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# Assignment 3 - Build CNN Model For Classification Of Flowers

# → 1. Unzip dataset

#### !unzip '/content/Flowers-Dataset.zip'

```
Archive: /content/Flowers-Dataset.zip
  inflating: flowers/daisy/100080576 f52e8ee070 n.jpg
  inflating: flowers/daisy/10140303196_b88d3d6cec.jpg
  inflating: flowers/daisy/10172379554_b296050f82_n.jpg
  inflating: flowers/daisy/10172567486_2748826a8b.jpg
  inflating: flowers/daisy/10172636503_21bededa75_n.jpg
  inflating: flowers/daisy/102841525_bd6628ae3c.jpg
  inflating: flowers/daisy/10300722094 28fa978807 n.jpg
  inflating: flowers/daisy/1031799732_e7f4008c03.jpg
  inflating: flowers/daisy/10391248763_1d16681106_n.jpg
  inflating: flowers/daisy/10437754174_22ec990b77_m.jpg
  inflating: flowers/daisy/10437770546_8bb6f7bdd3_m.jpg
  inflating: flowers/daisy/10437929963 bc13eebe0c.jpg
  inflating: flowers/daisy/10466290366_cc72e33532.jpg
  inflating: flowers/daisy/10466558316_a7198b87e2.jpg
  inflating: flowers/daisy/10555749515_13a12a026e.jpg
  inflating: flowers/daisy/10555815624_dc211569b0.jpg
  inflating: flowers/daisy/10555826524 423eb8bf71 n.jpg
  inflating: flowers/daisy/10559679065 50d2b16f6d.jpg
  inflating: flowers/daisy/105806915_a9c13e2106_n.jpg
  inflating: flowers/daisy/10712722853_5632165b04.jpg
  inflating: flowers/daisy/107592979_aaa9cdfe78_m.jpg
  inflating: flowers/daisy/10770585085 4742b9dac3 n.jpg
  inflating: flowers/daisy/10841136265 af473efc60.jpg
  inflating: flowers/daisy/10993710036 2033222c91.jpg
  inflating: flowers/daisy/10993818044 4c19b86c82.jpg
  inflating: flowers/daisy/10994032453_ac7f8d9e2e.jpg
  inflating: flowers/daisy/11023214096 b5b39fab08.jpg
  inflating: flowers/daisy/11023272144 fce94401f2 m.jpg
  inflating: flowers/daisy/11023277956_8980d53169_m.jpg
  inflating: flowers/daisy/11124324295_503f3a0804.jpg
  inflating: flowers/daisy/1140299375_3aa7024466.jpg
  inflating: flowers/daisy/11439894966_dca877f0cd.jpg
  inflating: flowers/daisy/1150395827_6f94a5c6e4_n.jpg
  inflating: flowers/daisy/11642632 1e7627a2cc.jpg
  inflating: flowers/daisy/11834945233 a53b7a92ac m.jpg
  inflating: flowers/daisy/11870378973_2ec1919f12.jpg
```

```
inflating: flowers/daisy/11891885265_ccefec7284_n.jpg
inflating: flowers/daisy/12193032636_b50ae7db35_n.jpg
inflating: flowers/daisy/12348343085 d4c396e5b5 m.jpg
inflating: flowers/daisy/12585131704 0f64b17059 m.jpg
inflating: flowers/daisy/12601254324_3cb62c254a_m.jpg
inflating: flowers/daisy/1265350143_6e2b276ec9.jpg
inflating: flowers/daisy/12701063955_4840594ea6_n.jpg
inflating: flowers/daisy/1285423653_18926dc2c8_n.jpg
inflating: flowers/daisy/1286274236 1d7ac84efb n.jpg
inflating: flowers/daisy/12891819633_e4c82b51e8.jpg
inflating: flowers/daisy/1299501272_59d9da5510_n.jpg
inflating: flowers/daisy/1306119996_ab8ae14d72_n.jpg
inflating: flowers/daisy/1314069875_da8dc023c6_m.jpg
inflating: flowers/daisy/1342002397 9503c97b49.jpg
inflating: flowers/daisy/134409839_71069a95d1_m.jpg
inflating: flowers/daisy/1344985627_c3115e2d71_n.jpg
inflating: flowers/daisy/13491959645_2cd9df44d6_n.jpg
inflating: flowers/daisy/1354396826_2868631432_m.jpg
inflating: flowers/daisy/1355787476_32e9f2a30b.jpg
inflating: flowers/daisy/13583238844_573df2de8e_m.jpg
inflating: flowers/daisv/1374193928 a52320eafa.ipg
```

# Importing Necessary Libraries

```
import warnings
warnings.filterwarnings("ignore")

import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense,Activation,Dropout,Conv2D,Flatten,MaxPool2D,Resh
from tensorflow.keras.applications.resnet50 import ResNet50
from tensorflow.keras.applications.resnet50 import preprocess_input
from tensorflow.keras.preprocessing import image
from tensorflow.keras.preprocessing.image import ImageDataGenerator,load_img,img_to_array
from tensorflow.keras.callbacks import EarlyStopping, ReduceLROnPlateau
```

# → 2. Image Augumentation

- · Dataset consist of 5 classes.
- **Daisy** European Species of Aster family.
- Sunflower Identified as the genus of Helianthus.
- Tulip It belongs to the species of spring blooming geophytes.
- Rose It belongs to the family of rosaceae.
- Dandelion Indentifies as the genus of Asterceae.

```
path = 'flowers/'
```

```
train_data_gen = ImageDataGenerator(rescale = 1./255,
                             shear_range = 0.2,
                             zoom_range = 0.2,
                             horizontal flip = True,
                             validation_split = 0.30)
test_data_gen = ImageDataGenerator(rescale = 1./255, validation_split = 0.30)
training_set = train_data_gen.flow_from_directory(path,
                                                  target_size=(64,64),
                                                  batch_size=100,
                                                  class_mode='categorical',
                                                  shuffle=True,
                                                  color_mode='rgb',
                                                  subset = 'training')
testing_set = test_data_gen.flow_from_directory(path,
                                                  target_size=(64,64),
                                                  batch_size=100,
                                                  class_mode='categorical',
                                                  shuffle=True,
                                                  color_mode='rgb',
                                                  subset = 'validation')
     Found 3024 images belonging to 5 classes.
     Found 1293 images belonging to 5 classes.
```

### → 3. Create the model

```
model = Sequential()
```

# 4. Add Layers (Convolution, MaxPooling, Flatten, Dense-Hidden Layers, Output)

```
#convolution and Pooling layer 1
model.add(Conv2D(filters=48,kernel_size=3,activation='relu',input_shape=(64,64,3)))
model.add(MaxPool2D(pool_size=2,strides=2))
model.add(Dropout(0.2))

#convolution and Pooling layer 2
model.add(Conv2D(filters=32,kernel_size=3,activation='relu'))
model.add(MaxPool2D(pool_size=2,strides=2))
model.add(Dropout(0.2))

#Flattening the images
model.add(Flatten())
```

```
#Fully Connected layers
model.add(Dense(64,activation='relu'))
model.add(Dropout(0.2))
model.add(Dense(5,activation='softmax'))
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 62, 62, 48)	1344
<pre>max_pooling2d (MaxPooling2D )</pre>	(None, 31, 31, 48)	0
dropout (Dropout)	(None, 31, 31, 48)	0
conv2d_1 (Conv2D)	(None, 29, 29, 32)	13856
<pre>max_pooling2d_1 (MaxPooling 2D)</pre>	(None, 14, 14, 32)	0
dropout_1 (Dropout)	(None, 14, 14, 32)	0
flatten (Flatten)	(None, 6272)	0
dense (Dense)	(None, 64)	401472
dropout_2 (Dropout)	(None, 64)	0
dense_1 (Dense)	(None, 5)	325
=======================================		

Total params: 416,997 Trainable params: 416,997 Non-trainable params: 0

# → 5. Compiling the Model

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

# → 6. Fitting the Model

```
early_stop = EarlyStopping(monitor='val_accuracy',
                           patience=5, verbose=1, mode='auto')
lr = ReduceLROnPlateau(monitor='val_accuracy',
                       factor=0.2,patience=5,
                       min_lr=0.00001)
```

#### Training the Model

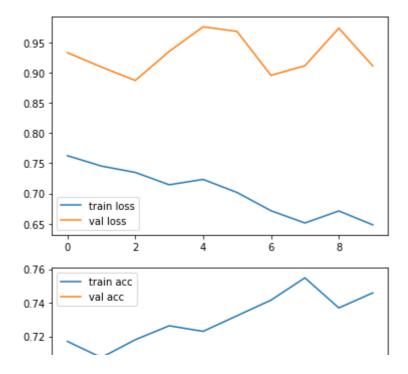
```
result = model.fit(x=training_set, validation_data=testing_set, epochs=10)
```

```
Epoch 1/10
Epoch 2/10
31/31 [============== ] - 30s 969ms/step - loss: 0.7454 - accuracy: 0
Epoch 3/10
31/31 [============== ] - 31s 985ms/step - loss: 0.7348 - accuracy: 0
Epoch 4/10
31/31 [============== ] - 30s 968ms/step - loss: 0.7144 - accuracy: 0
Epoch 5/10
31/31 [============== ] - 31s 992ms/step - loss: 0.7233 - accuracy: 0
Epoch 6/10
Epoch 7/10
Epoch 8/10
Epoch 9/10
Epoch 10/10
```

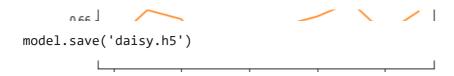
#### Loss and Accuracy check using plot

```
#plot the loss
plt.plot(result.history['loss'], label='train loss')
plt.plot(result.history['val_loss'], label='val loss')
plt.legend()
plt.show()

# plot the accuracy
plt.plot(result.history['accuracy'], label='train acc')
plt.plot(result.history['val_accuracy'], label='val acc')
plt.legend()
plt.show()
```



# → 7. Saving the Model



# ▼ 8. Testing the Model

```
training_set.class_indices

classes = ['Daisy','Dandelion','Rose','Sunflower','Tulip']

def testing(img):
    img = image.load_img(img,target_size=(64,64))
    x = image.img_to_array(img)
    x = np.expand_dims(x,axis=0)
    pred = np.argmax(model.predict(x))
    return print("Predicted class as:",classes[pred])

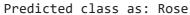
def img_show(img):
    img1 = image.load_img(img,target_size=(64,64))
    plt.imshow(img1)

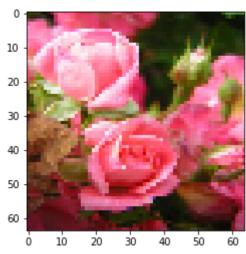
#test1
img_show('/content/flowers/sunflower/12471443383_b71e7a7480_m.jpg')
testing('/content/flowers/sunflower/12471443383_b71e7a7480_m.jpg')
```

#### Predicted class as: Sunflower



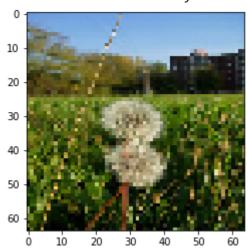
#test2
img\_show('/content/flowers/rose/323872063\_7264e7e018\_m.jpg')
testing('/content/flowers/rose/323872063\_7264e7e018\_m.jpg')





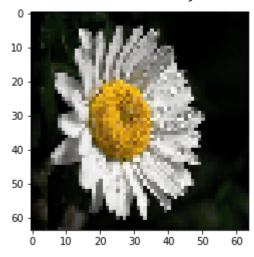
#test3
img\_show('/content/flowers/dandelion/2116997627\_30fed84e53\_m.jpg')
testing('/content/flowers/dandelion/2116997627\_30fed84e53\_m.jpg')

#### Predicted class as: Daisy



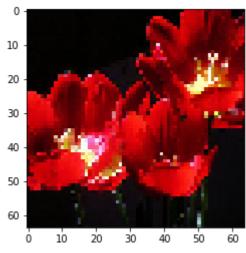
#test4
img\_show('/content/flowers/daisy/1314069875\_da8dc023c6\_m.jpg')
testing('/content/flowers/daisy/1314069875\_da8dc023c6\_m.jpg')

Predicted class as: Daisy



#test5
img\_show('/content/flowers/tulip/132538273\_335240fe5b\_n.jpg')
testing('/content/flowers/tulip/132538273\_335240fe5b\_n.jpg')

Predicted class as: Tulip



# **Conclusion:**

#### The dataset has about 4317 images from 5 different classes.

- 1. Each classes have more than 500 images for training the data.
- 2. 30% of the data taken for validation.
- 3. The accuracy of the model is around 80%.
- 4. The validation accuracy is around 70%.
- 5. The model is built with 2 layered convolutional network considering 1344 trainable parameters.
- 6. Testing the model with unknown images gives 95% accuracy.