

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

MessageError Traceback (most recent call last)

```
<ipython-input-2-d5df0069828e> in <module>
      1 from google.colab import drive
----> 2 drive.mount('/content/drive')
```

3 frames

```
/usr/local/lib/python3.7/dist-packages/google/colab/_message.py in read_reply_from_input(message_id, timeout_sec)
    100     reply.get('colab_msg_id') == message_id):
    101     if 'error' in reply:
--> 102         raise MessageError(reply['error'])
    103     return reply.get('data', None)
    104
```

MessageError: Error: credential propagation was unsuccessful

SEARCH STACK OVERFLOW

```
ls
```

 **drive/** **sample_data/**

```
cd /content/drive/MyDrive/IBM_NalaiyaThiran/images
```

```
/content/drive/MyDrive/IBM_NalaiyaThiran/images
```

```
pwd
```

```
'/content/drive/MyDrive/IBM_NalaiyaThiran/images'
```

```
!unzip flowers.zip
```

```
Archive: flowers.zip
  inflating: images/f1.jpg
  inflating: images/f2.jpg
  inflating: images/f3.jpg
  inflating: images/f4.jpg
  inflating: images/f5.jpg
```

Image Augmentation

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

```
train_datagen = ImageDataGenerator(rescale = 1./255, zoom_range= 0.3, horizontal_flip=True, vertical_flip=True)
```

```
test_datagen = ImageDataGenerator(rescale = 1./255)
```

```
x_train = train_datagen.flow_from_directory(r"/content/drive/MyDrive/IBM_NalaiyaThiran/images", target_size= (64,64), class_mode= "cate
```

```
    Found 5 images belonging to 1 classes.
```

```
x_test = test_datagen.flow_from_directory(r"/content/drive/MyDrive/IBM_NalaiyaThiran/images", target_size= (64,64), class_mode= "catego
```

```
    Found 5 images belonging to 1 classes.
```

```
x_train.class_indices
```

```
{'images': 0}
```

```
from tensorflow.keras.models import Sequential
```

```
from tensorflow.keras.layers import Dense, Convolution2D, MaxPooling2D, Flatten
```

```
model = Sequential()
```

```
model.add(Convolution2D(32,(3,3),activation="relu",strides=(1, 1),input_shape =(64,64,3)))
```

```
model.add(MaxPooling2D(strides=(1, 1)))
```

```
model.add(Flatten())
```

```
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 62, 62, 32)	896
max_pooling2d (MaxPooling2D)	(None, 61, 61, 32)	0
flatten (Flatten)	(None, 119072)	0
=====		
Total params: 896		
Trainable params: 896		
Non-trainable params: 0		

```
model.add(Dense(300,activation="relu"))
```

```
model.add(Dense(300,activation="relu"))
```

```
model.add(Dense(5,activation="softmax"))
```

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

```
len(x_train)
```

```
1
```

```
model.fit(x_train, epochs = 10, steps_per_epoch=len(x_train), validation_data=x_test, validation_steps=len(x_test))
```

```
Epoch 1/10
```

```
1/1 [=====] - 2s 2s/step - loss: 8.0750 - accuracy: 0.0000e+00 - val_loss: 20.8234 - val_accuracy: 0.0
```

```
Epoch 2/10
```

```
1/1 [=====] - 1s 848ms/step - loss: 19.1896 - accuracy: 0.0000e+00 - val_loss: 28.4673 - val_accuracy:
```

```
Epoch 3/10
```

```
1/1 [=====] - 1s 634ms/step - loss: 25.9591 - accuracy: 0.0000e+00 - val_loss: 42.4799 - val_accuracy:
```

```
Epoch 4/10
```

```
1/1 [=====] - 1s 684ms/step - loss: 37.1618 - accuracy: 0.0000e+00 - val_loss: 47.2677 - val_accuracy:
```

```
Epoch 5/10
```

```
1/1 [=====] - 1s 648ms/step - loss: 45.7534 - accuracy: 0.0000e+00 - val_loss: 51.9293 - val_accuracy:
```

```
Epoch 6/10
```

```
1/1 [=====] - 1s 717ms/step - loss: 51.0597 - accuracy: 0.0000e+00 - val_loss: 45.2593 - val_accuracy:
```

```
Epoch 7/10
```

```
1/1 [=====] - 1s 654ms/step - loss: 44.8140 - accuracy: 0.0000e+00 - val_loss: 50.0706 - val_accuracy:
```

```
Epoch 8/10
```

```
1/1 [=====] - 1s 718ms/step - loss: 50.1203 - accuracy: 0.0000e+00 - val_loss: 47.5019 - val_accuracy:
```

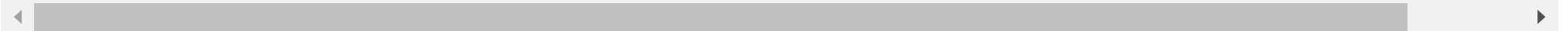
```
Epoch 9/10
```

```
1/1 [=====] - 1s 677ms/step - loss: 45.3977 - accuracy: 0.0000e+00 - val_loss: 81.6582 - val_accuracy:
```

```
Epoch 10/10
```

```
1/1 [=====] - 1s 748ms/step - loss: 78.7049 - accuracy: 0.0000e+00 - val_loss: 93.6470 - val_accuracy:
```

```
<keras.callbacks.History at 0x7fa0f6b393d0>
```



```
model.save("flower.h5")
```

```
import numpy as np
```

```
from tensorflow.keras.models import load_model
```

```
from tensorflow.keras.preprocessing import image
```

```
model = load_model("flower.h5")
```

```
img = image.load_img(r"/content/drive/MyDrive/IBM_NalaiyaThiran/images/f1.jpg")
```

```
img
```



```
img = image.load_img(r"/content/drive/MyDrive/IBM_NalaiyaThiran/images/f1.jpg",target_size=(64,64))
```

```
img
```



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

```
x
```

```
array([[[ 44.,  41.,   6.],
        [ 37.,  37.,   3.],
        [ 30.,  36.,   2.],
        ...,
        [ 39.,  40.,   9.],
        [ 84.,  71.,  29.],
        [116.,  92.,  56.]],

       [[ 49.,  45.,   8.],
        [ 42.,  39.,   4.],
        [ 27.,  32.,   2.],
        ...,
        [ 47.,  49.,   2.],
        [147., 137.,  88.],
        [ 50.,  43.,  14.]],

       [[ 56.,  53.,  12.],
        [ 38.,  42.,   9.],
        [ 27.,  32.,   2.],
        ...,
        [ 41.,  45.,  12.],
        [ 26.,  32.,   6.],
        [ 26.,  23.,   6.]],

       ...,

       [[ 45.,  28.,   8.],
        [ 50.,  32.,  10.],
        [ 55.,  43.,  17.],
        ...,
        [ 42.,  46.,  21.],
        [  4.,   7.,   0.],
        [  6.,   9.,   2.]])
```

```

[[ 38., 20., 6.],
 [ 37., 21., 8.],
 [ 40., 29., 7.],
 ...,
 [ 11., 14., 0.],
 [ 8., 11., 0.],
 [ 7., 10., 0.]],

[[ 26., 16., 4.],
 [ 28., 16., 4.],
 [ 29., 17., 1.],
 ...,
 [ 12., 14., 3.],
 [ 9., 12., 1.],
 [ 9., 12., 1.]]], dtype=float32)

```

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

```
x
```

```

array([[[[ 44., 41., 6.],
 [ 37., 37., 3.],
 [ 30., 36., 2.],
 ...,
 [ 39., 40., 9.],
 [ 84., 71., 29.],
 [116., 92., 56.]],

[[ 49., 45., 8.],
 [ 42., 39., 4.],
 [ 27., 32., 2.],
 ...,
 [ 47., 49., 2.],
 [147., 137., 88.],
 [ 50., 43., 14.]],

[[ 56., 53., 12.],
 [ 38., 42., 9.],
 [ 27., 32., 2.],

```

```

...,
[[ 41.,  45.,  12.],
 [ 26.,  32.,   6.],
 [ 26.,  23.,   6.]],

...,

[[ 45.,  28.,   8.],
 [ 50.,  32.,  10.],
 [ 55.,  43.,  17.],

...,
 [ 42.,  46.,  21.],
 [  4.,   7.,   0.],
 [  6.,   9.,   2.]],

[[ 38.,  20.,   6.],
 [ 37.,  21.,   8.],
 [ 40.,  29.,   7.],

...,
 [ 11.,  14.,   0.],
 [  8.,  11.,   0.],
 [  7.,  10.,   0.]],

[[ 26.,  16.,   4.],
 [ 28.,  16.,   4.],
 [ 29.,  17.,   1.],

...,
 [ 12.,  14.,   3.],
 [  9.,  12.,   1.],
 [  9.,  12.,   1.] ]], dtype=float32)

```

```
pred = model.predict(x)
```

```
1/1 [=====] - 0s 184ms/step
```

```
pred
```

```
array([[0., 0., 0., 1., 0.]], dtype=float32)
```



```
x_test.class_indices
```

```
index = ["", "images"]
```

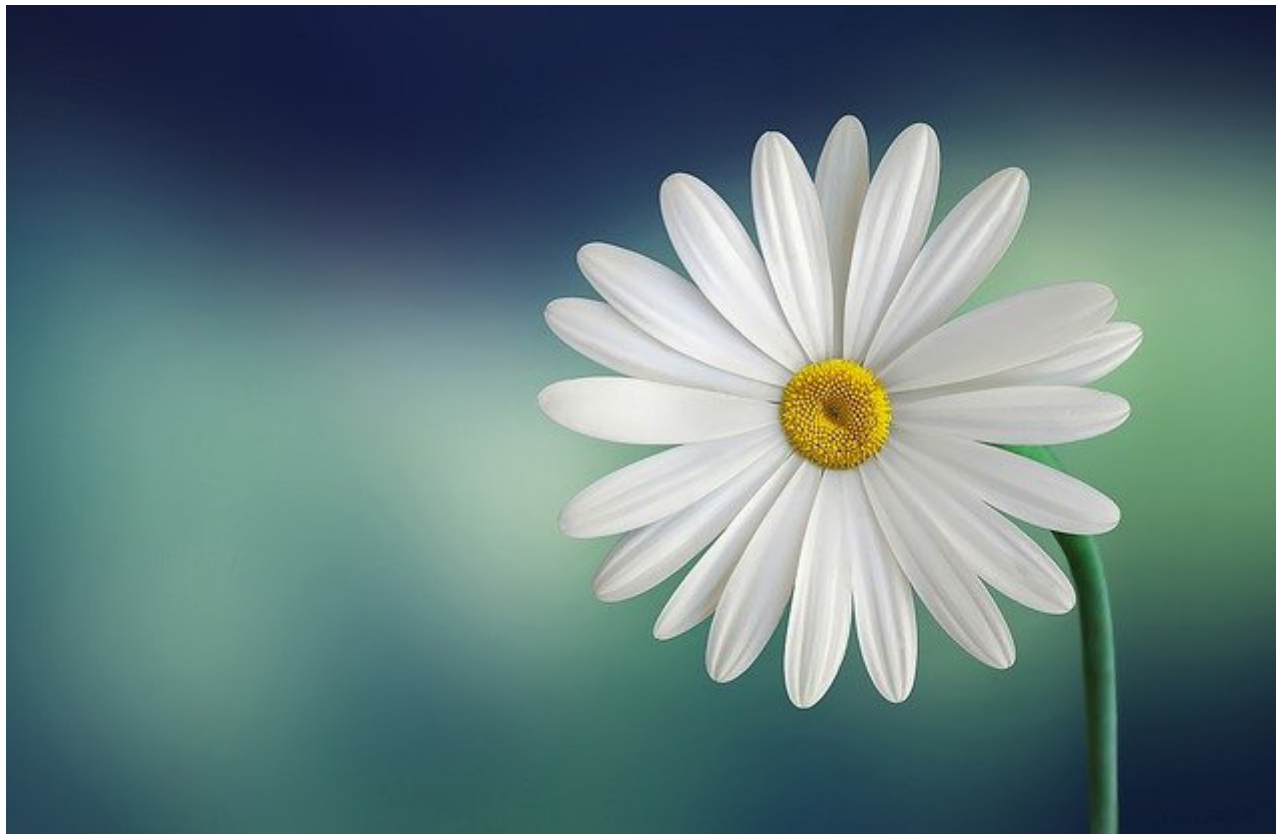
```
img = image.load_img(r"/content/drive/MyDrive/IBM_NalaiyaThiran/images/f1.jpg", target_size=(64, 64))
```

```
img
```



```
img = image.load_img(r"/content/drive/MyDrive/IBM_NalaiyaThiran/images/f4.jpg")
```

```
img
```



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