nb0a7xpxr

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[]: import os
     import numpy as np
     import cv2
     from glob import glob
     import tensorflow as tf
     from sklearn.model_selection import train_test_split
     import imgaug.augmenters as iaa
     from tensorflow.keras.preprocessing.image import ImageDataGenerator
     import skimage.io as io
     import skimage.transform as trans
     import tensorflow.keras.layers as layers
     from keras.models import *
     from keras.optimizers import *
     from keras.callbacks import ModelCheckpoint, LearningRateScheduler
     from keras import backend as keras
     from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint,
      →ReduceLROnPlateau, CSVLogger, TensorBoard
     from tensorflow.keras.metrics import Recall, Precision
     from tqdm import tqdm
     IMG_W = 512
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[]: IMG_H = 512
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[]: def load_dataset(path, split=0.1):
         """ Loading the images and masks """
         X = sorted(glob(os.path.join(path, "images", "*")))
         Y = sorted(glob(os.path.join(path, "masks", "*")))
         """ Spliting the data into training and testing """
         split_size = int(len(X) * split)
         train_x, valid_x = train_test_split(X, test_size=split_size,_
      →random state=42)
         train_y, valid_y = train_test_split(Y, test_size=split_size,__
      →random_state=42)
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train_y, test_y = train_test_split(train_y, test_size=split_size,_
      →random state=42)
         return (train_x, train_y), (valid_x, valid_y), (test_x, test_y)
     def read_image(path):
         path = path.decode()
         image = cv2.imread(path, cv2.IMREAD_COLOR)
         image = cv2.resize(image, (IMG_W, IMG_H))
         image = image / 255.0
         image = image.astype(np.float32)
         return image
     def read_mask(path):
         path = path.decode()
         mask = cv2.imread(path, cv2.IMREAD_GRAYSCALE)
         mask = cv2.resize(mask, (IMG_W, IMG_H))
         mask = mask / 255.0
         mask = mask.astype(np.float32)
         mask = np.expand_dims(mask, axis=-1)
         return mask
     def tf_parse(x, y):
         def _parse(x, y):
            x = read image(x)
             y = read_mask(y)
             return x, y
         x, y = tf.numpy_function(_parse, [x, y], [tf.float32, tf.float32])
         x.set_shape([IMG_H, IMG_W, 3])
         y.set_shape([IMG_H, IMG_W, 1])
         return x, y
     def tf_dataset(X, Y, batch=2):
         ds = tf.data.Dataset.from_tensor_slices((X, Y))
         ds = ds.map(tf_parse).batch(batch).prefetch(10)
         return ds
[]: def conv_block(x, kernelsize, filters, dropout, batchnorm=False):
         conv = layers.Conv2D(filters, (kernelsize, kernelsize),
      ⇔kernel_initializer='he_normal', padding="same")(x)
         if batchnorm is True:
             conv = layers.BatchNormalization(axis=3)(conv)
         conv = layers.Activation("relu")(conv)
         if dropout > 0:
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train_x, test_x = train_test_split(train_x, test_size=split_size,_

→random_state=42)

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conv = layers.Dropout(dropout)(conv)
    conv = layers.Conv2D(filters, (kernelsize, kernelsize),
 ⇔kernel_initializer='he_normal', padding="same")(conv)
    if batchnorm is True:
        conv = layers.BatchNormalization(axis=3)(conv)
   conv = layers.Activation("relu")(conv)
   return conv
def encoder_block(x, num_filters, dropout, batchnorm):
   x = conv_block(x, kernelsize=3, filters=num_filters, dropout=dropout,_
 ⇒batchnorm=batchnorm)
   x = conv block(x, kernelsize=3, filters=num filters, dropout=dropout,
 ⇒batchnorm=batchnorm)
   p = layers.MaxPool2D((2, 2))(x)
   return x, p
def unet3plus(input_shape, num_classes=1, dropout=0.3, batchnorm=True):
    """ Inputs """
    inputs = layers.Input(input_shape, name="input_layer")
    """ Encoder """
   e1, p1 = encoder block(inputs, 64, dropout=dropout, batchnorm=batchnorm)
   e2, p2 = encoder_block(p1, 128, dropout=dropout, batchnorm=batchnorm)
   e3, p3 = encoder_block(p2, 256, dropout=dropout, batchnorm=batchnorm)
   e4, p4 = encoder_block(p3, 512, dropout=dropout, batchnorm=batchnorm)
    """ Bottleneck """
   e5 = conv_block(p4, kernelsize=3, filters=1024, dropout=dropout,_
 ⇔batchnorm=batchnorm)
    e5 = conv_block(e5, kernelsize=3, filters=1024, dropout=dropout,__
 ⇒batchnorm=batchnorm)
    """ Decoder 4 """
   e1_d4 = layers.MaxPool2D((8, 8))(e1)
    e1_d4 = conv_block(e1_d4, kernelsize=3, filters=64, dropout=dropout,_
 ⇔batchnorm=batchnorm)
   e2_d4 = layers.MaxPool2D((4, 4))(e2)
   e2_d4 = conv_block(e2_d4, kernelsize=3, filters=64, dropout=dropout,_
 ⇒batchnorm=batchnorm)
   e3_d4 = layers.MaxPool2D((2, 2))(e3)
    e3_d4 = conv_block(e3_d4, kernelsize=3, filters=64, dropout=dropout,_
 ⇒batchnorm=batchnorm)
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e4_d4 = conv_block(e4, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
  e5_d4 = layers.UpSampling2D((2, 2), interpolation="bilinear")(e5)
  e5_d4 = conv_block(e5_d4, kernelsize=3, filters=64, dropout=dropout,_
⇔batchnorm=batchnorm)
  d4 = layers.Concatenate()([e1_d4, e2_d4, e3_d4, e4_d4, e5_d4])
  d4 = conv_block(d4, kernelsize=3, filters=64*5, dropout=dropout,_
⇒batchnorm=batchnorm)
   """ Decoder 3 """
  e1_d3 = layers.MaxPool2D((4, 4))(e1)
  e1_d3 = conv_block(e1_d3, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
  e2_d3 = layers.MaxPool2D((2, 2))(e2)
  e2_d3 = conv_block(e2_d3, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
  e3_d3 = conv_block(e3, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
  d4_d3 = layers.UpSampling2D((2, 2), interpolation="bilinear")(d4)
  d4_d3 = conv_block(d4_d3, kernelsize=3, filters=64, dropout=dropout,__
⇒batchnorm=batchnorm)
  e5_d3 = layers.UpSampling2D((4, 4), interpolation="bilinear")(e5)
  e5_d3 = conv_block(e5_d3, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
  d3 = layers.Concatenate()([e1 d3, e2 d3, e3 d3, d4 d3, e5 d3])
  d3 = conv_block(d3, kernelsize=3, filters=64*5, dropout=dropout,__
⇒batchnorm=batchnorm)
  """ Decoder 2 """
  e1_d2 = layers.MaxPool2D((2, 2))(e1)
  e1 d2 = conv block(e1 d2, kernelsize=3, filters=64, dropout=dropout,
⇒batchnorm=batchnorm)
  e2_d2 = conv_block(e2, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
  d3_d2 = layers.UpSampling2D((2, 2), interpolation="bilinear")(d3)
  d3_d2 = conv_block(d3_d2, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
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d4_d2 = layers.UpSampling2D((4, 4), interpolation="bilinear")(d4)
  d4_d2 = conv_block(d4_d2, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
  e5 d2 = layers.UpSampling2D((8, 8), interpolation="bilinear")(e5)
  e5_d2 = conv_block(e5_d2, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
  d2 = layers.Concatenate()([e1_d2, e2_d2, d3_d2, d4_d2, e5_d2])
  d2 = conv_block(d2, kernelsize=3, filters=64*5, dropout=dropout,__
⇒batchnorm=batchnorm)
  """ Decoder 1 """
  e1_d1 = conv_block(e1, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
  d2_d1 = layers.UpSampling2D((2, 2), interpolation="bilinear")(d2)
  d2_d1 = conv_block(d2_d1, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
  d3_d1 = layers.UpSampling2D((4, 4), interpolation="bilinear")(d3)
  d3_d1 = conv_block(d3_d1, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
  d4_d1 = layers.UpSampling2D((8, 8), interpolation="bilinear")(d4)
  d4_d1 = conv_block(d4_d1, kernelsize=3, filters=64, dropout=dropout,_
⇒batchnorm=batchnorm)
  e5_d1 = layers.UpSampling2D((16, 16), interpolation="bilinear")(e5)
  e5_d1 = conv_block(e5_d1, kernelsize=3, filters=64, dropout=dropout,__
⇒batchnorm=batchnorm)
  d1 = layers.Concatenate()([e1_d1, d2_d1, d3_d1, d4_d1, e5_d1])
  d1 = conv_block(d1, kernelsize=3, filters=64*5, dropout=dropout,__
⇒batchnorm=batchnorm)
  """ Final Output """
  # No deep supervision, just a single output
  y1 = layers.Conv2D(num_classes, kernel_size=1, padding="same")(d1)
  y1 = layers.Activation("sigmoid")(y1)
  outputs = [y1]
  model = tf.keras.Model(inputs, outputs)
  return model
```

```
if __name__ == "__main__":
   input_shape = (256, 256, 3)
   model = unet3plus(input_shape)
   model.summary()

smooth = 1e-15
```

```
def dice_coef(y_true, y_pred):
    intersection = tf.reduce_sum(y_true * y_pred)
    return (2.0 * intersection + smooth) / (tf.reduce_sum(y_true) + tf.
    reduce_sum(y_pred) + smooth)

def dice_loss(y_true, y_pred):
    return 1.0 - dice_coef(y_true, y_pred)
```

```
[]: def create_dir(path):
         if not os.path.exists(path):
             os.makedirs(path)
     if __name__ == "__main__":
         """ Seeding """
         np.random.seed(42)
         tf.random.set_seed(42)
         """ Directory for storing files """
         create_dir("files")
         """ Hyperparameters """
         batch_size = 2
         lr = 1e-4
         num_epochs = 50
         model_path = os.path.join("files", "model.keras")
         csv_path = os.path.join("files", "log.csv")
         """ Dataset """
         dataset path = "Kvasir-SEG"
         (train_x, train_y), (valid_x, valid_y), (test_x, test_y) = ___
      →load_dataset(dataset_path)
         print(f"Train: \t{len(train_x)} - {len(train_y)}")
         print(f"Valid: \t{len(valid_x)} - {len(valid_y)}")
         print(f"Test: \t{len(test_x)} - {len(test_y)}")
         train_dataset = tf_dataset(train_x, train_y, batch=batch_size)
         valid_dataset = tf_dataset(valid_x, valid_y, batch=batch_size)
```

```
""" Model """
  model = unet3plus((IMG_H, IMG_W, 3))
  metrics = ["acc", tf.keras.metrics.Recall(), tf.keras.metrics.Precision(),

dice_coef]

  model.compile(loss=dice loss, optimizer=Adam(lr), metrics=metrics)
  callbacks = [
      ModelCheckpoint(model_path, verbose=1, save_best_only=True),
      ReduceLROnPlateau(monitor='val_loss', factor=0.1, patience=2,__
→min_lr=1e-10, verbose=1),
      CSVLogger(csv path),
      TensorBoard(),
      EarlyStopping(monitor='val_loss', patience=5, restore_best_weights=True)
  1
  model.fit(
      train_dataset,
      validation_data=valid_dataset,
      epochs=num_epochs,
      steps_per_epoch=1000,
      callbacks=callbacks
  )
```

```
[]: def read image(path):
         x = cv2.imread(path, cv2.IMREAD_COLOR)
         x = cv2.resize(x, (IMG W, IMG H))
         x = x/255.0
         return x
     def read_mask(path):
        x = cv2.imread(y, cv2.IMREAD_GRAYSCALE)
         x = cv2.resize(x, (IMG_W, IMG_H))
        x = x / 255.0
         x = np.expand_dims(x, axis=-1)
         x = np.concatenate([x, x, x], axis=-1)
         return x
     if __name__ == "__main__":
        """ Seeding """
         np.random.seed(42)
         tf.random.set_seed(42)
         """ Directory for storing files """
         create_dir(f"results")
         """ Load the model """
```

```
model_path = os.path.join("files", "model.keras")
  model = tf.keras.models.load_model(model_path, custom_objects={"dice_loss":_

dice_loss, "dice_coef": dice_coef})
  """ Dataset """
  dataset path = "/content/drive/MyDrive/kvasir-seg/Kvasir-SEG"
  (train_x, train_y), (valid_x, valid_y), (test_x, test_y) = ___
→load_dataset(dataset_path)
  print(f"Train: \t{len(train_x)} - {len(train_y)}")
  print(f"Valid: \t{len(valid_x)} - {len(valid_y)}")
  print(f"Test: \t{len(test_x)} - {len(test_y)}")
  """ Prediction """
  for x, y in tqdm(zip(test_x, test_y), total=len(test_x)):
      """ Extracting the name """
      name = x.split("/")[-1].split(".")[0]
      """ Reading the image """
      x = read_image(x)
      """ Reading the mask """
      y = read_mask(y)
      """ Prediction """
      x = np.expand_dims(x, axis=0)
      pred = model.predict(x, verbose=0)[0]
      pred = np.concatenate([pred, pred, pred], axis=-1)
      \# pred = (pred > 0.5).astype(np.int32)
      """ Save final mask """
      line = np.ones((IMG_H, 10, 3)) * 255
      cat_images = np.concatenate([x[0], line, y*255, line, pred*255], axis=1)
      save_image_path = os.path.join("results", f"{name}.jpg")
      cv2.imwrite(save_image_path, cat_images)
```