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Assignment3-BuildCNNModelForClassificationOfFlowers

▼ 1. Unzipdataset

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

```
Archive:/content/Flowers-Dataset.zip
  inflating:
flowers/daisy/100080576_f52e8ee070_n.jpginflatinf:f
lowers/daisy/10140303196_b88d3d6cec.jpg
  inflating:
flowers/daisy/10172379554_b296050f82_n.jpginflatinf:
flowers/daisy/10172567486_2748826a8b.jpg
  inflating:
flowers/daisy/10172636503_21bededa75_n.jpginflatinf:
flowers/daisy/102841525_bd6628ae3c.jpg
  inflating: flowers/daisy/10300722094_28fa978807_n.jpg
  inflating: flowers/daisy/1031799732_e7f4008c03.jpg
  inflating:
flowers/daisy/10391248763_1d16681106_n.jpginflatinf:
flowers/daisy/10437754174_22ec990b77_m.jpginflatinf:
flowers/daisy/10437770546_8bb6f7bdd3_m.jpginflatinf:
flowers/daisy/10437929963_bc13eebe0c.jpg
  inflating:
flowers/daisy/10466290366_cc72e33532.jpginflatinf:
flowers/daisy/10466558316_a7198b87e2.jpginflatinf:
flowers/daisy/10555749515_13a12a026e.jpginflatinf:f
lowers/daisy/10555815624_dc211569b0.jpg
  inflating:
flowers/daisy/10555826524_423eb8bf71_n.jpginflatinf:
flowers/daisy/10559679065_50d2b16f6d.jpg
  inflating:
flowers/daisy/105806915_a9c13e2106_n.jpginflatinf:
flowers/daisy/10712722853_5632165b04.jpginflatinf:f
lowers/daisy/107592979_aaa9cdfef78_m.jpg
  inflating:
flowers/daisy/10770585085_4742b9dac3_n.jpginflatinf:
flowers/daisy/10841136265_af473efc60.jpg
  inflating: flowers/daisy/10993710036_2033222c91.jpg
  inflating:
flowers/daisy/10993818044_4c19b86c82.jpginflatinf:
flowers/daisy/10994032453_ac7f8d9e2e.jpginflatinf:f
lowers/daisy/11023214096_b5b39fab08.jpg
  inflating:
flowers/daisy/11023272144_fce94401f2_m.jpginflatinf:
```

flowers/daisy/11023277956_8980d53169_m.jpginflating:
flowers/daisy/11124324295_503f3a0804.jpg
inflating:
flowers/daisy/1140299375_3aa7024466.jpginflating:fl
owers/daisy/11439894966_dca877f0cd.jpg
inflating:
flowers/daisy/1150395827_6f94a5c6e4_n.jpginflating:
flowers/daisy/11642632_1e7627a2cc.jpg
inflating: flowers/daisy/11834945233_a53b7a92ac_m.jpg
inflating: flowers/daisy/11870378973_2ec1919f12.jpg

```

inflating:
flowers/daisy/11891885265_ccefec7284_n.jpginflating:
flowers/daisy/12193032636_b50ae7db35_n.jpginflating:
flowers/daisy/12348343085_d4c396e5b5_m.jpginflating:
flowers/daisy/12585131704_0f64b17059_m.jpginflating:
flowers/daisy/12601254324_3cb62c254a_m.jpginflating:
flowers/daisy/1265350143_6e2b276ec9.jpg
inflating:
flowers/daisy/12701063955_4840594ea6_n.jpginflating:
flowers/daisy/1285423653_18926dc2c8_n.jpginflating:
flowers/daisy/1286274236_1d7ac84efb_n.jpginflating:
flowers/daisy/12891819633_e4c82b51e8.jpg
inflating:
flowers/daisy/1299501272_59d9da5510_n.jpginflating:
flowers/daisy/1306119996_ab8ae14d72_n.jpginflating:
flowers/daisy/1314069875_da8dc023c6_m.jpginflating:
flowers/daisy/1342002397_9503c97b49.jpg
inflating:
flowers/daisy/134409839_71069a95d1_m.jpginflating:flowers/daisy/1344985627_c3115e2d71_n.jpg
inflating:
flowers/daisy/13491959645_2cd9df44d6_n.jpginflating:
flowers/daisy/1354396826_2868631432_m.jpginflating:
flowers/daisy/1355787476_32e9f2a30b.jpg
inflating:
flowers/daisy/13583238844_573df2de8e_m.jpginflating:
flowers/daisy/1374193928_a52320eafa.jpg

