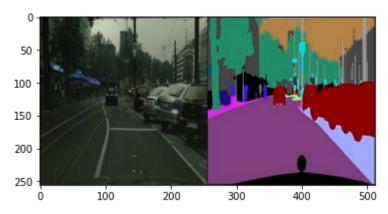
```
In [28]:
          from google.colab import drive
 In [ ]:
          drive.mount('/content/drive')
         Mounted at /content/drive
          import tensorflow as tf
In [29]:
          import tensorflow.keras.layers as tfl
          import os
          from tqdm import tqdm
          import cv2
          import matplotlib.pyplot as plt
          import numpy as np
          from sklearn.model selection import train test split
          from sklearn.cluster import KMeans
          from PIL import Image
          from os.path import splitext
          os.environ['TF CPP MIN LOG LEVEL'] = '3'
In [30]:
          path="/content/drive/MyDrive/cityscapes data"
          def images_upload(path):
              images=[]
              for root, subfolders, files in os.walk(path):
                  for file in tqdm(files):
                      filename=root+os.sep+file
                      if filename.endswith('jpg') or filename.endwith('png'):
                           images.append(filename)
              return images
          images=images_upload(path)
         0it [00:00, ?it/s]
         100%
                      | | 500/500 [00:00<00:00, 347440.69it/s]
         100%
                        | 1330/1330 [00:00<00:00, 681043.14it/s]
In [31]:
          def convert_image_rgb(data):
              imgs=[]
              for i in tqdm(data):
                  img = cv2.imread(i,cv2.COLOR BGR2RGB)
                  del i
                  imgs.append(img)
              return imgs
          img=convert_image_rgb(images)
         100% | 1830/1830 [00:17<00:00, 106.43it/s]
          plt.imshow(img[np.random.randint(0,len(img))]);
In [32]:
```



```
def split_input_mask(data):
In [34]:
               inputs=[]
               mask=[]
               for i in data:
                   a=i[:,:256]
                   inputs.append(a)
                   b=i[:,256:]
                   mask.append(b)
               return inputs,mask
          inputs,mask=split_input_mask(img)
In [35]:
          del images
          del img
          def show_images(data):
In [36]:
              plt.figure(figsize=(10,10))
               for i in range(9):
                   idx=np.random.randint(0,len(data))
```

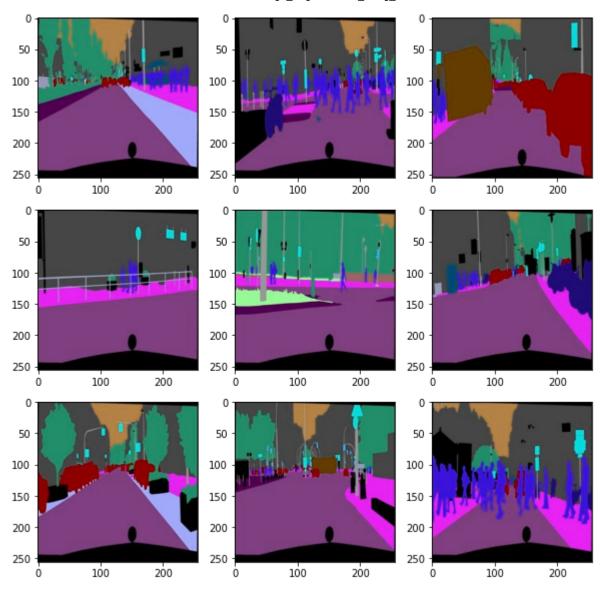
plt.subplot(3,3,i+1)

img=data[idx]
plt.imshow(img)

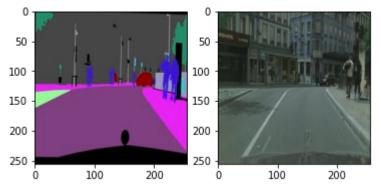
show\_images(inputs)



In [37]: show\_images(mask)

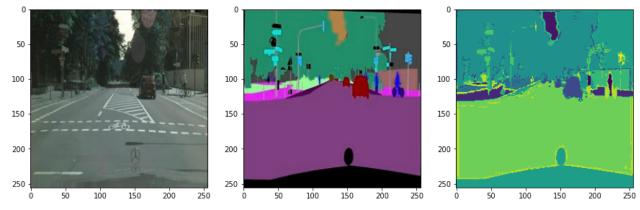


```
In [38]: def images_compare(inputs,mask):
    idx_new=np.random.randint(0,len(mask))
    fig = plt.figure()
    ax1 = fig.add_subplot(1,2,1)
    ax1.imshow(mask[idx_new])
    ax2 = fig.add_subplot(1,2,2)
    ax2.imshow(inputs[idx_new],cmap='gray')
    plt.show()
    images_compare(inputs,mask)
```



```
In [39]: num_items = 1000
```

```
color_array = np.random.choice(range(256), 3*num_items).reshape(-1,3)
num_classes = 20
label_model = KMeans(n_clusters = num_classes)
label_model.fit(color_array)
label_class = label_model.predict(mask[10].reshape(-1,3)).reshape(256,256)
fig, axes = plt.subplots(1,3,figsize=(15,5))
axes[0].imshow(inputs[10]);
axes[1].imshow(mask[10]);
axes[2].imshow(label_class);
```

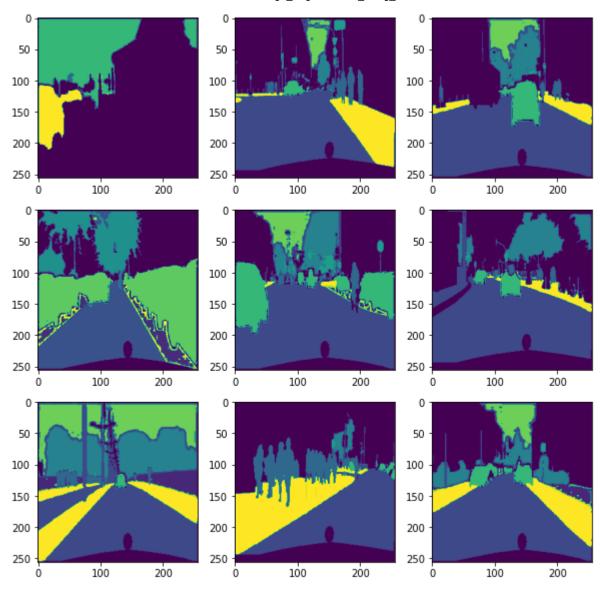


```
In [40]:
    def new_labels(mask):
        num_items = 1000
        color_array = np.random.choice(range(256), 3*num_items).reshape(-1,3)
        num_classes = 10
        label_model = KMeans(n_clusters = num_classes)
        label_model.fit(color_array)
        labels=[]
        for i in tqdm(range(len(mask))):
            label_class = label_model.predict(mask[i].reshape(-1,3)).reshape(256,256)
            labels.append(label_class)
        return labels
```

```
In [41]: labels=new_labels(mask)
    idx=np.random.randint(0,len(labels))
    classes,freq=np.unique(labels[idx],return_counts=True)
    print(f'number of classes :{len(classes)}')
```

100%| 1830/1830 [00:08<00:00, 224.34it/s] number of classes :10

```
In [42]: plt.figure(figsize=(10,10))
    for i in range(9):
        plt.subplot(3,3,i+1)
        im=labels[i]
        plt.imshow(im)
# plt.imshow(a)
```



```
In [43]:
    def rescale(data):
        rescaled=[]
        for i in tqdm(data):
            img=tf.image.convert_image_dtype(i, tf.float32)
            del i
            rescaled.append(img)
        return rescaled
```

```
In [44]: rescaled_input=rescale(inputs)
```

100%| 1830/1830 [00:01<00:00, 1513.65it/s]

```
In [45]: def build_model(inputsize=(256,256,3),classes=None):
    inputs = tf.keras.Input(shape=(inputsize))

conv = tfl.Conv2D(32, (3, 3), padding="same",kernel_initializer='he_normal', name='
    inputs)
    x=tfl.BatchNormalization()(conv)
    x=tfl.LeakyReLU()(x)
    x1 = tfl.Conv2D(32, (3, 3), padding="same",kernel_initializer='he_normal', name='Co
    x)
    x=tfl.BatchNormalization()(x1)
    x=tfl.LeakyReLU()(x)
```

```
x = tfl.MaxPool2D(pool_size=(2, 2), strides=(2, 2), name='MaxPool1')(x)
x = tfl.Conv2D(64, (3, 3), padding="same",kernel_initializer='he_normal', name='Con
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)
x2 = tfl.Conv2D(64, (3, 3), padding="same",kernel_initializer='he_normal', name='Co
x=tfl.BatchNormalization()(x2)
x=tfl.LeakyReLU()(x)
x = tfl.MaxPool2D(pool_size=(2, 2), name='MaxPool2')(x)
x = tfl.Conv2D(128, (3, 3), padding="same", kernel_initializer='he_normal', name='Co
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)
x3 = tfl.Conv2D(128, (3, 3), padding="same",kernel_initializer='he_normal', name='C
x=tfl.BatchNormalization()(x3)
x=tfl.LeakyReLU()(x)
x = tfl.MaxPool2D(pool_size=(2, 2), strides=(2, 2), name='MaxPool3')(x)
x = tfl.Conv2D(256, (3, 3), padding="same",kernel initializer='he normal', name='Co
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)
x = tfl.Conv2D(256, (3, 3), padding="same",kernel initializer='he normal', name='Co
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)
x = tfl.Conv2DTranspose(128, (3, 3), strides=2, padding="same")(x)
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)
x = tfl.concatenate([x, x3], axis=3)
x = tfl.Conv2D(128, (3, 3), padding="same",kernel initializer='he normal', name='Co
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)
x = tfl.Conv2D(128, (3, 3), padding="same", kernel initializer='he normal', name='Co
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)
x = tfl.Conv2DTranspose(64, (3, 3), strides=2, padding="same")(x)
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)
x = tfl.concatenate([x, x2], axis=3)
x = tfl.Conv2D(64, (3, 3), padding="same",kernel_initializer='he_normal', name='Con
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)
x = tfl.Conv2D(64, (3, 3), padding="same", kernel_initializer='he_normal', name='Con
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)
x = tfl.Conv2DTranspose(32, (3, 3), strides=2, padding="same")(x)
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)
x = tfl.concatenate([x, x1], axis=3)
x = tfl.Conv2D(32, (3, 3), padding="same",kernel_initializer='he_normal', name='Con
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)
```

```
x = tfl.Conv2D(32, (3, 3), padding="same",kernel_initializer='he_normal', name='Con
x=tfl.BatchNormalization()(x)
x=tfl.LeakyReLU()(x)

outputs = tfl.Conv2D(classes, (1, 1), padding="same", activation='softmax', name='0
final_model = tf.keras.Model(inputs=inputs, outputs=outputs)
final_model.summary()
return final_model
```

## In [46]: mymodel=build\_model(classes=10)

Model: "model\_1"

Layer (type)	Output Shape	Param #	Connected to
=======			
<pre>input_2 (InputLayer)</pre>	[(None, 256, 256, 3)]	0	[]
Conv1 (Conv2D)	(None, 256, 256, 32)	896	['input_2[0][0]']
<pre>batch_normalization_17 (BatchN ormalization)</pre>	(None, 256, 256, 32)	128	['Conv1[0][0]']
<pre>leaky_re_lu_17 (LeakyReLU) 7[0][0]']</pre>	(None, 256, 256, 32	0	['batch_normalization_1
Conv2 (Conv2D) [0]']	(None, 256, 256, 32	9248	['leaky_re_lu_17[0]
	)		
<pre>batch_normalization_18 (BatchN ormalization)</pre>	(None, 256, 256, 32)	128	['Conv2[0][0]']
leaky_re_lu_18 (LeakyReLU) 8[0][0]']	(None, 256, 256, 32	0	['batch_normalization_1
<pre>MaxPool1 (MaxPooling2D) [0]']</pre>	(None, 128, 128, 32)	0	['leaky_re_lu_18[0]
Conv3 (Conv2D)	(None, 128, 128, 64	18496	['MaxPool1[0][0]']
<pre>batch_normalization_19 (BatchN ormalization)</pre>	(None, 128, 128, 64)	256	['Conv3[0][0]']
leaky_re_lu_19 (LeakyReLU) 9[0][0]']	(None, 128, 128, 64	0	['batch_normalization_1
Conv4 (Conv2D) [0]']	(None, 128, 128, 64	36928	['leaky_re_lu_19[0]
batch_normalization_20 (BatchN	(None, 128, 128, 64	256	['Conv4[0][0]']

```
ormalization)
                                )
leaky_re_lu_20 (LeakyReLU)
                                (None, 128, 128, 64 0
                                                                 ['batch_normalization_2
0[0][0]']
MaxPool2 (MaxPooling2D)
                                (None, 64, 64, 64)
                                                                 ['leaky re lu 20[0]
[0]']
                                (None, 64, 64, 128) 73856
                                                                 ['MaxPool2[0][0]']
Conv5 (Conv2D)
batch normalization 21 (BatchN (None, 64, 64, 128) 512
                                                                 ['Conv5[0][0]']
ormalization)
leaky re lu 21 (LeakyReLU)
                                (None, 64, 64, 128) 0
                                                                 ['batch normalization 2
1[0][0]']
Conv6 (Conv2D)
                                (None, 64, 64, 128) 147584
                                                                 ['leaky_re_lu_21[0]
[0]']
batch normalization 22 (BatchN (None, 64, 64, 128) 512
                                                                 ['Conv6[0][0]']
ormalization)
leaky re lu 22 (LeakyReLU)
                                (None, 64, 64, 128) 0
                                                                 ['batch normalization 2
2[0][0]']
MaxPool3 (MaxPooling2D)
                                (None, 32, 32, 128) 0
                                                                 ['leaky_re_lu_22[0]
[0]']
Conv7 (Conv2D)
                                (None, 32, 32, 256) 295168
                                                                 ['MaxPool3[0][0]']
batch_normalization_23 (BatchN (None, 32, 32, 256)
                                                                 ['Conv7[0][0]']
                                                     1024
ormalization)
leaky_re_lu_23 (LeakyReLU)
                                (None, 32, 32, 256) 0
                                                                 ['batch_normalization_2
3[0][0]']
Conv8 (Conv2D)
                                (None, 32, 32, 256)
                                                     590080
                                                                 ['leaky re lu 23[0]
[0]']
batch normalization 24 (BatchN (None, 32, 32, 256)
                                                                 ['Conv8[0][0]']
                                                     1024
ormalization)
leaky re lu 24 (LeakyReLU)
                                (None, 32, 32, 256) 0
                                                                 ['batch normalization 2
4[0][0]']
conv2d transpose 3 (Conv2DTran (None, 64, 64, 128) 295040
                                                                 ['leaky re lu 24[0]
[0]']
spose)
batch normalization 25 (BatchN (None, 64, 64, 128) 512
                                                                 ['conv2d_transpose_3[0]
[0]']
ormalization)
leaky_re_lu_25 (LeakyReLU)
                                (None, 64, 64, 128) 0
                                                                 ['batch_normalization_2
5[0][0]']
concatenate_3 (Concatenate)
                                (None, 64, 64, 256) 0
                                                                 ['leaky_re_lu_25[0]
[0]',
                                                                   'Conv6[0][0]']
```

```
Conv9 (Conv2D)
                                (None, 64, 64, 128) 295040
                                                                  ['concatenate 3[0][0]']
batch_normalization_26 (BatchN (None, 64, 64, 128) 512
                                                                  ['Conv9[0][0]']
ormalization)
leaky re lu 26 (LeakyReLU)
                                (None, 64, 64, 128) 0
                                                                  ['batch normalization 2
6[0][0]']
Conv10 (Conv2D)
                                (None, 64, 64, 128) 147584
                                                                  ['leaky re lu 26[0]
[0]']
batch_normalization_27 (BatchN (None, 64, 64, 128)
                                                                  ['Conv10[0][0]']
                                                      512
ormalization)
leaky_re_lu_27 (LeakyReLU)
                                                                  ['batch normalization 2
                                (None, 64, 64, 128) 0
7[0][0]']
conv2d transpose 4 (Conv2DTran (None, 128, 128, 64 73792
                                                                  ['leaky re lu 27[0]
[0]']
spose)
                                )
batch normalization 28 (BatchN (None, 128, 128, 64 256
                                                                  ['conv2d transpose 4[0]
[0]']
ormalization)
leaky_re_lu_28 (LeakyReLU)
                                (None, 128, 128, 64 0
                                                                  ['batch_normalization_2
8[0][0]']
                                )
concatenate 4 (Concatenate)
                                (None, 128, 128, 12 0
                                                                  ['leaky_re_lu_28[0]
[0]',
                                8)
                                                                   'Conv4[0][0]']
Conv11 (Conv2D)
                                (None, 128, 128, 64 73792
                                                                  ['concatenate_4[0][0]']
batch normalization 29 (BatchN
                                 (None, 128, 128, 64 256
                                                                  ['Conv11[0][0]']
ormalization)
leaky re lu 29 (LeakyReLU)
                                (None, 128, 128, 64 0
                                                                  ['batch normalization 2
9[0][0]']
                                )
Conv12 (Conv2D)
                                (None, 128, 128, 64 36928
                                                                  ['leaky_re_lu_29[0]
[0]']
                                )
batch normalization 30 (BatchN
                                (None, 128, 128, 64
                                                     256
                                                                  ['Conv12[0][0]']
ormalization)
                                )
leaky re lu 30 (LeakyReLU)
                                (None, 128, 128, 64 0
                                                                  ['batch normalization 3
0[0][0]']
                                )
conv2d transpose 5 (Conv2DTran (None, 256, 256, 32 18464
                                                                  ['leaky_re_lu_30[0]
[0]']
                                )
spose)
batch_normalization_31 (BatchN (None, 256, 256, 32 128
                                                                  ['conv2d transpose 5[0]
```

```
[0]']
          ormalization)
                                          )
          leaky_re_lu_31 (LeakyReLU)
                                          (None, 256, 256, 32 0
                                                                           ['batch_normalization_3
         1[0][0]']
                                          )
          concatenate_5 (Concatenate)
                                          (None, 256, 256, 64 0
                                                                           ['leaky_re_lu_31[0]
         [0]',
                                                                             'Conv2[0][0]']
          Conv25 (Conv2D)
                                          (None, 256, 256, 32 18464
                                                                           ['concatenate_5[0][0]']
          batch normalization 32 (BatchN
                                          (None, 256, 256, 32 128
                                                                           ['Conv25[0][0]']
          ormalization)
          leaky_re_lu_32 (LeakyReLU)
                                          (None, 256, 256, 32 0
                                                                           ['batch_normalization_3
         2[0][0]']
          Conv26 (Conv2D)
                                          (None, 256, 256, 32 9248
                                                                           ['leaky_re_lu_32[0]
          [0]']
          batch_normalization_33 (BatchN (None, 256, 256, 32 128
                                                                           ['Conv26[0][0]']
          ormalization)
          leaky_re_lu_33 (LeakyReLU)
                                          (None, 256, 256, 32 0
                                                                           ['batch normalization 3
         3[0][0]']
          Outputs (Conv2D)
                                                                           ['leaky_re_lu_33[0]
                                          (None, 256, 256, 10 330
         [0]']
         Total params: 2,147,466
         Trainable params: 2,144,202
         Non-trainable params: 3,264
          img_file = './model_arch.png'
 In [ ]:
          tf.keras.utils.plot_model(mymodel, to_file=img_file, show_shapes=True, show_layer_names
         ('Failed to import pydot. You must `pip install pydot` and install graphviz (https://gra
         phviz.gitlab.io/download/), ', 'for `pydotprint` to work.')
In [47]:
          del inputs
In [48]:
          def split_data(x,y,test_size=0.2):
              x1=np.array(x)
              del x
              y1=np.array(y)
              del y
```

```
x_train, x_test, y_train, y_test = train_test_split(x1, y1, test_size=test_size)
         return x train, x test, y train, y test
       x train, x test, y train, y test=split data(rescaled input[:1300],labels[:1300],test s
In [49]:
In [50]:
      def callbacks(patience=5):
         checkpoint = tf.keras.callbacks.ModelCheckpoint('seg model.h5', monitor='loss', ver
         early=tf.keras.callbacks.EarlyStopping(monitor='loss',patience=patience)
         callbacks list=[checkpoint, early]
         return callbacks_list
In [51]:
      mymodel.compile(optimizer=tf.keras.optimizers.Adam(learning rate=0.01),loss=tf.keras.lo
      hist=mymodel.fit(x train,y train,batch size=16,epochs=200,callbacks=callbacks())
      Epoch 1/200
      65/65 [=================== ] - ETA: 0s - loss: 1.1677 - acc: 0.6240
      Epoch 00001: loss improved from inf to 1.16775, saving model to seg model.h5
      Epoch 2/200
      65/65 [============= ] - ETA: 0s - loss: 0.9437 - acc: 0.6896
      Epoch 00002: loss improved from 1.16775 to 0.94367, saving model to seg model.h5
      Epoch 3/200
      Epoch 00003: loss improved from 0.94367 to 0.84578, saving model to seg model.h5
      Epoch 4/200
      Epoch 00004: loss improved from 0.84578 to 0.78446, saving model to seg model.h5
      Epoch 5/200
      Epoch 00005: loss improved from 0.78446 to 0.75665, saving model to seg model.h5
      65/65 [============== ] - ETA: 0s - loss: 0.7119 - acc: 0.7608
      Epoch 00006: loss improved from 0.75665 to 0.71187, saving model to seg_model.h5
      65/65 [============= ] - 44s 674ms/step - loss: 0.7119 - acc: 0.7608
      Epoch 7/200
      65/65 [================ ] - ETA: 0s - loss: 0.6921 - acc: 0.7676
      Epoch 00007: loss improved from 0.71187 to 0.69205, saving model to seg model.h5
      Epoch 8/200
      Epoch 00008: loss improved from 0.69205 to 0.66852, saving model to seg model.h5
      Epoch 9/200
      65/65 [============= ] - ETA: 0s - loss: 0.6522 - acc: 0.7813
      Epoch 00009: loss improved from 0.66852 to 0.65219, saving model to seg model.h5
      Epoch 10/200
      65/65 [============== ] - ETA: 0s - loss: 0.6283 - acc: 0.7899
      Epoch 00010: loss improved from 0.65219 to 0.62832, saving model to seg model.h5
      Epoch 11/200
      65/65 [============== - - ETA: 0s - loss: 0.6177 - acc: 0.7942
      Epoch 00011: loss improved from 0.62832 to 0.61772, saving model to seg model.h5
      Epoch 12/200
```

```
65/65 [============= ] - ETA: 0s - loss: 0.6093 - acc: 0.7967
Epoch 00012: loss improved from 0.61772 to 0.60933, saving model to seg model.h5
Epoch 13/200
65/65 [==============] - ETA: 0s - loss: 0.5943 - acc: 0.8027
Epoch 00013: loss improved from 0.60933 to 0.59427, saving model to seg model.h5
Epoch 14/200
Epoch 00014: loss improved from 0.59427 to 0.57863, saving model to seg model.h5
Epoch 15/200
Epoch 00015: loss did not improve from 0.57863
Epoch 16/200
65/65 [============= ] - ETA: 0s - loss: 0.5723 - acc: 0.8088
Epoch 00016: loss improved from 0.57863 to 0.57225, saving model to seg_model.h5
65/65 [============= ] - ETA: 0s - loss: 0.5516 - acc: 0.8167
Epoch 00017: loss improved from 0.57225 to 0.55157, saving model to seg_model.h5
Epoch 18/200
65/65 [============= ] - ETA: 0s - loss: 0.5485 - acc: 0.8167
Epoch 00018: loss improved from 0.55157 to 0.54847, saving model to seg_model.h5
Epoch 19/200
Epoch 00019: loss did not improve from 0.54847
65/65 [============= ] - 44s 677ms/step - loss: 0.5526 - acc: 0.8159
Epoch 20/200
65/65 [=================== ] - ETA: 0s - loss: 0.5395 - acc: 0.8210
Epoch 00020: loss improved from 0.54847 to 0.53954, saving model to seg model.h5
Epoch 21/200
Epoch 00021: loss improved from 0.53954 to 0.53210, saving model to seg model.h5
Epoch 22/200
Epoch 00022: loss improved from 0.53210 to 0.52209, saving model to seg model.h5
Epoch 23/200
Epoch 00023: loss improved from 0.52209 to 0.51691, saving model to seg model.h5
Epoch 24/200
65/65 [================ ] - ETA: 0s - loss: 0.5107 - acc: 0.8305
Epoch 00024: loss improved from 0.51691 to 0.51066, saving model to seg_model.h5
Epoch 25/200
65/65 [============= ] - ETA: 0s - loss: 0.5078 - acc: 0.8319
Epoch 00025: loss improved from 0.51066 to 0.50784, saving model to seg model.h5
Epoch 26/200
65/65 [================ ] - ETA: 0s - loss: 0.5095 - acc: 0.8302
Epoch 00026: loss did not improve from 0.50784
65/65 [============= ] - 44s 676ms/step - loss: 0.5095 - acc: 0.8302
Epoch 27/200
```

```
65/65 [============ ] - ETA: 0s - loss: 0.4996 - acc: 0.8342
Epoch 00027: loss improved from 0.50784 to 0.49959, saving model to seg model.h5
Epoch 28/200
65/65 [============= ] - ETA: 0s - loss: 0.4870 - acc: 0.8382
Epoch 00028: loss improved from 0.49959 to 0.48703, saving model to seg_model.h5
Epoch 29/200
Epoch 00029: loss improved from 0.48703 to 0.47921, saving model to seg model.h5
Epoch 30/200
Epoch 00030: loss did not improve from 0.47921
Epoch 31/200
65/65 [============= ] - ETA: 0s - loss: 0.4840 - acc: 0.8385
Epoch 00031: loss did not improve from 0.47921
Epoch 32/200
65/65 [============= ] - ETA: 0s - loss: 0.4715 - acc: 0.8419
Epoch 00032: loss improved from 0.47921 to 0.47148, saving model to seg_model.h5
Epoch 33/200
65/65 [============= ] - ETA: 0s - loss: 0.4672 - acc: 0.8435
Epoch 00033: loss improved from 0.47148 to 0.46719, saving model to seg_model.h5
Epoch 34/200
Epoch 00034: loss did not improve from 0.46719
Epoch 35/200
65/65 [============== ] - ETA: 0s - loss: 0.4636 - acc: 0.8454
Epoch 00035: loss improved from 0.46719 to 0.46363, saving model to seg model.h5
Epoch 00036: loss improved from 0.46363 to 0.46267, saving model to seg model.h5
Epoch 37/200
Epoch 00037: loss improved from 0.46267 to 0.45387, saving model to seg model.h5
Epoch 38/200
Epoch 00038: loss improved from 0.45387 to 0.44832, saving model to seg model.h5
Epoch 39/200
65/65 [================ ] - ETA: 0s - loss: 0.4390 - acc: 0.8530
Epoch 00039: loss improved from 0.44832 to 0.43905, saving model to seg_model.h5
Epoch 40/200
65/65 [============= ] - ETA: 0s - loss: 0.4340 - acc: 0.8539
Epoch 00040: loss improved from 0.43905 to 0.43397, saving model to seg model.h5
Epoch 41/200
Epoch 00041: loss did not improve from 0.43397
65/65 [============= ] - 44s 677ms/step - loss: 0.4442 - acc: 0.8508
Epoch 42/200
```

```
65/65 [============= ] - ETA: 0s - loss: 0.4293 - acc: 0.8550
Epoch 00042: loss improved from 0.43397 to 0.42935, saving model to seg model.h5
Epoch 43/200
Epoch 00043: loss did not improve from 0.42935
65/65 [============== ] - 44s 677ms/step - loss: 0.4313 - acc: 0.8543
Epoch 44/200
Epoch 00044: loss did not improve from 0.42935
Epoch 45/200
Epoch 00045: loss improved from 0.42935 to 0.42099, saving model to seg model.h5
Epoch 46/200
Epoch 00046: loss improved from 0.42099 to 0.41741, saving model to seg_model.h5
Epoch 47/200
Epoch 00047: loss did not improve from 0.41741
Epoch 48/200
65/65 [============= ] - ETA: 0s - loss: 0.4167 - acc: 0.8586
Epoch 00048: loss improved from 0.41741 to 0.41672, saving model to seg_model.h5
Epoch 49/200
Epoch 00049: loss improved from 0.41672 to 0.40897, saving model to seg model.h5
Epoch 50/200
65/65 [============== ] - ETA: 0s - loss: 0.3977 - acc: 0.8648
Epoch 00050: loss improved from 0.40897 to 0.39766, saving model to seg model.h5
Epoch 00051: loss improved from 0.39766 to 0.39610, saving model to seg model.h5
Epoch 52/200
Epoch 00052: loss improved from 0.39610 to 0.39418, saving model to seg model.h5
Epoch 53/200
Epoch 00053: loss improved from 0.39418 to 0.39342, saving model to seg model.h5
Epoch 54/200
Epoch 00054: loss did not improve from 0.39342
Epoch 55/200
Epoch 00055: loss improved from 0.39342 to 0.37898, saving model to seg model.h5
65/65 [===========] - 44s 680ms/step - loss: 0.3790 - acc: 0.8703
Epoch 56/200
65/65 [============= ] - ETA: 0s - loss: 0.3771 - acc: 0.8713
Epoch 00056: loss improved from 0.37898 to 0.37713, saving model to seg_model.h5
Epoch 57/200
```

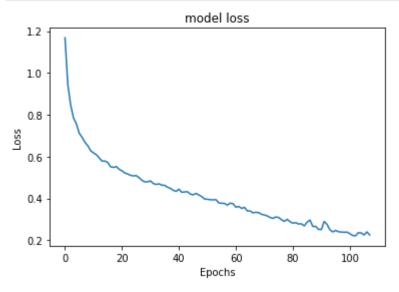
```
65/65 [============= ] - ETA: 0s - loss: 0.3762 - acc: 0.8713
Epoch 00057: loss improved from 0.37713 to 0.37617, saving model to seg model.h5
Epoch 58/200
65/65 [============= ] - ETA: 0s - loss: 0.3683 - acc: 0.8738
Epoch 00058: loss improved from 0.37617 to 0.36833, saving model to seg_model.h5
Epoch 59/200
Epoch 00059: loss did not improve from 0.36833
Epoch 60/200
Epoch 00060: loss did not improve from 0.36833
65/65 [=============== ] - 44s 677ms/step - loss: 0.3749 - acc: 0.8712
Epoch 61/200
65/65 [============== - - ETA: 0s - loss: 0.3581 - acc: 0.8773
Epoch 00061: loss improved from 0.36833 to 0.35810, saving model to seg_model.h5
Epoch 00062: loss did not improve from 0.35810
Epoch 63/200
65/65 [============= ] - ETA: 0s - loss: 0.3525 - acc: 0.8789
Epoch 00063: loss improved from 0.35810 to 0.35247, saving model to seg_model.h5
Epoch 64/200
Epoch 00064: loss did not improve from 0.35247
Epoch 65/200
65/65 [============= ] - ETA: 0s - loss: 0.3406 - acc: 0.8833
Epoch 00065: loss improved from 0.35247 to 0.34064, saving model to seg model.h5
Epoch 00066: loss improved from 0.34064 to 0.34008, saving model to seg model.h5
Epoch 67/200
Epoch 00067: loss improved from 0.34008 to 0.33154, saving model to seg model.h5
Epoch 68/200
Epoch 00068: loss did not improve from 0.33154
65/65 [============= ] - 44s 677ms/step - loss: 0.3339 - acc: 0.8849
Epoch 69/200
Epoch 00069: loss did not improve from 0.33154
Epoch 70/200
Epoch 00070: loss improved from 0.33154 to 0.32441, saving model to seg model.h5
65/65 [===========] - 44s 680ms/step - loss: 0.3244 - acc: 0.8884
Epoch 71/200
65/65 [================ ] - ETA: 0s - loss: 0.3211 - acc: 0.8895
Epoch 00071: loss improved from 0.32441 to 0.32106, saving model to seg_model.h5
65/65 [============ ] - 44s 679ms/step - loss: 0.3211 - acc: 0.8895
Epoch 72/200
```

```
65/65 [============= ] - ETA: 0s - loss: 0.3168 - acc: 0.8910
Epoch 00072: loss improved from 0.32106 to 0.31678, saving model to seg model.h5
Epoch 73/200
65/65 [============= ] - ETA: 0s - loss: 0.3093 - acc: 0.8935
Epoch 00073: loss improved from 0.31678 to 0.30926, saving model to seg model.h5
Epoch 74/200
Epoch 00074: loss improved from 0.30926 to 0.30487, saving model to seg model.h5
Epoch 75/200
Epoch 00075: loss did not improve from 0.30487
65/65 [=============== ] - 44s 679ms/step - loss: 0.3113 - acc: 0.8927
Epoch 76/200
Epoch 00076: loss did not improve from 0.30487
Epoch 77/200
65/65 [================ ] - ETA: 0s - loss: 0.2985 - acc: 0.8973
Epoch 00077: loss improved from 0.30487 to 0.29853, saving model to seg_model.h5
Epoch 78/200
65/65 [============= ] - ETA: 0s - loss: 0.2904 - acc: 0.9002
Epoch 00078: loss improved from 0.29853 to 0.29042, saving model to seg_model.h5
Epoch 79/200
Epoch 00079: loss did not improve from 0.29042
Epoch 80/200
65/65 [============= ] - ETA: 0s - loss: 0.2891 - acc: 0.9004
Epoch 00080: loss improved from 0.29042 to 0.28912, saving model to seg model.h5
Epoch 81/200
Epoch 00081: loss improved from 0.28912 to 0.28206, saving model to seg model.h5
Epoch 82/200
Epoch 00082: loss did not improve from 0.28206
65/65 [==========] - 44s 678ms/step - loss: 0.2845 - acc: 0.9016
Epoch 83/200
Epoch 00083: loss improved from 0.28206 to 0.27793, saving model to seg model.h5
65/65 [============ ] - 44s 680ms/step - loss: 0.2779 - acc: 0.9039
Epoch 84/200
65/65 [================ ] - ETA: 0s - loss: 0.2790 - acc: 0.9035
Epoch 00084: loss did not improve from 0.27793
Epoch 85/200
Epoch 00085: loss improved from 0.27793 to 0.26894, saving model to seg model.h5
Epoch 86/200
65/65 [=================== ] - ETA: 0s - loss: 0.2867 - acc: 0.9006
Epoch 00086: loss did not improve from 0.26894
65/65 [============= ] - 44s 678ms/step - loss: 0.2867 - acc: 0.9006
Epoch 87/200
```

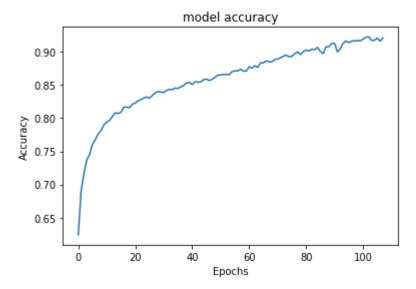
```
65/65 [=========== ] - ETA: 0s - loss: 0.2970 - acc: 0.8973
Epoch 00087: loss did not improve from 0.26894
Epoch 88/200
65/65 [============= ] - ETA: 0s - loss: 0.2665 - acc: 0.9078
Epoch 00088: loss improved from 0.26894 to 0.26647, saving model to seg_model.h5
Epoch 89/200
Epoch 00089: loss did not improve from 0.26647
Epoch 90/200
Epoch 00090: loss improved from 0.26647 to 0.25255, saving model to seg_model.h5
Epoch 91/200
Epoch 00091: loss improved from 0.25255 to 0.25125, saving model to seg_model.h5
Epoch 92/200
Epoch 00092: loss did not improve from 0.25125
Epoch 93/200
65/65 [============== ] - ETA: 0s - loss: 0.2762 - acc: 0.9047
Epoch 00093: loss did not improve from 0.25125
Epoch 94/200
Epoch 00094: loss improved from 0.25125 to 0.25037, saving model to seg model.h5
65/65 [============] - 44s 680ms/step - loss: 0.2504 - acc: 0.9131
Epoch 95/200
65/65 [================ ] - ETA: 0s - loss: 0.2401 - acc: 0.9166
Epoch 00095: loss improved from 0.25037 to 0.24014, saving model to seg model.h5
Epoch 00096: loss did not improve from 0.24014
65/65 [============== ] - 44s 677ms/step - loss: 0.2473 - acc: 0.9139
Epoch 97/200
Epoch 00097: loss did not improve from 0.24014
65/65 [==========] - 44s 677ms/step - loss: 0.2408 - acc: 0.9163
Epoch 98/200
Epoch 00098: loss improved from 0.24014 to 0.23961, saving model to seg model.h5
65/65 [============= ] - 44s 680ms/step - loss: 0.2396 - acc: 0.9167
Epoch 99/200
65/65 [================ ] - ETA: 0s - loss: 0.2383 - acc: 0.9170
Epoch 00099: loss improved from 0.23961 to 0.23828, saving model to seg_model.h5
Epoch 100/200
65/65 [============= ] - ETA: 0s - loss: 0.2393 - acc: 0.9166
Epoch 00100: loss did not improve from 0.23828
Epoch 101/200
65/65 [================ ] - ETA: 0s - loss: 0.2321 - acc: 0.9191
Epoch 00101: loss improved from 0.23828 to 0.23209, saving model to seg_model.h5
Epoch 102/200
```

```
65/65 [============= ] - ETA: 0s - loss: 0.2231 - acc: 0.9221
Epoch 00102: loss improved from 0.23209 to 0.22307, saving model to seg model.h5
Epoch 103/200
Epoch 00103: loss improved from 0.22307 to 0.22144, saving model to seg_model.h5
Epoch 104/200
Epoch 00104: loss did not improve from 0.22144
Epoch 105/200
Epoch 00105: loss did not improve from 0.22144
Epoch 106/200
65/65 [============= ] - ETA: 0s - loss: 0.2254 - acc: 0.9210
Epoch 00106: loss did not improve from 0.22144
Epoch 107/200
Epoch 00107: loss did not improve from 0.22144
Epoch 108/200
65/65 [============== ] - ETA: 0s - loss: 0.2253 - acc: 0.9213
Epoch 00108: loss did not improve from 0.22144
plt.plot(hist.history['loss'])
```





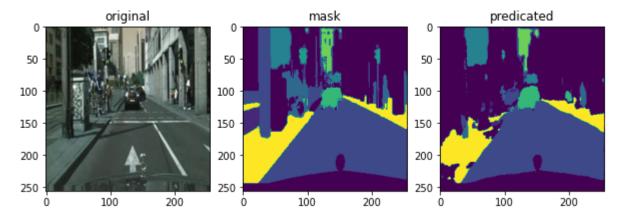
```
In [53]: plt.plot(hist.history["acc"])
    plt.title("model accuracy")
    plt.ylabel("Accuracy")
    plt.xlabel("Epochs")
    plt.show()
```

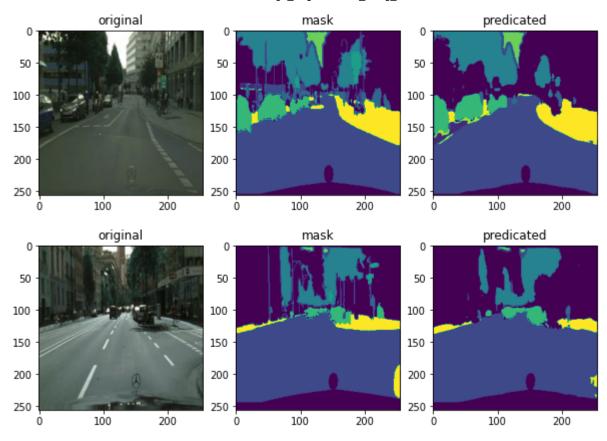


```
In [54]: pred=mymodel.predict(x_test)
    y_pred=tf.argmax(pred,axis=-1)
```

```
In [55]: def show_predications(x_test,y_test,y_pred):
    idx=np.random.randint(0,len(y_pred))
    fig, axes = plt.subplots(1,3,figsize=(10,10))
    axes[0].imshow(x_test[idx])
    axes[0].set_title("original")
    axes[1].imshow(y_test[idx])
    axes[1].set_title("mask")
    axes[2].imshow(y_pred[idx])
    axes[2].set_title("predicated")
```







In [58]: import pandas as pd

df = pd.DataFrame(hist.history)
df

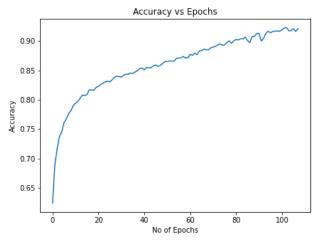
Out[58]:		loss	acc
	0	1.167745	0.624000
	1	0.943668	0.689557
	2	0.845775	0.716860
	3	0.784463	0.737166
	4	0.756650	0.745333
	•••		
	103	0.236433	0.917284
	104	0.235418	0.917608
	105	0.225444	0.920963
	106	0.240224	0.916070
	107	0.225275	0.921252

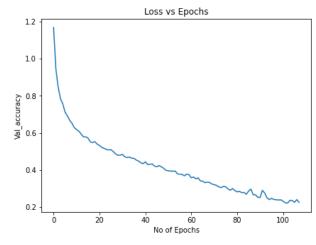
108 rows × 2 columns

```
In [64]: import matplotlib.pyplot as plt
fig, axes = plt.subplots(1,2, figsize=(15, 5))
```

```
axes[0].plot(df.acc)
axes[0].set_xlabel('No of Epochs')
axes[0].set_ylabel('Accuracy')
axes[0].set_title('Accuracy vs Epochs')

axes[1].plot(df.loss)
axes[1].set_xlabel('No of Epochs')
axes[1].set_ylabel('Val_accuracy')
axes[1].set_title('Loss vs Epochs')
plt.show()
```





In [78]:

In [79]:

```
InvalidArgumentError
                                          Traceback (most recent call last)
<ipython-input-79-6a1ed4a6fc89> in <module>()
---> 1 print(iou_coef(y_test,y_pred))
<ipython-input-78-007ee8784880> in iou_coef(y_true, y_pred, smooth)
      1 from keras import backend as K
      2 def iou coef(y true, y pred, smooth=1):
---> 3
          intersection = K.sum(K.abs(y_true * y_pred), axis=[1,2,3])
          union = K.sum(y_true)+K.sum(y_pred)-intersection
     4
          iou = K.mean((intersection + smooth) / (union + smooth), axis=3)
/usr/local/lib/python3.7/dist-packages/tensorflow/python/util/traceback utils.py in erro
r_handler(*args, **kwargs)
            except Exception as e:
   151
              filtered tb = process traceback frames(e. traceback )
    152
--> 153
              raise e.with traceback(filtered tb) from None
   154
            finally:
              del filtered tb
    155
/usr/local/lib/python3.7/dist-packages/keras/backend.py in sum(x, axis, keepdims)
   2453
              A tensor with sum of `x`.
   2454
-> 2455
          return tf.reduce_sum(x, axis, keepdims)
  2456
   2457
```

InvalidArgumentError: Invalid reduction dimension (3 for input with 3 dimension(s) [Op:S
um]