```
In [ ]: import tensorflow as tf
         import IPython.display as display
         from PIL import Image
         import numpy as np
         import matplotlib.pyplot as plt
         import os
         import pathlib
         from skimage import io, color
         import tensorflow as tf
         from tqdm import tqdm
In [0]: #Taking the paths for train images
         x tr = tf.data.Dataset.list files((str(pathlib.Path('idd20k lite/leftIm
         q8bit/train')/'*/*')),shuffle=False)
In [0]: #Taking the paths for train masks
         y tr = tf.data.Dataset.list files((str(pathlib.Path('idd20k lite/gtFin
         e/train')/'*/* inst label.png')),shuffle=False)
In [10]: for f in x tr.take(5):
             print(f.numpy())
         b'idd20k lite/leftImg8bit/train/0/024541 image.jpg'
         b'idd20k lite/leftImg8bit/train/0/024703 image.jpg'
         b'idd20k lite/leftImg8bit/train/1/092468 image.jpg'
         b'idd20k lite/leftImg8bit/train/1/340676 image.jpg'
         b'idd20k lite/leftImg8bit/train/1/502201 image.jpg'
In [11]: for f in y tr.take(5):
             print(f.numpy())
         b'idd20k lite/gtFine/train/0/024541 inst label.png'
         b'idd20k lite/gtFine/train/0/024703 inst label.png'
         b'idd20k lite/gtFine/train/1/092468 inst label.png'
```

```
b'idd20k lite/gtFine/train/1/340676 inst label.png'
         b'idd20k lite/gtFine/train/1/502201 inst label.png'
In [0]: #Taking the paths for validation images
         x val = tf.data.Dataset.list files((str(pathlib.Path('idd20k lite/leftI
         mg8bit/val')/'*/*')), shuffle=False)
In [0]: #Taking the paths for validation masks
         y val = tf.data.Dataset.list files((str(pathlib.Path('idd20k lite/gtFin
         e/val')/'*/* inst label.png')),shuffle=False)
In [14]: for f in x val.take(5):
             print(f.numpv())
         b'idd20k lite/leftImg8bit/val/119/903127 image.jpg'
         b'idd20k lite/leftImg8bit/val/132/475092 image.jpg'
         b'idd20k lite/leftImg8bit/val/132/489315 image.jpg'
         b'idd20k lite/leftImg8bit/val/132/874777 image.jpg'
         b'idd20k lite/leftImg8bit/val/147/425716 image.jpg'
In [15]: for f in y val.take(5):
             print(f.numpy())
         b'idd20k lite/gtFine/val/119/903127 inst label.png'
         b'idd20k lite/gtFine/val/132/475092 inst label.png'
         b'idd20k lite/gtFine/val/132/489315 inst label.png'
         b'idd20k lite/gtFine/val/132/874777 inst label.png'
         b'idd20k lite/gtFine/val/147/425716 inst label.png'
In [0]: #THis function return processed images
         def process path(file path):
             img = tf.io.read file(file path)
             img = tf.image.decode jpeg(img, channels=3)
             img = tf.image.convert image dtype(img, tf.float32)
             img = tf.image.resize(img, [224, 224])
             return imq
```

```
In [0]: #THis function return processed labels
        def segment(filename):
            def a1():
                c= tf.tensor scatter nd update(b, tf.constant([[i,j,0]]), tf.co
        nstant([1],dtype=tf.dtypes.float32))
                return c
            def b1():
                c= tf.tensor scatter nd update(b, tf.constant([[i,j,1]]), tf.co
        nstant([1],dtype=tf.dtypes.float32))
                return c
            def c1():
                c= tf.tensor scatter nd update(b, tf.constant([[i,j,2]]), tf.co
        nstant([1],dtype=tf.dtypes.float32))
                return c
            def d1():
                c= tf.tensor scatter nd update(b, tf.constant([[i,j,3]]), tf.co
        nstant([1],dtvpe=tf.dtvpes.float32))
                return c
            def e1():
                c= tf.tensor scatter nd update(b, tf.constant([[i,j,4]]), tf.co
        nstant([1],dtype=tf.dtypes.float32))
                return c
            def f1():
                c= tf.tensor scatter nd update(b, tf.constant([[i,j,5]]), tf.co
        nstant([1],dtype=tf.dtypes.float32))
                return c
            def a1():
                c= tf.tensor scatter nd update(b, tf.constant([[i,j,6]]), tf.co
        nstant([1],dtype=tf.dtypes.float32))
                return c
            def h1():
                c= tf.tensor scatter nd update(b, tf.constant([[i,j,6]]), tf.co
        nstant([0],dtype=tf.dtypes.float32))
                return c
            img = tf.io.read file(filename)
            img = tf.image.decode jpeg(img, channels=3)
            img = tf.image.resize(img, [224, 224])
            img = img[:, :, 0]
            img=tf.math.round(img)
```

```
a=imq
             b=tf.zeros([224,224,7], dtype=tf.dtypes.float32)
             for i in tqdm(range(0,224)):
                 for j in range(0,224):
                     b=tf.case([(tf.math.equal(a[i][j],0), a1), (tf.math.equal(a
         [i][j],1), b1),
                                (tf.math.equal(a[i][j],2), c1),(tf.math.equal(a[
         i][i],3), d1),
                               (tf.math.equal(a[i][j],4), e1), (tf.math.equal(a[
         i][j],5), f1),
                               (tf.math.equal(a[i][j],6), g1)],default=h1,exclus
         ive=True)
             return b
In [0]: #Mapping the function to transform all the train and validation images
         x tr 1=x tr.map(process path)
         x val 1=x val.map(process path)
In [0]: #Mapping the function to transform all the train and validation masks
         y tr 1=y tr.map(segment)
         y val 1=y val.map(segment)
In [0]: #Preparing train and validation datasets using zipping
         train = tf.data.Dataset.zip((x tr 1, y tr 1))
         val = tf.data.Dataset.zip((x val 1, y val 1))
In [0]: #Preparing dataset to feed to model by shuffling and batching
         train dataset = train.cache().shuffle(500).batch(32).repeat()
         train dataset = train dataset.prefetch(buffer size=tf.data.experimental
         . AUTOTUNE)
         test dataset = val.batch(32)
In [30]: train dataset
Out[30]: <PrefetchDataset shapes: ((None, 224, 224, 3), (None, 224, 224, 7)), ty
         pes: (tf.float32, tf.float64)>
```

```
In [31]: test dataset
Out[31]: <BatchDataset shapes: ((None, 224, 224, 3), (None, 224, 224, 7)), type
         s: (tf.float32, tf.float64)>
In [0]: IMAGE ORDERING = 'channels last'
         input height, input width = 224, 224
         output height, output width = 224, 224
         n classes=7
In [0]: #Creating model with keras subclass api
         class VGGNetModel(tf.keras.Model):
                 def init (self, classes, chanDim=-1):
                         super(VGGNetModel, self). init (name='my model')
                         self.conv1A = layers.Conv2D(64, (3, 3), activation = 'r
         elu', padding = 'same', name = 'block1 conv1', data format = IMAGE ORDE
         RING, input shape=(224,224,3))
                         self.act1A = layers.Conv2D(64, (3, 3), activation = 're
         lu', padding = 'same', name = 'block1 conv2', data format = IMAGE ORDER
         ING )
                         self.pool1 = layers.MaxPooling2D((2, 2), strides = (2,
         2), name = 'block1 pool', data format = IMAGE ORDERING )
                         self.conv2A = layers.Conv2D(128, (3, 3), activation =
         'relu', padding = 'same', name = 'block2 conv1', data format = IMAGE OR
         DERING )
                         self.act2A = layers.Conv2D(128, (3, 3), activation = 'r
         elu', padding = 'same', name = 'block2 conv2', data format = IMAGE ORDE
         RING )
                         self.bn2A = layers.MaxPooling2D((2, 2), strides = (2, 2))
         ), name = 'block2 pool', data format = IMAGE ORDERING )
                         self.conv2B = layers.Conv2D(256, (3, 3), activation =
         'relu', padding = 'same', name = 'block3 conv1', data format = IMAGE OR
         DERING )
                         self.act2B = layers.Conv2D(256, (3, 3), activation = 're
         lu', padding = 'same', name = 'block3 conv2', data format = IMAGE ORDER
         ING )
                         self.bn2B = layers.Conv2D(256, (3, 3), activation = 're
         lu', padding = 'same', name = 'block3 conv3', data format = IMAGE ORDER
         ING )
```

```
self.pool2 = layers.MaxPooling2D((2, 2), strides = (2, 2))
2), name = 'block3 pool', data format = IMAGE ORDERING )
                self.flatten = layers.Conv2D(512, (3, 3), activation =
'relu', padding = 'same', name = 'block4 conv1', data format = IMAGE OR
DERING )
                self.dense3 = layers.Conv2D(512, (3, 3), activation =
'relu', padding = 'same', name = 'block4 conv2', data format = IMAGE OR
DERING )
                self.act3 = layers.Conv2D(512, (3, 3), activation = 're
lu', padding = 'same', name = 'block4 conv3', data format = IMAGE ORDER
ING )
                self.bn3 = layers.MaxPooling2D((2, 2), strides = (2, 2))
), name = 'block4 pool', data format = IMAGE ORDERING )
                self.do3 = layers.Conv2D(512, (3, 3), activation = 'rel
u', padding = 'same', name = 'block5 conv1', data format = IMAGE ORDERI
NG )
                self.dense4 = layers.Conv2D(512, (3, 3), activation =
'relu', padding = 'same', name = 'block5 conv2', data format = IMAGE OR
DERING )
                self.softmax = layers.Conv2D(512, (3, 3), activation =
'relu', padding = 'same', name = 'block5 conv3', data_format = IMAGE_OR
DERING )
                self.test1=layers.MaxPooling2D((2, 2), strides = (2, 2
), name = 'block5 pool', data format = IMAGE ORDERING )
                self.test2=layers.Conv2D(4096, (7, 7), activation = 're
lu', padding = 'same', name = 'conv6', data format = IMAGE ORDERING)
                self.test3=layers.Conv2D(4096, (1, 1), activation = 're
lu', padding = 'same', name = 'conv7', data format = IMAGE ORDERING)
                self.test4=layers.Conv2DTranspose(nClasses, kernel size
= (4, 4), strides = (4, 4), use bias = False, data format = IMAGE ORDE
RING )
                self.test5=layers.Conv2D(nClasses, (1, 1), activation =
 'relu', padding = 'same', name = 'pool4 11', data format = IMAGE ORDER
ING)
                self.test6=layers.Conv2DTranspose(nClasses, kernel size
= (2, 2), strides = (2, 2), use bias = False, data format = IMAGE ORDE
RING )
                self.test7=layers.Conv2D(nClasses, (1, 1), activation =
```

```
'relu', padding = 'same', name = 'pool3 11', data format = IMAGE ORDER
ING)
                self.test8=layers.Add(name = 'add')
                self.test9=layers.Conv2DTranspose(nClasses, kernel size
= (8, 8), strides = (8, 8), use bias = False, data format = IMAGE ORDE
RING )
                self.test10=layers.Activation('softmax')
       def call(self, inputs):
                inp=Input(shape = (input_height, input_width, 3))
                x = self.conv1A(inputs)
                x = self.act1A(x)
                x = self.pool1(x)
                f1 = x
                x = self.conv2A(x)
                x = self.act2A(x)
                x = self.bn2A(x)
                f2 = x
                x = self.conv2B(x)
                x = self.act2B(x)
                x = self.bn2B(x)
                x = self.pool2(x)
                pool3 = x
                x = self.flatten(x)
                x = self.dense3(x)
                x = self.act3(x)
                x = self.bn3(x)
                pool4 = x
                x = self.do3(x)
                x = self.dense4(x)
                x = self.softmax(x)
                x=self.test1(x)
                pool5 = x
                o = self.test2(pool5)
                conv7 = (self.test3)(o)
                conv7 4 = (self.test4)(conv7)
                pool411 = (self.test5)(pool4)
```

```
pool411 2 = (self.test6)(pool411)
                          pool311 = (self.test7)(pool3)
                          o = self.test8([pool411 2, pool311, conv7 4])
                          o = self.test9(o)
                          o = (self.test10)(o)
                          return o
In [0]: model=VGGNetModel(classes=7)
In [0]: #building model with input shape
         model.build((32,224,224,3))
In [36]: model.summary()
         Model: "my model"
         Layer (type)
                                       Output Shape
                                                                 Param #
         block1 conv1 (Conv2D)
                                       multiple
                                                                 1792
         block1 conv2 (Conv2D)
                                       multiple
                                                                 36928
         block1 pool (MaxPooling2D)
                                      multiple
                                                                 0
         block2 conv1 (Conv2D)
                                       multiple
                                                                 73856
         block2 conv2 (Conv2D)
                                       multiple
                                                                 147584
         block2 pool (MaxPooling2D)
                                      multiple
                                                                 0
         block3 conv1 (Conv2D)
                                       multiple
                                                                 295168
         block3 conv2 (Conv2D)
                                       multiple
                                                                 590080
         block3 conv3 (Conv2D)
                                       multiple
                                                                 590080
         block3 pool (MaxPooling2D)
                                      multiple
                                                                 0
```

