from google.colab import drive
drive.mount('/content/drive')

Mounted at /content/drive

!unzip '/content/drive/MyDrive/Flowers-Dataset.zip'

```
inflating: flowers/tulip/8668973377 c69527db42 m.jpg
Г⇒
      inflating: flowers/tulip/8668974855 8389ecbdca m.jpg
      inflating: flowers/tulip/8669794378 97dda6036f n.jpg
      inflating: flowers/tulip/8673412732 f8fd690ee4 n.jpg
      inflating: flowers/tulip/8673416166_620fc18e2f_n.jpg
      inflating: flowers/tulip/8673416556_639f5c88f1_n.jpg
      inflating: flowers/tulip/8677713853 1312f65e71.jpg
      inflating: flowers/tulip/8681825637_837a63513a_n.jpg
      inflating: flowers/tulip/8686013485 3c4dfbfd1f n.jpg
      inflating: flowers/tulip/8686332852_c6dcb2e86b.jpg
      inflating: flowers/tulip/8687675254_c93f50d8b0_m.jpg
      inflating: flowers/tulip/8688502760 1c8d6de921 m.jpg
      inflating: flowers/tulip/8689672277_b289909f97_n.jpg
      inflating: flowers/tulip/8690789564 394eb04982 n.jpg
      inflating: flowers/tulip/8690791226 b1f015259f n.jpg
      inflating: flowers/tulip/8695367666 0809529eaf n.jpg
      inflating: flowers/tulip/8695372372 302135aeb2.jpg
      inflating: flowers/tulip/8697784345 e75913d220.jpg
      inflating: flowers/tulip/8702982836 75222725d7.jpg
      inflating: flowers/tulip/8706523526 a0f161b72b.jpg
      inflating: flowers/tulip/8708209606 d3aede4801.jpg
      inflating: flowers/tulip/8708856019 f3be2353a4 n.jpg
      inflating: flowers/tulip/8710148289_6fc196a0f8_n.jpg
      inflating: flowers/tulip/8711277462 b43df5454b m.jpg
      inflating: flowers/tulip/8712230357 1298b8513b.jpg
      inflating: flowers/tulip/8712243901_54d686319e_m.jpg
      inflating: flowers/tulip/8712244311_da8e90bf8e_n.jpg
      inflating: flowers/tulip/8712260079 c0ff42e0e2 n.jpg
      inflating: flowers/tulip/8712263493 3db76c5f82.jpg
      inflating: flowers/tulip/8712266605_3787e346cd_n.jpg
      inflating: flowers/tulip/8712267391 c756f18ee7 n.jpg
      inflating: flowers/tulip/8712267813_f7a9be2ec5.jpg
      inflating: flowers/tulip/8712268519 f4c2c39a06 n.jpg
      inflating: flowers/tulip/8712269349 2b933da2b8 n.jpg
      inflating: flowers/tulip/8712270243_8512cf4fbd.jpg
      inflating: flowers/tulip/8712270665 57b5bda0a2 n.jpg
      inflating: flowers/tulip/8712282563 3819afb7bc.jpg
      inflating: flowers/tulip/8713357842_9964a93473_n.jpg
      inflating: flowers/tulip/8713387500 6a9138b41b n.jpg
      inflating: flowers/tulip/8713388322 e5ae26263b n.jpg
      inflating: flowers/tulip/8713389178_66bceb71a8_n.jpg
      inflating: flowers/tulip/8713390684 041148dd3e n.jpg
      inflating: flowers/tulip/8713391394_4b679ea1e3_n.jpg
      inflating: flowers/tulip/8713392604_90631fb809_n.jpg
      inflating: flowers/tulip/8713394070 b24561b0a9.jpg
      inflating: flowers/tulip/8713396140_5af8136136.jpg
```

```
intrating: trowers/turip/&/r33a/32a_n2a2ccnr/p_n.jpg
       inflating: flowers/tulip/8713397694 bcbcbba2c2 n.jpg
       inflating: flowers/tulip/8713398114 bc96f1b624 n.jpg
       inflating: flowers/tulip/8713398614_88202e452e_n.jpg
       inflating: flowers/tulip/8713398906_28e59a225a_n.jpg
       inflating: flowers/tulip/8713407768 f880df361f.jpg
       inflating: flowers/tulip/8717900362 2aa508e9e5.jpg
       inflating: flowers/tulip/8722514702 7ecc68691c.jpg
       inflating: flowers/tulip/8723767533 9145dec4bd n.jpg
       inflating: flowers/tulip/8729501081 b993185542 m.jpg
       inflating: flowers/tulip/8733586143 3139db6e9e n.jpg
       inflating: flowers/tulip/8748266132 5298a91dcf n.jpg
import numpy as np
import tensorflow as tf
from tensorflow.keras import layers
from tensorflow.keras.models import Sequential
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import matplotlib.pyplot as plt
batch size = 32
img\ height = 180
img width = 180
data dir = "/content/flowers"
train datagen = ImageDataGenerator(rescale = 1./255, horizontal flip = True, vertical flip =
x train = train datagen.flow from directory('/content/flowers',
                                                target size=(64,64),
                                                class mode='categorical',
                                                batch size=100)
     Found 4317 images belonging to 5 classes.
data augmentation = Sequential(
  Γ
    layers.RandomFlip("vertical",input_shape=(img_height, img_width, 3)),
   layers.RandomRotation(0.1),
   layers.RandomZoom(0.1),
  ]
)
from tensorflow.keras.layers import Convolution2D, MaxPooling2D, Flatten, Dense
model = Sequential()
#Image Augumentation accuracy
data augmentation = Sequential(
  layers.RandomFlip("horizontal",input_shape=(img_height, img_width, 3)),
    layers.RandomRotation(0.1),
```

```
layers.RandomZoom(0.1),
 ]
)
training_ds = tf.keras.utils.image_dataset_from_directory(
  data dir,
 validation_split=0.2,
 subset="training",
 seed=57,
 image_size=(img_height, img_width),
 batch size=batch size)
     Found 4317 files belonging to 5 classes.
     Using 3454 files for training.
validation_ds = tf.keras.utils.image_dataset_from_directory(
  data dir,
 validation_split=0.2,
 subset="validation",
seed=107,
  image_size=(img_height, img_width),
 batch size=batch size)
     Found 4317 files belonging to 5 classes.
     Using 863 files for validation.
training ds.class names
     ['daisy', 'dandelion', 'rose', 'sunflower', 'tulip']
plt.figure(figsize=(7, 7))
for data, labels in training_ds.take(1):
 for i in range(6):
   ax = plt.subplot(3, 4, i + 1)
   plt.imshow(data[i].numpy().astype("uint8"))
   plt.title(training_ds.class_names[labels[i]])
   plt.axis("off")
```

```
dandelion
                            sunflower
     tulip
                     rose
model.add(Convolution2D(32, (3,3), activation = "relu", input shape = (64,64,3) ))
                           444
model.add(MaxPooling2D(pool size = (2,2)))
    dandelion
             daisy
model.add(Flatten())
   model.add(Dense(300, activation = "relu"))
model.add(Dense(150, activation = "relu"))
model.add(Dense(5, activation = "softmax"))
model.compile(optimizer='adam',loss='categorical crossentropy',metrics=['accuracy'])
model.fit(x train, epochs = 15, steps per epoch = len(x train))
   Epoch 1/15
  44/44 [============= ] - 30s 627ms/step - loss: 1.6264 - accuracy: 0.37!
  Epoch 2/15
  Epoch 3/15
  44/44 [============= ] - 27s 605ms/step - loss: 1.0621 - accuracy: 0.568
   Epoch 4/15
  44/44 [============= ] - 27s 598ms/step - loss: 1.0014 - accuracy: 0.603
   Epoch 5/15
  44/44 [============= ] - 27s 605ms/step - loss: 0.9545 - accuracy: 0.626
   Epoch 6/15
  44/44 [============== ] - 27s 599ms/step - loss: 0.8956 - accuracy: 0.656
   Epoch 7/15
  Epoch 8/15
  Epoch 9/15
  Epoch 10/15
  Epoch 11/15
  44/44 [=============== ] - 27s 598ms/step - loss: 0.7667 - accuracy: 0.701
   Epoch 12/15
  Epoch 13/15
  Epoch 14/15
  Epoch 15/15
  44/44 [============= ] - 27s 599ms/step - loss: 0.6840 - accuracy: 0.738
   <keras.callbacks.History at 0x7fba6a7bacd0>
```

tulip_img



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