```
In [1]:
         # 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
         # load an image from file
         image = load_img('download.jpg', target_size=(224, 224))
         # convert the image pixels to a numpy array
         image = img_to_array(image)
         # reshape data for the model
         image = image.reshape((1, image.shape[0], image.shape[1], image.shape[2]))
         # prepare the image for the VGG model
         image = preprocess input(image)
         # Load the model
         model = VGG16()
         # predict the probability across all output classes
         yhat = model.predict(image)
         # convert the probabilities to class labels
         label = decode_predictions(yhat)
         # retrieve the most likely result, e.g. highest probability
         label = label[0][0]
         # print the classification
         print('%s (%.2f%%)' % (label[1], label[2]*100))
        Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/v
        gg16/vgg16_weights_tf_dim_ordering_tf_kernels.h5
        553467904/553467096 [============] - 228s Ous/step
        Downloading data from https://storage.googleapis.com/download.tensorflow.org/data/im
        agenet class index.json
        castle (34.03%)
In [5]:
        # Load an image from file
         image = load_img('download2.png', target_size=(224, 224))
         # convert the image pixels to a numpy array
         image = img_to_array(image)
         # reshape data for the model
         image = image.reshape((1, image.shape[0], image.shape[1], image.shape[2]))
         # prepare the image for the VGG model
         image = preprocess input(image)
         # Load the model
         model = VGG16()
         # predict the probability across all output classes
         yhat = model.predict(image)
         # convert the probabilities to class labels
         label = decode predictions(yhat)
         # retrieve the most likely result, e.g. highest probability
         label = label[0][0]
         # print the classification
         print('%s (%.2f%%)' % (label[1], label[2]*100))
        valley (44.85%)
In [6]:
         # load an image from file
         image = load img('download3.jpg', target size=(224, 224))
         # convert the image pixels to a numpy array
         image = img_to_array(image)
         # reshape data for the model
```

image = image.reshape((1, image.shape[0], image.shape[1], image.shape[2]))

# prepare the image for the VGG model

```
image = preprocess_input(image)
# load the model
model = VGG16()
# predict the probability across all output classes
yhat = model.predict(image)
# convert the probabilities to class labels
label = decode_predictions(yhat)
# retrieve the most likely result, e.g. highest probability
label = label[0][0]
# print the classification
print('%s (%.2f%%)' % (label[1], label[2]*100))
```

WARNING:tensorflow:5 out of the last 5 calls to <function Model.make\_predict\_function n.<locals>.predict\_function at 0x0000021D82424EE8> triggered tf.function retracing. Tracing is expensive and the excessive number of tracings 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 experimental\_relax\_shapes=True option that relaxes argument shapes that can avoid unnecessary retracing. For (3), please refer to https://www.tensorflow.org/tutorials/customization/performance#python\_or\_tensor\_args and https://www.tensorflow.org/api\_docs/python/tf/function for more detail s.

golden\_retriever (84.78%)