

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

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.

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

```

inflating: flowers/tulip/8614237582_74417799f4_m.jpg
inflating: flowers/tulip/8619064872_dea79a9eb9.jpg
inflating: flowers/tulip/8622237974_b362574785_n.jpg
inflating: flowers/tulip/8623170936_83f4152431.jpg
inflating: flowers/tulip/8623173256_3f0eb4c506.jpg
inflating: flowers/tulip/8628453641_6f87755815_m.jpg
inflating: flowers/tulip/8659691170_09db83d023.jpg
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/8713391374_40079e1e5_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

```

## Image augmentation

```

import tensorflow as tensorflow
from tensorflow.keras.preprocessing.image import ImageDataGenerator

train_data_aug = ImageDataGenerator(rescale=1./255, zoom_range=0.2, horizontal_flip=True)

train_data_aug = ImageDataGenerator(rescale=1./255, zoom_range=0.2, horizontal_flip=True)

test_data_aug = ImageDataGenerator(rescale=1./255)

xtrain = train_data_aug.flow_from_directory('/content/flowers', target_size=(64,64), class_n

Found 4317 images belonging to 5 classes.

xtest = test_data_aug.flow_from_directory('/content/flowers', target_size=(64,64), class_moc

Found 4317 images belonging to 5 classes.

```

## Create Model

```

from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Convolution2D, MaxPooling2D, Flatten, Dense

```

## Add Layers (Convolution, Max-Pooling, Flatten, Dense-(Hidden Layers), Output)

```

model = Sequential()
model.add(Convolution2D(32, (3,3), activation='relu', input_shape=(64,64,3))) # Convolution 1
model.add(MaxPooling2D(pool_size=(2,2))) # Max pooling layer
model.add(Flatten()) # Flatten layer
# Fully connected layers (ANN)
model.add(Dense(300, activation='relu')) # Hidden layer 1
model.add(Dense(150, activation='relu')) # Hidden layer 2

```

```
model.add(Dense(5,activation='softmax')) # Output layer
```

## Compile the Model

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

## Fit the Model

```
model.fit_generator(xtrain,
                    steps_per_epoch=len(xtrain),
                    epochs=10,
                    validation_data=xtest,
                    validation_steps=len(xtest))
```

```
Epoch 1/10
44/44 [=====] - 48s 1s/step - loss: 1.5197 - accuracy: 0.40
Epoch 2/10
44/44 [=====] - 45s 1s/step - loss: 1.1072 - accuracy: 0.54
Epoch 3/10
44/44 [=====] - 45s 1s/step - loss: 1.0195 - accuracy: 0.60
Epoch 4/10
44/44 [=====] - 47s 1s/step - loss: 0.9714 - accuracy: 0.62
Epoch 5/10
44/44 [=====] - 45s 1s/step - loss: 0.9062 - accuracy: 0.65
Epoch 6/10
44/44 [=====] - 45s 1s/step - loss: 0.8899 - accuracy: 0.65
Epoch 7/10
44/44 [=====] - 45s 1s/step - loss: 0.8113 - accuracy: 0.69
Epoch 8/10
44/44 [=====] - 45s 1s/step - loss: 0.7739 - accuracy: 0.70
Epoch 9/10
44/44 [=====] - 45s 1s/step - loss: 0.7225 - accuracy: 0.72
Epoch 10/10
44/44 [=====] - 45s 1s/step - loss: 0.7019 - accuracy: 0.73
<keras.callbacks.History at 0x7fa7110c9090>
```

Double-click (or enter) to edit

## Save the Model

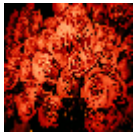
```
model.save('/content/flowers')
```

## Test the Model

```
import numpy as np
from tensorflow.keras.preprocessing import image
```

```
img = image.load_img('/content/flowers/rose/10503217854_e66a804309.jpg',target_size=(64,64))
```

```
img
```



```
x = image.img_to_array(img)
x
```

```
array([[ 0.,  2.,  0.],
       [ 0.,  2.,  0.],
       [ 0.,  2.,  0.],
       ...,
       [ 92., 14.,  0.],
       [ 61., 13.,  9.],
       [ 17.,  7.,  5.]],

      [[ 0.,  2.,  0.],
       [ 0.,  2.,  0.],
       [ 0.,  2.,  0.],
       ...,
       [150.,  3.,  0.],
       [ 85., 10.,  7.],
       [119.,  4.,  1.]],

      [[ 0.,  2.,  0.],
       [ 0.,  2.,  0.],
       [ 0.,  2.,  0.],
       ...,
       [ 88.,  9.,  0.],
       [207.,  7., 10.],
       [152.,  0.,  0.]],

      ...,

      [[ 0.,  4.,  0.],
       [ 1.,  3.,  0.],
       [ 0.,  2.,  0.],
       ...,
       [ 2.,  2.,  4.],
       [ 0.,  2.,  5.],
       [ 51., 10.,  6.]],

      [[ 0.,  2.,  0.],
       [ 1.,  3.,  0.],
       [ 1.,  3.,  0.],
       ...,
       [ 0.,  3.,  1.],
       [ 0.,  3.,  4.],
       [ 0.,  5.,  3.]],

      [[ 1.,  3.,  0.]])
```

```
[ 0.,  2.,  0.],
[ 1.,  1.,  0.],
...,
[ 29.,  5.,  1.],
[ 41., 13.,  0.],
[ 5.,  4.,  0.]]], dtype=float32)
```

```
x = np.expand_dims(x,axis=0)
```

```
x
```

```
array([[[[ 0.,  2.,  0.],
          [ 0.,  2.,  0.],
          [ 0.,  2.,  0.],
          ...,
          [ 92., 14.,  0.],
          [ 61., 13.,  9.],
          [ 17.,  7.,  5.]],

        [[ 0.,  2.,  0.],
          [ 0.,  2.,  0.],
          [ 0.,  2.,  0.],
          ...,
          [150.,  3.,  0.],
          [ 85., 10.,  7.],
          [119.,  4.,  1.]],

        [[ 0.,  2.,  0.],
          [ 0.,  2.,  0.],
          [ 0.,  2.,  0.],
          ...,
          [ 88.,  9.,  0.],
          [207.,  7., 10.],
          [152.,  0.,  0.]],

        ...,

        [[ 0.,  4.,  0.],
          [ 1.,  3.,  0.],
          [ 0.,  2.,  0.],
          ...,
          [ 2.,  2.,  4.],
          [ 0.,  2.,  5.],
          [ 51., 10.,  6.]],

        [[ 0.,  2.,  0.],
          [ 1.,  3.,  0.],
          [ 1.,  3.,  0.],
          ...,
          [ 0.,  3.,  1.],
          [ 0.,  3.,  4.],
          [ 0.,  5.,  3.]],

        [[ 1.,  3.,  0.],
          [ 0.,  2.,  0.],
          [ 1.,  1.,  0.],
          ...,
          [ 29.,  5.,  1.],
```

```
[ 41., 13.,  0.],
 [  5.,  4.,  0.] ]], dtype=float32)
```

```
model.predict(x)
```

```
1/1 [=====] - 0s 121ms/step
array([[0., 0., 1., 0., 0.]], dtype=float32)
```

```
xtrain.class_indices
```

```
{'daisy': 0, 'dandelion': 1, 'rose': 2, 'sunflower': 3, 'tulip': 4}
```

```
op = ['daisy','dandelion','rose','sunflower','tulip']
pred = np.argmax(model.predict(x))
op[pred]
```

```
1/1 [=====] - 0s 26ms/step
'rose'
```

```
img = image.load_img('/content/flowers/dandelion/10043234166_e6dd915111_n.jpg',target_size=(256,256))
x = image.img_to_array(img)
x = np.expand_dims(x,axis=0)
pred = np.argmax(model.predict(x))
op[pred]
```

```
1/1 [=====] - 0s 27ms/step
'daisy'
```