

Mask-RCNN using TensorFlow on Windows

The steps for Mask-RCNN are similar to TFOD

Download the file and extract the file

Download link:

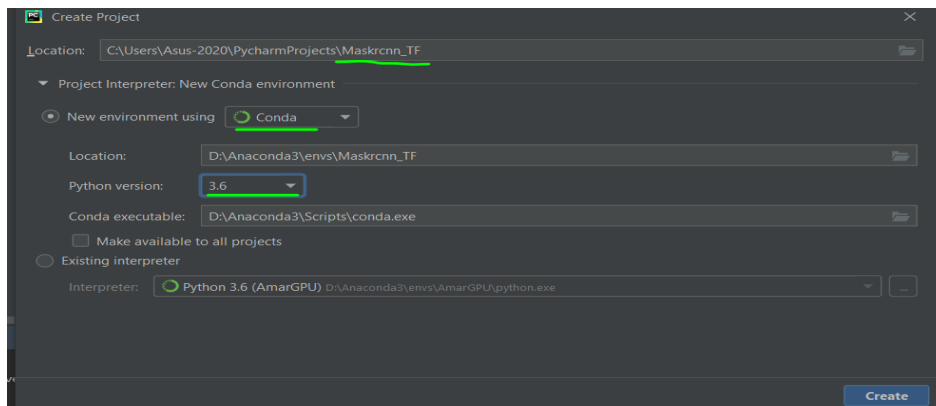
<https://drive.google.com/file/d/1j2a6mvg4dLdm3vTKNRQITmihc3Xf4hDi/view>

Once the file is downloaded then do the extraction process from the zip file.

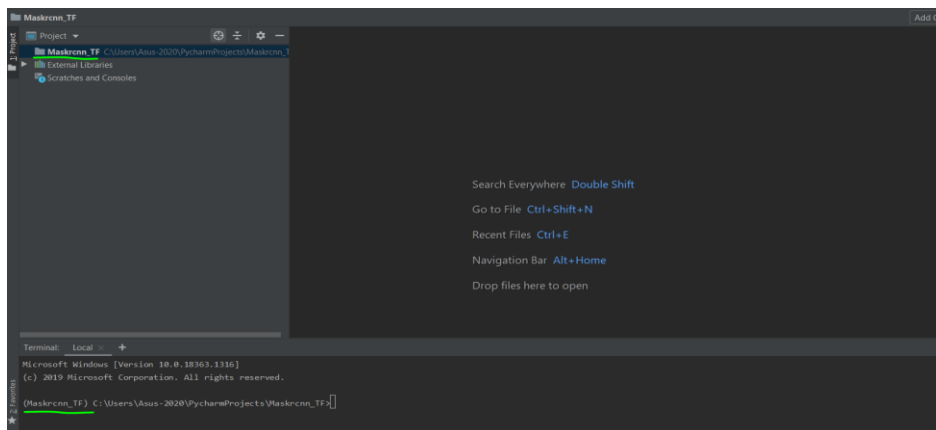
data	6/13/2020 9:19 PM	File folder	
models	6/13/2020 5:53 PM	File folder	
requirements	6/14/2020 4:19 PM	Text Document	2 KB

Open PyCharm

File --> New Project --> Project name --> conda environment --> python version --> click create

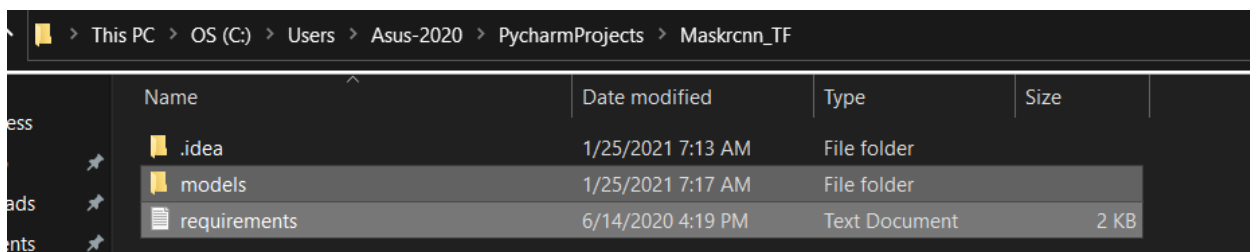
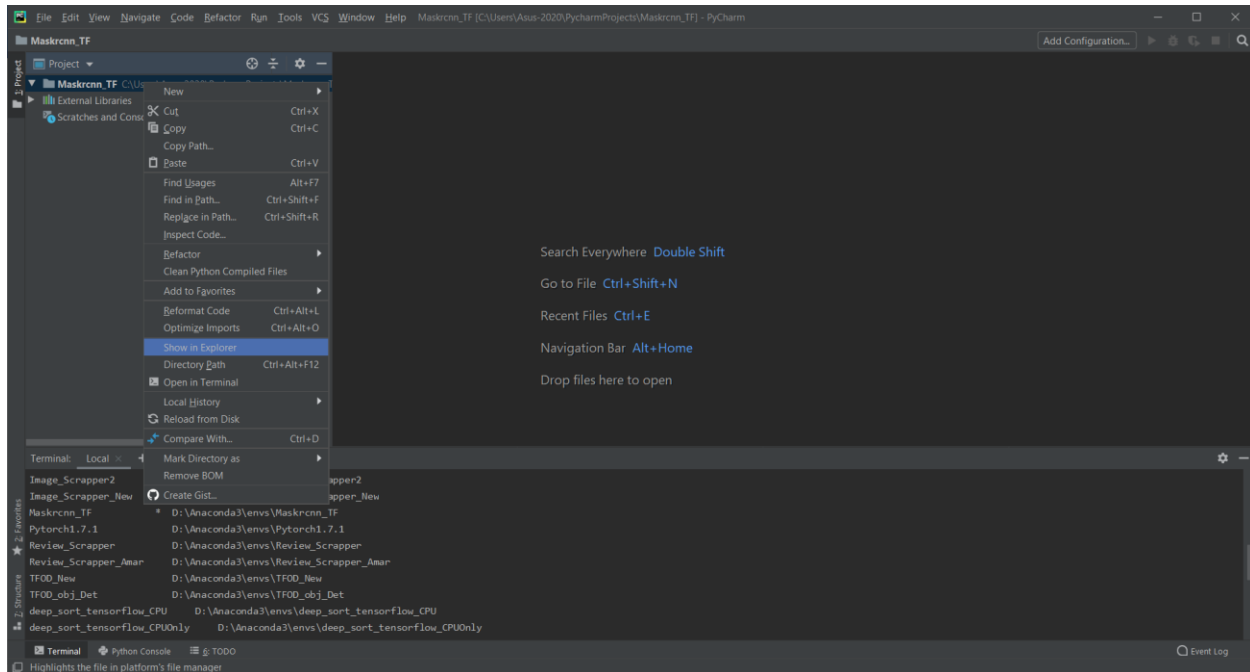


New environment and project are created.

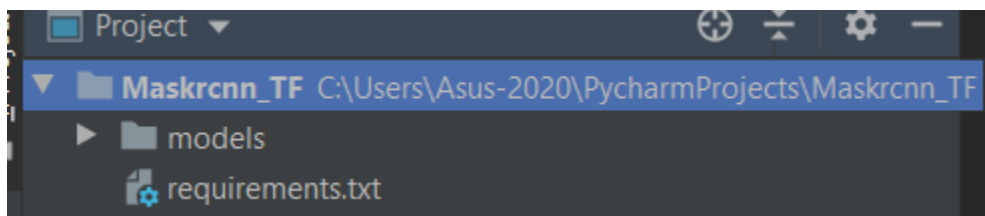


Now copy the two files models folder and requirements.txt and paste it into the python project.

Right click on the Project -- > then select show in explorer



I have moved those files in to the Maskrcnn project location.



In PyCharm also it has updated the folders in our project section.

Now we need to install the project dependencies, in order to do that go the terminal in PyCharm and execute the command pip install -r requirements.txt

```
(Maskrcnn_TF) C:\Users\Asus-2020\PycharmProjects\Maskrcnn_TF>pip install -r requirements.txt
```

```
Successfully installed Cython-0.29.20 Jinja2-2.11.2 Keras-Applications-1.0.8 Keras-Preprocessing-1.1.2 Markdown-3.2.2 MarkupSafe-1.1.1 Pillow-7.1.2 Pygments-2.6.1 QtPy-1.9.0 Send2Trash-1.5.0 Werkzeug-1.0.1 absl-
py-0.9.0 astor-0.8.1 attrrs-19.3.0 backcall-0.2.0 bleach-3.1.5 certifi-2020.4.5.1 colorama-0.4.4 contextlib2-0.6.0.post1 cyciler-0.10.0 decorator-4.4.2 defusedxml-0.6.0 entrypoints-0.3 gast-0.3.3 google-pasta-0.2.
0 grpcio-1.29.0 h5py-2.10.0 importlib-metadata-1.6.1 ipykernel-5.3.0 ipython-7.15.0 ipython-genutils-0.2.0 ipywidgets-7.5.1 jedi-0.17.0 jsonschema-3.2.0 jupyter-1.0.0 jupyter-client-6.1.3 jupyter-console-6.1.0 j
upyter-core-4.6.3 kiwisolver-1.2.0 lml-4.5.1 matplotlib-3.2.1 mistune-0.8.4 nbconvert-5.6.1 nbformat-5.0.7 notebook-6.0.3 numpy-1.18.5 opencv-python-4.2.0.34 packaging-20.4 pandas-1.0.4 pandocfilters-1.4.2 pars
o-0.7.0 pexpect-4.8.0 pickleshare-0.7.5 prometheus-client-0.8.0 prompt-toolkit-3.0.5 protobuf-3.12.2 ptyprocess-0.6.0 pyparsing-2.4.7 pyrsistent-0.16.0 python-dateutil-2.8.1 pytz-2020.1 pywin32-300 pywinpty-0.5.
7 pyzmq-19.0.1 qtconsole-4.7.4 six-1.15.0 tensorboard-1.14.0 tensorflow-1.14.0 tensorflow-estimator-1.14.0 termcolor-1.1.0 terminado-0.8.3 testpath-0.4.4 tornado-6.0.4 traitlets-4.3.3 wcwidth-0.2.4 webencodings-
0.5.1 widgetsnbextension-3.5.1 wrapt-1.12.1 zipp-3.1.0

(Maskrcnn_TF) C:\Users\Asus-2020\PycharmProjects\Maskrcnn_TF>
```

Here the installation of packages are installed successfully in our environment.

Setup.py

There is a setup.py file inside the research folder.

Navigate to the path cd/models/research inside the terminal

```
(Maskrcnn_TF) C:\Users\Asus-2020\PycharmProjects\Maskrcnn_TF>cd models/research
(Maskrcnn_TF) C:\Users\Asus-2020\PycharmProjects\Maskrcnn_TF\models\research>python setup.py install
```

Successfully done

```
Using d:\anaconda3\envs\maskrcnn_tf\lib\site-packages
Finished processing dependencies for object-detection==0.1
```

Annotation:

Name of the tool: LabelMe

A screenshot of a Google search interface. The search bar contains the text "labelme annotation tool github". Below the search bar, there are tabs for "All", "Videos", "Images", "News", "Maps", and "More". The "All" tab is selected. Below the tabs, it says "About 6,66,000 results (0.90 seconds)". The first search result is from "github.com" and is titled "wkentaro/labelme: Image Polygonal Annotation with ... - GitHub". The description says "Labelme is a graphical image annotation tool inspired by http://labelme.csail.mit.edu. It is written in Python and uses Qt ...". Below the description are links for "README.md", "Wkentaro/labelme", "Labelme.spec", and "Pull requests".

Link to download the specific versions:

<https://github.com/wkentaro/labelme/releases>

Don't download the latest version.

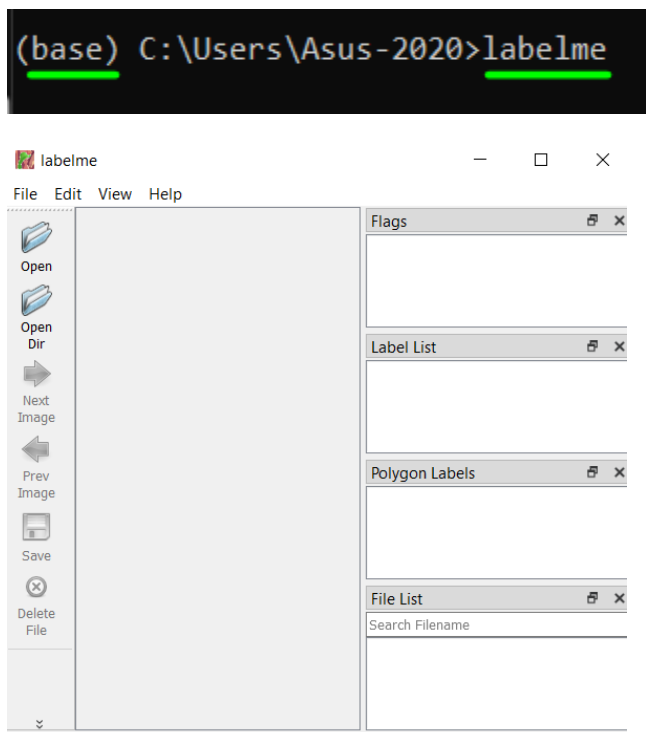
Here I'm going to install the specific version 3.16.7

In order to install the package of this LabelMe, create a new anaconda environment or you can proceed inside the base environment.

Here I have installed into my base environment

```
(base) C:\Users\Asus-2020>pip install labelme==3.16.7
Collecting labelme==3.16.7
  Downloading https://files.pythonhosted.org/packages/1d/7e/41bd6f37f42c289164029bd10c5de82d85b95f087257ec6eae9befa/labelme-3.16.7.tar.gz (1.5MB)
    | 1.5MB 1.6MB/s
Requirement already satisfied: matplotlib in d:\anaconda3\lib\site-packages (from labelme==3.16.7) (3.1.1)
Requirement already satisfied: numpy in d:\anaconda3\lib\site-packages (from labelme==3.16.7) (1.16.5)
Requirement already satisfied: Pillow>=2.8.0 in d:\anaconda3\lib\site-packages (from labelme==3.16.7) (6.2.0)
Requirement already satisfied: PyYAML in d:\anaconda3\lib\site-packages (from labelme==3.16.7) (5.1.2)
Requirement already satisfied: qtpy in d:\anaconda3\lib\site-packages (from labelme==3.16.7) (1.9.0)
Requirement already satisfied: termcolor in d:\anaconda3\lib\site-packages (from labelme==3.16.7) (1.1.0)
Requirement already satisfied: cyclor>=0.10 in d:\anaconda3\lib\site-packages (from matplotlib->labelme==3.16.7) (0.1.0)
Requirement already satisfied: kiwisolver>=1.0.1 in d:\anaconda3\lib\site-packages (from matplotlib->labelme==3.16.7) (1.0.1)
Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in d:\anaconda3\lib\site-packages (from matplotlib->labelme==3.16.7) (2.4.2)
```

Open the LabelMe tool

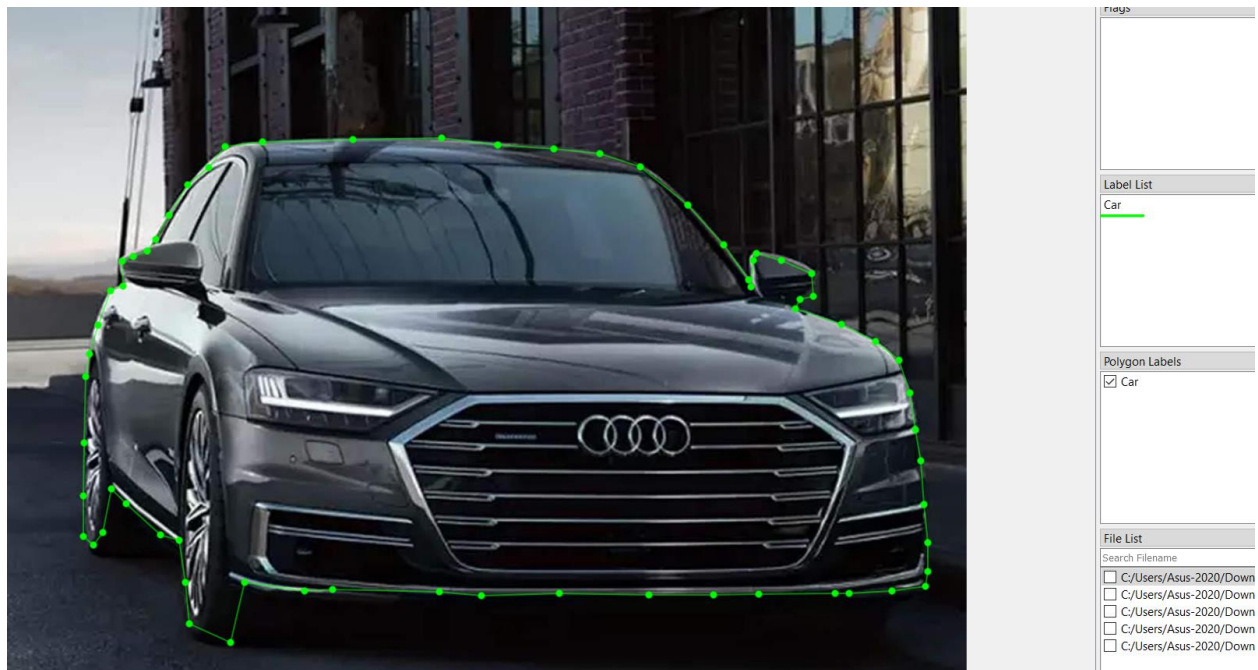
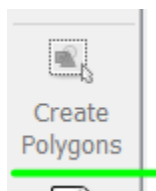


Open dir -- > select the folder to annotate

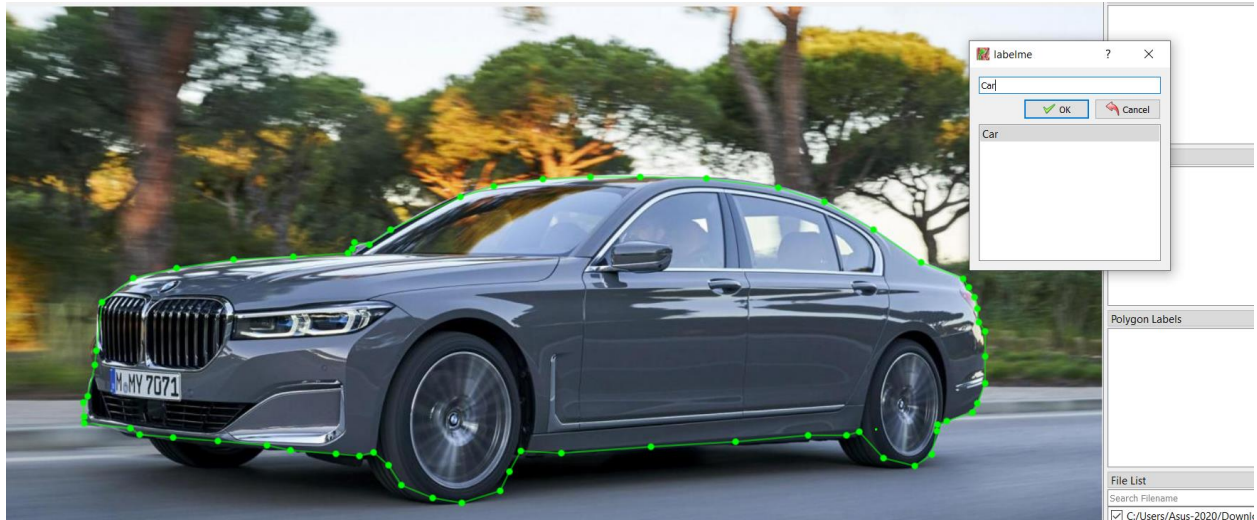
Sample annotations:



To do the annotation click on create polygons



Save the file



Once you annotated for all the images in that folder then corresponding .json files are created for each and every image.

This PC > Downloads > Cars					
	Name	Date modified	Type	Size	
ss	Audi	1/25/2021 8:31 AM	JPG File	89 KB	
ds	Audi	1/25/2021 10:01 AM	JSON File	129 KB	
ts	bmw	1/25/2021 8:33 AM	JPG File	111 KB	
	bmw	1/25/2021 10:05 AM	JSON File	153 KB	
	mercedes-benz	1/25/2021 8:29 AM	JPG File	59 KB	
	mercedes-benz	1/25/2021 10:07 AM	JSON File	97 KB	
ogonizati	rolls-royce	1/25/2021 10:09 AM	JSON File	1,169 KB	
	rolls-royce	1/25/2021 8:32 AM	PNG File	873 KB	
	tata-altroz	1/25/2021 8:31 AM	JPG File	40 KB	
	tata-altroz	1/25/2021 10:11 AM	JSON File	59 KB	

```

{
  "version": "3.16.7",
  "flags": {},
  "shapes": [
    {
      "label": "Car",
      "line_color": null,
      "fill_color": null,
      "points": [
        [
          293.4761904761905,
          180.9047619047619
        ],
        [
          339.19047619047615,
          175.19047619047618
        ],
        [
          449.66666666666663,
          172.33333333333331
        ],
        [
          558.2380952380952,
          170.42857142857142
        ],
        [
          626.8095238095237,
          179.0
        ],
        [
          695.3809523809523,
          183.76190476190476
        ],
        [
          751.5714285714286,
          189.47619047619048
        ]
      ]
    }
  ]
}

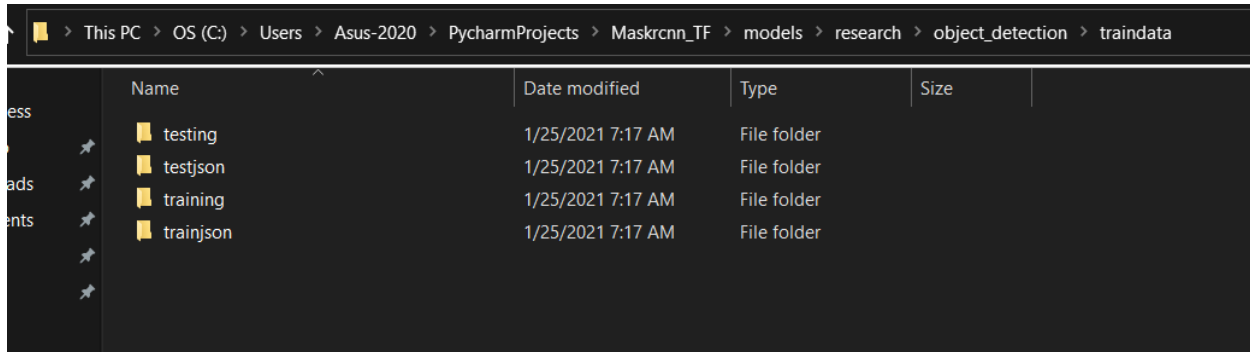
```

Three files are important – 1. TF record file, 2. Labelmap file, 3. dataset

The data's inside the JSON file looks like above.

Actually our dataset should be in the form of training and test, which is located or created inside the

pycharmProjects --> Projectname (Maskrcnn_TF) --> models --> research --> object_detection --> traindata



Here we have 4 folders 2 for test and 2 for train (dog class)

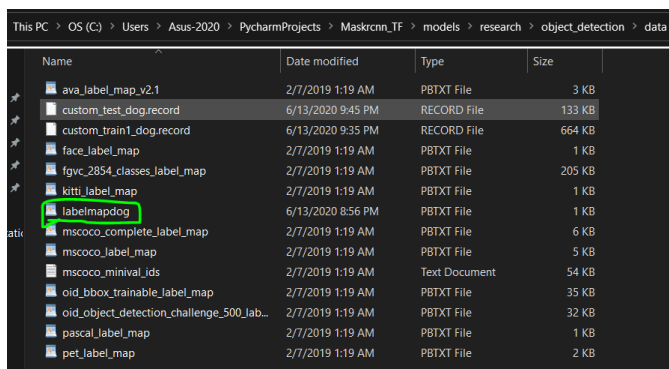
Testing and training folder contains only images

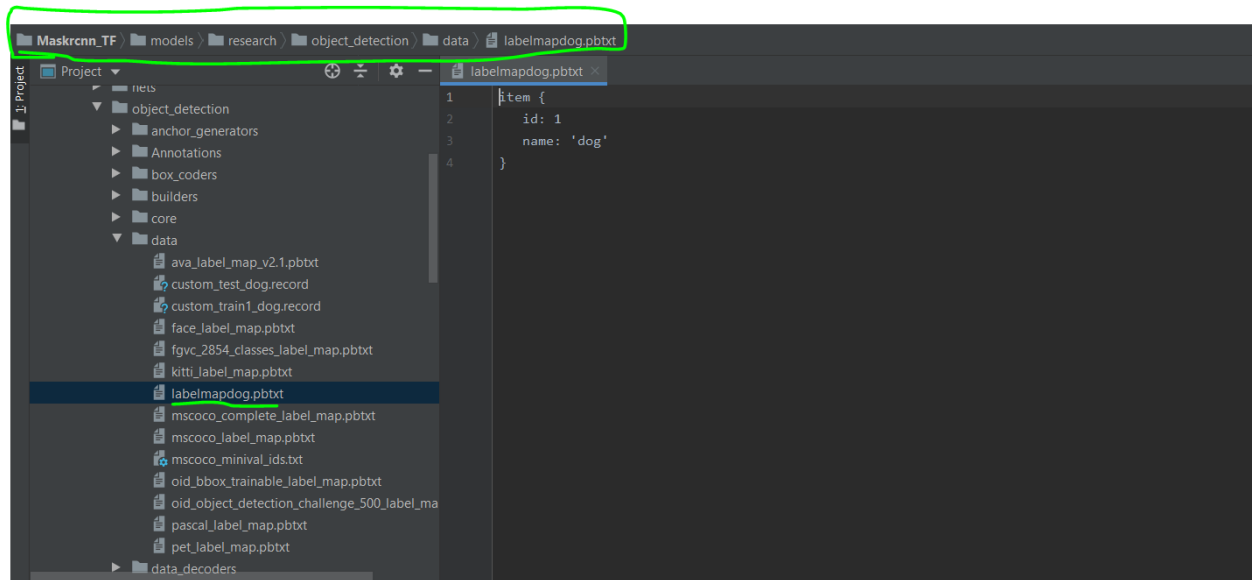
Testjson and trainjson contains the json files

According to the community of tensorflow all the lablemap.pbtxt are available inside the data folder.

pycharmProjects --> Projectname (Maskrcnn_TF) --> models --> research --> object_detection --> data --> lablemapdog.pbtxt

In that data folder we can able to see many .pbtxt files and I have used only the lablemapdog.pbtxt file.

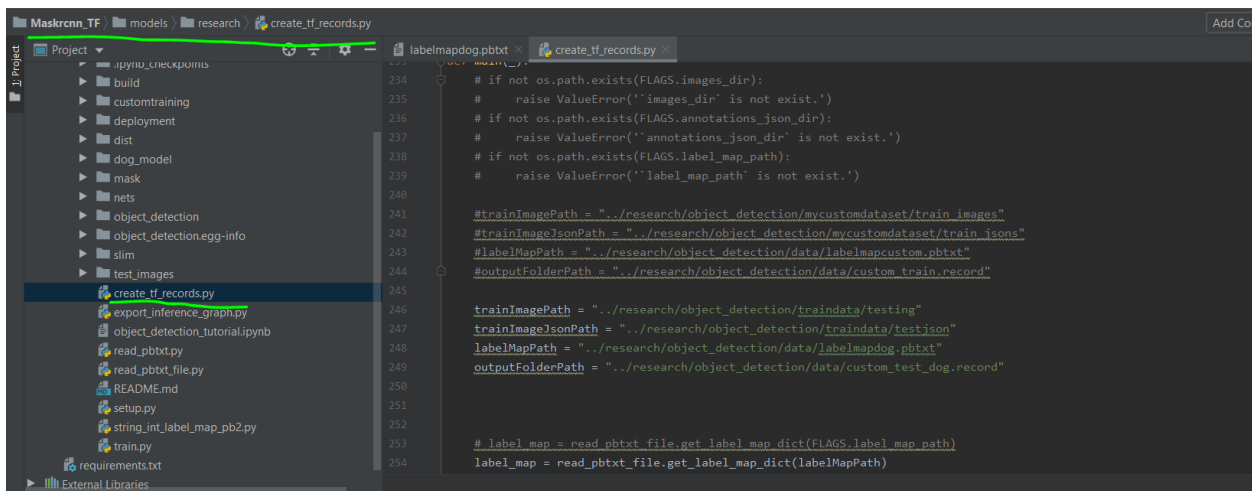




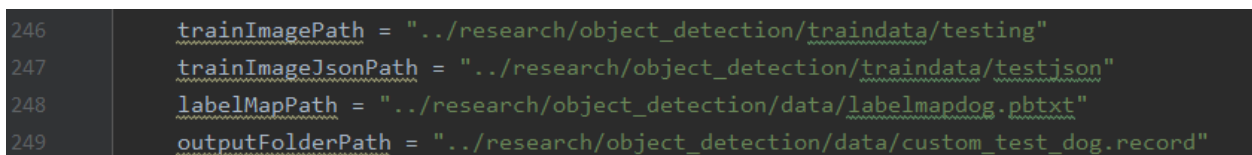
In this file we need to mention the number of classes.

Generate TF record:

pycharmProjects --> Projectname (Maskrcnn_TF) --> models --> research --> create_tf_records.py



In line number 246 to 249 this are the utmost important in the create_tf_records.py file

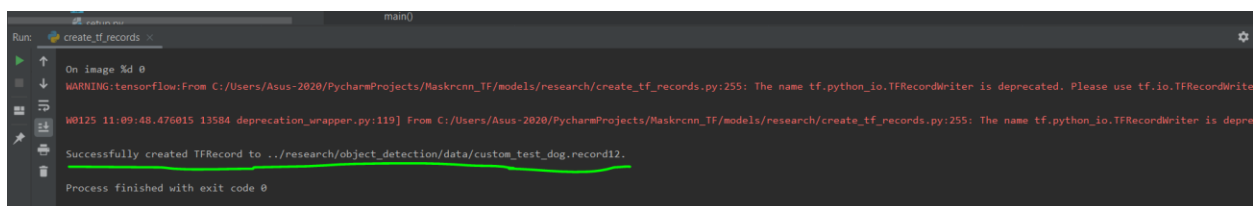
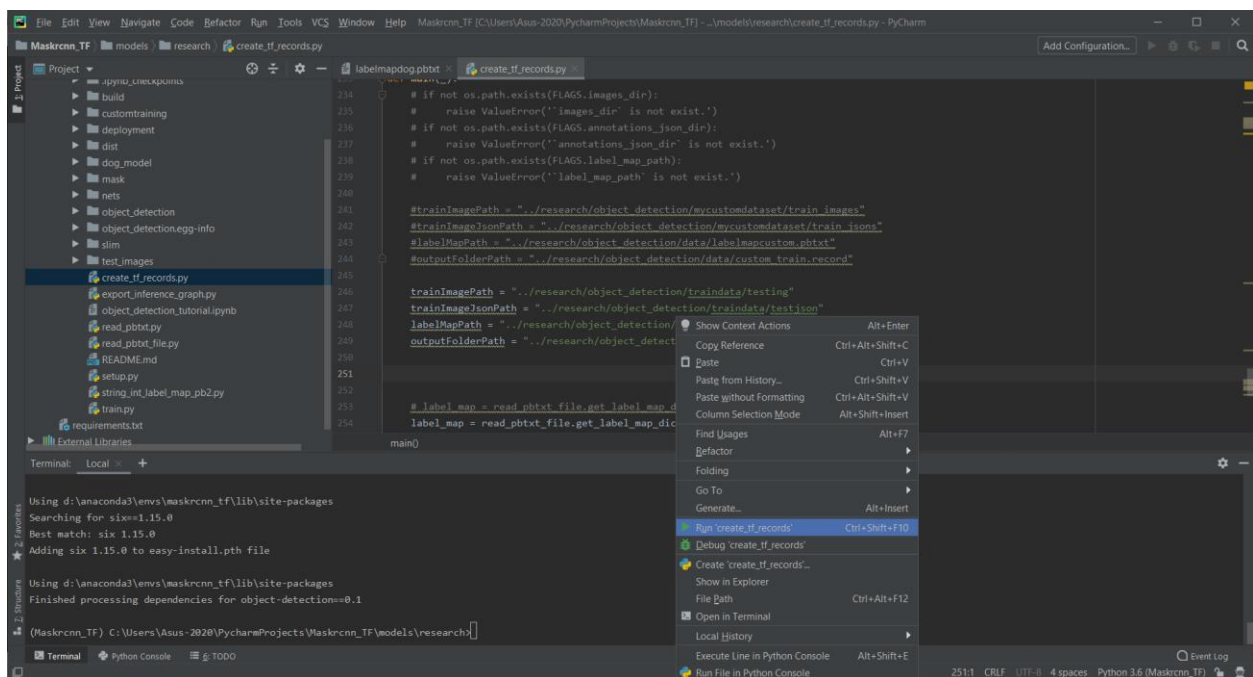


Here the output file (tf record) will be created or generated for the test data on the path which we have specified above

```
trainImagePath = "../research/object_detection/traindata/testing"  
trainImageJsonPath = "../research/object_detection/traindata/testjson"  
labelMapPath = "../research/object_detection/data/labelmapdog.pbtxt"  
outputFolderPath = "../research/object_detection/data/custom_test_dog.record12"
```

Here I have changed the name of the tf record like "custom_test_dog.record12"

Now run the create_tf_records.py file



Tf record is successfully created.

I have created for test and now I have to create tf record for training and do some changes like below.

Save and run the file.

```

trainImagePath = "../research/object_detection/traindata/training"
trainImageJsonPath = "../research/object_detection/traindata/trainjson"
labelMapPath = "../research/object_detection/data/labelmapdog.pbtxt"
outputFolderPath = "../research/object_detection/data/custom_train_dog.record12"

```

```

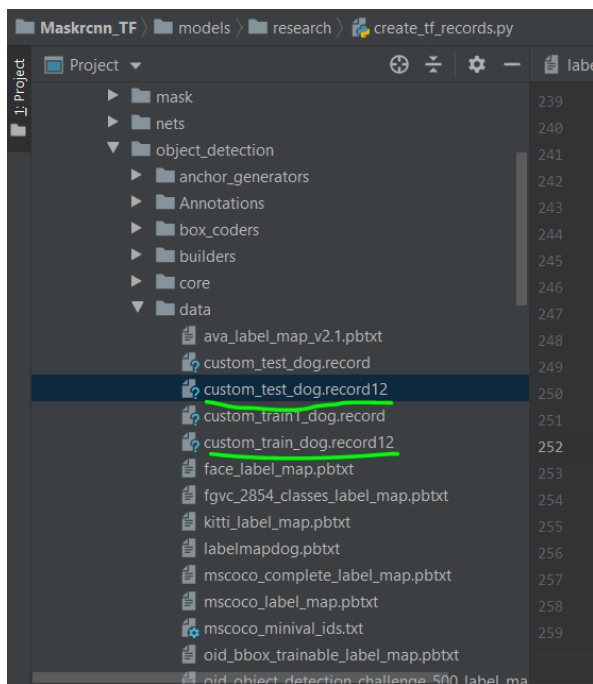
WARNING:tensorflow:From C:/Users/Asus-2020/PycharmProjects/Maskrcnn_TF/models/research/create_tf_records.py:255: The name tf.python_io.TFRecordWriter is deprecated. Please use tf.io.TFRecordWriter instead.
W0125 11:17:58.622582 12512 deprecation_wrapper.py:119] From C:/Users/Asus-2020/PycharmProjects/Maskrcnn_TF/models/research/create_tf_records.py:255: The name tf.python_io.TFRecordWriter is deprecated. Please use tf.io.TFRecordWriter instead.

On image %d 0
Successfully created TFRecord to ../research/object_detection/data/custom_train_dog.record12.

Process finished with exit code 0

```








Two files are created.



Files needed for Training:

Go to the model zoo

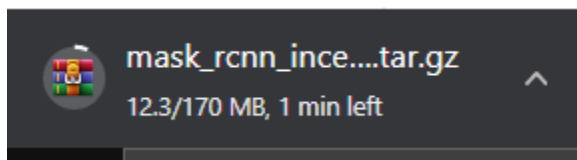
https://github.com/tensorflow/models/tree/master/research/object_detection/g3doc

 running_pets.md	Merged commit includes the following changes: (#8830)	7 months ago
 tf1.md	Use 2020-resolver with pip.	5 months ago
 <u>tf1_detection_zoo.md</u>	remove not guaranteed	6 months ago
 tf1_training_and_evaluation.md	Merged commit includes the following changes: (#8830)	7 months ago
 tf2.md	Add an end-to-end Colab for few-shot object detection with TFLite inf...	4 months ago
 tf2_classification_zoo.md	Merged commit includes the following changes: (#8830)	7 months ago
 tf2_detection_zoo.md	Fix broken link in Object Detection Model Zoo	5 months ago

faster_rcnn_nas	1833	43	Boxes
faster_rcnn_nas_lowproposals_coco	540		Boxes
mask_rcnn_inception_resnet_v2_atrous_coco	771	36	Masks
mask_rcnn_inception_v2_coco	79	25	Masks
mask_rcnn_resnet101_atrous_coco	470	33	Masks
mask_rcnn_resnet50_atrous_coco	343	29	Masks

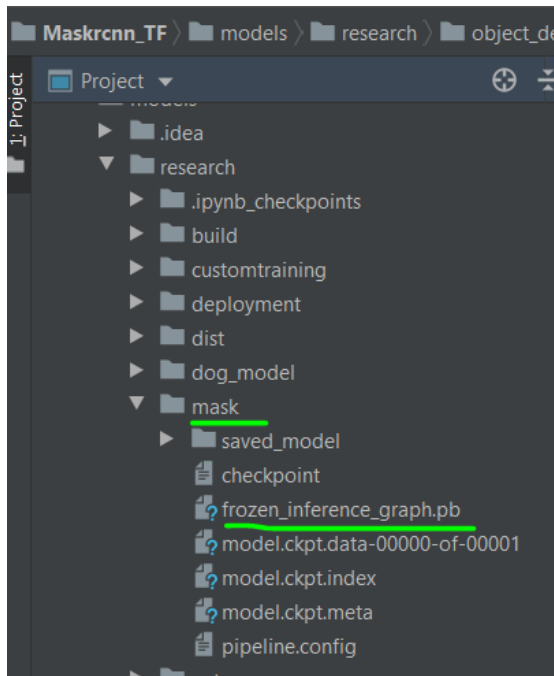
Here I'm going to use `mask_rcnn_inception_v2_coco`

Click the second architecture, then it start the downloading process.



Once the file is downloaded then extract the file and if u want to rename the folder as well.

Here I have changed the name as mask and move the folder into the research folder.



Config file:

For corresponding each and every architecture we have dedicated config files are available.

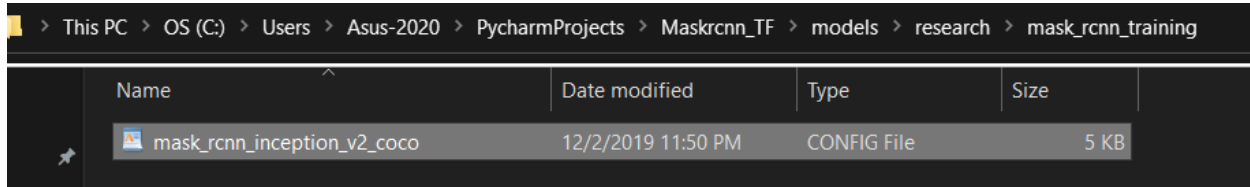
pycharmProjects --> Projectname (Maskrcnn_TF) --> models --> research --> object_detection --> samples --> configs

	Name	Date modified	Type	Size
	embedded_ssd_mobilenet_v1_coco	2/7/2019 1:19 AM	CONFIG File	5 KB
	facesd_mobilenet_v2_quantized_320x32...	2/7/2019 1:19 AM	CONFIG File	5 KB
	faster_rcnn_inception_resnet_v2_atrous_c...	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_inception_resnet_v2_atrous_c...	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_inception_resnet_v2_atrous_oid	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_inception_resnet_v2_atrous_p...	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_inception_v2_coco	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_inception_v2_pets	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_nas_coco	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_resnet50_coco	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_resnet50_fgvc	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_resnet50_pets	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_resnet101_atrous_coco	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_resnet101_ava_v2.1	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_resnet101_coco	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_resnet101_fgvc	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_resnet101_kitti	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_resnet101_pets	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_resnet101_voc07	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_resnet152_coco	2/7/2019 1:19 AM	CONFIG File	4 KB
	faster_rcnn_resnet152_pets	2/7/2019 1:19 AM	CONFIG File	4 KB
	mask_rcnn_inception_resnet_v2_atrous_co...	2/7/2019 1:19 AM	CONFIG File	5 KB
	<u>mask_rcnn_inception_v2_coco</u>	12/2/2019 11:50 PM	CONFIG File	5 KB
	mask_rcnn_resnet50_atrous_coco	2/7/2019 1:19 AM	CONFIG File	5 KB

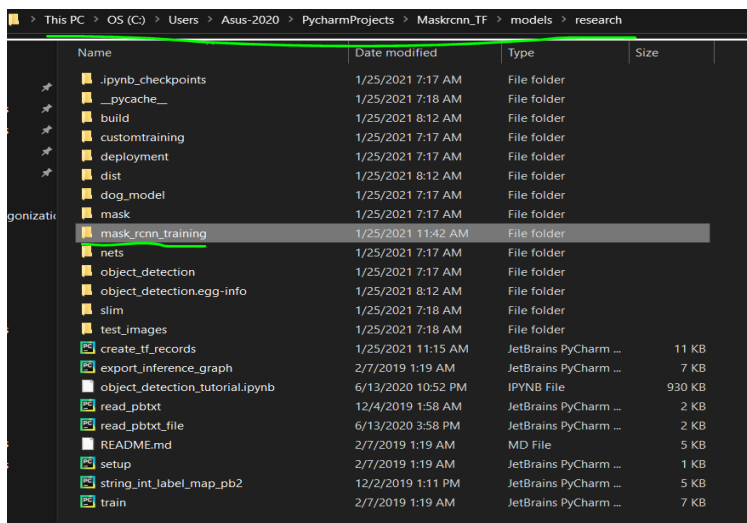
The above marked one is the config file iam going to use.

Now we need to create a new folder called **"mask_rcnn_training"**

pycharmProjects --> Projectname (Maskrcnn_TF) --> models --> research --> **mask_rcnn_training (folder)**

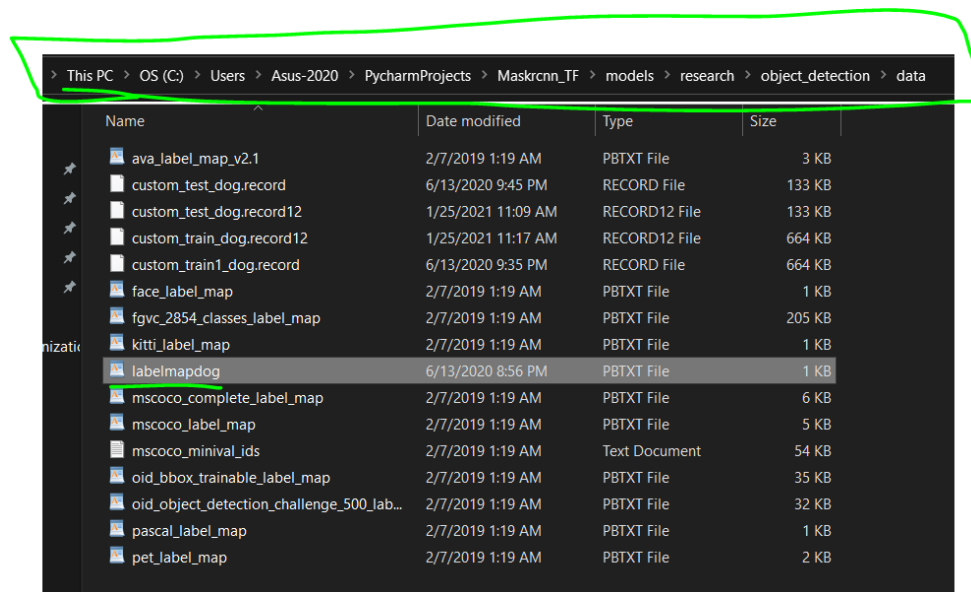


Copy the labelmap.pbtxt and paste into the mask_rcnn_training folder

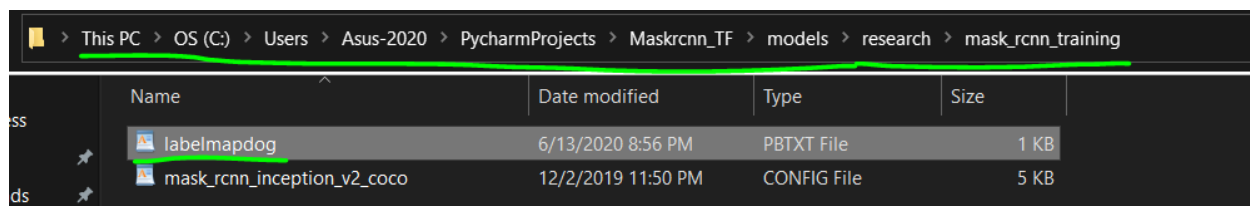


Copy the config file and paste into the mask_rcnn_training folder

Copy the file from below location



Paste the file in to the mask_rcnn_training



Now open the config file and we need to do some changes on that file

Changes 1:

```

1  | # Mask R-CNN with Inception V2
2  | # Configured for MSCOCO Dataset.
3  | # Users should configure the fine_tune_checkpoint field in the train config as
4  | # well as the label_map_path and input_path fields in the train_input_reader and
5  | # eval_input_reader. Search for "PATH_TO_BE_CONFIGURED" to find the fields that
6  | # should be configured.
7  |
8  | model {
9  |   faster_rcnn {
10 |     num_classes: 1
11 |     image_resizer {
12 |       keep_aspect_ratio_resizer {
13 |         min_dimension: 800
14 |         max_dimension: 1365

```

Changes 2:

In line no 127

```

127 | fine_tune_checkpoint: "PATH_TO_BE_CONFIGURED/model.ckpt"

```

Change the path

```

123 |   }
124 |   use_moving_average: false
125 | }
126 | gradient_clipping_by_norm: 10.0
127 | fine_tune_checkpoint: "mask/model.ckpt"
128 | from_detection_checkpoint: true
129 | # Note: The below line limits the training process to 200K steps, which we
130 | # empirically found to be sufficient enough to train the pets dataset. This
131 | # effectively bypasses the learning rate schedule (the learning rate will
132 | # never decay). Remove the below line to train indefinitely.

```

Changes 3:

Train and test record

Line no : 142, 144, 158, 160

Train:


```

140   train_input_reader: {
141     tf_record_input_reader {
142       input_path: "PATH_TO_BE_CONFIGURED/test.record"
143     }
144     label_map_path: "PATH_TO_BE_CONFIGURED/mscoco_label.pbtxt"
145     load_instance_masks: true
146     mask_type: PNG_MASKS
147   }

```

Change the location of the train record and labelmap path

```

140   train_input_reader: {
141     tf_record_input_reader {
142       input_path: "object_detection/data/custom_train_dog.record12"
143     }
144     label_map_path: "mask_rcnn_training/labelmapdog.pbtxt"
145     load_instance_masks: true
146     mask_type: PNG_MASKS
147   }
148

```

Test:

```

156   eval_input_reader: {
157     tf_record_input_reader {
158       input_path: "PATH_TO_BE_CONFIGURED/test2.record"
159     }
160     label_map_path: "PATH_TO_BE_CONFIGURED/mscoco_label_map.pbtxt"
161     load_instance_masks: true
162     mask_type: PNG_MASKS
163     shuffle: false
164     num_readers: 1
165   }
166

```

Change the test and label map path

```

156   eval_input_reader: {
157     tf_record_input_reader {
158       input_path: "object_detection/data/custom_test_dog.record12"
159     }
160     label_map_path: "mask_rcnn_training/labelmapdog.pbtxt"
161     load_instance_masks: true
162     mask_type: PNG_MASKS
163     shuffle: false
164     num_readers: 1
165   }

```

Changes 4:

Line no : 133

Number of steps : I have changed the value to 10000 from 100

```
131     # effectively bypasses the learning rate schedule (the learning rate will
132     # never decay). Remove the below line to train indefinitely.
133     num_steps: 10000
134     data_augmentation_options {
135         random_horizontal_flip {
```

Start the training:

Training command:

open the terminal window and navigate to the research folder.

Default command:

```
python train.py --logtostderr --train_dir=training/ --
pipeline_config_path=training/YOUR_MODEL.config
```

I have done some changes on the above command:

```
python train.py --logtostderr --train_dir=mask_rcnn_training/ --
pipeline_config_path=mask_rcnn_training/mask_rcnn_inception_v2_coco.config
```

Now the training is going on

```
I0125 12:49:31.076041 7432 learning.py:507] global step 21: loss = 0.7536 (16.949 sec/step)
INFO:tensorflow:global step 22: loss = 0.7489 (16.227 sec/step)
I0125 12:49:47.304175 7432 learning.py:507] global step 22: loss = 0.7489 (16.227 sec/step)
INFO:tensorflow:global step 23: loss = 3.7717 (15.760 sec/step)
I0125 12:50:03.067027 7432 learning.py:507] global step 23: loss = 3.7717 (15.760 sec/step)
INFO:tensorflow:global step 24: loss = 0.7328 (15.269 sec/step)
I0125 12:50:18.339195 7432 learning.py:507] global step 24: loss = 0.7328 (15.269 sec/step)
INFO:tensorflow:global step 25: loss = 3.7595 (15.199 sec/step)
I0125 12:50:33.540703 7432 learning.py:507] global step 25: loss = 3.7595 (15.199 sec/step)
```

Consider now the training has completed,

```
1. # Copy the file from object_detection to research
2. ## Replace the XXXX with the last generated ckpt file inside the training folder.
3.
4. python export_inference_graph.py --input_type image_tensor --pipeline_config_path
   training/faster_rcnn_inception_v2_coco.config --trained_checkpoint_prefix
   training/model.ckpt-1000 --output_directory inference_graph
```

Command:

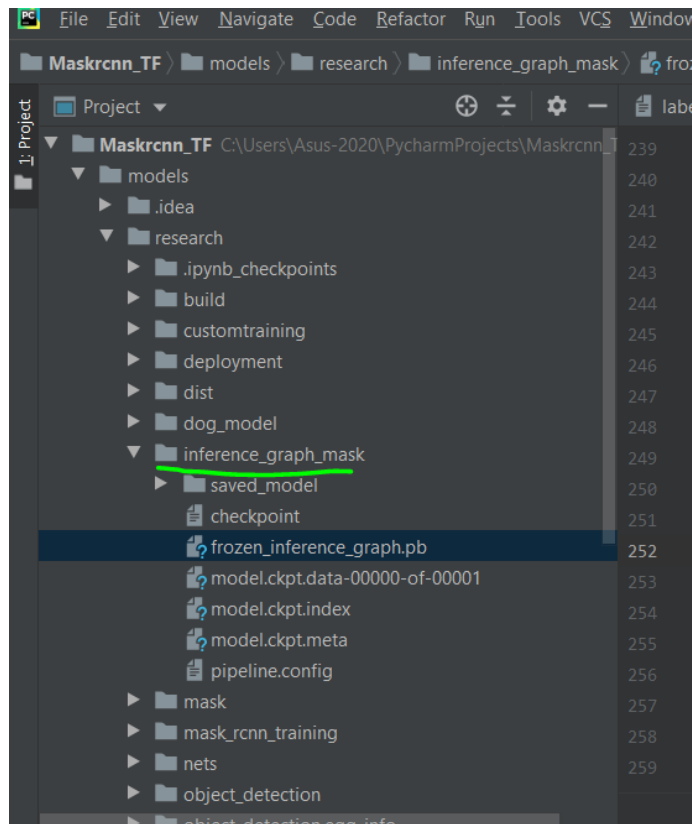
```
python export_inference_graph.py --input_type image_tensor --  
pipeline_config_path mask_rcnn_training/mask_rcnn_inception_v2_coco.config --  
trained_checkpoint_prefix mask_rcnn_training/model.ckpt-1000 --output_directory  
inference_graph
```

```
python export_inference_graph.py --input_type image_tensor --  
pipeline_config_path mask_rcnn_training/mask_rcnn_inception_v2_coco.conf  
g --trained_checkpoint_prefix mask_rcnn_training/model.ckpt-1972 --  
output_directory inference_graph_mask
```

```
(Maskrcnn_TF) C:\Users\Asus-2020\PycharmProjects\Maskrcnn_TF\models\research>python export_inference_graph.py --input_type image_tensor --pipeline_config_path mask_rcnn_training/mask_rcnn_inception_v2_coco.conf  
g --trained_checkpoint_prefix mask_rcnn_training/model.ckpt-1972 --output_directory inference_graph_mask
```

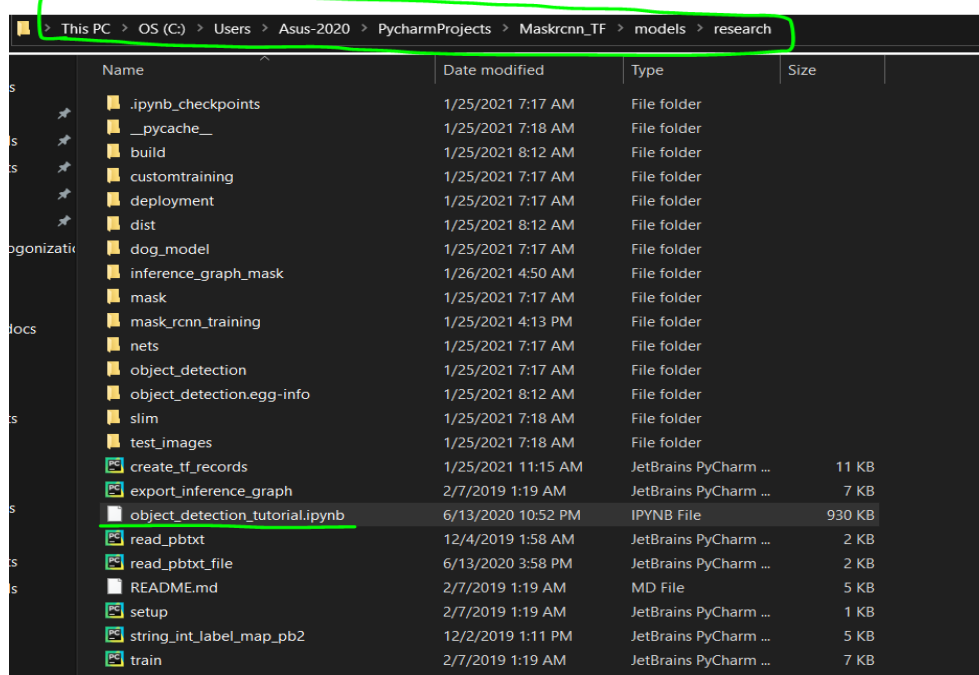
Completed successfully

```
I0126 04:50:48.782864 18296 builder_impl.py:421] SavedModel written to: inference_graph_mask\saved_model\saved_model.pb  
WARNING:tensorflow:From C:\Users\Asus-2020\PycharmProjects\Maskrcnn_TF\models\research\object_detection\utils\config_util.py:180: The name tf.gfile.Open is deprecated. Please use tf.io.gfile.GFile instead.  
I0126 04:50:48.851675 18296 deprecation_wrapper.py:119] From C:\Users\Asus-2020\PycharmProjects\Maskrcnn_TF\models\research\object_detection\utils\config_util.py:180: The name tf.gfile.Open is deprecated. Please  
use tf.io.gfile.GFile instead.  
INFO:tensorflow:Writing pipeline config file to inference_graph_mask\pipeline.config  
I0126 04:50:48.852678 18296 config_util.py:182] Writing pipeline config file to inference_graph_mask\pipeline.config
```



Now the .pb model is ready and we do the prediction.

Prediction:



Open the file

```
import numpy as np
import os
import six.moves.urllib as urllib
import sys
import tarfile
import tensorflow as tf
import zipfile

from distutils.version import StrictVersion
from collections import defaultdict
from io import StringIO
from matplotlib import pyplot as plt
from PIL import Image

# This is needed since the notebook is stored in the object_detection folder.
sys.path.append("..")
from object_detection.utils import ops as utils_ops

if StrictVersion(tf.__version__) < StrictVersion('1.9.0'):
    raise ImportError('Please upgrade your TensorFlow installation to v1.9.* or later!')
```

Model preparation

Variables

Any model exported using the `export_inference_graph.py` tool can be loaded here simply by changing `PATH_TO_FROZEN_GRAPH` to point to a new .pb file.

By default we use an "SSD with Mobilenet" model here. See the [detection model zoo](#) for a list of other models that can be run out-of-the-box with varying speeds and accuracies.

```
: # What model to download.
MODEL_NAME = 'inference_graph_mask'
#MODEL_FILE = MODEL_NAME + ".tar.gz"
#DOWNLOAD_BASE = 'http://download.tensorflow.org/models/object_detection/'

# Path to frozen detection graph. This is the actual model that is used for the object detection.
PATH_TO_FROZEN_GRAPH = MODEL_NAME + '/frozen_inference_graph.pb'

# List of the strings that is used to add correct label for each box.
PATH_TO_LABELS = os.path.join('mask_rcnn_training', 'labelmapdog.pbtxt')
```

In this above section do some changes accordingly to your folder and file location.

Comment this section

Download Model

```
: '''opener = urllib.request.URLopener()
opener.retrieve(DOWNLOAD_BASE + MODEL_FILE, MODEL_FILE)
tar_file = tarfile.open(MODEL_FILE)
for file in tar_file.getmembers():
    file_name = os.path.basename(file.name)
    if 'frozen_inference_graph.pb' in file_name:
        tar_file.extract(file, os.getcwd())'''

: "opener = urllib.request.URLopener()\nopener.retrieve(DOWNLOAD_BASE + MODEL_FILE, MODEL_FILE)\ntar_file = tarfile.open(MODEL_FILE)\nfor file in tar_file.getmembers():\n    file_name = os.path.basename(file.name)\n    if 'frozen_inference_graph.pb' in file_name:\n        tar_file.extract(file, os.getcwd())"
```

Run all the rest of cells inside the notebook file

Final prediction looks like below.

```
] : %matplotlib inline
plt.imshow(image_np)
```

```
] : <matplotlib.image.AxesImage at 0x16300643b70>
```

