Assignment -3 Build CNN model for classification of Flowers

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Project Name	AI BASED DISCOURSE FOR BANKING INDUSTRY
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Question-1. Load the dataset

Solution:

!unzip Flowers-Dataset.zip

```
inflating: flowers/daisy/1396526833 fb867165be n.jpg
inflating: flowers/daisy/13977181862_f8237b6b52.jpg
inflating: flowers/daisy/14021430525_e06baf93a9.jpg
inflating: flowers/daisy/14073784469 ffb12f3387 n.jpg
inflating: flowers/daisy/14087947408 9779257411 n.jpg
inflating: flowers/daisy/14088053307_1a13a0bf91_n.jpg
inflating: flowers/daisy/14114116486_0bb6649bc1_m.jpg
inflating: flowers/daisy/14147016029_8d3cf2414e.jpg
inflating: flowers/daisy/14163875973_467224aaf5_m.jpg
inflating: flowers/daisy/14167534527_781ceb1b7a_n.jpg
inflating: flowers/daisy/14167543177 cd36b54ac6 n.jpg
inflating: flowers/daisy/14219214466 3ca6104eae m.jpg
inflating: flowers/daisy/14221836990_90374e6b34.jpg
inflating: flowers/daisy/14221848160_7f0a37c395.jpg
inflating: flowers/daisy/14245834619_153624f836.jpg
inflating: flowers/daisy/14264136211 9531fbc144.jpg
inflating: flowers/daisy/14272874304 47c0a46f5a.jpg
inflating: flowers/daisy/14307766919 fac3c37a6b m.jpg
inflating: flowers/daisy/14330343061_99478302d4_m.jpg
inflating: flowers/daisy/14332947164_9b13513c71_m.jpg
inflating: flowers/daisy/14333681205 a07c9f1752 m.jpg
inflating: flowers/daisy/14350958832_29bdd3a254.jpg
inflating: flowers/daisy/14354051035_1037b30421_n.jpg
inflating: flowers/daisy/14372713423_61e2daae88.jpg
inflating: flowers/daisy/14399435971_ea5868c792.jpg
inflating: flowers/daisy/14402451388 56545a374a n.jpg
inflating: flowers/daisy/144076848_57e1d662e3_m.jpg
```

```
inflating: flowers/daisy/14372713423_61e2daae88.jpg
inflating: flowers/daisy/14399435971_ea5868c792.jpg
inflating: flowers/daisy/14402451388 56545a374a n.jpg
inflating: flowers/daisy/144076848 57e1d662e3 m.jpg
inflating: flowers/daisy/144099102 bf63a41e4f n.jpg
inflating: flowers/daisy/1441939151 b271408c8d n.jpg
inflating: flowers/daisy/14421389519_d5fd353eb4.jpg
inflating: flowers/daisy/144603918_b9de002f60_m.jpg
inflating: flowers/daisy/14471433500 cdaa22e3ea m.jpg
inflating: flowers/daisy/14485782498_fb342ec301.jpg
inflating: flowers/daisy/14507818175_05219b051c_m.jpg
inflating: flowers/daisy/14523675369_97c31d0b5b.jpg
inflating: flowers/daisy/14551098743_2842e7a004_n.jpg
inflating: flowers/daisy/14554906452_35f066ffe9_n.jpg
inflating: flowers/daisy/14564545365 1f1d267bf1 n.jpg
inflating: flowers/daisy/14569895116_32f0dcb0f9.jpg
inflating: flowers/daisy/14591326135 930703dbed m.jpg
inflating: flowers/daisy/14600779226_7bbc288d40_m.jpg
inflating: flowers/daisy/14613443462_d4ed356201.jpg
inflating: flowers/daisy/14621687774 ec52811acd n.jpg
inflating: flowers/daisy/14674743211_f68b13f6d9.jpg
inflating: flowers/daisy/14698531521_0c2f0c6539.jpg
inflating: flowers/daisy/147068564_32bb4350cc.jpg
inflating: flowers/daisy/14707111433_cce08ee007.jpg
inflating: flowers/daisy/14716799982 ed6d626a66.jpg
inflating: flowers/daisy/14816364517_2423021484_m.jpg
inflating: flowers/daisy/14866200659_6462c723cb_m.jpg
```

```
#importing required libraries to build a CNN classification model with accuracy
import numpy as np
import tensorflow as tf
from tensorflow.keras import layers
from tensorflow.keras.models import Sequential
import matplotlib.pyplot as plt
batch_size = 32
img_height = 180
img_width = 180
data_dir = "/content/flowers"
```

Solution:

 $from\ tensor flow. keras. preprocessing. image\ import\ Image Data Generator$

 $train_datagen = ImageDataGenerator(rescale = 1./255, horizontal_flip = True, vertical_flip = True, z \\ oom_range = 0.2)$

 $x_train = train_datagen.flow_from_directory(r''/content/flowers'', target_size = (64,64) \ , class_mode = ''categorical'', batch_size = 100)$

Found 4317 images belonging to 5 classes.

```
#Image Augumentation accuracy
data_augmentation = Sequential(
   [
        layers.RandomFlip("horizontal",input_shape=(img_height, img_width, 3)),
        layers.RandomRotation(0.1),
        layers.RandomZoom(0.1),
    ]
)
```

Question-3. Create model - Model Building and also Split dataset into training and testing sets

Solution:

from tensorflow.keras.models import Sequential

from tensorflow.keras.layers import Convolution2D,MaxPooling2D,Flatten,Dense model = Sequential()

```
train_ds = tf.keras.utils.image_dataset_from_directory(
  data_dir,
  validation_split=0.2,
  subset="training",
  seed=123,
  image_size=(img_height, img_width),
  batch_size=batch_size)
```

```
Found 4317 files belonging to 5 classes.
    Using 3454 files for training.
val_ds = tf.keras.utils.image_dataset_from_directory(
data_dir,
validation_split=0.2,
subset="validation",
seed=123,
image_size=(img_height, img_width),
batch_size=batch_size)
 Found 4317 files belonging to 5 classes.
 Using 863 files for validation.
class_names = train_ds.class_names
print(class_names)
['daisy', 'dandelion', 'rose', 'sunflower', 'tulip']
plt.figure(figsize=(10, 10))
for images, labels in train_ds.take(1):
for i in range(9):
 ax = plt.subplot(3, 3, i + 1)
 plt.imshow(images[i].numpy().astype("uint8"))
 plt.title(class_names[labels[i]])
```

plt.axis("off")



Question-4. Add the layers (Convolution, MaxPooling, Flatten, Dense-(HiddenLayers), Output)

Solution:

```
model.add(Convolution2D(32, (3,3), activation = "relu", input_shape = (64,64,3) ))
model.add(MaxPooling2D(pool_size = (2,2)))
model.add(Flatten())
model.add(Dense(300, activation = "relu"))
model.add(Dense(150, activation = "relu")) #mulitple dense layers
model.add(Dense(5, activation = "softmax")) #output layer
```

```
#Adding the layers for accuracy
num_classes = len(class_names)

model = Sequential([
   data_augmentation,
   layers.Rescaling(1./255, input_shape=(img_height, img_width, 3)),
   layers.Conv2D(16, 3, padding='same', activation='relu'),
   layers.MaxPooling2D(),
   layers.Conv2D(32, 3, padding='same', activation='relu'),
   layers.MaxPooling2D(),
   layers.Conv2D(64, 3, padding='same', activation='relu'),
   layers.MaxPooling2D(),
   layers.Flatten(),
   layers.Dense(128, activation='relu'),
   layers.Dense(num_classes)
])
```

Question-5. Compile The Model

Solution:

```
model.compile(loss = "categorical_crossentropy", metrics = ["accuracy"], optimizer = "adam") len(x_train)
```

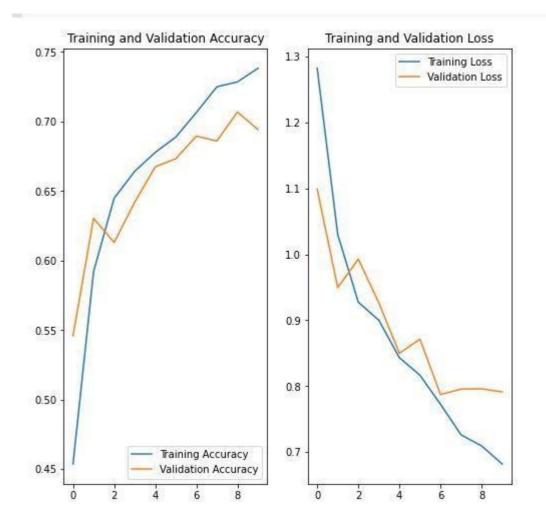
44

#Compile the model for further accuracy

```
Epoch 1/10
108/108 [==
                                  ===] - 132s 1s/step - loss: 1.2821 - accuracy: 0.4537 - val_loss: 1.0988 - val_accuracy: 0.5458
   Epoch 2/10
                                  ==] - 130s 1s/step - loss: 1.0298 - accuracy: 0.5921 - val_loss: 0.9494 - val_accuracy: 0.6304
   Epoch 3/10
   108/108 [==
Epoch 4/10
                            ======] - 129s 1s/step - loss: 0.9274 - accuracy: 0.6448 - val_loss: 0.9927 - val_accuracy: 0.6130
                             =======] - 129s 1s/step - loss: 0.9000 - accuracy: 0.6642 - val_loss: 0.9264 - val_accuracy: 0.6419
   Epoch 5/10
   108/108 [==
Epoch 6/10
                                  ==] - 136s 1s/step - loss: 0.8432 - accuracy: 0.6778 - val_loss: 0.8499 - val_accuracy: 0.6674
   108/108 [==
Epoch 7/10
                            =======] - 130s 1s/step - loss: 0.8166 - accuracy: 0.6888 - val_loss: 0.8714 - val_accuracy: 0.6732
   108/108 [==
                             =======] - 130s 1s/step - loss: 0.7726 - accuracy: 0.7064 - val_loss: 0.7873 - val_accuracy: 0.6895
                         ========] - 130s 1s/step - loss: 0.7262 - accuracy: 0.7250 - val_loss: 0.7957 - val_accuracy: 0.6860
   108/108 [==
                             108/108 [==
```

#To find the Training and Validation- Accuracy & Loss (Visualization)

```
acc = history.history['accuracy']
val_acc = history.history['val_accuracy']
loss = history.history['loss']
val_loss = history.history['val_loss']
epochs_range = range(epochs)
plt.figure(figsize=(8, 8))
plt.subplot(1, 2, 1)
plt.plot(epochs_range, acc, label='Training Accuracy')
plt.plot(epochs_range, val_acc, label='Validation Accuracy')
plt.legend(loc='lower right')
plt.title('Training and Validation Accuracy')
plt.subplot(1, 2, 2)
plt.plot(epochs_range, loss, label='Training Loss')
plt.plot(epochs_range, val_loss, label='Validation Loss')
plt.legend(loc='upper right')
plt.title('Training and Validation Loss')
plt.show()
```



Question-6. Fit The Model

Solution:

model.fit(x_train, epochs = 15, steps_per_epoch = len(x_train))

```
Epoch 1/15
   44/44 [=========== ] - 31s 684ms/step - loss: 1.7914 - accuracy: 0.3588
   Epoch 2/15
   44/44 [============ ] - 29s 648ms/step - loss: 1.1730 - accuracy: 0.5045
   Epoch 3/15
   44/44 [=========== ] - 29s 650ms/step - loss: 1.0967 - accuracy: 0.5529
   Epoch 4/15
   44/44 [=========== ] - 29s 648ms/step - loss: 1.0351 - accuracy: 0.5939
   Epoch 5/15
   44/44 [============ - 29s 645ms/step - loss: 0.9920 - accuracy: 0.6127
   Epoch 6/15
   44/44 [============] - 30s 677ms/step - loss: 0.9659 - accuracy: 0.6259
   Epoch 7/15
   44/44 [============ ] - 29s 648ms/step - loss: 0.9129 - accuracy: 0.6426
   Epoch 8/15
   44/44 [========== ] - 29s 647ms/step - loss: 0.9085 - accuracy: 0.6433
   Epoch 9/15
   44/44 [============ ] - 32s 717ms/step - loss: 0.8597 - accuracy: 0.6620
   Epoch 10/15
   44/44 [============ ] - 30s 674ms/step - loss: 0.8350 - accuracy: 0.6824
   Epoch 11/15
   44/44 [=========== - 295 648ms/step - loss: 0.8420 - accuracy: 0.6718
   Epoch 12/15
   44/44 [===========] - 29s 650ms/step - loss: 0.7857 - accuracy: 0.7030
   Epoch 13/15
   44/44 [======== - 29s 649ms/step - loss: 0.7868 - accuracy: 0.7000
   Epoch 14/15
   44/44 [============= ] - 29s 650ms/step - loss: 0.7542 - accuracy: 0.7132
   Epoch 15/15
   44/44 [=========== ] - 30s 676ms/step - loss: 0.7467 - accuracy: 0.7107
   <keras.callbacks.History at 0x7f602ce90090>
```

Question-7. Save The Model

Solution:

model.save("flowers.h1")

model.save("flowers.m5")#another model to show the accuracy

Question-8. Test The Model

Solution:

from tensorflow.keras.models import load_model from tensorflow.keras.preprocessing import image import numpy as np

```
model = load_model("/content/flowers.h1")
# Testing with a random rose image from Google
img = image.load_img("/content/rose.gif", target_size = (64,64) )
img
x = image.img_to_array(img)
x.ndim
 3
x = np.expand_dims(x,axis = 0)
x.ndim
4
pred = model.predict(x)
pred
 array([[0., 0., 1., 0., 0.]], dtype=float32)
labels = ['daisy','dandelion','roses','sunflowers','tulips']
labels[np.argmax(pred)]
'roses'
```

```
sunflower_url = "https://storage.googleapis.com/download.tensorflow.org/example_images/592
px-Red_sunflower.jpg"
sunflower_path = tf.keras.utils.get_file('Red_sunflower', origin=sunflower_url)
img = tf.keras.utils.load_img(
  sunflower_path, target_size=(img_height, img_width)
)
img_array = tf.keras.utils.img_to_array(img)
img_array = tf.expand_dims(img_array, 0) # Create a batch
predictions = model.predict(img_array)
score = tf.nn.softmax(predictions[0])
print(
  "This image most likely belongs to {} with a {:.2f} percent confidence."
  .format(class_names[np.argmax(score)], 100 * np.max(score))
)
 {\tt Downloading\ data\ from\ \underline{https://storage.googleapis.com/download.tensorflow.org/example\_images/592px-Red\_sunflower.jpg}}
 122880/117948 [=========] - 0s Ous/step
 131072/117948 [===========] - 0s Ous/step
 This image most likely belongs to sunflower with a 99.85 percent confidence.
```