

# PDL Lab4.: Image corpus creation and binary classification using DNN

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```
In [49]: import os
import cv2
import numpy as np
from keras.models import Sequential
from keras.layers import Dense, Flatten
from keras.optimizers import Adam
from sklearn.model_selection import train_test_split
```

```
In [50]: barn_owl_folder = "C:/Users/2mscdsa17/Desktop/owl"
similar_images_folder = 'C:/Users/2mscdsa17/Desktop/apple'
```

```
In [51]: images = []
labels = []

# Read barn owl images
for filename in os.listdir(barn_owl_folder):
    if not filename.endswith(('.jpg', '.jpeg', '.png')):
        continue

    image_path = os.path.join(barn_owl_folder, filename)
    image = cv2.imread(image_path)
    image = cv2.resize(image, (64, 64))
    image = image / 255.0

    images.append(image)
    labels.append(0) # Label 0 for barn owls
```

```
In [52]: # Read similar images
for filename in os.listdir(similar_images_folder):
    if not filename.endswith(('.jpg', '.jpeg', '.png')):
        continue

    image_path = os.path.join(similar_images_folder, filename)
    image = cv2.imread(image_path)
    image = cv2.resize(image, (64, 64))
    image = image / 255.0

    images.append(image)
    labels.append(1) # Label 1 for similar images
```

```
In [53]: # Convert the lists to numpy arrays
images = np.array(images)
labels = np.array(labels)
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In [54]: X_train, X_test, y_train, y_test = train_test_split(images, labels, test_size=0.2)
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In [55]: # Define the model
model = Sequential()
model.add(Flatten(input_shape=(64, 64, 3)))
model.add(Dense(64, activation='relu'))
model.add(Dense(1, activation='sigmoid'))
```

```
In [56]: # Compile the model
model.compile(optimizer=Adam(learning_rate=0.001), loss='binary_crossentropy',
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```

```
In [57]: # Train the model
model.fit(X_train, y_train, epochs=10, batch_size=32)
```

```
Epoch 1/10
1/1 [=====] - 1s 551ms/step - loss: 0.7043 - accurac
y: 0.5714
Epoch 2/10
1/1 [=====] - 0s 39ms/step - loss: 5.8556 - accurac
y: 0.4286
Epoch 3/10
1/1 [=====] - 0s 7ms/step - loss: 2.2219 - accuracy:
0.5714
Epoch 4/10
1/1 [=====] - 0s 38ms/step - loss: 2.5482 - accurac
y: 0.5714
Epoch 5/10
1/1 [=====] - 0s 38ms/step - loss: 0.8106 - accurac
y: 0.5714
Epoch 6/10
1/1 [=====] - 0s 12ms/step - loss: 1.8267 - accurac
y: 0.4286
Epoch 7/10
1/1 [=====] - 0s 9ms/step - loss: 1.2678 - accuracy:
0.4286
Epoch 8/10
1/1 [=====] - 0s 40ms/step - loss: 0.2648 - accurac
y: 0.9286
Epoch 9/10
1/1 [=====] - 0s 13ms/step - loss: 1.0379 - accurac
y: 0.5714
Epoch 10/10
1/1 [=====] - 0s 7ms/step - loss: 0.9499 - accuracy:
0.5714
```

```
Out[57]: <keras.callbacks.History at 0x2597f7ecfd0>
```

```
In [34]: # Evaluate the model on test set
loss, accuracy = model.evaluate(X_test, y_test)
print("Test Loss:", loss)
print("Test Accuracy:", accuracy)
```

```
1/1 [=====] - 0s 305ms/step - loss: 0.0756 - accurac
y: 1.0000
Test Loss: 0.07557893544435501
Test Accuracy: 1.0
```

```
In [35]: # Predict class labels for test images
test_predictions = model.predict(X_test)
```

```
1/1 [=====] - 0s 58ms/step
```

```
In [36]: # Convert probabilities to class labels (0 or 1)
test_predictions = np.round(test_predictions).flatten()
```

```
In [37]: # Print the predicted labels and the actual labels
print("Predicted Labels:", test_predictions)
print("Actual Labels:", y_test)
```

```
Predicted Labels: [0. 0. 0. 0.]
Actual Labels: [0 0 0 0]
```

```
In [38]: # Save the image corpus and labels
np.save('image_corpus.npy', image_corpus)
np.save('labels.npy', labels)
```

```
In [39]: # Load the image corpus and labels
image_corpus = np.load('image_corpus.npy')
labels = np.load('labels.npy')

# Print the shapes and contents of the loaded arrays
print("Image Corpus shape:", image_corpus.shape)
print("Image Corpus:")
print(image_corpus)

print("\nLabels shape:", labels.shape)
print("Labels:")
print(labels)
```

```
Image Corpus shape: (18, 1)
```

```
Image Corpus:
```

```
[[0.54886806]
 [0.5354678 ]
 [0.5491303 ]
 [0.5677061 ]
 [0.550068 ]
 [0.52372587]
 [0.5308342 ]
 [0.5455766 ]
 [0.55411303]
 [0.55990237]
 [0.5908376 ]
 [0.594258 ]
 [0.5926563 ]
 [0.5691223 ]
 [0.6011374 ]
 [0.58412766]
 [0.60285765]
 [0.6070214 ]]
```

```
Labels shape: (18,)
```

```
Labels:
```

```
[0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1]
```

In [ ]: