CamVidSegmenAttentionUNet

February 25, 2023

```
[2]: # Common
     import os
     import keras
     import pandas as pd
     import numpy as np
     from glob import glob
     from tqdm import tqdm
     import tensorflow as tf
     # Data
     import tensorflow.image as tfi
     from tensorflow.keras.utils import Sequence
     from tensorflow.keras.preprocessing.image import load_img, img_to_array
     # Data Viz
     import plotly.express as px
     import matplotlib.pyplot as plt
     # Model
     from keras.models import Model
     from keras.layers import Input
     from keras.layers import Layer
     from keras.layers import Conv2D
     from keras.layers import UpSampling2D
     from keras.layers import BatchNormalization
     from keras.layers import Dropout
     from keras.layers import Add
     from keras.layers import Multiply
     from keras.layers import MaxPool2D
     from keras.layers import concatenate as c_
     from keras.layers import ReLU
     # Model Learning Viz
     from tf_explain.core.grad_cam import GradCAM
     # Callbacks
```

```
from keras.callbacks import Callback
from keras.callbacks import EarlyStopping
from keras.callbacks import ModelCheckpoint
```

```
[3]: def show_image(image, title=None, cmap=None, alpha=1):
    plt.imshow(image, cmap=cmap, alpha=alpha)
    if title is not None:
        plt.title(title)
    plt.axis('off')

def load_image(path, SIZE):
    img = load_img(path)
    img = img_to_array(img)
    img = tfi.resize(img, (SIZE, SIZE))
    img = tf.cast(img, tf.float32)
    img = img/255.
    return img
```

```
train_images = np.zeros(shape=(len(train_image_paths), SIZE, SIZE, 3))
train_masks = np.zeros(shape=(len(train_image_paths), SIZE, SIZE, 3))

val_images = np.zeros(shape=(len(val_image_paths), SIZE, SIZE, 3))
val_masks = np.zeros(shape=(len(val_image_paths), SIZE, SIZE, 3))

test_images = np.zeros(shape=(len(test_image_paths), SIZE, SIZE, 3))

test_masks = np.zeros(shape=(len(test_image_paths), SIZE, SIZE, 3))

for i,path in tqdm(enumerate(train_image_paths), desc="Train Images"):
    image = load_image(path, SIZE=SIZE)
    train_images[i] = image

for i,path in tqdm(enumerate(train_mask_paths), desc="Train Masks"):
    image = load_image(path, SIZE=SIZE)
    train_masks[i] = image
```

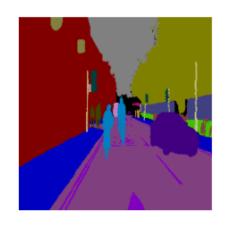
```
for i,path in tqdm(enumerate(val_image_paths), desc="Valid Masks"):
         image = load_image(path, SIZE=SIZE)
         val_images[i] = image
     for i,path in tqdm(enumerate(val_mask_paths), desc="Valid Masks"):
         image = load_image(path, SIZE=SIZE)
         val_masks[i] = image
     for i,path in tqdm(enumerate(test_image_paths), desc="Test Masks"):
         image = load image(path, SIZE=SIZE)
         test_images[i] = image
     for i,path in tqdm(enumerate(test_mask_paths), desc="Test Masks"):
         image = load_image(path, SIZE=SIZE)
         test_masks[i] = image
    Train Images: Oit [00:00, ?it/s]2023-02-25 02:58:29.436047: E
    tensorflow/compiler/xla/stream executor/cuda/cuda driver.cc:267] failed call to
    cuInit: CUDA_ERROR_NO_DEVICE: no CUDA-capable device is detected
    2023-02-25 02:58:29.436088: I
    tensorflow/compiler/xla/stream executor/cuda/cuda diagnostics.cc:156] kernel
    driver does not appear to be running on this host (picox):
    /proc/driver/nvidia/version does not exist
    2023-02-25 02:58:29.493821: I tensorflow/core/platform/cpu_feature_guard.cc:193]
    This TensorFlow binary is optimized with oneAPI Deep Neural Network Library
    (oneDNN) to use the following CPU instructions in performance-critical
    operations: AVX2 FMA
    To enable them in other operations, rebuild TensorFlow with the appropriate
    compiler flags.
    Train Images: 369it [00:13, 27.18it/s]
    Train Masks: 369it [00:04, 81.20it/s]
    Valid Masks: 100it [00:03, 27.42it/s]
    Valid Masks: 100it [00:01, 75.77it/s]
    Test Masks: 232it [00:08, 27.29it/s]
    Test Masks: 232it [00:03, 69.19it/s]
[6]: plt.figure(figsize=(8,12))
     for i in range(8):
         plt.subplot(4,2,i+1)
         if (i+1)\%2!=0:
             rand id = np.random.randint(len(train image paths))
             rand_image = train_images[rand_id]
             rand_mask = train_masks[rand_id]
             show_image(rand_image)
         elif (i+1)\%2==0:
             show_image(rand_mask)
```

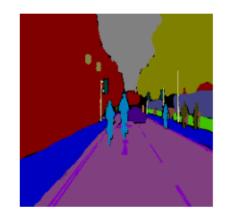


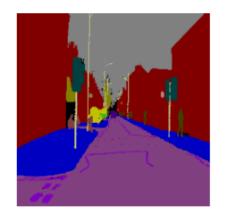


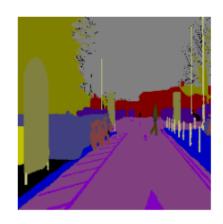












```
[7]: # Contraction Block
     class EncoderBlock(Layer):
         def __init__(self, filters, rate, pooling=True,**kwargs):
             super(EncoderBlock, self).__init__(**kwargs)
             self.filters = filters
             self.rate = rate
             self.pooling = pooling
             self.c1 = Conv2D(filters, kernel_size=3, strides=1, padding='same',
      →activation='relu', kernel_initializer='he_normal')
             self.c2 = Conv2D(filters, kernel_size=3, strides=1, padding='same',
      →activation='relu', kernel_initializer='he_normal')
             self.drop = Dropout(rate)
             self.pool = MaxPool2D()
             self.bn = BatchNormalization()
         def call(self, X):
             x = self.c1(X)
             x = self.drop(x)
             x = self.c2(x)
             if self.pooling:
                 y = self.pool(x)
                 return y, x
             else:
                 return x
         def get_config(self):
             base_config = super().get_config()
             return base_config.update({
                 "filters":self.filters,
                 "rate":self.rate,
                 "pooling":self.pooling
             })
[8]: class DecoderBlock(Layer):
         def __init__(self, filters, rate,**kwargs):
             super(DecoderBlock, self).__init__(**kwargs)
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self.filters = filters

self.rate = rate

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self.up = UpSampling2D()
self.net = EncoderBlock(filters, rate, pooling=False)
self.bn = BatchNormalization()

def call(self, X):
    org_x, att_x = X
    x = self.up(org_x)
    x = c_([x, att_x])
    x = self.net(x)
    return x

def get_config(self):
    base_config = super().get_config()
    return base_config.update({
        "filters":self.filters,
        "rate":self.rate
})
```

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[9]: class AttentionGate(Layer):
         def __init__(self, filters, rate, **kwargs):
             super(AttentionGate, self).__init__(**kwargs)
             self.filters = filters
             self.rate = rate
             self.down = Conv2D(filters, kernel_size=3, strides=2, padding='same')
             self.normal = Conv2D(filters, kernel_size=3, strides=1, padding='same')
             self.add = Add()
             self.act = ReLU()
             self.learn = Conv2D(1,kernel_size=1,padding='same',__
      ⇔activation='sigmoid')
             self.resample = UpSampling2D()
             self.apply = Multiply()
             self.drop = Dropout(rate)
         def call(self, X):
             net_X, skip_X = X
             g = self.normal(net_X)
             h = self.drop(self.down(skip_X)) # this will force it to learn
             added = self.add([g,h])
             x = self.act(added)
             x = self.learn(x)
             x = self.resample(x)
```

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f = self.apply([x, skip_X])
    return f

def get_config(self):
    base_estimator = super().get_config()
    return base_estimator.update({
        "filters":self.filters,
        "rate":self.rate
    })
# Input
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[10]: # Input
      input_layer = Input(shape=(SIZE, SIZE, 3), name="Input_layer")
      # Encoder
      p1, c1 = EncoderBlock(32,0.1, name="Encoder1")(input_layer)
      p2, c2 = EncoderBlock(64,0.1, name="Encoder2")(p1)
      p3, c3 = EncoderBlock(128,0.2, name="Encoder3")(p2)
      p4, c4 = EncoderBlock(256,0.2, name="Encoder4")(p3)
      # Encodings
      encoding = EncoderBlock(512,0.3, pooling=False, name="Encoding")(p4)
      # Attention + Decoder
      a1 = AttentionGate(256,0, name="Attention1")([encoding, c4])
      d1 = DecoderBlock(256,0.2, name="Decoder1")([encoding, a1])
      a2 = AttentionGate(128,0, name="Attention2")([d1, c3])
      d2 = DecoderBlock(128,0.2, name="Decoder2")([d1, a2])
      a3 = AttentionGate(64,0, name="Attention3")([d2, c2])
      d3 = DecoderBlock(64,0.1, name="Decoder3")([d2, a3])
      a4 = AttentionGate(32,0, name="Attention4")([d3, c1])
      d4 = DecoderBlock(32,0.1, name="Decoder4")([d3, a4])
      # Outputs
      output_layer = Conv2D(3, kernel_size=1, strides=1, padding='same',_
       →activation='sigmoid', name="Output_layer")(d4)
      # Model
      att_unet = Model( inputs=[input_layer], outputs=[output_layer],)
```

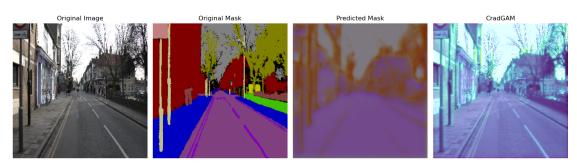
```
[11]: class ShowProgress(Callback):
    def on_epoch_end(self, epoch, logs=None):
        i = np.random.randint(len(val_images))

    image, mask = val_images[i], val_masks[i]
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pred_mask = self.model.predict(image[np.newaxis,...])[0]
              exp = GradCAM()
              cam = exp.explain(
                  validation_data=(image[np.newaxis,...], mask[np.newaxis,...]),
                  model=self.model,
                  class_index=None,
                  layer_name='Attention4'
              )
              plt.figure(figsize=(15,8))
              plt.subplot(1,4,1)
              show_image(image, title="Original Image")
              plt.subplot(1,4,2)
              show_image(mask, title="Original Mask")
              plt.subplot(1,4,3)
              show_image(pred_mask, title="Predicted Mask")
              plt.subplot(1,4,4)
              show_image(cam, title="CradGAM")
              plt.tight_layout()
              plt.show()
[12]: class_path = 'CamVid/class_dict.csv'
      class_dict = pd.read_csv(class_path)
      n_classes = len(class_dict['name'])
[13]: # Compiling
      att unet.compile(
          loss='binary_crossentropy',
          optimizer='adam',
          metrics=['accuracy', keras.metrics.MeanIoU(num_classes=n_classes,_

¬name="IoU")]
 []: # training
      nepochs=25
      nbatch_size=8
      results = att_unet.fit(
          train_images, train_masks,
          validation_data=(val_images, val_masks),
          epochs=nepochs,
          callbacks=[ EarlyStopping(patience=3, monitor='IoU', mode='max',__
       →restore_best_weights=True), ShowProgress()],
          batch size=nbatch size
```

)



```
0.6862 - IoU: 0.4996 - val_loss: 0.6199 - val_accuracy: 0.4110 - val_IoU: 1.0000
   Epoch 2/5
   19/93 [====>...] - ETA: 4:55 - loss: 0.6015 - accuracy:
   0.6323 - IoU: 1.0000
[]: plt.figure(figsize=(10,24))
    n = 0
    for i in range(1,(10*3)+1):
       plt.subplot(10,3,i)
       if n == 0:
           id = np.random.randint(len(test_images))
           image = test_images[id][np.newaxis,...]
           mask = test_masks[id]
           pred_mask = att_unet.predict(image)[0]
           image = image[0]
           show_image(image)
           n+=1
       elif n==1:
           show_image(pred_mask)
           n+=1
       elif n==2:
           show_image(mask)
           n=0
    plt.tight_layout()
    plt.show()
```

[]: