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

Mounted at /content/drive

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

```
intiating: tiowers/tulip/8/122/0243 8512Ct4tbd.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
inflating: flowers/tulip/8713397358_0505cc0176_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
inflating: flowers/tulip/8750288831 5e49a9f29b.jpg
inflating: flowers/tulip/8757486380 90952c5377.jpg
inflating: flowers/tulip/8758464923 75a5ffe320 n.jpg
inflating: flowers/tulip/8758519201 16e8d2d781 n.jpg
inflating: flowers/tulip/8759594528_2534c0ec65_n.jpg
inflating: flowers/tulip/8759597778 7fca5d434b n.jpg
inflating: flowers/tulip/8759601388 36e2a50d98 n.jpg
inflating: flowers/tulip/8759606166_8e475013fa_n.jpg
inflating: flowers/tulip/8759618746 f5e39fdbf8 n.jpg
inflating: flowers/tulip/8762189906 8223cef62f.jpg
inflating: flowers/tulip/8762193202 0fbf2f6a81.jpg
inflating: flowers/tulip/8768645961 8f1e097170 n.jpg
inflating: flowers/tulip/8817622133 a42bb90e38 n.jpg
inflating: flowers/tulip/8838347159 746d14e6c1 m.jpg
inflating: flowers/tulip/8838354855 c474fc66a3 m.jpg
inflating: flowers/tulip/8838914676 8ef4db7f50 n.jpg
inflating: flowers/tulip/8838975946 f54194894e m.jpg
inflating: flowers/tulip/8838983024 5c1a767878 n.jpg
inflating: flowers/tulip/8892851067 79242a7362 n.jpg
inflating: flowers/tulip/8904780994 8867d64155 n.jpg
```

```
inflating: flowers/tulip/8908062479 449200a1b4.jpg
       inflating: flowers/tulip/8908097235 c3e746d36e n.jpg
       inflating: flowers/tulip/9019694597 2d3bbedb17.jpg
       inflating: flowers/tulip/9030467406 05e93ff171 n.jpg
       inflating: flowers/tulip/9048307967 40a164a459 m.jpg
       inflating: flowers/tulip/924782410 94ed7913ca m.jpg
       inflating: flowers/tulip/9378657435 89fabf13c9 n.jpg
       inflating: flowers/tulip/9444202147_405290415b_n.jpg
       inflating: flowers/tulip/9446982168 06c4d71da3 n.jpg
       inflating: flowers/tulip/9831362123 5aac525a99 n.jpg
       inflating: flowers/tulip/9870557734_88eb3b9e3b_n.jpg
       inflating: flowers/tulip/9947374414 fdf1d0861c n.jpg
       inflating: flowers/tulip/9947385346 3a8cacea02 n.jpg
       inflating: flowers/tulip/9976515506_d496c5e72c.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(
  Γ
    layers.RandomFlip("horizontal",input_shape=(img_height, img_width, 3)),
    layers.RandomRotation(0.1),
    layers.RandomZoom(0.1),
  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")
```

tulip

dandelion

rose

sunflower

```
model.add(Convolution2D(32, (3,3), activation = "relu", input_shape = (64,64,3) ))
    warranerrerr
model.add(MaxPooling2D(pool_size = (2,2)))
   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
  Epoch 2/15
  Epoch 3/15
  Epoch 4/15
  Epoch 5/15
  Epoch 6/15
  44/44 [=================== ] - 31s 692ms/step - loss: 0.9209 - accuracy: 0.64
  Epoch 7/15
  44/44 [=================== ] - 30s 677ms/step - loss: 0.8930 - accuracy: 0.64
  Epoch 8/15
  Epoch 9/15
  Epoch 10/15
  Epoch 11/15
  44/44 [==================== ] - 30s 677ms/step - loss: 0.7859 - accuracy: 0.70
  Epoch 12/15
  44/44 [==================== ] - 30s 683ms/step - loss: 0.7977 - accuracy: 0.69
  Epoch 13/15
  44/44 [==================== ] - 30s 672ms/step - loss: 0.7781 - accuracy: 0.70
  Epoch 14/15
```

x = Image.Img_co_dray(ddisy_img)
x = np.expand_dims(x,axis=0)
predicted_class=model.predict(x)

labels = ['daisy','dandelion','roses','sunflowers','tulips']
labels[np.argmax(predicted_class)]

「→ 'daisy'

daisy_img



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