

Preparing Inputs

[TOC]

To use your own dataset in TensorFlow Object Detection API, you must convert it into the TFRecord file format. This document outlines how to write a script to generate the TFRecord file.

Label Maps

Each dataset is required to have a label map associated with it. This label map defines a mapping from string class names to integer class Ids. The label map should be a **StringIntLabelMap** text protobuf. Sample label maps can be found in `object_detection/data`. Label maps should always start from id 1.

Dataset Requirements

For every example in your dataset, you should have the following information:

1. An RGB image for the dataset encoded as jpeg or png.
2. A list of bounding boxes for the image. Each bounding box should contain:
 1. A bounding box coordinates (with origin in top left corner) defined by 4 floating point numbers [ymin, xmin, ymax, xmax]. Note that we store the *normalized* coordinates (x / width, y / height) in the TFRecord dataset.
 2. The class of the object in the bounding box.

Example Image

Consider the following image:

Example Image

with the following label map:

```
item {
  id: 1
  name: 'Cat'
}
```

```
item {
  id: 2
  name: 'Dog'
}
```

We can generate a `tf.Example` proto for this image using the following code:

```

def create_cat_tf_example(encoded_cat_image_data):
    """Creates a tf.Example proto from sample cat image.

    Args:
        encoded_cat_image_data: The jpg encoded data of the cat image.

    Returns:
        example: The created tf.Example.
    """

    height = 1032.0
    width = 1200.0
    filename = 'example_cat.jpg'
    image_format = b'jpg'

    xmin = [322.0 / 1200.0]
    xmax = [1062.0 / 1200.0]
    ymin = [174.0 / 1032.0]
    ymax = [761.0 / 1032.0]
    classes_text = ['Cat']
    classes = [1]

    tf_example = tf.train.Example(features=tf.train.Features(feature={
        'image/height': dataset_util.int64_feature(height),
        'image/width': dataset_util.int64_feature(width),
        'image/filename': dataset_util.bytes_feature(filename),
        'image/source_id': dataset_util.bytes_feature(filename),
        'image/encoded': dataset_util.bytes_feature(encoded_image_data),
        'image/format': dataset_util.bytes_feature(image_format),
        'image/object/bbox/xmin': dataset_util.float_list_feature(xmin),
        'image/object/bbox/xmax': dataset_util.float_list_feature(xmax),
        'image/object/bbox/ymin': dataset_util.float_list_feature(ymin),
        'image/object/bbox/ymax': dataset_util.float_list_feature(ymax),
        'image/object/class/text': dataset_util.bytes_list_feature(classes_text),
        'image/object/class/label': dataset_util.int64_list_feature(classes),
    }))
    return tf_example

```

Conversion Script Outline

A typical conversion script will look like the following:

```
import tensorflow as tf
```

```

from object_detection.utils import dataset_util

flags = tf.app.flags
flags.DEFINE_string('output_path', '', 'Path to output TFRecord')
FLAGS = flags.FLAGS

def create_tf_example(example):
    # TODO(user): Populate the following variables from your example.
    height = None # Image height
    width = None # Image width
    filename = None # Filename of the image. Empty if image is not from file
    encoded_image_data = None # Encoded image bytes
    image_format = None # b'jpeg' or b'png'

    xmin = [] # List of normalized left x coordinates in bounding box (1 per box)
    xmax = [] # List of normalized right x coordinates in bounding box
                # (1 per box)
    ymin = [] # List of normalized top y coordinates in bounding box (1 per box)
    ymax = [] # List of normalized bottom y coordinates in bounding box
                # (1 per box)
    classes_text = [] # List of string class name of bounding box (1 per box)
    classes = [] # List of integer class id of bounding box (1 per box)

    tf_example = tf.train.Example(features=tf.train.Features(feature={
        'image/height': dataset_util.int64_feature(height),
        'image/width': dataset_util.int64_feature(width),
        'image/filename': dataset_util.bytes_feature(filename),
        'image/source_id': dataset_util.bytes_feature(filename),
        'image/encoded': dataset_util.bytes_feature(encoded_image_data),
        'image/format': dataset_util.bytes_feature(image_format),
        'image/object/bbox/xmin': dataset_util.float_list_feature(xmin),
        'image/object/bbox/xmax': dataset_util.float_list_feature(xmax),
        'image/object/bbox/ymin': dataset_util.float_list_feature(ymin),
        'image/object/bbox/ymax': dataset_util.float_list_feature(ymax),
        'image/object/class/text': dataset_util.bytes_list_feature(classes_text),
        'image/object/class/label': dataset_util.int64_list_feature(classes),
    }))
    return tf_example

def main(_):
    writer = tf.python_io.TFRecordWriter(FLAGS.output_path)

    # TODO(user): Write code to read in your dataset to examples variable

```

```

for example in examples:
    tf_example = create_tf_example(example)
    writer.write(tf_example.SerializeToString())

writer.close()

if __name__ == '__main__':
    tf.app.run()

```

Note: You may notice additional fields in some other datasets. They are currently unused by the API and are optional.

Note: Please refer to the section on Running an Instance Segmentation Model for instructions on how to configure a model that predicts masks in addition to object bounding boxes.

Sharding datasets

When you have more than a few thousand examples, it is beneficial to shard your dataset into multiple files:

- tf.data.Dataset API can read input examples in parallel improving throughput.
- tf.data.Dataset API can shuffle the examples better with sharded files which improves performance of the model slightly.

Instead of writing all tf.Example protos to a single file as shown in conversion script outline, use the snippet below.

```

import contextlib2
from object_detection.dataset_tools import tf_record_creation_util

num_shards=10
output_filebase='/path/to/train_dataset.record'

with contextlib2.ExitStack() as tf_record_close_stack:
    output_tfrecords = tf_record_creation_util.open_sharded_output_tfrecords(
        tf_record_close_stack, output_filebase, num_shards)
    for index, example in examples:
        tf_example = create_tf_example(example)
        output_shard_index = index % num_shards
        output_tfrecords[output_shard_index].write(tf_example.SerializeToString())

```

This will produce the following output files

```

/path/to/train_dataset.record-00000-00010
/path/to/train_dataset.record-00001-00010

```

```
...  
/path/to/train_dataset.record-00009-00010  
which can then be used in the config file as below.  
tf_record_input_reader {  
  input_path: "/path/to/train_dataset.record-?????-of-00010"  
}
```