

Load Train Model From Checkpoint

```
In [1]: import os
import tensorflow as tf
from object_detection.utils import label_map_util
from object_detection.utils import visualization_utils as viz_utils
from object_detection.builders import model_builder
from object_detection.utils import config_util
```

```
In [40]: # Load pipeline config and build a detection model
MODEL_DIR = r'C:\Users\pui_s\Downloads\bird_box\trained_ssd_mobnet'
configs = config_util.get_configs_from_pipeline_file(MODEL_DIR + r'\pipeline.config')
detection_model = model_builder.build(model_config=configs['model'], is_training=False)

# Restore checkpoint
ckpt = tf.compat.v2.train.Checkpoint(model=detection_model)
ckpt.restore(os.path.join(MODEL_DIR, r'checkpoint', 'ckpt-0')).expect_partial()

@tf.function
def detect_fn(image):
    image, shapes = detection_model.preprocess(image)
    prediction_dict = detection_model.predict(image, shapes)
    detections = detection_model.postprocess(prediction_dict, shapes)
    return detections
```

Detect from an Image

```
In [21]: import cv2
import numpy as np
from matplotlib import pyplot as plt
from datetime import datetime
%matplotlib inline
```

```
In [5]: # category_index = label_map_util.create_category_index_from_labelmap(files['LABELMAP'])
category_index = label_map_util.create_category_index_from_labelmap(r'C:\Users\pui_s\Downloads\bird_box\trained_ssd_mobnet\labelmap.pbtxt')
```

```
In [34]: from glob import glob
images = glob(r'C:\Users\pui_s\Downloads\bird_pics_free\*')
```

```
In [41]: t0 = datetime.now()
for IMAGE_PATH in images:
    img = cv2.imread(IMAGE_PATH)
    image_np = np.array(img)

    input_tensor = tf.convert_to_tensor(np.expand_dims(image_np, 0), dtype=tf.float32)
    detections = detect_fn(input_tensor)

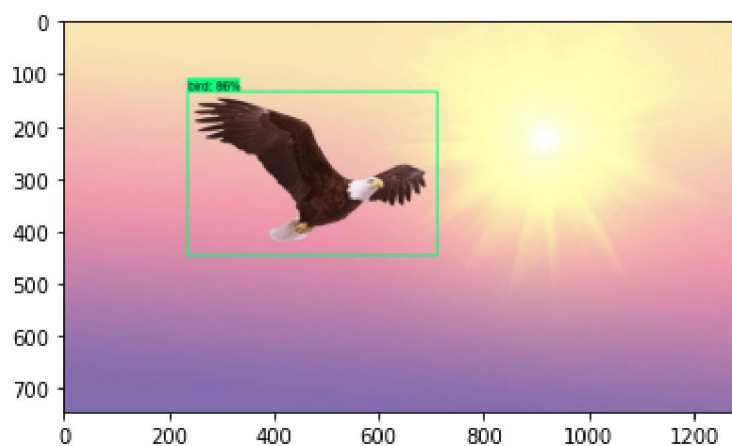
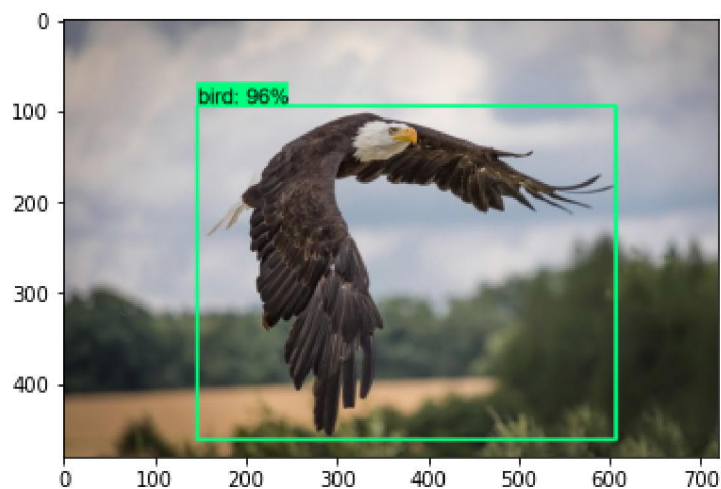
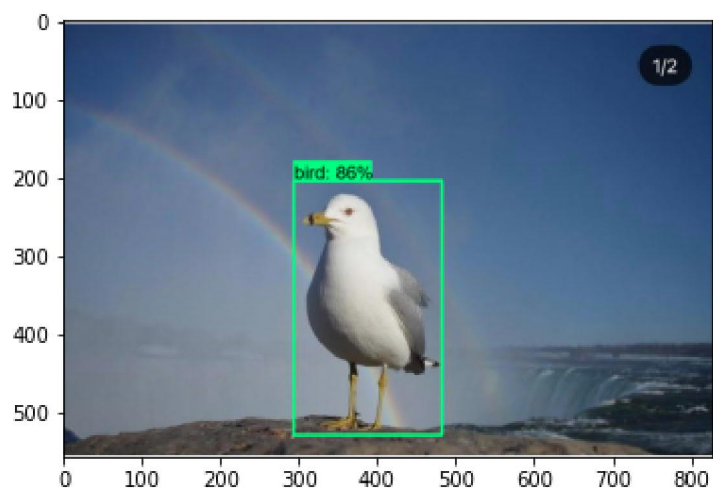
    num_detections = int(detections.pop('num_detections'))
    detections = {key: value[0, :num_detections].numpy()
                  for key, value in detections.items()}
    detections['num_detections'] = num_detections

    # detection_classes should be ints.
    detections['detection_classes'] = detections['detection_classes'].astype(np.int64)

    label_id_offset = 1
    image_np_with_detections = image_np.copy()

    viz_utils.visualize_boxes_and_labels_on_image_array(
        image_np_with_detections,
        detections['detection_boxes'],
        detections['detection_classes']+label_id_offset,
        detections['detection_scores'],
        category_index,
        use_normalized_coordinates=True,
        max_boxes_to_draw=5,
        min_score_thresh=.4,
        agnostic_mode=False)

    plt.imshow(cv2.cvtColor(image_np_with_detections, cv2.COLOR_BGR2RGB))
    plt.show()
print(f'Total Time: {(datetime.now()-t0).total_seconds()} seconds')
```

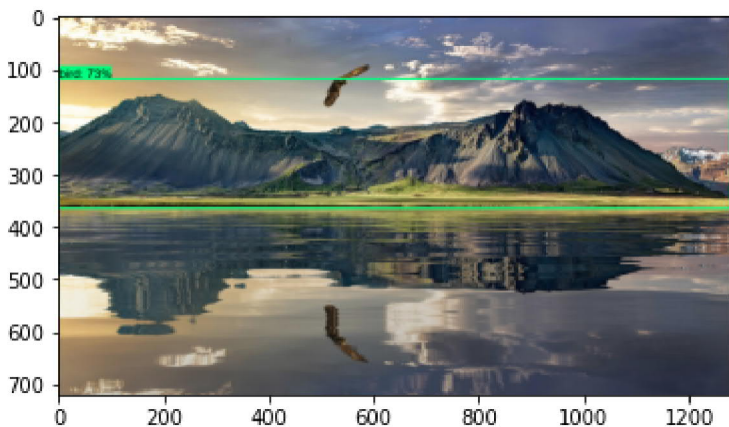
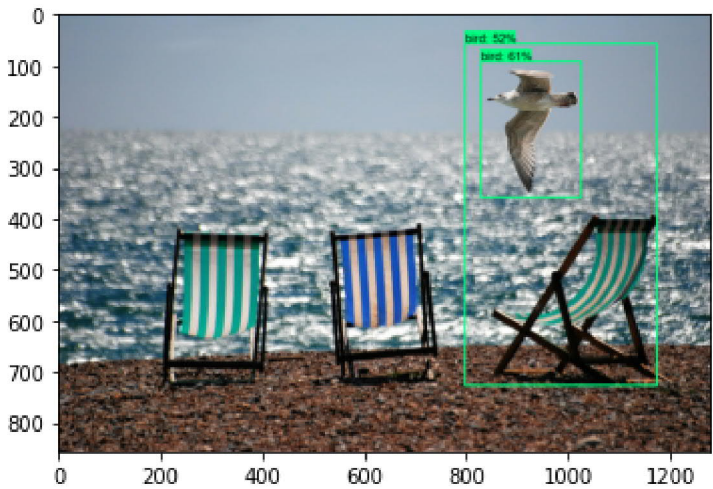
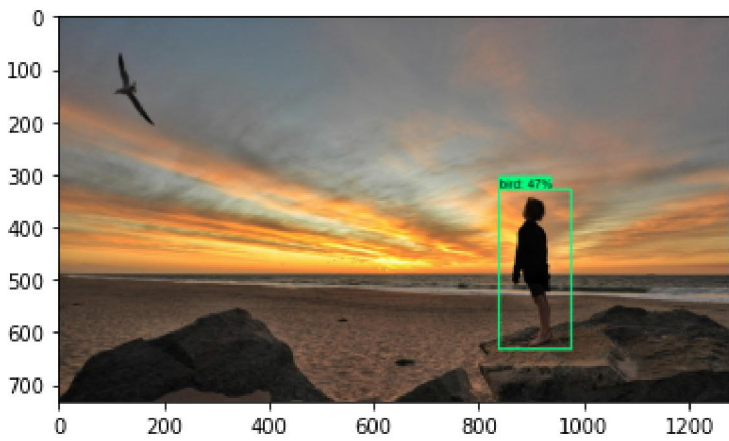
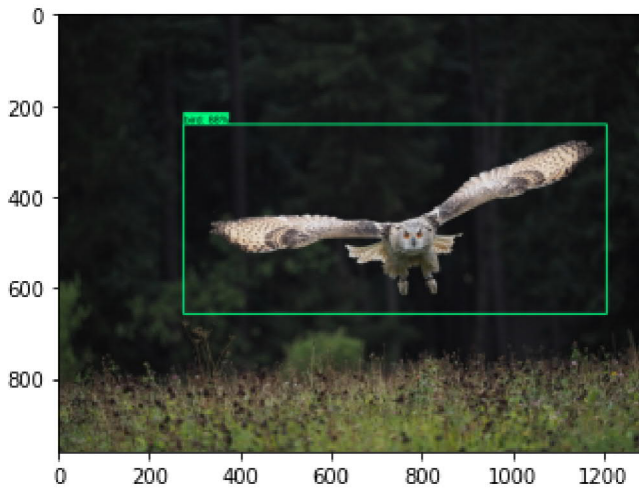


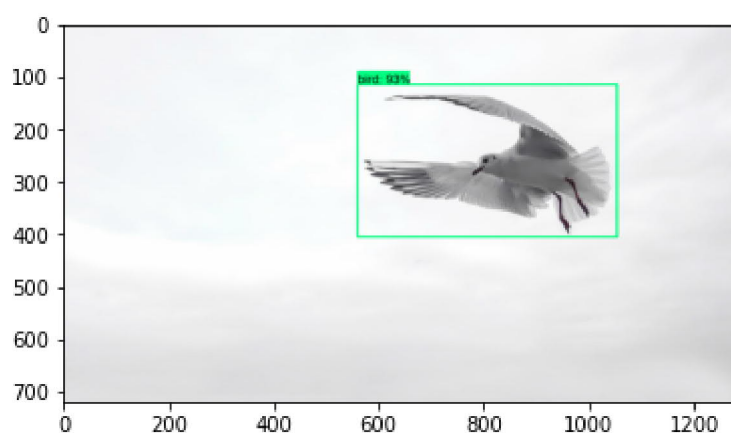
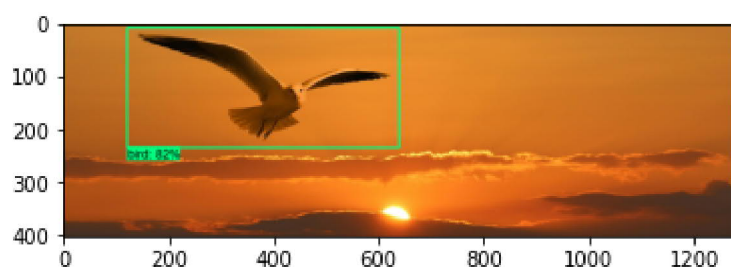
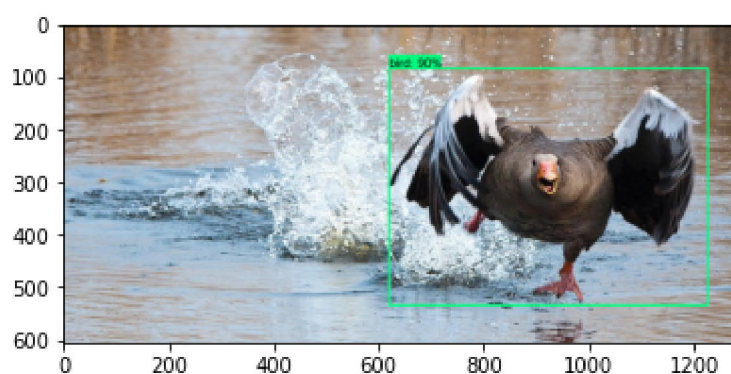
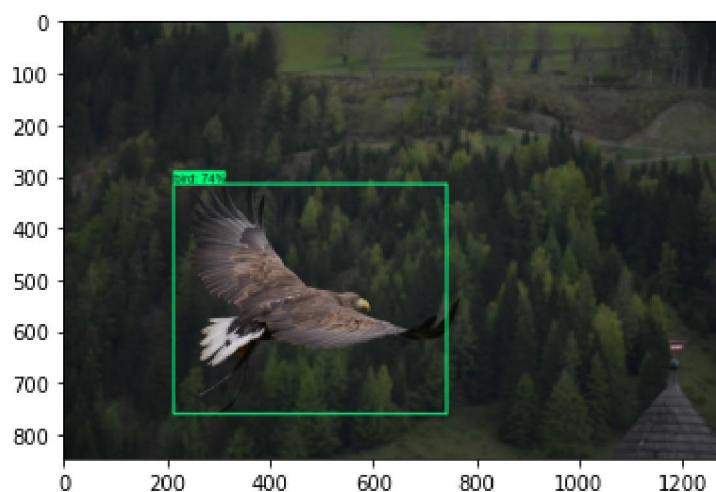
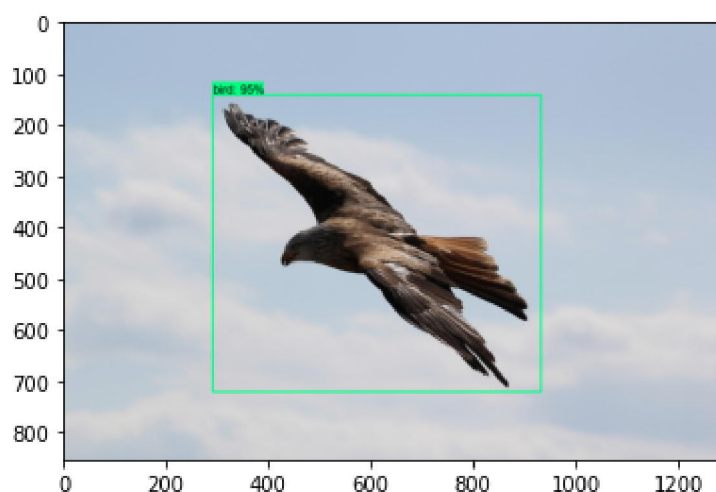
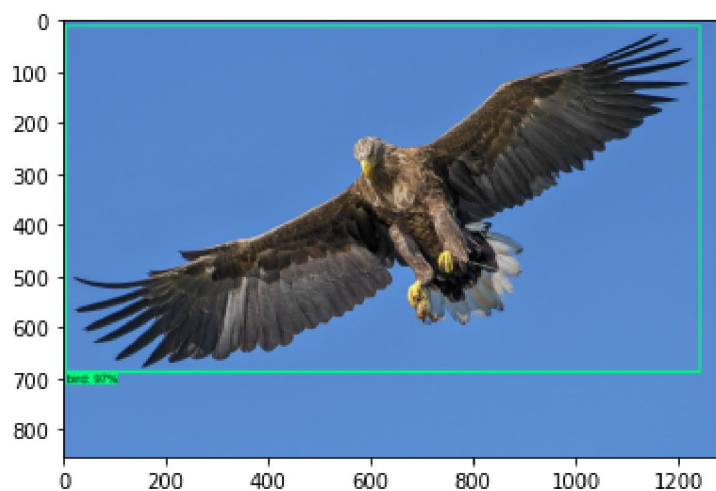
WARNING:tensorflow:5 out of the last 5 calls to <function detect_fn at 0x0000028A10503F70> triggered tf.function retracing. Tracing is expensive and the excessive number of tracings could be due to (1) creating @tf.function repeatedly in a loop, (2) passing tensors with different shapes, (3) passing 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/guide/function#controlling_retracing and https://www.tensorflow.org/api_docs/python/tf/function for more details.

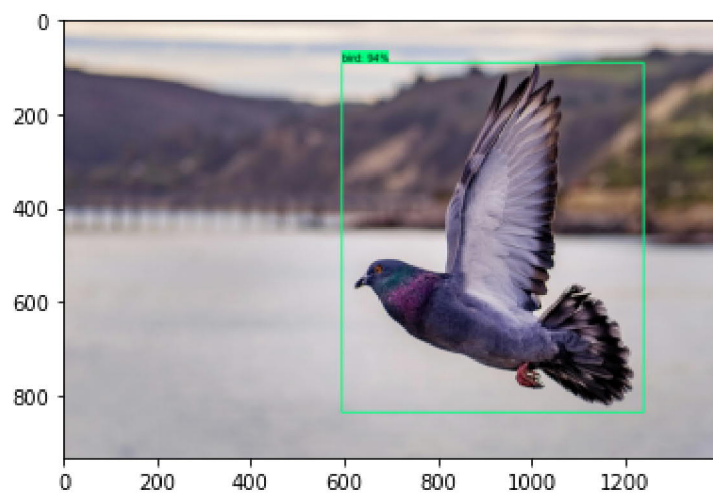
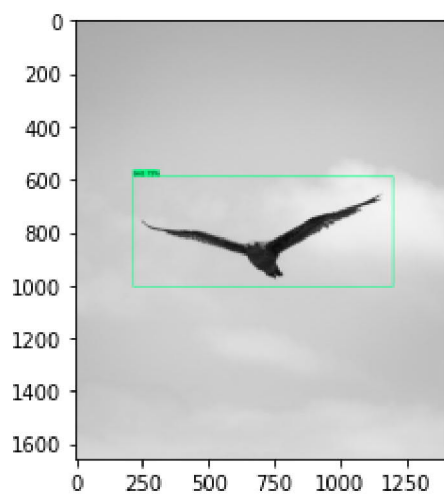
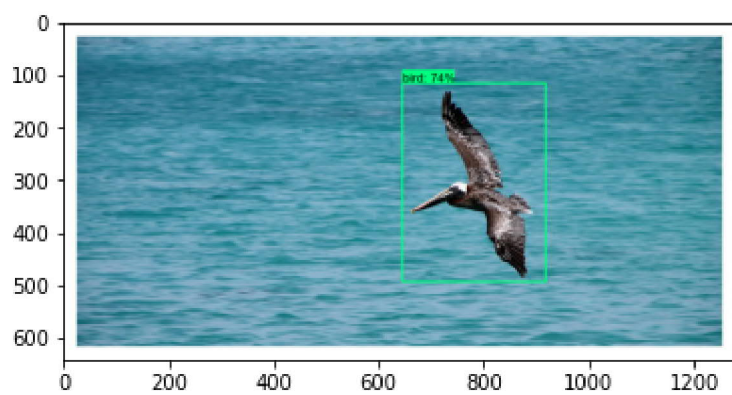
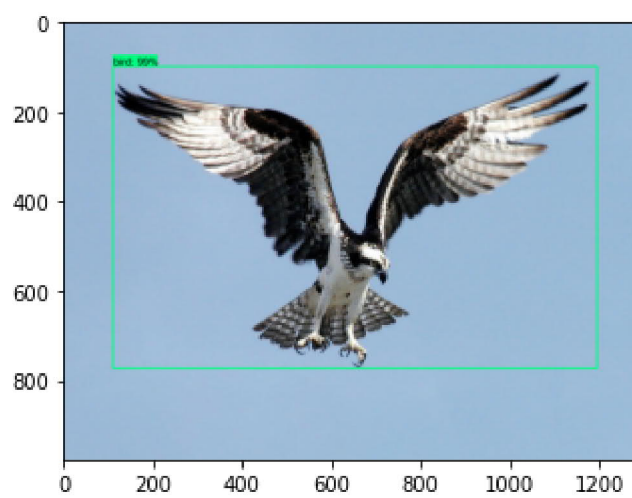
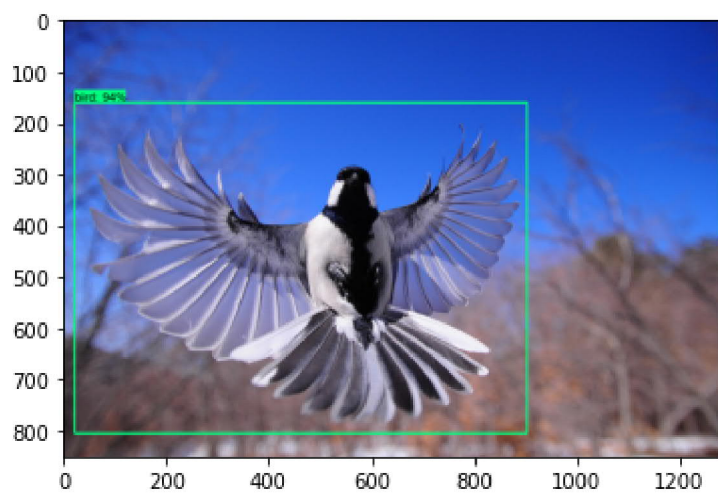


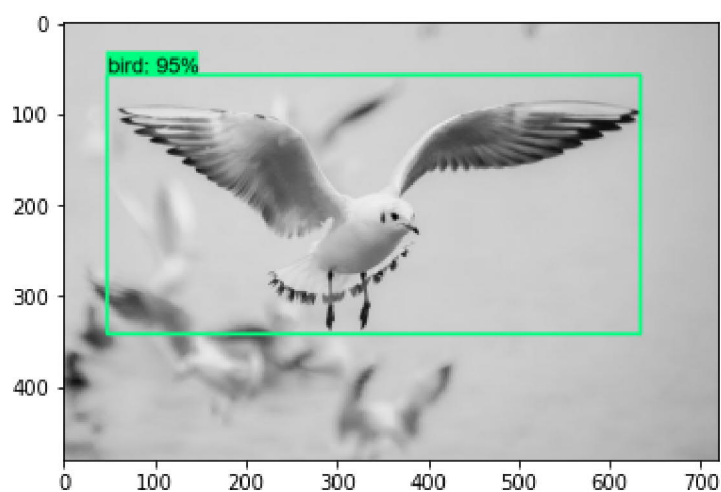
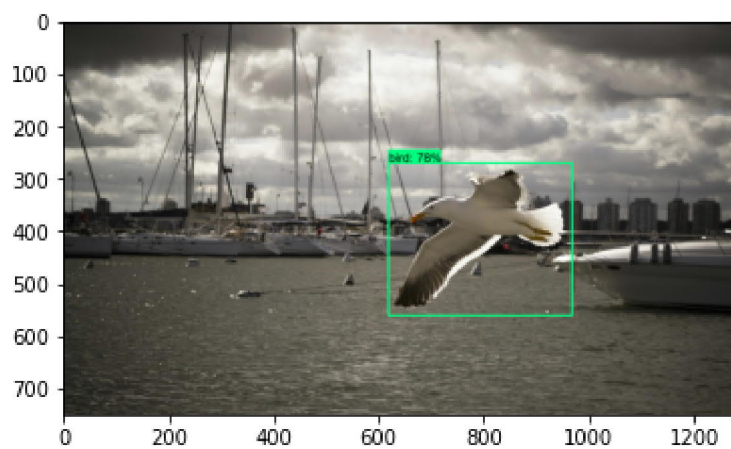
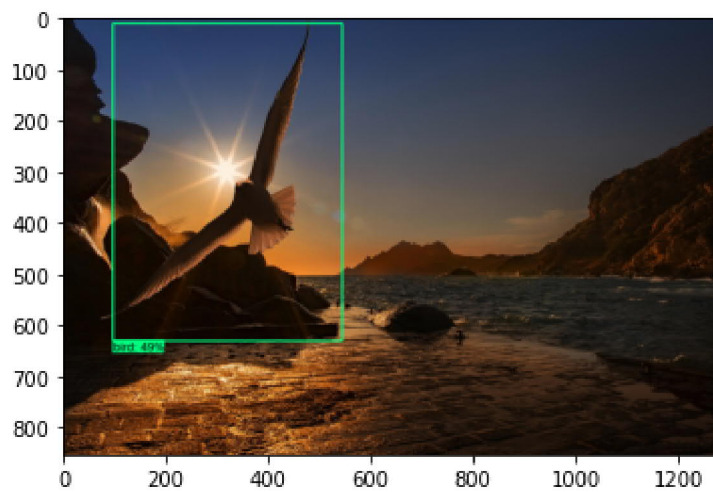
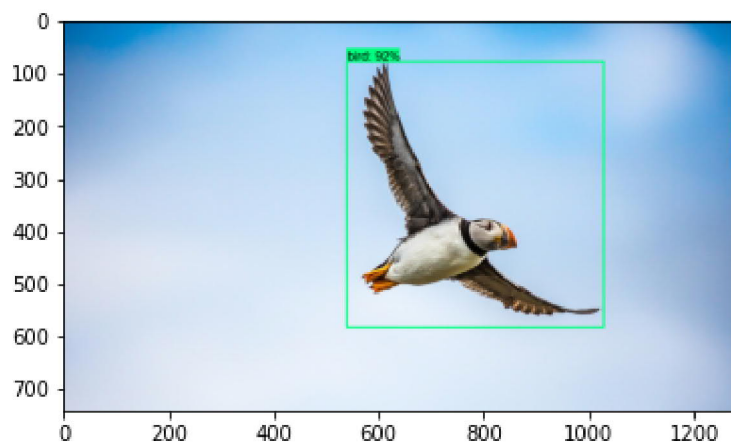
WARNING:tensorflow:6 out of the last 6 calls to <function detect_fn at 0x0000028A10503F70> trigger

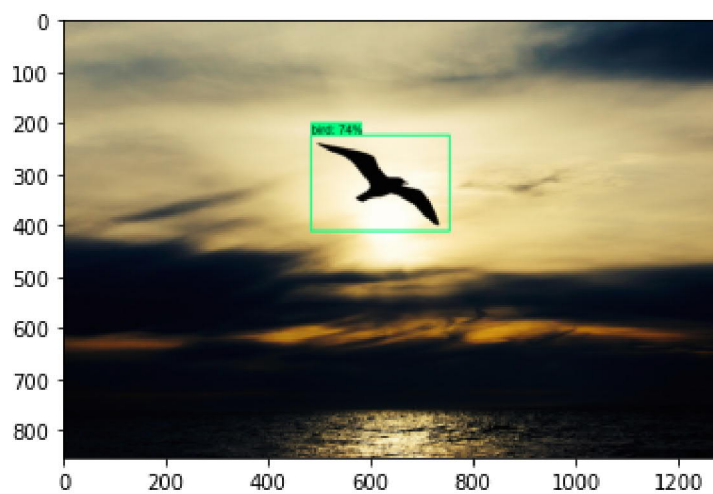
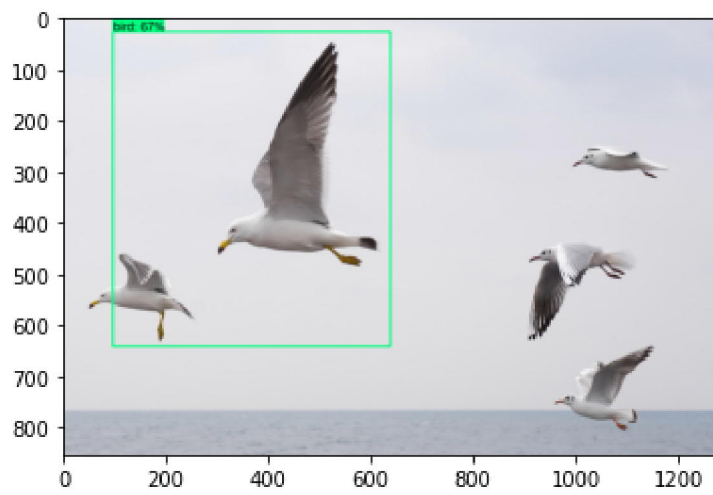
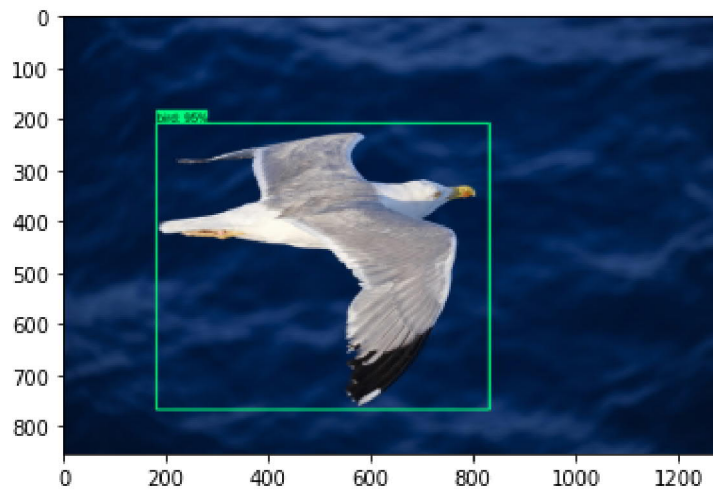
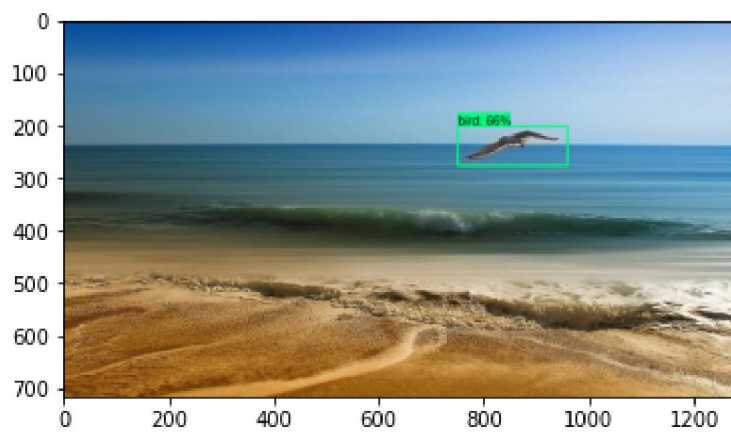
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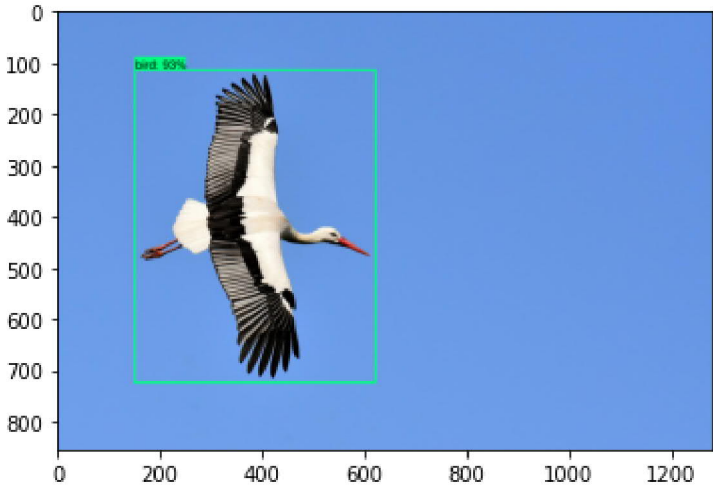
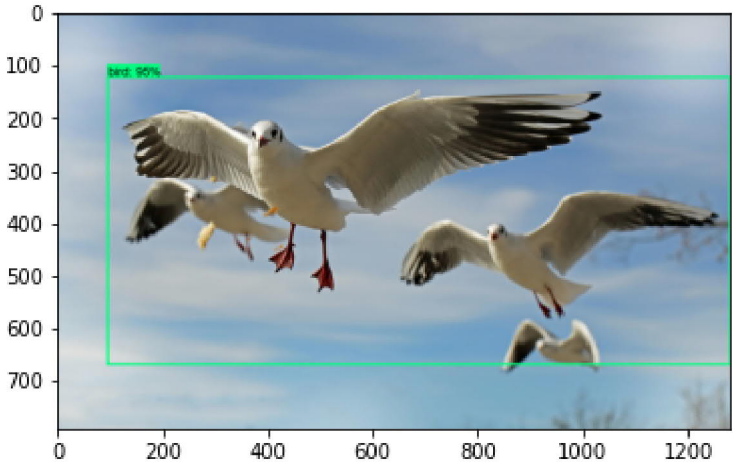
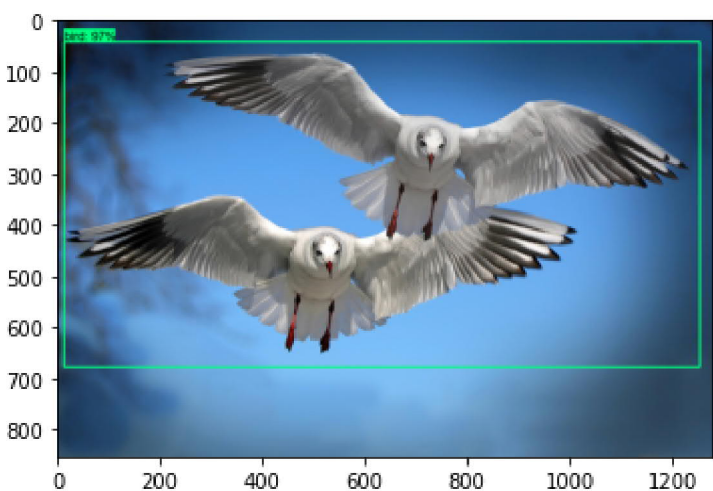












Total Time: 139.650361 seconds

In []: