load and resize images from the folder
images = load_images(image_folder)
print("Image shape:", images.shape)

Image shape: (25, 300, 300, 3)

In []: plt.figure(figsize=[12,8])
    plt.imshow(cv2.cvtColor(images[1], cv2.COLOR_BGR2RGB)) # OpenCV reads images in BGR format, so we convert it to RGB for displ
plt.title('Original Image')
    plt.show()
```



```
In [ ]: images flattened = images.reshape(images.shape[0], -1)
        print("Flattened image shape:", images flattened.shape)
       Flattened image shape: (25, 270000)
In [ ]: pca = PCA()
        pca.fit(images flattened)
Out[ ]:
       ▼ PCA
        PCA()
In [ ]: var cumu = np.cumsum(pca.explained variance ratio ) * 100
        var cumu
Out[]: array([22.34683724, 37.60041994, 46.79205472, 54.60804188,
                60.32295644, 65.3722507, 69.45593445, 73.24308353,
                76.10939619, 78.73105463, 81.27402831, 83.63704056,
                85.85876853, 87.73831425, 89.46827525, 91.1226531,
                92.4911216 , 93.78163845, 95.00327912, 96.17728962,
                97.27897343, 98.30540158, 99.22200835, 100.
               100.
                           1)
In [ ]: k = np.argmax(var cumu > 95)
        print("Number of components explaining 95% variance:", k)
       Number of components explaining 95% variance: 18
In [ ]: from sklearn.decomposition import IncrementalPCA
        ipca = IncrementalPCA(n components=k)
        images recon = ipca.inverse transform(ipca.fit transform(images flattened))
In [ ]: # Reshape the reconstructed images
        images recon = images recon.reshape(images.shape)
        images recon
```

```
Out[]: array([[[ 3.52100102e+01, 3.14943246e+01, 3.01547374e+01],
                 [ 2.94531816e+01, 2.71103233e+01, 2.48999544e+01],
                 [ 2.27601252e+01, 1.93329153e+01, 1.71886221e+01],
                 [ 1.63859138e+02, 1.63400109e+02, 1.19807893e+02],
                 [ 1.61793819e+02, 1.61018137e+02, 1.17012614e+02],
                 [1.61677404e+02, 1.60623053e+02, 1.16655791e+02]],
                [ [ 4.36869624e+01, 4.02895239e+01, 3.86166746e+01], 
                 [ 4.22620659e+01, 3.96633003e+01, 3.78274917e+01],
                 [ 3.65152012e+01, 3.34862946e+01, 3.12190489e+01],
                 [1.63777126e+02, 1.63059107e+02, 1.20119104e+02],
                 [ 1.63385490e+02, 1.62306579e+02, 1.19281019e+02],
                 [ 1.58322929e+02, 1.58281898e+02, 1.15144081e+02]],
                [ 4.11710477e+01, 4.03606524e+01, 3.63787868e+01],
                 [ 4.68378577e+01, 4.62356282e+01, 4.22045559e+01],
                 [ 4.48697411e+01, 4.29070674e+01, 3.94815737e+01],
                 . . . ,
                 [ 1.64929948e+02, 1.64443478e+02, 1.22929552e+02],
                 [1.63145859e+02, 1.62469388e+02, 1.19129199e+02],
                 [ 1.58554209e+02, 1.57719639e+02, 1.15316283e+02]],
                . . . ,
                [1.00415654e+02, 1.50035152e+02, 1.66814985e+02],
                 [ 1.00901428e+02, 1.50611112e+02, 1.66631373e+02],
                 [ 9.43360428e+01, 1.43936507e+02, 1.59958548e+02],
                 [1.06861393e+01, 1.69925965e+02, 1.74683784e+02],
                 [ 9.28458307e+00, 1.59627259e+02, 1.64897263e+02],
                 [ 1.14845910e+01, 1.51828889e+02, 1.58837186e+02]],
                [[ 9.73999440e+01, 1.40516877e+02, 1.57904204e+02],
                 [ 8.88443137e+01, 1.33786879e+02, 1.51253844e+02],
                 [ 7.46890108e+01, 1.22203294e+02, 1.38853134e+02],
                 [ 9.37447267e+00, 1.72844550e+02, 1.76813061e+02],
                 [ 7.68176707e+00, 1.64551097e+02, 1.69086423e+02],
```

```
[ 4.15628710e+00, 1.56494257e+02, 1.61869096e+02]],
[ 9.44987276e+01, 1.36591628e+02, 1.54130943e+02],
 [8.92104440e+01, 1.32217322e+02, 1.49892379e+02],
 [ 7.98433809e+01, 1.25406538e+02, 1.43107242e+02],
 [ 1.20192081e+01, 1.78428526e+02, 1.81084421e+02],
 [ 8.09448321e+00, 1.69847130e+02, 1.73287187e+02],
 [ 5.97304493e+00, 1.61466292e+02, 1.65824023e+02]]],
[[ 8.78215819e+01, 1.16348701e+02, 1.39912922e+02],
 [ 8.84154039e+01, 1.16703316e+02, 1.40210518e+02],
 [ 8.83538274e+01, 1.16267443e+02, 1.40274570e+02],
 [ 5.52688864e+01, 7.70274320e+01, 1.26109899e+02],
 [ 4.43371681e+01, 6.45510989e+01, 1.11992824e+02],
 [ 3.92217603e+01, 5.77812369e+01, 1.04122950e+02]],
[[8.54581964e+01, 1.14099131e+02, 1.38264151e+02],
 [ 9.09580095e+01, 1.19243385e+02, 1.43297214e+02],
 [ 8.74994961e+01, 1.15280370e+02, 1.39627250e+02],
 [ 5.81896379e+01, 7.97532187e+01, 1.28234440e+02],
 [ 4.27227652e+01, 6.35650184e+01, 1.10117124e+02],
 [ 3.79454654e+01, 5.70812694e+01, 1.02776969e+02]],
[[ 8.29107316e+01, 1.11211574e+02, 1.35656579e+02],
 [ 8.65030259e+01, 1.14737653e+02, 1.39114961e+02],
 [ 8.86077699e+01, 1.16398762e+02, 1.40835331e+02],
 [ 5.87878712e+01, 8.16006808e+01, 1.29709924e+02],
 [ 4.02337823e+01, 6.05446345e+01, 1.06866827e+02],
 [ 3.29396123e+01, 5.20756379e+01, 9.75228151e+01]],
 . . . ,
[[ 5.81953722e+01, 6.45556797e+01, 7.06713955e+01],
 [ 5.71787919e+01, 6.34266620e+01, 6.94750895e+01],
 [ 5.35302602e+01, 5.99860021e+01, 6.59116575e+01],
  . . . ,
```

```
[ 6.32874545e+01, 9.11591547e+01, 9.84197469e+01],
 [ 5.96264346e+01, 8.69711017e+01, 9.37404109e+01],
 [ 5.84544908e+01, 8.52261496e+01, 9.20576274e+01]],
[5.43552700e+01, 6.12932521e+01, 6.74238732e+01],
 [ 5.38129920e+01, 6.00201298e+01, 6.62829375e+01],
 [ 5.09614238e+01, 5.74647327e+01, 6.34959170e+01],
 . . . ,
 [ 5.99227575e+01, 8.82875830e+01, 9.54806841e+01],
 [ 5.84622310e+01, 8.57415381e+01, 9.25325389e+01],
 [ 5.56835405e+01, 8.27600834e+01, 8.94950669e+01]],
[ 5.07439393e+01, 5.67192523e+01, 6.30375865e+01],
 [ 4.95267664e+01, 5.55812007e+01, 6.20649220e+01],
 [ 4.79095136e+01, 5.41313609e+01, 6.05654413e+01],
 [ 5.76394739e+01, 8.58573609e+01, 9.31773345e+01],
 [5.50159901e+01, 8.21774533e+01, 8.90186230e+01],
 [ 5.56366625e+01, 8.24613941e+01, 8.92323002e+01]]],
[[2.04193385e+01, 1.93610698e+01, 2.11616114e+01],
 [ 1.98492676e+01, 1.89214916e+01, 2.05460841e+01],
 [ 1.56983844e+01, 1.42970839e+01, 1.60760324e+01],
  . . . ,
 [ 2.69077288e+01, 4.95244782e+01, 6.40915254e+01],
 [ 2.45306691e+01, 4.79671951e+01, 6.32040818e+01],
 [ 2.61461002e+01, 4.99020301e+01, 6.43718201e+01]],
[ 1.67697280e+01, 1.52231200e+01, 1.61985770e+01],
 [ 1.25217267e+01, 1.06245930e+01, 1.18361826e+01],
 [ 1.67999873e+01, 1.49811266e+01, 1.61721121e+01],
 [ 2.59341043e+01, 4.85591542e+01, 6.36628394e+01],
 [ 2.40206901e+01, 4.72670125e+01, 6.22597405e+01],
 [2.53796841e+01, 4.85862604e+01, 6.29609848e+01]],
[[ 2.14298468e+01, 1.92410103e+01, 1.98621732e+01],
 [ 2.41794738e+01, 2.23465227e+01, 2.32947967e+01],
 [ 2.51460761e+01, 2.27692954e+01, 2.38488429e+01],
  . . . ,
```

file:///C:/Users/Deepthi/OneDrive/Desktop/programs/mlt lab works/mlt lab 8/mlt lab 8.html

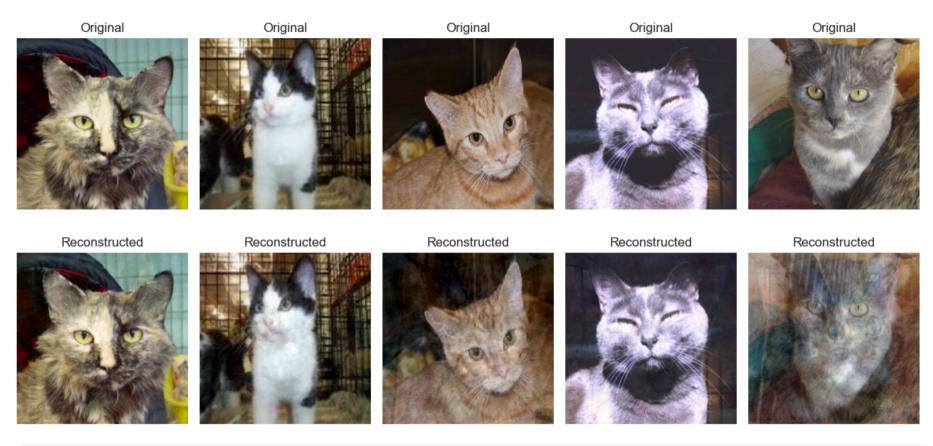
```
[ 2.80505817e+01, 4.91316693e+01, 6.46389529e+01],
 [ 2.70609926e+01, 4.91998671e+01, 6.45644004e+01],
 [ 2.81416804e+01, 5.02926703e+01, 6.51780071e+01]],
 . . . ,
[ 7.02691567e+01, 8.38833094e+01, 1.10244255e+02 ],
 [7.21251055e+01, 8.53973451e+01, 1.11427641e+02],
 [ 7.50034746e+01, 8.60227539e+01, 1.12975352e+02],
 [ 7.43834740e+01, 8.89433541e+01, 1.06422639e+02],
 [ 6.76451996e+01, 8.19577143e+01, 9.79596394e+01],
 [ 6.70685829e+01, 8.13546935e+01, 9.75377266e+01]],
[ 7.06012503e+01, 8.37693472e+01, 1.09459176e+02 ],
 [ 7.17605974e+01, 8.44182691e+01, 1.10802612e+02],
 [ 7.43705731e+01, 8.44300841e+01, 1.10838711e+02],
 [ 9.26925942e+01, 1.07630995e+02, 1.23729697e+02],
 [ 8.79956776e+01, 1.02230739e+02, 1.17382573e+02],
 [8.36485737e+01, 9.74933710e+01, 1.13649204e+02]],
[ [ 6.46856413e+01, 7.79383971e+01, 1.03734472e+02 ], 
 [ 6.82550824e+01, 8.16151815e+01, 1.07160033e+02],
 [ 7.36874359e+01, 8.38825703e+01, 1.10112756e+02],
 [ 1.05557155e+02, 1.19267296e+02, 1.34703827e+02],
 [ 9.49215643e+01, 1.07610669e+02, 1.23198994e+02],
 [ 9.20195454e+01, 1.04707384e+02, 1.20503357e+02]]],
. . . ,
[[[ 1.22364947e+02, 1.33871473e+02, 1.41036548e+02],
 [ 1.24028187e+02, 1.35847731e+02, 1.42742173e+02],
 [ 1.18513901e+02, 1.29368840e+02, 1.37692595e+02],
 [ 9.67879140e+01, 1.17093362e+02, 1.38778732e+02],
 [ 9.43272522e+01, 1.15448412e+02, 1.38914856e+02],
 [ 9.43418889e+01, 1.15597533e+02, 1.39676513e+02]],
```

```
[[1.14262099e+02, 1.25361654e+02, 1.31607328e+02],
[ 1.11171386e+02, 1.21586938e+02, 1.28621537e+02],
 [ 1.18635977e+02, 1.28862528e+02, 1.36466992e+02],
 [ 9.51531261e+01, 1.15091103e+02, 1.36677833e+02],
 [ 9.48081531e+01, 1.14730639e+02, 1.37670140e+02],
 [ 9.37406668e+01, 1.14308637e+02, 1.37921061e+02]],
[[1.19384032e+02, 1.29491989e+02, 1.35670462e+02],
 [1.27484124e+02, 1.37915648e+02, 1.44186364e+02],
 [ 1.30613039e+02, 1.40188920e+02, 1.47420034e+02],
 . . . ,
 [ 9.64022438e+01, 1.15002268e+02, 1.37202698e+02],
 [ 9.72647434e+01, 1.15740924e+02, 1.38545372e+02],
 [ 9.65496056e+01, 1.15477297e+02, 1.38985444e+02]],
. . . ,
[[ 1.37556103e+02, 1.51616617e+02, 1.64058487e+02],
[ 1.31979932e+02, 1.46778913e+02, 1.58761238e+02],
 [ 1.31369954e+02, 1.46426280e+02, 1.58402587e+02],
 [ 3.42485687e+01, 6.50682583e+01, 8.48126358e+01],
 [ 3.35134243e+01, 6.23974151e+01, 8.27404235e+01],
 [ 3.60862778e+01, 6.37544678e+01, 8.45534576e+01]],
[[ 1.33801986e+02, 1.46986252e+02, 1.58215451e+02],
 [ 1.26544012e+02, 1.40088448e+02, 1.50978650e+02],
 [ 1.25902426e+02, 1.40169073e+02, 1.50821966e+02],
 [ 2.20679371e+01, 5.27663106e+01, 7.27668917e+01],
 [ 2.22037309e+01, 5.23477487e+01, 7.32391918e+01],
 [ 2.69844121e+01, 5.64208331e+01, 7.70482074e+01]],
[1.31786331e+02, 1.44053753e+02, 1.54880043e+02],
[ 1.26708540e+02, 1.38230152e+02, 1.49628111e+02],
 [ 1.26755358e+02, 1.38974534e+02, 1.49508161e+02],
 [ 2.17773882e+01, 5.30104957e+01, 7.37503087e+01],
 [ 2.28306984e+01, 5.31125425e+01, 7.49213729e+01],
```

```
[ 2.95816203e+01, 5.97916068e+01, 8.11078560e+01]]],
[[[ 1.83532211e+02, 1.26769359e+02, 6.69169738e+00],
 [ 1.83177423e+02, 1.26172926e+02, 5.99211177e+00],
 [ 1.85432257e+02, 1.28423610e+02, 7.51805422e+00],
 [ 8.01319718e+01, 8.96984336e+01, 9.13192043e+01],
 [ 8.38039861e+01, 9.16908533e+01, 9.43661740e+01],
 [ 8.80591039e+01, 9.32208914e+01, 9.79851523e+01]],
[[ 1.81597583e+02, 1.24964202e+02, 2.98618376e+00],
 [ 1.84076444e+02, 1.27491607e+02, 5.26280121e+00],
 [ 1.81423741e+02, 1.24802301e+02, 2.06564372e+00],
 7.55755591e+01, 8.48688561e+01, 8.81486734e+01],
 [ 7.76263717e+01, 8.50884245e+01, 9.04284298e+01],
 [ 8.47434086e+01, 8.95442136e+01, 9.51797157e+01]],
[[1.78822893e+02, 1.23178913e+02, 4.48239279e-01],
 [ 1.78416607e+02, 1.22681908e+02, -2.35121379e-01],
 [ 1.79095321e+02, 1.23291919e+02, -4.98043128e-02],
 [ 6.84862872e+01, 7.75904253e+01, 8.54856335e+01],
 [ 6.92521010e+01, 7.68370025e+01, 8.60876094e+01],
 [ 7.69532616e+01, 8.31024506e+01, 8.96889957e+01]],
. . . ,
[ 7.77269824e+01, 9.98134959e+01, 1.38986003e+02 ],
 [ 7.71021250e+01, 9.91681045e+01, 1.37493406e+02],
 [ 6.98094143e+01, 9.19537826e+01, 1.30165626e+02],
 [ 2.11546061e+01, 4.42296984e+01, 7.29547784e+01],
 [ 2.24919287e+01, 4.65549611e+01, 7.46337858e+01],
 [2.46202434e+01, 4.74338094e+01, 7.51790446e+01]],
[[ 7.59247653e+01, 9.81464103e+01, 1.36832570e+02],
 [ 7.93213196e+01, 1.01675738e+02, 1.40467307e+02],
 [ 7.60591812e+01, 9.81197040e+01, 1.36993336e+02],
  . . . ,
```

```
[ 1.85481378e+01, 4.30986315e+01, 7.12375152e+01],
 [ 2.20218496e+01, 4.55897903e+01, 7.38713808e+01],
 [ 2.41931085e+01, 4.92076664e+01, 7.64111700e+01]],
[6.80938891e+01, 9.03345185e+01, 1.29275476e+02],
 [ 7.75609621e+01, 9.97834142e+01, 1.38949626e+02],
 [7.92288907e+01, 1.01084963e+02, 1.40260649e+02],
 [1.99003514e+01, 4.62781216e+01, 7.44067065e+01],
 [ 2.13836899e+01, 4.57522673e+01, 7.43086868e+01],
 [ 2.70790144e+01, 5.20378720e+01, 8.05125880e+01]]],
[[[1.09945290e+02, 1.41847899e+02, 2.06508441e+02],
 [ 1.10617576e+02, 1.43290704e+02, 2.08147082e+02],
 [ 1.09206127e+02, 1.42774491e+02, 2.11246894e+02],
 [ 6.58647972e+01, 7.47994296e+01, 8.32717489e+01],
 [ 6.67362171e+01, 7.53093482e+01, 8.27273838e+01],
 [ 7.03839022e+01, 7.78690069e+01, 8.50694988e+01]],
[[1.11371393e+02, 1.43730309e+02, 2.07814309e+02],
 [ 1.03886910e+02, 1.37017107e+02, 2.01193389e+02],
 [ 1.11891556e+02, 1.45627181e+02, 2.12762637e+02],
  . . . ,
 [6.38254633e+01, 7.24337136e+01, 8.17126335e+01],
 [ 6.74428332e+01, 7.55160620e+01, 8.39235481e+01],
 [ 6.62809592e+01, 7.35150542e+01, 8.16415396e+01]],
[[1.08196350e+02, 1.40724406e+02, 2.04259247e+02],
 [ 1.07680950e+02, 1.40637707e+02, 2.04197037e+02],
 [ 1.08914188e+02, 1.42363527e+02, 2.08910364e+02],
 [ 6.50816449e+01, 7.31584929e+01, 8.34052619e+01],
 [ 6.63794597e+01, 7.41356206e+01, 8.35192928e+01],
 [ 6.56171200e+01, 7.26750608e+01, 8.19285308e+01]],
 . . . ,
[[ 1.89668256e+02, 1.92336082e+02, 1.88318731e+02],
 [ 1.83777636e+02, 1.86513353e+02, 1.82529116e+02],
```

```
[1.76937754e+02, 1.79603575e+02, 1.75709040e+02],
                 [ 1.15308761e+02, 1.14152901e+02, 1.41383383e+02],
                 [ 1.18586402e+02, 1.18428138e+02, 1.42493819e+02],
                 [ 1.21329960e+02, 1.21454671e+02, 1.45564049e+02]],
                [[1.90172999e+02, 1.93090716e+02, 1.89365167e+02],
                 [ 1.87192511e+02, 1.90008460e+02, 1.86269373e+02],
                 [ 1.81563960e+02, 1.84218075e+02, 1.80614684e+02],
                 [ 1.13558487e+02, 1.13283798e+02, 1.37815054e+02],
                 [ 1.15348483e+02, 1.14950425e+02, 1.39322000e+02],
                 [ 1.18311000e+02, 1.16915042e+02, 1.41329950e+02]],
                [[1.11540967e+02, 1.14522347e+02, 1.10851700e+02],
                 [1.08902684e+02, 1.11797124e+02, 1.08137305e+02],
                 [ 1.04901170e+02, 1.07462781e+02, 1.03958011e+02],
                 . . . ,
                 [ 5.71598560e+01, 5.69496450e+01, 8.15004754e+01],
                 [ 5.94006282e+01, 5.87595404e+01, 8.34306409e+01],
                 [ 6.21571109e+01, 6.06090554e+01, 8.42852770e+01]]]])
In [ ]: # Plot some of the original and reconstructed images
        n images = 5 # Number of images to plot
        plt.figure(figsize=[12, 6])
        for i in range(n images):
            plt.subplot(2, n images, i + 1)
            plt.imshow(cv2.cvtColor(images[i], cv2.COLOR BGR2RGB))
            plt.title('Original')
            plt.axis('off')
            plt.subplot(2, n images, n images + i + 1)
            reconstructed image = cv2.convertScaleAbs(images recon[i]) # Convert reconstructed image to compatible format
            plt.imshow(cv2.cvtColor(reconstructed image, cv2.COLOR BGR2RGB))
            plt.title('Reconstructed')
            plt.axis('off')
        plt.tight layout()
        plt.show()
```



In [ ]: