

## ▼ Road Finder

### ▼ Initial Config

```
from google.colab import drive
drive.mount('/content/gdrive')
base_path = 'gdrive/My\ Drive/road_finder_data/'
%cd gdrive/My\ Drive/road_finder_data/
```

```
Drive already mounted at /content/gdrive; to attempt to forcibly remount, call drive
[Errno 2] No such file or directory: 'gdrive/My Drive/road_finder_data/'
/content/gdrive/My Drive/road_finder_data
```

```
import cv2
import h5py
import random
import numpy as np
import tensorflow as tf
import keras
import time
from matplotlib import pyplot as plt
from keras import backend as K
from keras.models import Model, load_model
from keras.metrics import MeanIoU
from keras.layers import Input
from keras.layers.core import Lambda
from keras.layers.convolutional import Conv2D, Conv2DTranspose
from keras.layers.pooling import MaxPooling2D
from keras.layers.merge import concatenate
from keras import optimizers
from keras.callbacks import EarlyStopping, ModelCheckpoint, ReduceLROnPlateau
```

#### Pathes to datasets

```
train_file = './big/train.hdf5'
```

#### Constants

```
ACT_FUNCTION = 'relu'
KERNEL_INIT = 'he_normal'
PADDING_TYPE = 'same'
EPOCHS = 100
LEARNING_RATE = 0.0001
BATCH_SIZE = 16
CHECKPOINT_MODEL_PATH = "./Models/road_mapper_2.h5"
```

```
FINAL_MODEL_PATH = "./Models/road_mapper_final_relu_1000"
```

## Loss function

```
def soft_dice_loss(y_true, y_pred, smooth = 1):
    y_true_f = K.flatten(y_true)
    y_pred_f = K.flatten(y_pred)
    intersection = K.sum(y_true_f * y_pred_f)
    return 1 - (2. * intersection + smooth) / (K.sum(y_true_f) + K.sum(y_pred_f) + smooth)
```

## Intersection over Union

```
def iou_coef(y_true, y_pred, smooth = 1):
    I = K.sum(K.abs(y_true * y_pred), axis=[1,2,3])
    U = K.sum(y_true,[1,2,3]) + K.sum(y_pred, [1,2,3]) - I
    iou = K.mean((I + smooth) / (U + smooth), axis=0)
    return iou
```

## ▼ Load Train images

```
hfile = h5py.File(train_file, 'r')

train_images = np.array(hfile.get('images'))[:1000]
print(train_images.shape)

train_masks = np.array(hfile.get('masks'))[:1000]
print(train_masks.shape)

hfile.close()

☞ (1000, 256, 256, 3)
   (1000, 256, 256)
```

## ▼ View samples

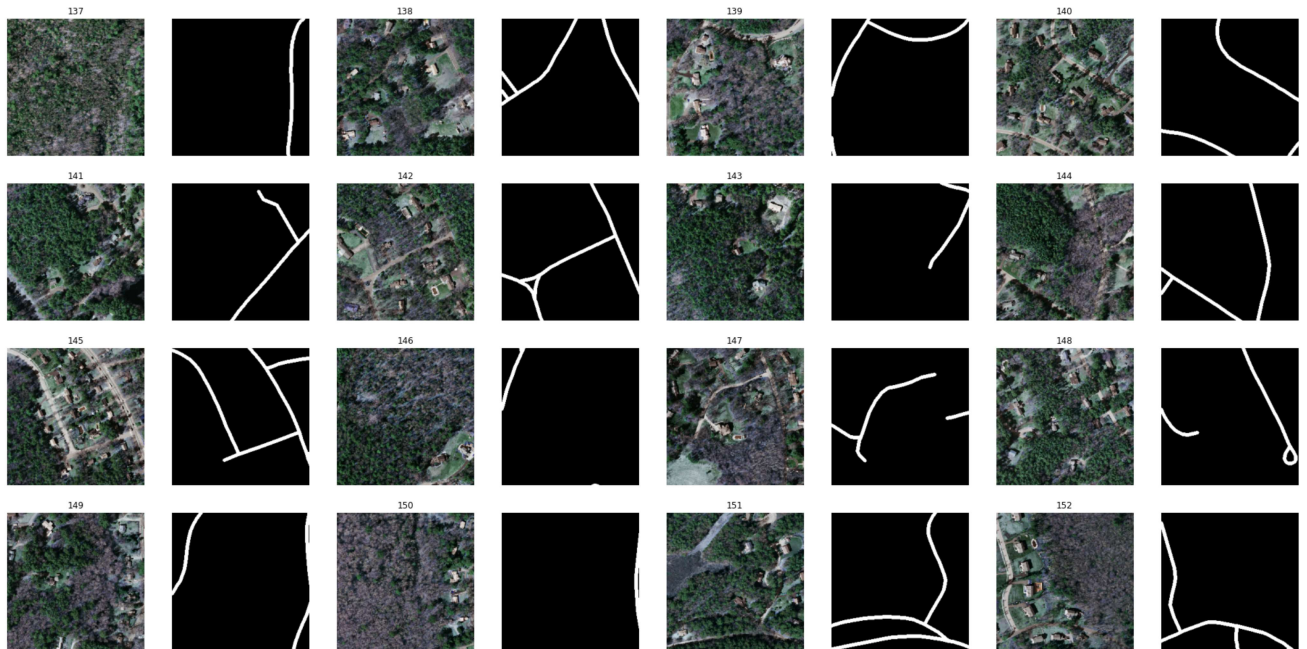
```
random.seed(1)
plt.figure(figsize=(32,16))
x, y = 4, 4
ipos = random.randint(0, len(train_images)-x*y)
for i in range(y):
    for j in range(x):

        pos = ipos + i*x + j

        plt.subplot(y, x*2, i*x*2+j*2+1)
        plt.imshow(train_images[pos])
```

```
plt.title(pos)
plt.axis('off')

plt.subplot(y, x*2, i*x*2+(j*2)+2)
plt.imshow(train_masks[pos], cmap='gray', vmin=0, vmax=255)
plt.axis('off')
plt.show()
```



## ▼ Create Model

```
inputs = Input((256, 256, 3))
s = Lambda(lambda x: x / 255) (inputs)
```

```
conv1 = Conv2D(16, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PAD)
conv1 = Conv2D(16, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PAD)
pooling1 = MaxPooling2D() (conv1)
```

```
conv2 = Conv2D(32, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PAD)
conv2 = Conv2D(32, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PAD)
pooling2 = MaxPooling2D() (conv2)
```

```
conv3 = Conv2D(64, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PAD)
conv3 = Conv2D(64, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PAD)
pooling3 = MaxPooling2D() (conv3)
```

```
conv4 = Conv2D(128, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PA)
conv4 = Conv2D(128, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PA)
pooling4 = MaxPooling2D() (conv4)
```

```

conv5 = Conv2D(256, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PA
conv5 = Conv2D(256, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PA

upsample6 = Conv2DTranspose(128, 2, strides=(2,2), padding=PADDING_TYPE) (conv5)
upsample6 = concatenate([upsample6, conv4])
conv6 = Conv2D(128, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PA
conv6 = Conv2D(128, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PA

upsample7 = Conv2DTranspose(64, 2, strides=(2, 2), padding=PADDING_TYPE) (conv6)
upsample7 = concatenate([upsample7, conv3])
conv7 = Conv2D(64, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PAD
conv7 = Conv2D(64, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PAD

upsample8 = Conv2DTranspose(32, 2, strides=(2, 2), padding=PADDING_TYPE) (conv7)
upsample8 = concatenate([upsample8, conv2])
conv8 = Conv2D(32, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PAD
conv8 = Conv2D(32, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PAD

upsample9 = Conv2DTranspose(16, 2, strides=(2, 2), padding=PADDING_TYPE) (conv8)
upsample9 = concatenate([upsample9, conv1], axis=3)
conv9 = Conv2D(16, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PAD
conv9 = Conv2D(16, 3, activation=ACT_FUNCTION, kernel_initializer=KERNEL_INIT, padding=PAD

outputs = Conv2D(1, 1, activation='sigmoid') (conv9)

model = Model(inputs=[inputs], outputs=[outputs])
model.summary()

```

Model: "model\_2"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_3 (InputLayer)	[(None, 256, 256, 3)]	0	
conv2d_38 (Conv2D)	(None, 256, 256, 16)	448	input_3[0][0]
conv2d_39 (Conv2D)	(None, 256, 256, 16)	2320	conv2d_38[0][0]
max_pooling2d_8 (MaxPooling2D)	(None, 128, 128, 16)	0	conv2d_39[0][0]
conv2d_40 (Conv2D)	(None, 128, 128, 32)	4640	max_pooling2d_8[0]
conv2d_41 (Conv2D)	(None, 128, 128, 32)	9248	conv2d_40[0][0]
max_pooling2d_9 (MaxPooling2D)	(None, 64, 64, 32)	0	conv2d_41[0][0]
conv2d_42 (Conv2D)	(None, 64, 64, 64)	18496	max_pooling2d_9[0]
conv2d_43 (Conv2D)	(None, 64, 64, 64)	36928	conv2d_42[0][0]
max_pooling2d_10 (MaxPooling2D)	(None, 32, 32, 64)	0	conv2d_43[0][0]
conv2d_44 (Conv2D)	(None, 32, 32, 128)	73856	max_pooling2d_10[0]
conv2d_45 (Conv2D)	(None, 32, 32, 128)	147584	conv2d_44[0][0]
max_pooling2d_11 (MaxPooling2D)	(None, 16, 16, 128)	0	conv2d_45[0][0]

conv2d_46 (Conv2D)	(None, 16, 16, 256)	295168	max_pooling2d_11[0][0]
conv2d_47 (Conv2D)	(None, 16, 16, 256)	590080	conv2d_46[0][0]
conv2d_transpose_8 (Conv2DTrans	(None, 32, 32, 128)	131200	conv2d_47[0][0]
concatenate_8 (Concatenate)	(None, 32, 32, 256)	0	conv2d_transpose_8[0][0]
conv2d_48 (Conv2D)	(None, 32, 32, 128)	295040	concatenate_8[0][0]
conv2d_49 (Conv2D)	(None, 32, 32, 128)	147584	conv2d_48[0][0]
conv2d_transpose_9 (Conv2DTrans	(None, 64, 64, 64)	32832	conv2d_49[0][0]
concatenate_9 (Concatenate)	(None, 64, 64, 128)	0	conv2d_transpose_9[0][0]
conv2d_50 (Conv2D)	(None, 64, 64, 64)	73792	concatenate_9[0][0]
conv2d_51 (Conv2D)	(None, 64, 64, 64)	36928	conv2d_50[0][0]
conv2d_transpose_10 (Conv2DTran	(None, 128, 128, 32)	8224	conv2d_51[0][0]
concatenate_10 (Concatenate)	(None, 128, 128, 64)	0	conv2d_transpose_10[0][0]
conv2d_52 (Conv2D)	(None, 128, 128, 32)	18464	concatenate_10[0][0]

## ▼ Compile model

```

train_masks = np.expand_dims(train_masks, -1)
train_masks.shape

(1000, 256, 256, 1)

opt = keras.optimizers.Adam(LEARNING_RATE)
model.compile(
    optimizer=opt,
    loss=soft_dice_loss,
    metrics=[iou_coef]) #MeanIoU(num_classes=2)

tf.test.gpu_device_name()

'/device:GPU:0'

start_time = time.time()

history = model.fit(train_images,
                    train_masks/255,
                    validation_split = 0.1,
                    epochs=EPOCHS,

```

```

batch_size = BATCH_SIZE,
callbacks = [
    ModelCheckpoint(CHECKPOINT_MODEL_PATH,
                    monitor="val_loss",
                    mode="min",
                    save_best_only = True,
                    verbose=1),
    EarlyStopping(monitor = 'val_loss',
                  min_delta = 0,
                  patience = 5,
                  verbose = 1,
                  restore_best_weights = True),
    ReduceLROnPlateau(monitor='val_loss',
                      factor=0.1,
                      patience=4,
                      verbose=1,
                      min_delta=1e-4)
]
)

```

Epoch 1/100

57/57 [=====] - 10s 154ms/step - loss: 0.9566 - iou\_coef: 0

Epoch 00001: val\_loss improved from inf to 0.89923, saving model to ./Models/road\_map

Epoch 2/100

57/57 [=====] - 7s 121ms/step - loss: 0.8837 - iou\_coef: 0.6

Epoch 00002: val\_loss improved from 0.89923 to 0.89732, saving model to ./Models/road\_map

Epoch 3/100

57/57 [=====] - 7s 121ms/step - loss: 0.8828 - iou\_coef: 0.6

Epoch 00003: val\_loss improved from 0.89732 to 0.89729, saving model to ./Models/road\_map

Epoch 4/100

57/57 [=====] - 7s 121ms/step - loss: 0.8832 - iou\_coef: 0.6

Epoch 00004: val\_loss did not improve from 0.89729

Epoch 5/100

57/57 [=====] - 7s 122ms/step - loss: 0.8807 - iou\_coef: 0.6

Epoch 00005: val\_loss did not improve from 0.89729

Epoch 6/100

57/57 [=====] - 7s 122ms/step - loss: 0.8832 - iou\_coef: 0.6

Epoch 00006: val\_loss did not improve from 0.89729

Epoch 00006: ReduceLROnPlateau reducing learning rate to 9.999999747378752e-06.

Epoch 7/100

57/57 [=====] - 7s 123ms/step - loss: 0.8846 - iou\_coef: 0.6

Epoch 00007: val\_loss did not improve from 0.89729

Epoch 8/100

57/57 [=====] - 7s 123ms/step - loss: 0.8814 - iou\_coef: 0.6

Epoch 00008: val\_loss did not improve from 0.89729

Restoring model weights from the end of the best epoch.

Epoch 00008: early stopping

```
end_time = time.time()  
total_time = end_time - start_time  
print("Total training time: {}".format(total_time))
```

```
Total training time: 61.07701826095581s
```

```
model.save(FINAL_MODEL_PATH)
```

```
INFO:tensorflow:Assets written to: ./Models/road_mapper_final_relu_1000/assets
```

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