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

```
inflating: flowers/tulip/8668973377_c69527db42_m.jpg
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
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, flavons/tulin/0717000000 2000000 f

```
intiating: tiowers/tulip/&/i/900362_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
from tensorflow.keras.preprocessing.image import ImageDataGenerator
train datagen = ImageDataGenerator(rescale=1./255,
                                   zoom range=0.2,
                                   horizontal flip=True, vertical flip=True)
test datagen = ImageDataGenerator(rescale=1./255)
xtrain = train datagen.flow from directory('/content/flowers',
                                           target size=(64,64),
                                           class mode='categorical',
                                           batch size=24)
xtest = test_datagen.flow_from_directory('/content/flowers',
                                         target size=(64,64),
                                         class mode='categorical',
                                         batch size=124)
     Found 4317 images belonging to 5 classes.
     Found 4317 images belonging to 5 classes.
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Convolution2D, MaxPooling2D, Flatten, Dense
model = Sequential()
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'))
model.add(Dense(4,activation='softmax'))
model.compile(optimizer='adam',loss='categorical crossentropy',metrics=['accuracy'])
len(xtrain)
```

180

```
model.compile(optimizer='adam',loss='categorical_crossentropy',metrics=['accuracy'])
model.fit(xtrain,
          steps_per_epoch=len(xtrain),
          epochs=5,
          validation data=xtest,
          validation_steps=len(xtest))
model.save('daisy.h5')
from tensorflow.keras.models import load model
from tensorflow.keras.preprocessing import image
import numpy as np
model=load model("/content/daisy.h5")
img = image.load_img('/content/flowers/daisy/100080576_f52e8ee070_n.jpg',target_size=(64,64))
img
x = image.img_to_array(img)
Х
     array([[[141., 141., 139.],
             [149., 149., 149.],
             [152., 152., 154.],
             [162., 161., 166.],
             [154., 154., 152.],
             [153., 153., 153.]],
            [[136., 135., 131.],
             [146., 145., 143.],
             [169., 168., 174.],
             [159., 158., 163.],
             [155., 155., 153.],
```

[149., 149., 149.]],

x.ndim

x.ndim

pred

3

4

```
[[125., 125., 117.],
             [138., 140., 137.],
             [152., 152., 152.],
             . . . ,
             [156., 156., 156.],
             [157., 157., 155.],
             [143., 142., 140.]],
            . . . ,
            [[ 41., 44., 23.],
             [ 43., 46., 25.],
             [ 49., 51., 37.],
             [128., 124., 121.],
             [125., 121., 118.],
             [125., 122., 117.]],
           [[ 43., 46., 25.],
             [ 43., 46., 25.],
             [ 54., 55., 37.],
             . . . ,
             [130., 126., 125.],
             [129., 125., 124.],
             [127., 123., 122.]],
            [[ 44., 47., 26.],
            [ 45., 48., 27.],
             [ 53.,
                   55., 34.],
             [137., 133., 132.],
             [133., 129., 128.],
             [130., 126., 125.]]], dtype=float32)
x = np.expand dims(x,axis=0)
pred = model.predict(x)
     1/1 [======= ] - 0s 384ms/step
```

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✓ 0s completed at 11:06 AM

X