block4_conv1 (Conv2D)	multiple	1180160
block4_conv2 (Conv2D)	multiple	2359808
block4_conv3 (Conv2D)	multiple	2359808
block4_pool (MaxPooling2D)	multiple	0
block5_conv1 (Conv2D)	multiple	2359808
block5_conv2 (Conv2D)	multiple	2359808
block5_conv3 (Conv2D)	multiple	2359808
block5_pool (MaxPooling2D)	multiple	0
conv6 (Conv2D)	multiple	102764544
conv7 (Conv2D)	multiple	16781312
<pre>conv2d_transpose (Conv2DTran</pre>	multiple	458752
pool4_11 (Conv2D)	multiple	3591
<pre>conv2d_transpose_1 (Conv2DTr</pre>	multiple	196
pool3_11 (Conv2D)	multiple	1799
add (Add)	multiple	0
<pre>conv2d_transpose_2 (Conv2DTr</pre>	multiple	3136
activation (Activation)	multiple	0
Total params: 134,728,018		

Trainable params: 134,728,018 Non-trainable params: 0

```
In [0]: from keras.callbacks import ModelCheckpoint
      CHECKPOINT FILE PATH = "model main.hdf5"
      checkpoint = ModelCheckpoint(CHECKPOINT FILE PATH, monitor='val acc', v
      erbose=1, save best only=True, mode='max', period=1)
In [0]: sqd = optimizers.SGD(lr = 0.01, decay = 5 ** (-4), momentum = 0.9, nest
      erov = True)
      model.compile(loss = 'categorical crossentropy',
                optimizer = 'adam'.
                metrics = [tf.keras.metrics.MeanIoU(num classes=7)])
In [38]: hist2 = model.fit(train dataset,
                   validation data = test dataset,epochs = 300, verbose
      = 1,steps per epoch=43,validation steps=6,
                  callbacks=[checkpoint])
      Epoch 1/300
      - mean io u: 0.4286 - val loss: 1.9172 - val mean io u: 0.4286
      Epoch 2/300
      - mean io u: 0.4286 - val loss: 1.2350 - val mean io u: 0.4286
      Epoch 3/300
      - mean io u: 0.4286 - val loss: 1.0791 - val mean io u: 0.4286
      Epoch 4/300
      - mean io u: 0.4286 - val loss: 1.0215 - val mean io u: 0.4286
      Epoch 5/300
      - mean io u: 0.4286 - val loss: 0.8803 - val mean io u: 0.4286
      Epoch 6/300
      - mean io u: 0.4286 - val loss: 0.8139 - val mean io u: 0.4286
      Epoch 7/300
      - mean io u: 0.4286 - val loss: 0.7486 - val mean io u: 0.4286
```

```
Epoch 8/300
- mean io u: 0.4286 - val loss: 0.7106 - val mean io u: 0.4286
Epoch 9/300
- mean io u: 0.4286 - val loss: 0.7065 - val mean io u: 0.4286
Epoch 10/300
- mean io u: 0.4286 - val loss: 0.6839 - val mean io u: 0.4286
Epoch 11/300
- mean io u: 0.4287 - val loss: 0.6788 - val mean io u: 0.4286
Epoch 12/300
- mean io u: 0.4289 - val loss: 0.6986 - val mean io u: 0.4292
Epoch 13/300
- mean io u: 0.4295 - val loss: 0.6950 - val mean io u: 0.4301
Epoch 14/300
- mean io u: 0.4337 - val loss: 0.7035 - val mean io u: 0.4347
Epoch 15/300
- mean io u: 0.4403 - val loss: 0.7110 - val mean io u: 0.4396
Epoch 16/300
- mean io u: 0.4468 - val loss: 0.7404 - val mean io u: 0.4533
Epoch 17/300
- mean io u: 0.4553 - val loss: 0.7564 - val mean io u: 0.4523
Epoch 18/300
- mean io u: 0.4619 - val loss: 0.7979 - val mean io u: 0.4621
Epoch 19/300
- mean io u: 0.4684 - val loss: 0.8017 - val mean io u: 0.4659
Epoch 20/300
- mean io u: 0.4761 - val loss: 0.8263 - val mean io u: 0.4735
```

```
Epoch 21/300
- mean io u: 0.4829 - val loss: 0.8822 - val mean io u: 0.4792
Epoch 22/300
- mean io u: 0.4896 - val loss: 0.8858 - val mean io u: 0.4839
Epoch 23/300
- mean io u: 0.4942 - val loss: 0.9242 - val mean io u: 0.4938
Epoch 24/300
- mean io u: 0.5027 - val loss: 0.9292 - val mean io u: 0.4952
Epoch 25/300
- mean io u: 0.5058 - val loss: 0.9691 - val mean io u: 0.5030
Epoch 26/300
- mean io u: 0.5124 - val loss: 0.9761 - val mean io u: 0.5034
Epoch 27/300
- mean io u: 0.5217 - val loss: 1.0279 - val mean io u: 0.5071
Epoch 28/300
- mean io u: 0.5233 - val loss: 1.0324 - val mean io u: 0.5139
Epoch 29/300
- mean io u: 0.5300 - val loss: 1.0453 - val mean io u: 0.5215
Epoch 30/300
- mean io u: 0.5362 - val loss: 1.1009 - val mean io u: 0.5308
Epoch 31/300
- mean io u: 0.5418 - val loss: 1.1367 - val mean io u: 0.5270
Epoch 32/300
- mean io u: 0.5456 - val loss: 1.1451 - val mean io u: 0.5300
Epoch 33/300
- mean io u: 0.5483 - val loss: 1.1384 - val mean io u: 0.5257
```

```
Epoch 34/300
- mean io u: 0.5523 - val loss: 1.1998 - val mean io u: 0.5349
Epoch 35/300
- mean io u: 0.5513 - val loss: 1.1595 - val mean io u: 0.5321
Epoch 36/300
- mean io u: 0.5494 - val loss: 1.2266 - val mean io u: 0.5368
Epoch 37/300
- mean io u: 0.5580 - val loss: 1.2597 - val mean io u: 0.5407
Epoch 38/300
- mean io u: 0.5666 - val loss: 1.2556 - val mean io u: 0.5455
Epoch 39/300
- mean io u: 0.5604 - val loss: 1.2634 - val mean io u: 0.5367
Epoch 40/300
- mean io u: 0.5620 - val loss: 1.2969 - val mean io u: 0.5415
Epoch 41/300
- mean io u: 0.5713 - val loss: 1.3003 - val mean io u: 0.5374
Epoch 42/300
- mean io u: 0.5734 - val loss: 1.3217 - val mean io u: 0.5519
Epoch 43/300
- mean io u: 0.5819 - val loss: 1.3368 - val mean io u: 0.5523
Epoch 44/300
- mean io u: 0.5801 - val loss: 1.3814 - val mean io u: 0.5582
Epoch 45/300
- mean io u: 0.5789 - val loss: 1.3514 - val mean io u: 0.5487
Epoch 46/300
- mean io u: 0.5839 - val loss: 1.3587 - val mean io u: 0.5559
```

```
Epoch 47/300
- mean io u: 0.5875 - val loss: 1.4301 - val mean io u: 0.5574
Epoch 48/300
- mean io u: 0.5905 - val loss: 1.3808 - val mean io u: 0.5558
Epoch 49/300
- mean io u: 0.5905 - val loss: 1.3938 - val mean io u: 0.5581
Epoch 50/300
- mean io u: 0.5897 - val loss: 1.4091 - val mean io u: 0.5544
Epoch 51/300
- mean io u: 0.5907 - val loss: 1.3920 - val mean io u: 0.5551
Epoch 52/300
- mean io u: 0.5920 - val loss: 1.4241 - val mean io u: 0.5654
Epoch 53/300
- mean io u: 0.5969 - val loss: 1.4216 - val mean io u: 0.5629
Epoch 54/300
- mean io u: 0.5972 - val loss: 1.4417 - val mean io u: 0.5643
Epoch 55/300
- mean io u: 0.5984 - val loss: 1.4249 - val mean io u: 0.5601
Epoch 56/300
- mean io u: 0.6022 - val loss: 1.4482 - val mean io u: 0.5667
Epoch 57/300
- mean io u: 0.5986 - val loss: 1.4306 - val mean io u: 0.5567
Epoch 58/300
- mean io u: 0.5922 - val loss: 1.4203 - val_mean_io_u: 0.5556
Epoch 59/300
- mean io u: 0.5902 - val loss: 1.4081 - val mean io u: 0.5558
```

```
Epoch 60/300
- mean io u: 0.5986 - val loss: 1.4601 - val mean io u: 0.5610
Epoch 61/300
- mean io u: 0.6034 - val loss: 1.4650 - val mean io u: 0.5616
Epoch 62/300
- mean io u: 0.6081 - val loss: 1.4609 - val mean io u: 0.5610
Epoch 63/300
- mean io u: 0.5917 - val loss: 1.4252 - val mean io u: 0.5585
Epoch 64/300
- mean io u: 0.5991 - val loss: 1.4765 - val mean io u: 0.5685
Epoch 65/300
- mean io u: 0.6105 - val loss: 1.4928 - val mean io u: 0.5693
Epoch 66/300
- mean io u: 0.6157 - val loss: 1.4846 - val mean io u: 0.5697
Epoch 67/300
- mean io u: 0.6104 - val loss: 1.4891 - val mean io u: 0.5659
Epoch 68/300
- mean io u: 0.6044 - val loss: 1.4219 - val mean io u: 0.5521
Epoch 69/300
- mean io u: 0.6060 - val loss: 1.4752 - val mean io u: 0.5686
Epoch 70/300
- mean io u: 0.6093 - val loss: 1.4941 - val mean io u: 0.5660
Epoch 71/300
- mean io u: 0.6096 - val loss: 1.4884 - val mean io u: 0.5649
Epoch 72/300
- mean io u: 0.6120 - val loss: 1.4583 - val mean io u: 0.5642
```

```
Epoch 73/300
- mean io u: 0.6102 - val loss: 1.4747 - val mean io u: 0.5629
Epoch 74/300
- mean io u: 0.6146 - val loss: 1.4616 - val mean io u: 0.5624
Epoch 75/300
- mean io u: 0.6191 - val loss: 1.5262 - val mean io u: 0.5759
Epoch 76/300
- mean io u: 0.6180 - val loss: 1.5554 - val mean io u: 0.5767
Epoch 77/300
- mean io u: 0.6218 - val loss: 1.4982 - val mean io u: 0.5764
Epoch 78/300
- mean io u: 0.6083 - val loss: 1.4678 - val mean io u: 0.5672
Epoch 79/300
- mean io u: 0.6060 - val loss: 1.4769 - val mean io u: 0.5640
Epoch 80/300
- mean io u: 0.6107 - val loss: 1.4753 - val mean io u: 0.5632
Epoch 81/300
- mean io u: 0.6203 - val loss: 1.5208 - val mean io u: 0.5728
Epoch 82/300
- mean io u: 0.6233 - val loss: 1.5087 - val mean io u: 0.5726
Epoch 83/300
- mean io u: 0.6215 - val loss: 1.5242 - val mean io u: 0.5755
Epoch 84/300
- mean io u: 0.6212 - val loss: 1.5322 - val mean io u: 0.5709
Epoch 85/300
- mean io u: 0.6241 - val loss: 1.5224 - val mean io u: 0.5773
```

```
Epoch 86/300
- mean io u: 0.6251 - val loss: 1.4912 - val mean io u: 0.5703
Epoch 87/300
- mean io u: 0.6114 - val loss: 1.4815 - val mean io u: 0.5620
Epoch 88/300
- mean io u: 0.6083 - val loss: 1.4294 - val mean io u: 0.5608
Epoch 89/300
- mean io u: 0.6072 - val loss: 1.4690 - val mean io u: 0.5580
Epoch 90/300
- mean io u: 0.6220 - val loss: 1.5135 - val mean io u: 0.5716
Epoch 91/300
- mean io u: 0.6334 - val loss: 1.5418 - val mean io u: 0.5780
Epoch 92/300
- mean io u: 0.6359 - val loss: 1.5724 - val mean io u: 0.5780
Epoch 93/300
- mean io u: 0.6315 - val loss: 1.5554 - val mean io u: 0.5791
Epoch 94/300
- mean io u: 0.6348 - val loss: 1.5567 - val mean io u: 0.5786
Epoch 95/300
- mean io u: 0.6326 - val loss: 1.5498 - val mean io u: 0.5752
Epoch 96/300
- mean io u: 0.6308 - val loss: 1.5081 - val mean io u: 0.5709
Epoch 97/300
- mean io u: 0.6282 - val loss: 1.5159 - val_mean_io_u: 0.5731
Epoch 98/300
- mean io u: 0.6331 - val loss: 1.5677 - val mean io u: 0.5822
```

```
Epoch 99/300
- mean io u: 0.6390 - val loss: 1.5107 - val mean io u: 0.5766
Epoch 100/300
- mean io u: 0.6344 - val loss: 1.5549 - val mean io u: 0.5813
Epoch 101/300
- mean io u: 0.6296 - val loss: 1.5448 - val mean io u: 0.5814
Epoch 102/300
- mean io u: 0.6218 - val loss: 1.5331 - val mean io u: 0.5679
Epoch 103/300
- mean io u: 0.6259 - val loss: 1.4987 - val mean io u: 0.5655
Epoch 104/300
- mean io u: 0.6258 - val loss: 1.5328 - val mean io u: 0.5801
Epoch 105/300
- mean io u: 0.6348 - val loss: 1.5913 - val mean io u: 0.5831
Epoch 106/300
- mean io u: 0.6434 - val loss: 1.5372 - val mean io u: 0.5827
Epoch 107/300
- mean io u: 0.6364 - val loss: 1.5349 - val mean io u: 0.5710
Epoch 108/300
- mean io u: 0.6402 - val loss: 1.6100 - val mean io u: 0.5872
Epoch 109/300
- mean io u: 0.6456 - val loss: 1.5593 - val mean io u: 0.5843
Epoch 110/300
- mean io u: 0.6403 - val loss: 1.5229 - val_mean_io_u: 0.5656
Epoch 111/300
- mean io u: 0.6316 - val loss: 1.5470 - val mean io u: 0.5819
```

```
Epoch 112/300
- mean io u: 0.6352 - val loss: 1.5250 - val mean io u: 0.5797
Epoch 113/300
- mean io u: 0.6344 - val loss: 1.5139 - val mean io u: 0.5738
Epoch 114/300
- mean io u: 0.6349 - val loss: 1.5760 - val mean io u: 0.5775
Epoch 115/300
- mean io u: 0.6421 - val loss: 1.5565 - val mean io u: 0.5735
Epoch 116/300
- mean io u: 0.6448 - val loss: 1.5490 - val mean io u: 0.5812
Epoch 117/300
- mean io u: 0.6373 - val loss: 1.5623 - val mean io u: 0.5799
Epoch 118/300
- mean io u: 0.6363 - val loss: 1.5363 - val mean io u: 0.5769
Epoch 119/300
- mean io u: 0.6346 - val loss: 1.5243 - val mean io u: 0.5775
Epoch 120/300
- mean io u: 0.6314 - val loss: 1.5108 - val mean io u: 0.5720
Epoch 121/300
- mean io u: 0.6399 - val loss: 1.5646 - val mean io u: 0.5865
Epoch 122/300
- mean io u: 0.6431 - val loss: 1.5420 - val mean io u: 0.5830
Epoch 123/300
- mean io u: 0.6443 - val loss: 1.5511 - val mean io u: 0.5842
Epoch 124/300
- mean io u: 0.6432 - val loss: 1.6020 - val mean io u: 0.5889
```

```
Epoch 125/300
- mean io u: 0.6442 - val loss: 1.5344 - val mean io u: 0.5757
Epoch 126/300
- mean io u: 0.6368 - val loss: 1.5237 - val mean io u: 0.5784
Epoch 127/300
- mean io u: 0.6441 - val loss: 1.5707 - val mean io u: 0.5845
Epoch 128/300
- mean io u: 0.6483 - val loss: 1.5459 - val mean io u: 0.5772
Epoch 129/300
- mean io u: 0.6496 - val loss: 1.5388 - val mean io u: 0.5814
Epoch 130/300
- mean io u: 0.6476 - val loss: 1.5640 - val mean io u: 0.5832
Epoch 131/300
- mean io u: 0.6536 - val loss: 1.5593 - val mean io u: 0.5823
Epoch 132/300
- mean io u: 0.6439 - val loss: 1.5279 - val mean io u: 0.5751
Epoch 133/300
- mean io u: 0.6403 - val loss: 1.5454 - val mean io u: 0.5807
Epoch 134/300
- mean io u: 0.6483 - val loss: 1.5533 - val mean io u: 0.5726
Epoch 135/300
- mean io u: 0.6529 - val loss: 1.5685 - val mean io u: 0.5846
Epoch 136/300
- mean io u: 0.6544 - val loss: 1.5349 - val mean io u: 0.5727
Epoch 137/300
- mean io u: 0.6489 - val loss: 1.5381 - val mean io u: 0.5788
```

```
Epoch 138/300
- mean io u: 0.6493 - val loss: 1.5580 - val mean io u: 0.5870
Epoch 139/300
- mean io u: 0.6550 - val loss: 1.5593 - val mean io u: 0.5818
Epoch 140/300
- mean io u: 0.6551 - val loss: 1.5592 - val mean io u: 0.5761
Epoch 141/300
- mean io u: 0.6453 - val loss: 1.4867 - val mean io u: 0.5737
Epoch 142/300
- mean io u: 0.6418 - val loss: 1.5158 - val mean io u: 0.5701
Epoch 143/300
- mean io u: 0.6405 - val loss: 1.5752 - val mean io u: 0.5788
Epoch 144/300
- mean io u: 0.6438 - val loss: 1.5065 - val mean io u: 0.5716
Epoch 145/300
- mean io u: 0.6541 - val loss: 1.5558 - val mean io u: 0.5863
Epoch 146/300
- mean io u: 0.6512 - val loss: 1.5535 - val mean io u: 0.5804
Epoch 147/300
- mean io u: 0.6380 - val loss: 1.5551 - val mean io u: 0.5785
Epoch 148/300
- mean io u: 0.6516 - val loss: 1.5881 - val mean io u: 0.5899
Epoch 149/300
- mean io u: 0.6612 - val loss: 1.5911 - val mean io u: 0.5904
Epoch 150/300
- mean io u: 0.6628 - val loss: 1.6030 - val mean io u: 0.5898
```

```
Epoch 151/300
- mean io u: 0.6633 - val loss: 1.5277 - val mean io u: 0.5787
Epoch 152/300
- mean io u: 0.6536 - val loss: 1.5678 - val mean io u: 0.5854
Epoch 153/300
- mean io u: 0.6528 - val loss: 1.5487 - val mean io u: 0.5850
Epoch 154/300
- mean io u: 0.6528 - val loss: 1.5604 - val mean io u: 0.5880
Epoch 155/300
- mean io u: 0.6503 - val loss: 1.5382 - val mean io u: 0.5722
Epoch 156/300
- mean io u: 0.6464 - val loss: 1.5623 - val mean io u: 0.5781
Epoch 157/300
- mean io u: 0.6446 - val loss: 1.5423 - val mean io u: 0.5827
Epoch 158/300
- mean io u: 0.6585 - val loss: 1.5985 - val mean io u: 0.5887
Epoch 159/300
- mean io u: 0.6612 - val loss: 1.5954 - val mean io u: 0.5945
Epoch 160/300
- mean io u: 0.6605 - val loss: 1.5636 - val mean io u: 0.5809
Epoch 161/300
- mean io u: 0.6632 - val loss: 1.5843 - val mean io u: 0.5903
Epoch 162/300
- mean io u: 0.6658 - val loss: 1.6257 - val mean io u: 0.5892
Epoch 163/300
- mean io u: 0.6522 - val loss: 1.5017 - val mean io u: 0.5794
```

```
Epoch 164/300
- mean io u: 0.6443 - val loss: 1.5261 - val mean io u: 0.5783
Epoch 165/300
- mean io u: 0.6546 - val loss: 1.5942 - val mean io u: 0.5880
Epoch 166/300
- mean io u: 0.6559 - val loss: 1.5416 - val mean io u: 0.5796
Epoch 167/300
- mean io u: 0.6575 - val loss: 1.5623 - val mean io u: 0.5807
Epoch 168/300
- mean io u: 0.6566 - val loss: 1.5149 - val mean io u: 0.5805
Epoch 169/300
- mean io u: 0.6538 - val loss: 1.5165 - val mean io u: 0.5795
Epoch 170/300
- mean io u: 0.6622 - val loss: 1.5880 - val mean io u: 0.5878
Epoch 171/300
- mean io u: 0.6727 - val loss: 1.5961 - val mean io u: 0.5951
Epoch 172/300
- mean io u: 0.6732 - val loss: 1.5915 - val mean io u: 0.5960
Epoch 173/300
- mean io u: 0.6670 - val loss: 1.5332 - val mean io u: 0.5853
Epoch 174/300
- mean io u: 0.6687 - val loss: 1.5660 - val mean io u: 0.5870
Epoch 175/300
- mean io u: 0.6682 - val loss: 1.5823 - val mean io u: 0.5915
Epoch 176/300
- mean io u: 0.6679 - val loss: 1.5590 - val mean io u: 0.5890
```

```
Epoch 177/300
- mean io u: 0.6608 - val loss: 1.5408 - val mean io u: 0.5817
Epoch 178/300
- mean io u: 0.6521 - val loss: 1.5173 - val mean io u: 0.5772
Epoch 179/300
- mean io u: 0.6546 - val loss: 1.5530 - val mean io u: 0.5839
Epoch 180/300
- mean io u: 0.6666 - val loss: 1.5937 - val mean io u: 0.5954
Epoch 181/300
- mean io u: 0.6716 - val loss: 1.6100 - val mean io u: 0.5950
Epoch 182/300
- mean io u: 0.6722 - val loss: 1.5891 - val mean io u: 0.5884
Epoch 183/300
- mean io u: 0.6765 - val loss: 1.5975 - val mean io u: 0.5954
Epoch 184/300
- mean io u: 0.6768 - val loss: 1.5748 - val mean io u: 0.5901
Epoch 185/300
- mean io u: 0.6720 - val loss: 1.6126 - val mean io u: 0.5982
Epoch 186/300
- mean io u: 0.6575 - val loss: 1.5746 - val mean io u: 0.5831
Epoch 187/300
- mean io u: 0.6481 - val loss: 1.4956 - val mean io u: 0.5751
Epoch 188/300
- mean io u: 0.6604 - val loss: 1.5943 - val mean io u: 0.5975
Epoch 189/300
- mean io u: 0.6697 - val loss: 1.5632 - val mean io u: 0.5884
```

```
Epoch 190/300
- mean io u: 0.6767 - val loss: 1.5914 - val mean io u: 0.5928
Epoch 191/300
- mean io u: 0.6789 - val loss: 1.6066 - val mean io u: 0.5951
Epoch 192/300
- mean io u: 0.6706 - val loss: 1.5510 - val mean io u: 0.5844
Epoch 193/300
- mean io u: 0.6616 - val loss: 1.5810 - val mean io u: 0.5883
Epoch 194/300
- mean io u: 0.6678 - val loss: 1.5636 - val mean io u: 0.5865
Epoch 195/300
- mean io u: 0.6741 - val loss: 1.5545 - val mean io u: 0.5895
Epoch 196/300
- mean io u: 0.6766 - val loss: 1.5576 - val mean io u: 0.5932
Epoch 197/300
- mean io u: 0.6779 - val loss: 1.5947 - val mean io u: 0.5949
Epoch 198/300
- mean io u: 0.6783 - val loss: 1.5962 - val mean io u: 0.5909
Epoch 199/300
- mean io u: 0.6745 - val loss: 1.5331 - val mean io u: 0.5863
Epoch 200/300
- mean io u: 0.6743 - val loss: 1.5803 - val mean io u: 0.5941
Epoch 201/300
- mean io u: 0.6755 - val loss: 1.6395 - val mean io u: 0.6014
Epoch 202/300
- mean io u: 0.6798 - val loss: 1.5453 - val mean io u: 0.5843
```

```
Epoch 203/300
- mean io u: 0.6751 - val loss: 1.5771 - val mean io u: 0.5904
Epoch 204/300
- mean io u: 0.6730 - val loss: 1.5967 - val mean io u: 0.5966
Epoch 205/300
- mean io u: 0.6725 - val loss: 1.5494 - val mean io u: 0.5923
Epoch 206/300
- mean io u: 0.6682 - val loss: 1.5947 - val mean io u: 0.5882
Epoch 207/300
- mean io u: 0.6711 - val loss: 1.5954 - val mean io u: 0.5946
Epoch 208/300
- mean io u: 0.6819 - val loss: 1.5548 - val mean io u: 0.5906
Epoch 209/300
- mean io u: 0.6817 - val loss: 1.5876 - val mean io u: 0.5927
Epoch 210/300
- mean io u: 0.6831 - val loss: 1.6139 - val mean io u: 0.6005
Epoch 211/300
- mean io u: 0.6831 - val loss: 1.5830 - val mean io u: 0.5921
Epoch 212/300
- mean io u: 0.6861 - val loss: 1.5998 - val mean io u: 0.5961
Epoch 213/300
- mean io u: 0.6832 - val loss: 1.5818 - val mean io u: 0.5929
Epoch 214/300
- mean io u: 0.6807 - val loss: 1.5591 - val mean io u: 0.5931
Epoch 215/300
- mean io u: 0.6712 - val loss: 1.5660 - val mean io u: 0.5888
```

```
Epoch 216/300
- mean io u: 0.6794 - val loss: 1.5820 - val mean io u: 0.5977
Epoch 217/300
- mean io u: 0.6869 - val loss: 1.6140 - val mean io u: 0.5983
Epoch 218/300
- mean io u: 0.6867 - val loss: 1.5996 - val mean io u: 0.5952
Epoch 219/300
- mean io u: 0.6829 - val loss: 1.5570 - val mean io u: 0.5917
Epoch 220/300
- mean io u: 0.6820 - val loss: 1.6200 - val mean io u: 0.5965
Epoch 221/300
- mean io u: 0.6836 - val loss: 1.5908 - val mean io u: 0.5898
Epoch 222/300
- mean io u: 0.6772 - val loss: 1.5544 - val mean io u: 0.5870
Epoch 223/300
- mean io u: 0.6723 - val loss: 1.5780 - val mean io u: 0.5925
Epoch 224/300
- mean io u: 0.6784 - val loss: 1.5644 - val mean io u: 0.5937
Epoch 225/300
- mean io u: 0.6806 - val loss: 1.5660 - val mean io u: 0.5889
Epoch 226/300
- mean io u: 0.6809 - val loss: 1.6261 - val mean io u: 0.6045
Epoch 227/300
- mean io u: 0.6842 - val loss: 1.5936 - val mean io u: 0.5972
Epoch 228/300
- mean io u: 0.6824 - val loss: 1.5772 - val mean io u: 0.5942
```

```
Epoch 229/300
- mean io u: 0.6786 - val loss: 1.5476 - val mean io u: 0.5932
Epoch 230/300
- mean io u: 0.6549 - val loss: 1.4755 - val mean io u: 0.5678
Epoch 231/300
- mean io u: 0.6600 - val loss: 1.5490 - val mean io u: 0.5884
Epoch 232/300
- mean io u: 0.6798 - val loss: 1.5879 - val mean io u: 0.5942
Epoch 233/300
- mean io u: 0.6906 - val loss: 1.6193 - val mean io u: 0.5991
Epoch 234/300
- mean io u: 0.6876 - val loss: 1.5939 - val mean io u: 0.5951
Epoch 235/300
- mean io u: 0.6899 - val loss: 1.5826 - val mean io u: 0.5984
Epoch 236/300
- mean io u: 0.6958 - val loss: 1.6216 - val mean io u: 0.6036
Epoch 237/300
- mean io u: 0.6947 - val loss: 1.6133 - val mean io u: 0.6042
Epoch 238/300
- mean io u: 0.6981 - val loss: 1.5973 - val mean io u: 0.6024
Epoch 239/300
- mean io u: 0.6947 - val loss: 1.6257 - val mean io u: 0.6022
Epoch 240/300
- mean io u: 0.6930 - val loss: 1.5645 - val mean io u: 0.5970
Epoch 241/300
- mean io u: 0.6891 - val loss: 1.5885 - val mean io u: 0.5983
```

```
Epoch 242/300
- mean io u: 0.6889 - val loss: 1.5802 - val mean io u: 0.6014
Epoch 243/300
- mean io u: 0.6851 - val loss: 1.5510 - val mean io u: 0.5889
Epoch 244/300
- mean io u: 0.6794 - val loss: 1.5777 - val mean io u: 0.5875
Epoch 245/300
- mean io u: 0.6819 - val loss: 1.6010 - val mean io u: 0.6027
Epoch 246/300
- mean io u: 0.6865 - val loss: 1.5883 - val mean io u: 0.5959
Epoch 247/300
- mean io u: 0.6910 - val loss: 1.6060 - val mean io u: 0.6026
Epoch 248/300
- mean io u: 0.6980 - val loss: 1.6038 - val mean io u: 0.6003
Epoch 249/300
- mean io u: 0.6916 - val loss: 1.5837 - val mean io u: 0.5972
Epoch 250/300
- mean io u: 0.6913 - val loss: 1.5531 - val mean io u: 0.5900
Epoch 251/300
- mean io u: 0.6901 - val loss: 1.5723 - val mean io u: 0.5909
Epoch 252/300
- mean io u: 0.6861 - val loss: 1.6026 - val mean io u: 0.6008
Epoch 253/300
- mean io u: 0.6877 - val loss: 1.5789 - val_mean_io_u: 0.5924
Epoch 254/300
- mean io u: 0.6919 - val loss: 1.6054 - val mean io u: 0.6017
```

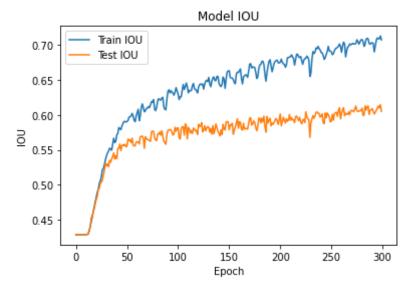
```
Epoch 255/300
- mean io u: 0.6945 - val loss: 1.6083 - val mean io u: 0.6028
Epoch 256/300
- mean io u: 0.6958 - val loss: 1.6097 - val mean io u: 0.6039
Epoch 257/300
- mean io u: 0.6965 - val loss: 1.5840 - val mean io u: 0.5979
Epoch 258/300
- mean io u: 0.6986 - val loss: 1.6218 - val mean io u: 0.6064
Epoch 259/300
- mean io u: 0.7015 - val loss: 1.6427 - val mean io u: 0.6101
Epoch 260/300
- mean io u: 0.6974 - val loss: 1.5691 - val mean io u: 0.5943
Epoch 261/300
- mean io u: 0.6896 - val loss: 1.5900 - val mean io u: 0.6020
Epoch 262/300
- mean io u: 0.6987 - val loss: 1.6249 - val mean io u: 0.6072
Epoch 263/300
- mean io u: 0.7041 - val loss: 1.6157 - val mean io u: 0.6102
Epoch 264/300
- mean io u: 0.7012 - val loss: 1.5659 - val mean io u: 0.5988
Epoch 265/300
- mean io u: 0.6829 - val loss: 1.5995 - val mean io u: 0.6025
Epoch 266/300
- mean io u: 0.6916 - val loss: 1.5879 - val mean io u: 0.6008
Epoch 267/300
- mean io u: 0.6971 - val loss: 1.5893 - val mean io u: 0.6007
```

```
Epoch 268/300
- mean io u: 0.6981 - val loss: 1.6049 - val mean io u: 0.6043
Epoch 269/300
- mean io u: 0.7016 - val loss: 1.5784 - val mean io u: 0.6002
Epoch 270/300
- mean io u: 0.7005 - val loss: 1.6034 - val mean io u: 0.6057
Epoch 271/300
- mean io u: 0.7012 - val loss: 1.6296 - val mean io u: 0.6066
Epoch 272/300
- mean io u: 0.7024 - val loss: 1.6229 - val mean io u: 0.6056
Epoch 273/300
- mean io u: 0.7013 - val loss: 1.5944 - val mean io u: 0.6044
Epoch 274/300
- mean io u: 0.7014 - val loss: 1.6156 - val mean io u: 0.6089
Epoch 275/300
- mean io u: 0.7022 - val loss: 1.6443 - val mean io u: 0.6127
Epoch 276/300
- mean io u: 0.7076 - val loss: 1.6174 - val mean io u: 0.6059
Epoch 277/300
- mean io u: 0.7048 - val loss: 1.5904 - val mean io u: 0.6049
Epoch 278/300
- mean io u: 0.7013 - val loss: 1.6032 - val mean io u: 0.6051
Epoch 279/300
- mean io u: 0.7043 - val loss: 1.6180 - val mean io u: 0.6064
Epoch 280/300
- mean io u: 0.7074 - val loss: 1.6092 - val mean io u: 0.6060
```

```
Epoch 281/300
- mean io u: 0.7079 - val loss: 1.6337 - val mean io u: 0.6096
Epoch 282/300
- mean io u: 0.7054 - val loss: 1.5776 - val mean io u: 0.5992
Epoch 283/300
- mean io u: 0.6981 - val loss: 1.5912 - val mean io u: 0.5997
Epoch 284/300
- mean io u: 0.7009 - val loss: 1.6255 - val mean io u: 0.6128
Epoch 285/300
- mean io u: 0.7081 - val loss: 1.6134 - val mean io u: 0.6068
Epoch 286/300
- mean io u: 0.7097 - val loss: 1.6254 - val mean io u: 0.6111
Epoch 287/300
- mean io u: 0.7088 - val loss: 1.6292 - val mean io u: 0.6127
Epoch 288/300
- mean io u: 0.7022 - val loss: 1.5814 - val mean io u: 0.6005
Epoch 289/300
- mean io u: 0.7010 - val loss: 1.6056 - val mean io u: 0.6068
Epoch 290/300
- mean io u: 0.7029 - val loss: 1.6117 - val mean io u: 0.6074
Epoch 291/300
- mean io u: 0.7033 - val loss: 1.6059 - val mean io u: 0.6091
Epoch 292/300
- mean io u: 0.7024 - val loss: 1.5841 - val mean io u: 0.6047
Epoch 293/300
- mean io u: 0.6901 - val loss: 1.5771 - val mean io u: 0.6012
```

```
Epoch 294/300
      - mean io u: 0.6989 - val loss: 1.6102 - val mean io u: 0.6037
      Epoch 295/300
      - mean io u: 0.7069 - val loss: 1.6138 - val mean io u: 0.6067
      Epoch 296/300
      - mean io u: 0.7102 - val loss: 1.6232 - val mean io u: 0.6099
      Epoch 297/300
      - mean io u: 0.7089 - val loss: 1.6489 - val mean io u: 0.6128
      Epoch 298/300
      - mean io u: 0.7088 - val loss: 1.6266 - val mean io u: 0.6098
      Epoch 299/300
      - mean io u: 0.7130 - val loss: 1.6464 - val mean io u: 0.6146
      Epoch 300/300
      - mean io u: 0.7073 - val loss: 1.5954 - val mean io u: 0.6051
In [42]: # Saving model
      from keras.models import load model
      model.save('my model.h5py') # creates a h5py file 'my model.h5'
      WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorfl
      ow/python/ops/resource variable ops.py:1817: calling BaseResourceVariab
      le. init (from tensorflow.python.ops.resource variable ops) with con
      straint is deprecated and will be removed in a future version.
      Instructions for updating:
      If using Keras pass * constraint arguments to layers.
      INFO:tensorflow:Assets written to: my model.h5py/assets
In [43]: print(hist2.history.keys())
      dict keys(['loss', 'mean io u', 'val loss', 'val mean io u'])
```

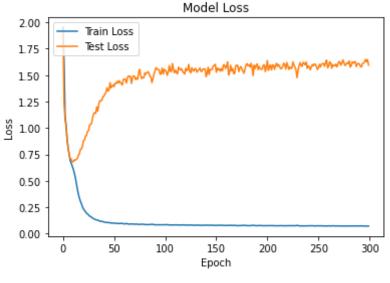
```
In [44]: plt.plot(hist2.history['mean_io_u'])
   plt.plot(hist2.history['val_mean_io_u'])
   plt.title('Model IOU')
   plt.ylabel('IOU')
   plt.xlabel('Epoch')
   plt.legend(['Train IOU', 'Test IOU'], loc='upper left')
   plt.show()
```



Conclusion:

1.Looks good as we are getting 60% Mean IOU after 300 epochs

```
In [45]: plt.plot(hist2.history['loss'])
   plt.plot(hist2.history['val_loss'])
   plt.title('Model Loss')
   plt.ylabel('Loss')
   plt.xlabel('Epoch')
   plt.legend(['Train Loss', 'Test Loss'], loc='upper left')
   plt.show()
```



```
In [0]: #Prediction on test dataset
        y pred = model.predict(test dataset)
 In [0]: y predi = tf.argmax(y pred, axis=3)
In [78]: y_predi.shape
Out[78]: TensorShape([204, 224, 224])
 In [2]: import numpy as np
        # Assigning some RGB colors for the 7 + 1 (Misc) classes
         colors = np.array([
             [255, 192, 203],
                              # Drivable
             [244, 35, 232], # Non Drivable
             [220, 20, 60], # Living Things
             [0, 0, 230],
                             # Vehicles
             [220, 190, 40], # Road Side Objects
            [70, 70, 70], # Far Objects
             [70, 130, 180],
                             # Sky
```

```
[0, 0, 0]
                                # Misc
        1, dtvpe=np.int)
In [0]: import matplotlib.pyplot as plt
        from matplotlib.pyplot import imread
        import numpy as np
        from glob import glob
        image paths = glob('idd20k lite/leftImg8bit/train/*/* image.jpg')
        label paths = [p.replace('leftImg8bit', 'gtFine').replace(' image.jpg',
         ' label.png') for p in image paths]
        image paths val = glob('idd20k lite/leftImg8bit/val/*/* image.jpg')
        label paths val = [p.replace('leftImg8bit', 'gtFine').replace(' image.j
        pg', 'label.png') for p in image paths val]
        #This function returns the original image, ground truth and predicted o
        utput
        def output(image no):
            label map = imread(label paths val[image no])
            image frame = imread(image paths val[image no])
            color image = np.zeros((label map.shape[0], label map.shape[1], 3),
         dtype=np.int)
            for i in range(7):
                color image[label map == i] = colors[i]
            color image[label map == 255] = colors[7]
            plt.imshow(image frame)
            #plt.imshow(color image, alpha=0.5)
            print('Original image is')
            plt.show()
            label map=label paths val[image no]
            image frame = imread(image paths val[image no])
            color image = np.zeros((label map.shape[0], label map.shape[1], 3),
         dtype=np.int)
            for i in range(7):
                color image[label map == i] = colors[i]
            color image[label map == 255] = colors[7]
            plt.imshow(image frame)
            plt.imshow(color image, alpha=0.8)
```

```
print('Original masked image is')
plt.show()

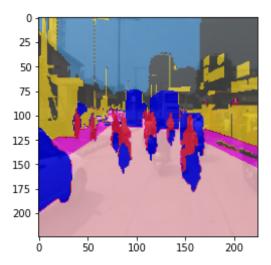
label_map=y_predi[image_no]
image_frame = imread(image_paths_val[image_no])
color_image = np.zeros((label_map.shape[0], label_map.shape[1], 3),
dtype=np.int)
for i in range(7):
    color_image[label_map == i] = colors[i]
color_image[label_map == 255] = colors[7]
plt.imshow(image_frame)
plt.imshow(color_image, alpha=0.8)
print('Predicted masked image')
plt.show()
```

In [126]: output(1)

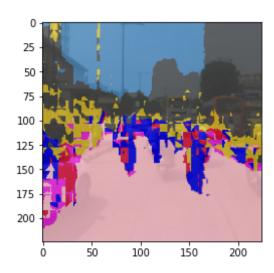
Original image is



Original masked image is

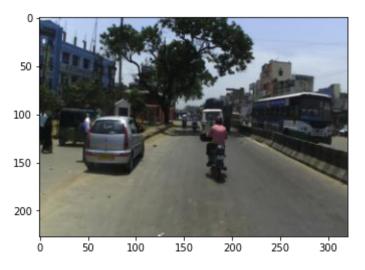


Predicted masked image

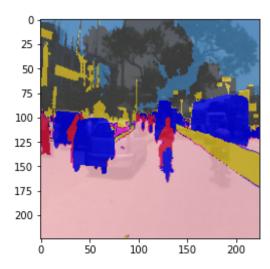


In [128]: output(106)

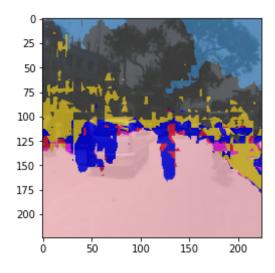
Original image is



Original masked image is



Predicted masked image



Conclusion:

1. Good predictions for highly weighted classes

In []: