```
import matplotlib.pyplot as plt
from matplotlib.image import imread
folder ='/content/drive/MyDrive/AI_baocao/images/images_flowers/'
for i in range(6):
 plt.subplot(1,6,i+1)
  filename = folder + 'camtucau'+str(i+1)+'.jpg'
  img = plt.imread(filename)
 plt.imshow(img)
plt.show()
C→
folder ='/content/drive/MyDrive/AI_baocao/images/images_flowers/'
for i in range(6):
 plt.subplot(1,6,i+1)
  filename = folder + 'hong'+str(i+1)+'.jpg'
  img = plt.imread(filename)
 plt.imshow(img)
plt.show()
       0
      20
                         25
                            0
                                       0
                                              25
```





```
folder ='/content/drive/MyDrive/AI_baocao/images/images_flowers/'
for i in range(6):
   plt.subplot(1,6,i+1)
   filename = folder + 'sen'+str(i+1)+'.jpg'
   img = plt.imread(filename)
   plt.imshow(img)
plt.show()
```



```
folder ='/content/drive/MyDrive/AI_baocao/images/images_flowers/'
for i in range(6):
   plt.subplot(1,6,i+1)
   filename = folder + 'tulip'+str(i+1)+'.jpg'
   img = plt.imread(filename)
   plt.imshow(img)
plt.show()
```

```
from os import listdir
from numpy import asarray
from numpy import save
from keras.utils import load_img, img_to_array
folder = '/content/drive/MyDrive/AI_baocao/images/images_flowers/'
photos, labels = list(), list()
for file in listdir(folder):
  output = 0.0
  if file.startswith('camtucau'):
    output = 1.0
  if file.startswith('hduong'):
    output = 2.0
  if file.startswith('hong'):
    output = 3.0
  if file.startswith('sen'):
    output = 4.0
  if file.startswith('tulip'):
    output = 5.0
  img =load_img(folder+file,target_size=(40,30))
  photo = img_to_array(img)
  photos.append(photo)
  labels.append(output)
photos = asarray(photos)
labels = asarray(labels)
print(photos.shape, labels.shape)
save('hooa_photos.npy', photos)
save('hooa_labels.npy', labels)
     (50, 40, 30, 3) (50,)
from google.colab import drive
drive.mount('/content/drive')
    Mounted at /content/drive
import numpy as no
x_train = np.load('hooa_photos.npy')
y_train = np.load('hooa_labels.npy')
print(x_train.shape)
print(y_train.shape)
     (50, 40, 30, 3)
     (50,)
x_train = x_train.astype('float32')/255
from keras.utils import to_categorical
y_train = to categorical(y train,10)
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten, Conv2D, MaxPooling2D, Normalization, LeakyReLU
from keras.optimizers import Adam
#32 lan tich chap
model = Sequential()
model.add(Conv2D(32,kernel_size = (3,3),activation = 'relu',input_shape=(40,30,3),padding='Same'))
model.add(MaxPooling2D((2,2),padding='same'))
model.add(Dropout(0.25))
#64 lan tich chap
model.add(Conv2D(64,(3,3),activation ='relu',padding ='same'))
model.add(MaxPooling2D((2,2),padding='same'))
model.add(Dropout(0.25))
model.add(Flatten())
model.add(Dense(64,activation = 'relu'))
model.add(Dropout(0.25))
model.add(Dense(10,activation='softmax'))
from keras.losses import categorical_crossentropy
model.compile(loss = categorical_crossentropy,optimizer = Adam(),metrics=['accuracy'])
model.summary()
```

train = model.fit(x_train,y_train,batch_size=250,epochs = 40,verbose = 1)

```
=======] - Os 14ms/step - loss: 1.2011 - accuracy: 0.5400
    Epoch 14/40
                                ======] - 0s 14ms/step - loss: 1.2051 - accuracy: 0.6000
    1/1 [==
    Epoch 15/40
                               ======] - 0s 14ms/step - loss: 1.2490 - accuracy: 0.5600
    1/1 [=====
    Epoch 16/40
    1/1 [==:
                                      ==] - 0s 13ms/step - loss: 1.1364 - accuracy: 0.6400
    Epoch 17/40
    1/1 [==
                                 ======] - 0s 11ms/step - loss: 1.0307 - accuracy: 0.7200
    Epoch 18/40
                                           Os 12ms/step - loss: 0.9173 - accuracy: 0.7600
    Epoch 19/40
    1/1 [==
                                      ==] - 0s 12ms/step - loss: 0.8415 - accuracy: 0.7800
    Epoch 20/40
                                ======] - 0s 18ms/step - loss: 0.8475 - accuracy: 0.7400
    1/1 [=====
    Epoch 21/40
                                    ====] - 0s 14ms/step - loss: 0.6881 - accuracy: 0.8600
    1/1 [==
    Epoch 22/40
    1/1 [===
                                         - 0s 13ms/step - loss: 0.7309 - accuracy: 0.8000
    Epoch 23/40
    1/1 [==
                                         - 0s 12ms/step - loss: 0.7278 - accuracy: 0.7400
    Epoch 24/40
    1/1 [==
                                         - 0s 13ms/step - loss: 0.7100 - accuracy: 0.7600
    Epoch 25/40
                                         - 0s 12ms/step - loss: 0.6569 - accuracy: 0.6800
    1/1 [===
    Epoch 26/40
                                         - 0s 12ms/step - loss: 0.7897 - accuracy: 0.7400
    1/1 [==
    Epoch 27/40
    1/1 [==:
                                         - 0s 12ms/step - loss: 0.5542 - accuracy: 0.8400
    Epoch 28/40
    1/1 [==
                                         - 0s 12ms/step - loss: 0.4567 - accuracy: 0.8800
    Epoch 29/40
    1/1 [==
                                         - 0s 13ms/step - loss: 0.5901 - accuracy: 0.8400
    Epoch 30/40
                                ======] - 0s 12ms/step - loss: 0.4271 - accuracy: 0.8600
    1/1 [=====
    Epoch 31/40
                                      ==] - 0s 18ms/step - loss: 0.4874 - accuracy: 0.9000
    1/1 [:
    Epoch 32/40
    1/1 [===
                             =======] - 0s 13ms/step - loss: 0.4097 - accuracy: 0.8800
    Epoch 33/40
                                           Os 12ms/step - loss: 0.3522 - accuracy: 0.9200
     1/1 [==:
    Epoch 34/40
                                         - 0s 12ms/step - loss: 0.3014 - accuracy: 0.8800
    1/1 [==:
    Epoch 35/40
                               ======] - 0s 17ms/step - loss: 0.4051 - accuracy: 0.8600
    1/1 [=====
    Epoch 36/40
                                    ====] - 0s 13ms/step - loss: 0.3808 - accuracy: 0.8600
    1/1 [==:
    Epoch 37/40
                            =======] - Os 12ms/step - loss: 0.3537 - accuracy: 0.8600
    1/1 [======
    Epoch 38/40
    1/1 [==
                                     ===] - 0s 13ms/step - loss: 0.2743 - accuracy: 0.9200
    Epoch 39/40
    1/1 [==
                                 ======] - Os 13ms/step - loss: 0.2795 - accuracy: 0.9000
    Epoch 40/40
                             =======] - Os 13ms/step - loss: 0.2394 - accuracy: 0.9400
    1/1 [=====
import matplotlib.pyplot as plt
from keras.utils import load_img
from keras.utils.image_utils import img_to_array
import numpy as np
vat = {1: 'cam tu cau',2:'huong duong',3:'hoa hong',4:'hoa sen',5:'hoa tulip'}
img = load_img("hoa1.jpg",target_size=(40,30))
plt.imshow(ima)
img = img_to_array(img)
img=img.reshape(1,40,30,3)
img = img.astype('float32')
img = img/255
result = np.argmax(model.predict(img),axis=1)
vat[result[0]]
```

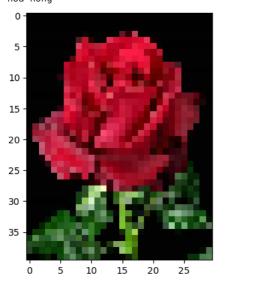
:======] - Os 12ms/step - loss: 1.2575 - accuracy: 0.5200

Epoch 12/40 1/1 [===

Epoch 13/40

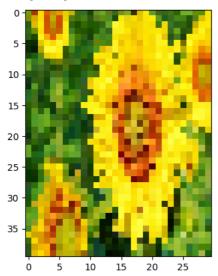
1/1 [===:

vat[result[0]]



```
img = load_img("hoa3.jpg",target_size=(40,30))
plt.imshow(img)
img = img_to_array(img)
img=img.eshape(1,40,30,3)
img = img.astype('float32')
img =img/255
result = np.argmax(model.predict(img),axis=1)
vat[result[0]]
```

1/1 [======] - 0s 19ms/step huong duong $^{\circ}$



```
img = load_img("hoa5.jpeg",target_size=(40,30))
plt.imshow(img)
```

```
img = img_to_array(img)
img=img.reshape(1,40,30,3)
img = img.astype('float32')
img =img/255
result = np.argmax(model.predict(img),axis=1)
vat[result[0]]
```

1/1 [======] - 0s 31ms/step 'hoa tulip'

