In [5]:
```
# example of using a pre-trained model as a classifier
from tensorflow.keras.preprocessing.image import load_img
from tensorflow.keras.preprocessing.image import img_to_array
from keras.applications.vgg16 import preprocess_input
from keras.applications.vgg16 import decode_predictions
from keras.applications.vgg16 import VGG16
```

In [19]:
```
# load an image from file
image = load_img('download.jpg', target_size=(224, 224))
```

In [20]:
```
# convert the image pixels to a numpy array
image = img_to_array(image)
```

In [21]:
```
# reshape data for the model
image = image.reshape((1, image.shape[0], image.shape[1], image.shape[2]))
```

In [22]:
```
# prepare the image for the VGG model
image = preprocess_input(image)
```

In [23]:
```
# load the model
model = VGG16()
```

In [24]:
```
# predict the probability across all output classes
yhat = model.predict(image)
```

```
WARNING:tensorflow:5 out of the last 5 calls to <function Model.make_predict_function.<locals>.predict_function
at 0x000001F2498996C0> triggered tf.function retracing. Tracing is expensive and the excessive number of tracing
s could be due to (1) creating @tf.function repeatedly in a loop, (2) passing tensors with different shapes, (3)
passing Python objects instead of tensors. For (1), please define your @tf.function outside of the loop. For (2)
, @tf.function has reduce_retracing=True option that can avoid unnecessary retracing. For (3), please refer to h
ttps://www.tensorflow.org/guide/function#controlling_retracing and https://www.tensorflow.org/api_docs/python/tf
/function for  more details.
1/1 [==============================] - 1s 726ms/step
```

In [25]:
```
# convert the probabilities to class labels
label = decode_predictions(yhat)
# convert the probabilities to class labels
label = decode_predictions(yhat)
```

In [26]:
```
# retrieve the most likely result, e.g. highest probability
label = label[0][0]
```

In [27]:
```
# print the classification
print('%s (%.2f%%)' % (label[1], label[2]*100))
```

```
castle (34.03%)
```

In [28]:
```
# load an image from file
image = load_img('download2.png', target_size=(224, 224))
```

In [29]:
```
# convert the image pixels to a numpy array
image = img_to_array(image)
```

In [30]:
```
# reshape data for the model
image = image.reshape((1, image.shape[0], image.shape[1], image.shape[2]))
```

In [31]:
```
# prepare the image for the VGG model
image = preprocess_input(image)
```

In [32]:
```
# load the model
model = VGG16()
```

In [33]:
```
# predict the probability across all output classes
yhat = model.predict(image)
```

```
WARNING:tensorflow:6 out of the last 6 calls to <function Model.make_predict_function.<locals>.predict_function
at 0x000001F249898860> triggered tf.function retracing. Tracing is expensive and the excessive number of tracing
s could be due to (1) creating @tf.function repeatedly in a loop, (2) passing tensors with different shapes, (3)
passing Python objects instead of tensors. For (1), please define your @tf.function outside of the loop. For (2)
, @tf.function has reduce_retracing=True option that can avoid unnecessary retracing. For (3), please refer to h
ttps://www.tensorflow.org/guide/function#controlling_retracing and https://www.tensorflow.org/api_docs/python/tf
/function for  more details.
1/1 [==============================] - 1s 1000ms/step
```

In [34]:
```
# convert the probabilities to class labels
label = decode_predictions(yhat)
```

In [35]:
```
# retrieve the most likely result, e.g. highest probability
label = label[0][0]
```

```
label
```

Out[35]: ('n09468604', 'valley', 0.448454)

In [36]:
```
# print the classification
print('%s (%.2f%%)' % (label[1], label[2]*100))
```

```
valley (44.85%)
```

In [37]:
```
# load an image from file
image = load_img('download3.jpg', target_size=(224, 224))
```

In [38]:
```
# convert the image pixels to a numpy array
image = img_to_array(image)
```

In [39]:
```
# reshape data for the model
image = image.reshape((1, image.shape[0], image.shape[1], image.shape[2]))
```

In [46]:
```
# prepare the image for the VGG model
image = preprocess_input(image)
```

In [47]:
```
# load the model
model = VGG16()
```

In [48]:
```
# predict the probability across all output classes
yhat = model.predict(image)
```

```
1/1 [==============================] - 2s 2s/step
```

In [49]:
```
# convert the probabilities to class labels
label = decode_predictions(yhat)
```

In [50]:
```
# retrieve the most likely result, e.g. highest probability
label = label[0][0]
```

In [51]:
```
# print the classification
print('%s (%.2f%%)' % (label[1], label[2]*100))
```

```
golden_retriever (77.67%)
```

In [52]:
```
model.summary()
```

```
Model: "vgg16"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 input_8 (InputLayer)        [(None, 224, 224, 3)]     0

 block1_conv1 (Conv2D)       (None, 224, 224, 64)      1792

 block1_conv2 (Conv2D)       (None, 224, 224, 64)      36928

 block1_pool (MaxPooling2D)  (None, 112, 112, 64)      0

 block2_conv1 (Conv2D)       (None, 112, 112, 128)     73856

 block2_conv2 (Conv2D)       (None, 112, 112, 128)     147584

 block2_pool (MaxPooling2D)  (None, 56, 56, 128)       0

 block3_conv1 (Conv2D)       (None, 56, 56, 256)       295168

 block3_conv2 (Conv2D)       (None, 56, 56, 256)       590080

 block3_conv3 (Conv2D)       (None, 56, 56, 256)       590080

 block3_pool (MaxPooling2D)  (None, 28, 28, 256)       0

 block4_conv1 (Conv2D)       (None, 28, 28, 512)       1180160

 block4_conv2 (Conv2D)       (None, 28, 28, 512)       2359808

 block4_conv3 (Conv2D)       (None, 28, 28, 512)       2359808

 block4_pool (MaxPooling2D)  (None, 14, 14, 512)       0

 block5_conv1 (Conv2D)       (None, 14, 14, 512)       2359808

 block5_conv2 (Conv2D)       (None, 14, 14, 512)       2359808

 block5_conv3 (Conv2D)       (None, 14, 14, 512)       2359808

 block5_pool (MaxPooling2D)  (None, 7, 7, 512)         0

 flatten (Flatten)           (None, 25088)             0

 fc1 (Dense)                 (None, 4096)              102764544

 fc2 (Dense)                 (None, 4096)              16781312

 predictions (Dense)         (None, 1000)              4097000

=================================================================
Total params: 138357544 (527.79 MB)
Trainable params: 138357544 (527.79 MB)
Non-trainable params: 0 (0.00 Byte)
_________________________________________________________________
```

In [ ]: