# Daniil's blog

## Machine Learning and Computer Vision artisan.

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## Tfrecords Guide

Tensorflow and TF-Slim | Dec 21, 2016

A post showing how to convert your dataset to *.tfrecords* file and later on use it as a part of a computational graph.

#### Introduction

In this post we will cover how to convert a dataset into .tfrecord file. Binary files are sometimes easier to use, because you don't have to specify different directories for images and groundtruth annotations. While storing your data in binary file, you have your data in one block of memory, compared to storing each image and annotation separately. Openning a file is a considerably time-consuming operation especially if you use hdd and not ssd, because it involves moving the disk reader head and that takes quite some time. Overall, by using binary files you make it easier to distribute and make the data better aligned for efficient reading.

The post consists of tree parts: in the first part, we demonstrate how you can get raw data bytes of any image using *numpy* which is in some sense similar to what you do when converting your dataset to binary format. Second part shows how to convert a dataset to *tfrecord* file without defining a computational graph and only by employing some built-in *tensorflow* functions. Third part explains how to define a model for reading your data from created binary file and batch it in a random manner, which is necessary during training.

The blog post is created using jupyter notebook. After each chunk of a code you can see the result of its evaluation. You can also get the notebook file from here.

## Getting raw data bytes in numpy

Here we demonstrate how you can get raw data bytes of an image (any ndarray) and how to restore the image back. We point out that during this operation the information about the dimensions of the image is lost and we have to use it to recover the original image. This is one of the reasons why we will have to store the raw image representation along with the dimensions of the original image.

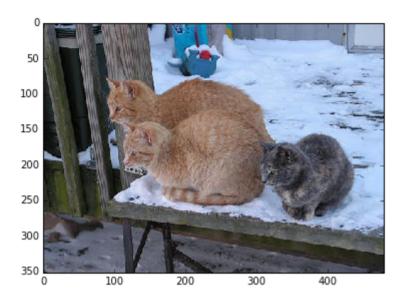
In the following examples, we convert the image into the raw representation, restore it and make sure that the original image and the restored one are the same.

%matplotlib inline

```
import numpy as np
import skimage.io as io

cat_img = io.imread('cat.jpg')
io.imshow(cat img)
```

<matplotlib.image.AxesImage at 0x7f8dc8cb3310>



```
# Let's convert the picture into string representation
# using the ndarray.tostring() function
cat_string = cat_img.tostring()
# Now let's convert the string back to the image
# Important: the dtype should be specified
# otherwise the reconstruction will be errorness
# Reconstruction is 1d, so we need sizes of image
# to fully reconstruct it.
reconstructed_cat_1d = np.fromstring(cat_string, dtype=np.uint8)
# Here we reshape the 1d representation
# This is the why we need to store the sizes of image
# along with its serialized representation.
reconstructed_cat_img = reconstructed_cat_1d.reshape(cat_img.shape)
# Let's check if we got everything right and compare
# reconstructed array to the original one.
np.allclose(cat_img, reconstructed_cat_img)
```

## True

## Creating a .tfrecord file and reading it without defining a graph

Here we show how to write a small dataset (three images/annotations from *PASCAL VOC*) to .tfrrecord file and read it without defining a computational graph.

We also make sure that images that we read back from *.tfrecord* file are equal to the original images. Pay attention that we also write the sizes of the images along with the

image in the raw format. We showed an example on why we need to also store the size in the previous section.

```
# Get some image/annotation pairs for example
filename_pairs = [
('/home/dpakhom1/tf_projects/segmentation/V0Cdevkit/V0Cdevkit/V0C2012/JPEGImages/2007_
000032.jpg',
'/home/dpakhom1/tf_projects/segmentation/V0Cdevkit/V0Cdevkit/V0C2012/SegmentationClas
s/2007_000032.png'),
('/home/dpakhom1/tf_projects/segmentation/V0Cdevkit/V0Cdevkit/V0C2012/JPEGImages/2007_
000039.jpg',
'/home/dpakhom1/tf_projects/segmentation/V0Cdevkit/V0Cdevkit/V0C2012/SegmentationClas
s/2007_000039.png'),
('/home/dpakhom1/tf_projects/segmentation/V0Cdevkit/V0Cdevkit/V0C2012/JPEGImages/2007_
000063.jpg',
'/home/dpakhom1/tf_projects/segmentation/V0Cdevkit/V0Cdevkit/V0C2012/SegmentationClas
s/2007_000063.png')
]
```

```
%matplotlib inline
# Important: We are using PIL to read .png files later.
# This was done on purpose to read indexed png files
# in a special way -- only indexes and not map the indexes
# to actual rgb values. This is specific to PASCAL VOC
# dataset data. If you don't want thit type of behaviour
# consider using skimage.io.imread()
from PIL import Image
import numpy as np
import skimage.io as io
import tensorflow as tf
def _bytes_feature(value):
    return tf.train.Feature(bytes_list=tf.train.BytesList(value=[value]))
def _int64_feature(value):
    return tf.train.Feature(int64_list=tf.train.Int64List(value=[value]))
tfrecords filename = 'pascal voc segmentation.tfrecords'
writer = tf.python_io.TFRecordWriter(tfrecords_filename)
# Let's collect the real images to later on compare
# to the reconstructed ones
original_images = []
for img path, annotation path in filename pairs:
    img = np.array(Image.open(img path))
    annotation = np.array(Image.open(annotation path))
    # The reason to store image sizes was demonstrated
    # in the previous example -- we have to know sizes
    # of images to later read raw serialized string,
    # convert to 1d array and convert to respective
    # shape that image used to have.
    height = img.shape[0]
    width = img.shape[1]
    # Put in the original images into array
    # Just for future check for correctness
```

```
original_images.append((img, annotation))

img_raw = img.tostring()
annotation_raw = annotation.tostring()

example = tf.train.Example(features=tf.train.Features(feature={
    'height': _int64_feature(height),
    'width': _int64_feature(width),
    'image_raw': _bytes_feature(img_raw),
    'mask_raw': _bytes_feature(annotation_raw)}))

writer.write(example.SerializeToString())
```

```
reconstructed images = []
record iterator = tf.python io.tf record iterator(path=tfrecords filename)
for string record in record iterator:
    example = tf.train.Example()
    example.ParseFromString(string record)
    height = int(example.features.feature['height']
                                 .int64 list
                                 .value[0])
    width = int(example.features.feature['width']
                                 .int64 list
                                .value[0])
    img string = (example.features.feature['image raw']
                                  .bytes list
                                  .value[0])
    annotation string = (example.features.feature['mask raw']
                                .bytes list
                                .value[0])
    img_1d = np.fromstring(img_string, dtype=np.uint8)
    reconstructed_img = img_1d.reshape((height, width, -1))
    annotation 1d = np.fromstring(annotation string, dtype=np.uint8)
    # Annotations don't have depth (3rd dimension)
    reconstructed annotation = annotation 1d.reshape((height, width))
    reconstructed_images.append((reconstructed_img, reconstructed_annotation))
```

### Defining the graph to read and batch images from .tfrecords

Here we define a graph to read and batch images from the file that we have created previously. It is very important to randomly shuffle images during training and depending on the application we have to use different batch size.

It is very important to point out that if we use batching - we have to define the sizes of images beforehand. This may sound like a limitation, but actually in the Image Classification and Image Segmentation fields the training is performed on the images of the same size.

The code provided here is partially based on this official example and code from this stackoverflow question. Also if you want to know how you can control the batching according to your need read these docs .

```
%matplotlib inline
import tensorflow as tf
import skimage.io as io
IMAGE HEIGHT = 384
IMAGE WIDTH = 384
tfrecords_filename = 'pascal_voc_segmentation.tfrecords'
def read and decode(filename queue):
    reader = tf.TFRecordReader()
    _, serialized_example = reader.read(filename_queue)
    features = tf.parse_single_example(
      serialized_example,
      # Defaults are not specified since both keys are required.
      features={
        'height': tf.FixedLenFeature([], tf.int64),
        'width': tf.FixedLenFeature([], tf.int64),
        'image_raw': tf.FixedLenFeature([], tf.string),
        'mask_raw': tf.FixedLenFeature([], tf.string)
        })
    # Convert from a scalar string tensor (whose single string has
    # length mnist.IMAGE PIXELS) to a uint8 tensor with shape
    # [mnist.IMAGE PIXELS].
    image = tf.decode_raw(features['image_raw'], tf.uint8)
    annotation = tf.decode_raw(features['mask_raw'], tf.uint8)
    height = tf.cast(features['height'], tf.int32)
    width = tf.cast(features['width'], tf.int32)
    image_shape = tf.pack([height, width, 3])
    annotation_shape = tf.pack([height, width, 1])
```

```
image = tf.reshape(image, image shape)
   annotation = tf.reshape(annotation, annotation shape)
   image size const = tf.constant((IMAGE HEIGHT, IMAGE WIDTH, 3), dtype=tf.int32)
   annotation size const = tf.constant((IMAGE HEIGHT, IMAGE WIDTH, 1),
dtype=tf.int32)
   # Random transformations can be put here: right before you crop images
   # to predefined size. To get more information look at the stackoverflow
   # question linked above.
   resized_image = tf.image.resize_image_with_crop_or_pad(image=image,
                                           target_height=IMAGE_HEIGHT,
                                           target width=IMAGE WIDTH)
    resized_annotation = tf.image.resize_image_with_crop_or_pad(image=annotation,
                                           target height=IMAGE HEIGHT,
                                           target width=IMAGE WIDTH)
    images, annotations = tf.train.shuffle_batch( [resized_image, resized_annotation],
                                                 batch size=2,
                                                 capacity=30,
                                                 num threads=2,
                                                 min after dequeue=10)
   return images, annotations
```

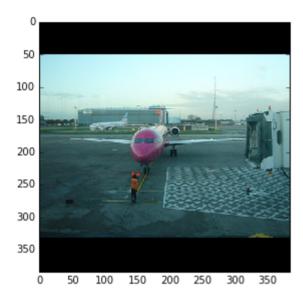
```
filename queue = tf.train.string input producer(
    [tfrecords filename], num epochs=10)
# Even when reading in multiple threads, share the filename
image, annotation = read_and_decode(filename_queue)
# The op for initializing the variables.
init op = tf.group(tf.global variables initializer(),
                   tf.local variables initializer())
with tf.Session() as sess:
    sess.run(init_op)
    coord = tf.train.Coordinator()
    threads = tf.train.start queue runners(coord=coord)
    # Let's read off 3 batches just for example
    for i in xrange(3):
        img, anno = sess.run([image, annotation])
        print(img[0, :, :, :].shape)
        print('current batch')
        # We selected the batch size of two
        # So we should get two image pairs in each batch
        # Let's make sure it is random
        io.imshow(img[\emptyset, :, :, :])
        io.show()
        io.imshow(anno[0, :, :, 0])
        io.show()
```

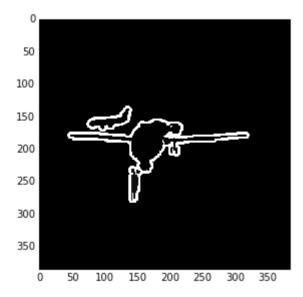
```
io.imshow(img[1, :, :, :])
io.show()

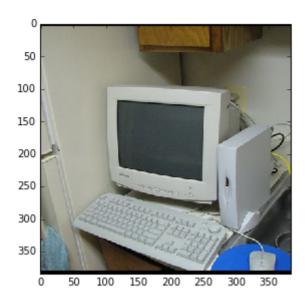
io.imshow(anno[1, :, :, 0])
io.show()

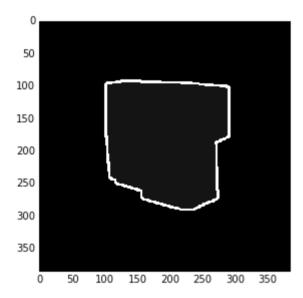
coord.request_stop()
coord.join(threads)
```

(384, 384, 3) current batch

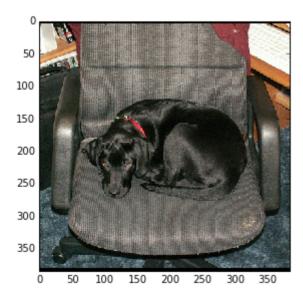


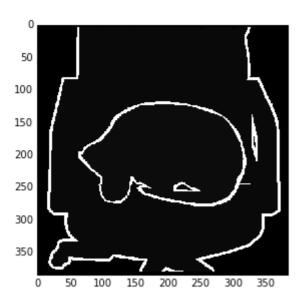


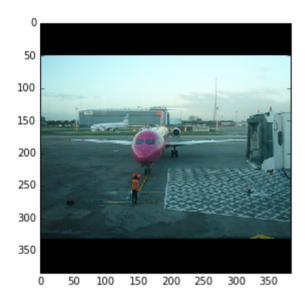


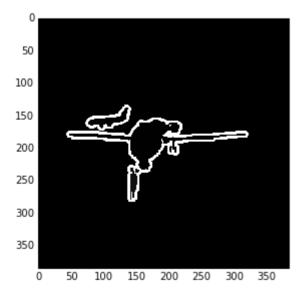


(384, 384, 3) current batch

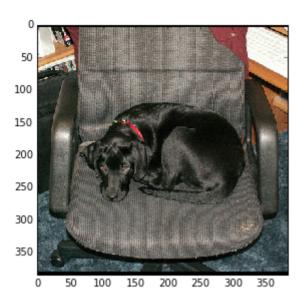


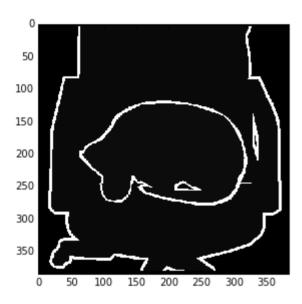


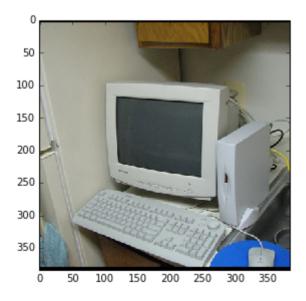


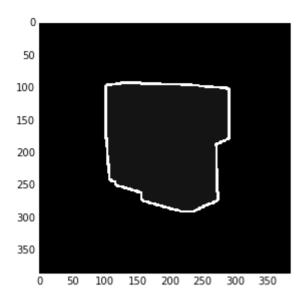


(384, 384, 3) current batch



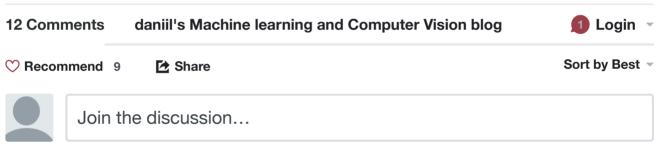






#### Conclusion and Discussion

In this post we covered how to convert a dataset into *.tfrecord* format, made sure that we didn't corrupt the data and saw how to define a graph to read and batch files from the created file.



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# speedplane • 5 months ago

Hi, It looks like tf.pack no longer exists in Tensorflow 1.0. Instead, it should be changed to tf.stack.

2 ^ Reply · Share ›



## BumperDogg • 5 months ago

Can you comment on the choice of data type for writing your record? Why write the image in bytes/string format, and not int64 or something else? And why do you write height/width to record in int64 then recast to int32 when decoding the record?

1 ^ V · Reply · Share ›



Alex Ponamarev → BumperDogg • 8 days ago

int64\_list allows you to serialize integers or a list of integers. However, it will fail if

you try to serialize something more complex - e.g. a list of lists (image). Therefore, converting your images into a string is a better choice. The only problem with this approach is restoration process. Effectively, once you convert you string back into integers, you get a list of integers without any shape. However, you can restore the shape, if you know dimensions! Thus, we need to store the dimensions of the image.

tf.reshape takes images dimensions only in tf.int32 format.

∧ V • Reply • Share ›



#### kirito • 4 months ago

I also build my tfrecords like you did,but, wan i get my image from my tfrecords there is a error like this:ensorflow.python.framework.errors\_impl.OutOfRangeError:

RandomShuffleQueue '\_1\_shuffle\_batch/random\_shuffle\_queue' is closed and has insufficient elements (requested 4, current size 0)

[[Node: shuffle\_batch = QueueDequeueManyV2[component\_types=[DT\_FLOAT, DT\_INT32], timeout\_ms=-1, \_device="/job:localhost/replica:0/task:0/cpu:0"] (shuffle\_batch/random\_shuffle\_queue, shuffle\_batch/n)]]

coulid you please tell me how to slove it?

1 ^ V · Reply · Share ·



## Neil Dencklau → kirito • 4 months ago

I had this same issue, and just figured out what was causing it. I am guessing you changed the number of batches you were looking at, ie changed: for i in xrange(3) to for i in xrange(10). Basically when the FIFOQueue object is defined: filename\_queue = tf.train.string\_input\_producer([tfrecords\_filename], num\_epochs=10) is called, it only allocated enough space for 10 epochs. Once the number of epochs hits the limit defined when you created the FIFOQueue object, the queue closes and cannot provide the images for the next call. You can set the num\_epoch to None and it should work as you expected, or make sure that num\_epoch is the same as the number of times you call the read\_and\_decode function.

∧ V • Reply • Share •



kirito → Neil Dencklau • 3 months ago

i did as you said.but it still not work

1 ^ V · Reply · Share ›



Sophie → Neil Dencklau • 25 days ago

I'm having same issue. T.T

∧ ∨ • Reply • Share •



## Youngwook Paul Kwon • 2 months ago

Thanks for your helpful post. What if I have multiple TFRecord files? In 'filename\_queue = tf.train.string\_input\_producer(

[tfrecords\_filename], num\_epochs=10)', do you think I can use a list of TFRecore files such as ['a.tfrecord', 'b.tfrecord']?

∧ V • Reply • Share ›



Alexander Gomez Villa · 4 months ago

Thank you! This post was very helpful



applegreen • 5 months ago

Thank you for nice article.

Would you mind if I ask you a question?

I got error on "img = tf.Session.run([image, annotation])".

The error is "TypeError: run() missing 1 required positional argument: 'fetches'".

I wonder why this error occurs.

∧ V • Reply • Share >



satwik kondamudi → applegreen • 6 days ago

in tf.Session() you are sending in two arguments(image and annotation). You need to have two objects for collecting this output but you are having only one output i.e. img.

∧ V • Reply • Share •



Ishrat Badami • 5 months ago

Excellent Article! thank you for explaining it so simply!

∧ V • Reply • Share •

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1 comment • 9 months ago•

AvatarAndrew Loerch — Very nice and wellexplained tutorial! Thank you for the clear explanations on how to accomplish

landaria (Carana and Albardaria) Tanadaria

16 comments ● 7 months ago ●

**Image Segmentation** 

AvatarArun M. Krishnakumar — FWIW, I made some changes and pushed into a fork: https://github.com/arunmk/t...The fork has

**Fully Convolutional Networks (FCNs) for** 

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