

```

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

from tensorflow.keras.preprocessing.image import ImageDataGenerator

train_data=ImageDataGenerator(rescale=1./255,shear_range=0.2,zoom_range=0.2,horizontal_flip=True,vertical
_flip=True)
test_data=ImageDataGenerator(rescale=1./255)

x_train=train_data.flow_from_directory(r"E:\assignment3\dataset\Training",target_size=(64,64),batch_size=32,
class_mode="categorical")
x_test=test_data.flow_from_directory(r"E:\assignment3\dataset\Testing",target_size=(64,64),batch_size=32,clas
s_mode="categorical")

Found 3453 images belonging to 5 classes.
Found 864 images belonging to 5 classes.

x_train.class_indices

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

model=Sequential()

model.add(Convolution2D(64,(3,3),input_shape=(64,64,3),activation='relu'))

model.add(MaxPooling2D(pool_size=(3,3)))

model.add(Flatten())

model.add(Dense(units=5,kernel_initializer="random_uniform",activation="softmax"))

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

model.fit_generator(x_train,steps_per_epoch=108,epochs=30,validation_data=x_test,validation_steps=27)

Epoch 1/30
108/108 [=====] - 88s 798ms/step - loss: 1.6050 - accuracy: 0.2299 - val_lo
ss: 1.6017 - val_accuracy: 0.2431
Epoch 2/30
108/108 [=====] - 70s 646ms/step - loss: 1.6005 - accuracy: 0.2438 - val_lo
ss: 1.5994 - val_accuracy: 0.2431
Epoch 3/30
108/108 [=====] - 71s 654ms/step - loss: 1.5992 - accuracy: 0.2438 - val_lo
ss: 1.5989 - val_accuracy: 0.2431
Epoch 4/30
108/108 [=====] - 70s 648ms/step - loss: 1.5988 - accuracy: 0.2438 - val_lo
ss: 1.5987 - val_accuracy: 0.2431
Epoch 5/30

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108/108 [=====] - 74s 685ms/step - loss: 1.5988 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 6/30
108/108 [=====] - 77s 712ms/step - loss: 1.5986 - accuracy: 0.2438 - val_loss: 1.5985 - val_accuracy: 0.2431
Epoch 7/30
108/108 [=====] - 80s 739ms/step - loss: 1.5985 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 8/30
108/108 [=====] - 76s 702ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 9/30
108/108 [=====] - 75s 692ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 10/30
108/108 [=====] - 78s 721ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 11/30
108/108 [=====] - 75s 693ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 12/30
108/108 [=====] - 75s 695ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 13/30
108/108 [=====] - 81s 745ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 14/30
108/108 [=====] - 76s 700ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 15/30
108/108 [=====] - 80s 742ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 16/30
108/108 [=====] - 75s 693ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 17/30
108/108 [=====] - 75s 692ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 18/30
108/108 [=====] - 75s 696ms/step - loss: 1.5986 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 19/30
108/108 [=====] - 75s 694ms/step - loss: 1.5986 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 20/30
108/108 [=====] - 75s 694ms/step - loss: 1.5988 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 21/30
108/108 [=====] - 75s 690ms/step - loss: 1.5986 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 22/30
108/108 [=====] - 76s 702ms/step - loss: 1.5986 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 23/30
108/108 [=====] - 76s 705ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 24/30

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108/108 [=====] - 75s 692ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 25/30
108/108 [=====] - 75s 690ms/step - loss: 1.5988 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 26/30
108/108 [=====] - 75s 695ms/step - loss: 1.5986 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 27/30
108/108 [=====] - 76s 699ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 28/30
108/108 [=====] - 75s 692ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 29/30
108/108 [=====] - 75s 690ms/step - loss: 1.5987 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431
Epoch 30/30
108/108 [=====] - 75s 696ms/step - loss: 1.5988 - accuracy: 0.2438 - val_loss: 1.5987 - val_accuracy: 0.2431

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```
<keras.callbacks.History at 0x23bcb0e0d60>
```

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

```

from tensorflow.keras.models import load_model
from tensorflow.keras.preprocessing import image

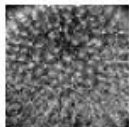
```

```
import numpy as np
```

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

```
img=image.load_img("dandelion.jpg",target_size=(64,64))
```

```
img
```



```
type(img)
```

```
PIL.Image.Image
```

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

```
x
```

```

array([[[ 4.,  4.,  4.],
         [ 11., 11., 11.],
         [ 9.,  9.,  9.],
         ...,
         [ 11., 11., 11.],
         [ 9.,  9.,  9.]

```

```

[ 10., 10., 10.]],

[[ 12., 12., 12.],
 [ 21., 21., 21.],
 [ 19., 19., 19.],
 ...,
 [ 15., 15., 15.],
 [ 23., 23., 23.],
 [ 2., 2., 2.]],

[[ 23., 23., 23.],
 [ 52., 52., 52.],
 [ 41., 41., 41.],
 ...,
 [ 49., 49., 49.],
 [ 34., 34., 34.],
 [ 42., 42., 42.]],

...,

[[ 56., 56., 56.],
 [121., 121., 121.],
 [ 44., 44., 44.],
 ...,
 [113., 113., 113.],
 [ 24., 24., 24.],
 [ 12., 12., 12.]],

[[ 54., 54., 54.],
 [ 30., 30., 30.],
 [ 36., 36., 36.],
 ...,
 [102., 102., 102.],
 [ 42., 42., 42.],
 [ 10., 10., 10.]],

[[ 23., 23., 23.],
 [ 20., 20., 20.],
 [ 11., 11., 11.],
 ...,
 [ 77., 77., 77.],
 [ 3., 3., 3.],
 [ 10., 10., 10.]], dtype=float32)

x.shape

(64, 64, 3)

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

pred_prob=model.predict(x)
1/1 [=====] - 0s 63ms/step

pred_prob

```

```
array([[0.17660806, 0.24450503, 0.18085377, 0.17052399, 0.22750916]],  
      dtype=float32)
```

```
class_name=["Daisy","Dandelion","Rose","Sunflower","Tulip"]  
pred_id=pred_prob.argmax(axis=1)[0]
```

```
pred_id
```

```
1
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```
print("Predicted animalis ",str(class_name[pred_id]))
```

```
Predicted animalis Dandelion
```