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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/vg
        g16/vgg16_weights_tf_dim_ordering_tf_kernels.h5
        553467904/553467096 [===============] - 228s Ous/step
        Downloading data from https://storage.googleapis.com/download.tensorflow.org/data/ima
        genet 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
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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. T racing 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) pas sing Python objects instead of tensors. For (1), please define your @tf.function outs ide 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\_arg s and https://www.tensorflow.org/api\_docs/python/tf/function for more details. golden\_retriever (84.78%)