```

▼ 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, Reshape
from tensorflow.keras.applications import ResNet50
from tensorflow.keras.preprocessing import image
from tensorflow.keras.callbacks import EarlyStopping, ReduceLROnPlateau

```

▼ 2. Image Augmentation

- Dataset consists 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**-IdentifiesasthegenusofAsterceae.

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.Found1293imagesbelongingto5classes.

```

▼ 3. Create the model

```
model = Sequential()
```

▼ 4. Add Layers(Convolution,Max Pooling,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
imagesmodel.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
max_pooling2d (MaxPooling2D)	(None, 31, 31, 48)	0
dropout (Dropout)	(None, 31, 31, 48)	0
conv2d_1 (Conv2D)	(None, 29, 29, 32)	13856
max_pooling2d_1 (MaxPooling2D)	(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)
```

```
callback = [early_stop,lr]
```

Training the Model

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

```
Epoch 1/10
31/31 [=====] - 30s 966ms/step - loss: 0.7625 - accuracy: 0
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
31/31 [=====] - 32s 1s/step - loss: 0.7017 - accuracy: 0.73
Epoch 7/10
31/31 [=====] - 30s 963ms/step - loss: 0.6715 - accuracy: 0
Epoch 8/10
31/31 [=====] - 31s 978ms/step - loss: 0.6512 - accuracy: 0
Epoch 9/10
31/31 [=====] - 31s 982ms/step - loss: 0.6711 - accuracy: 0
Epoch 10/10
31/31 [=====] - 30s 974ms/step - loss: 0.6481 - accuracy: 0
```

Loss and Accuracy check using plot

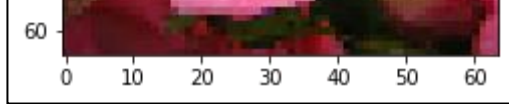
```
#plot the loss
plt.plot(result.history['loss'],label='trainloss')
plt.plot(result.history['val_loss'], label='val
loss')plt.legend()
plt.show()

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

▼ 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  
    how(img1)  
  
# test 1  
img_show('/content/flowers/sunflower/12471443383_b71e7a7480_m.jpg')  
testing('/content/flowers/sunflower/12471443383_b71e7a7480_m.jpg')
```

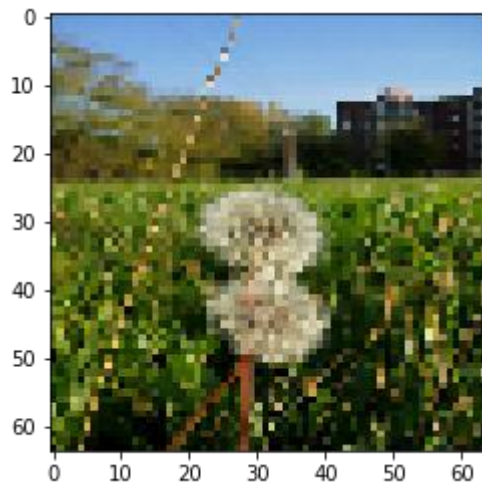



```
#test3
```

```
img_show('/content/flowers/dandelion/2116997627_30fed84e53_m.jpg')
```

```
testing('/content/flowers/dandelion/2116997627_30fed84e53_m.jpg')
```

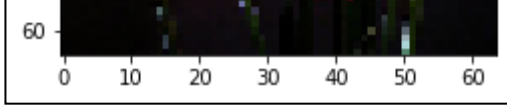
Predicted class as: Daisy



```
#test4
```

```
img_show('/content/flowers/daisy/1314069875_da8dc023c6_m.jpg')
```

```
testing('/content/flowers/daisy/1314069875_da8dc023c6_m.jpg')
```



Conclusion:

The dataset has about 4317 images from 5 different classes.

1. Each class has 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.