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

```
train_datagen = ImageDataGenerator(rescale= 1./255,horizontal_flip = True,vertical_flip = True,zoom_ran
```

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

```
x_train = train_datagen.flow_from_directory(r"C:\Users\deva\Desktop\data_for_ibm\Flowers-Dataset\flower
                                            class_mode = "categorical",batch_size = 24)
```

```
Found 4317 images belonging to 5 classes.
```

```
x_test = test_datagen.flow_from_directory(r"C:\Users\deva\Desktop\data_for_ibm\Flowers-Dataset\flowers"
                                                                                      class_mode = "cat
```

```
Found 4317 images belonging to 5 classes.
```

```
x_train.class_indices
```

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

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

```
model=Sequential()
```

```
model.add(Convolution2D(32,(3,3),input_shape=(64,64,3),activation='relu'))
model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Flatten())
model.summary()
```

```
Model: "sequential"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 conv2d (Conv2D)             (None, 62, 62, 32)        896

 max_pooling2d (MaxPooling2D  (None, 31, 31, 32)       0
 )

 flatten (Flatten)           (None, 30752)             0

=================================================================
Total params: 896
Trainable params: 896
Non-trainable params: 0
_________________________________________________________________
```

```
model.add(Dense(300,activation='relu'))
model.add(Dense(150,activation='relu'))

model.add(Dense(5,activation='softmax'))
len(x_train)
```

```
180
```

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

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

```
Epoch 1/10
180/180 [==============================] - 33s 183ms/step - loss: 1.3003 - accuracy: 0.4691 - val_
Epoch 2/10
180/180 [==============================] - 28s 157ms/step - loss: 1.0616 - accuracy: 0.5812 - val_
Epoch 3/10
180/180 [==============================] - 28s 157ms/step - loss: 0.9799 - accuracy: 0.6185 - val_
Epoch 4/10
180/180 [==============================] - 29s 161ms/step - loss: 0.9217 - accuracy: 0.6366 - val_
Epoch 5/10
180/180 [==============================] - 28s 158ms/step - loss: 0.8893 - accuracy: 0.6583 - val_
Epoch 6/10
180/180 [==============================] - 29s 162ms/step - loss: 0.8509 - accuracy: 0.6755 - val_
Epoch 7/10
180/180 [==============================] - 30s 169ms/step - loss: 0.8274 - accuracy: 0.6755 - val_
Epoch 8/10
180/180 [==============================] - 30s 166ms/step - loss: 0.7923 - accuracy: 0.6965 - val_
Epoch 9/10
180/180 [==============================] - 28s 157ms/step - loss: 0.7745 - accuracy: 0.7072 - val_
Epoch 10/10
180/180 [==============================] - 28s 158ms/step - loss: 0.7363 - accuracy: 0.7192 - val_
<keras.callbacks.History at 0x16061cf68f0>
```

```
model.save('IBM_flowers.h5')
```

```
pwd
```

```
'C:\\Users\\maris_q3mm6nk\\Desktop\\data_for_ibm'
```

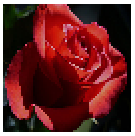
```
import numpy as np
from tensorflow.keras.models import load_model
from tensorflow.keras.preprocessing import image
```

```
model=load_model('IBM_flowers.h5')
```

```
img=image.load_img(r'C:\Users\deva\Desktop\data_for_ibm\Flowers-Dataset\flowers\rose/394990940_7af082cf
```

```
img
```

```
img=image.load_img(r'C:\Users\deva\Desktop\data_for_ibm\Flowers-Dataset\flowers\rose/394990940_7af082cf
img
```



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

```
array([[[ 4., 14.,  3.],
        [ 4., 15.,  0.],
        [ 7., 10.,  3.],
        ...,
        [ 1.,  1.,  1.],
        [ 1.,  1.,  1.],
        [ 3.,  3.,  3.]],

       [[21., 37.,  8.],
        [ 7., 18.,  1.],
        [ 5., 11.,  1.],
        ...,
        [ 1.,  1.,  3.],
        [ 1.,  1.,  1.],
        [ 2.,  2.,  2.]],

       [[15., 34.,  4.],
        [ 5., 18.,  0.],
        [ 6., 14.,  3.],
        ...,
        [ 1.,  2.,  4.],
        [ 0.,  0.,  0.],
        [ 1.,  1.,  1.]],

       ...,

       [[ 7., 11., 10.],
        [ 7., 16., 15.],
        [17., 23., 21.],
        ...,
        [ 1.,  1.,  1.],
        [ 2.,  2.,  2.],
        [ 0.,  0.,  0.]],
       [[ 9., 18., 15.],
        [ 2.,  7.,  3.],
        [ 5., 11.,  7.],
        ...,
        [ 0.,  0.,  0.],
        [ 1.,  1.,  1.],
        [ 1.,  1.,  1.]],

       [[18., 26., 28.],
        [ 0., 10.,  2.],
        [ 8., 14., 10.],
        ...,
        [ 2.,  6.,  9.],
        [ 1.,  1.,  1.],
        [ 1.,  1.,  1.]]], dtype=float32)
```

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

```
array([[[[ 4., 14.,  3.],
         [ 4., 15.,  0.],
         [ 7., 10.,  3.],
         ...,
         [ 1.,  1.,  1.],
         [ 1.,  1.,  1.],
         [ 3.,  3.,  3.]],

        [[21., 37.,  8.],
         [ 7., 18.,  1.],
         [ 5., 11.,  1.],
         ...,
         [ 1.,  1.,  3.],
         [ 1.,  1.,  1.],
         [ 2.,  2.,  2.]],

        [[15., 34.,  4.],
         [ 5., 18.,  0.],
         [ 6., 14.,  3.],
         ...,
         [ 1.,  2.,  4.],
         [ 0.,  0.,  0.],
         [ 1.,  1.,  1.]],

        ...,

        [[ 7., 11., 10.],
         [ 7., 16., 15.],
         [17., 23., 21.],
         ...,
         [ 1.,  1.,  1.],
         [ 2.,  2.,  2.],
         [ 0.,  0.,  0.]],

        [[ 9., 18., 15.],
         [ 2.,  7.,  3.],
         [ 5., 11.,  7.],
         ...,
         [ 0.,  0.,  0.],
         [ 1.,  1.,  1.],
         [ 1.,  1.,  1.]],

        [[18., 26., 28.],
         [ 0., 10.,  2.],
         [ 8., 14., 10.],
         ...,
         [ 2.,  6.,  9.],
         [ 1.,  1.,  1.],
         [ 1.,  1.,  1.]]]], dtype=float32)
```

```
y=np.argmax(model.predict(x),axis=1)
y
```

```
1/1 [==============================] - 0s 74ms/step
array([2], dtype=int64)
```

```
x_train.class_indices
```

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

```
index=['daisy','dandelion','rose','sunflower','tulip']
```

```
index[y[0]]
```

```
'rose'
```