

Untitled3

July 27, 2024

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[44]: import os
import numpy as np
import cv2
from glob import glob
import tensorflow as tf
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split

from tensorflow.keras.layers import Conv2D, Activation, BatchNormalization,
    ↳UpSampling2D, Input, Concatenate, Cropping2D
from tensorflow.keras.models import Model
from tensorflow.keras.applications import MobileNetV2
from tensorflow.keras.callbacks import EarlyStopping, ReduceLROnPlateau
from tensorflow.keras import backend as K

# Set NumPy random seed for reproducibility
np.random.seed(42)

# Set TensorFlow random seed for reproducibility
tf.random.set_seed(42)

# Define image size for resizing input images and masks
WIDTH = 672
HEIGHT = 504

# Define number of epochs for training
EPOCHS = 5

# Define batch size
BATCH = 4

# Define learning rate for the optimizer
LR = 1e-4

# Define output path for saving training results and model
PATH = "/root/61541v001/V-03"
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# Define label colors
colors = [
    (0, 0, 0), (128, 0, 0), (0, 128, 0), (128, 128, 0), (0, 0, 128),
    (128, 0, 128), (0, 128, 128), (128, 128, 128), (64, 0, 0), (192, 0, 0),
    (64, 128, 0), (192, 128, 0), (64, 0, 128), (192, 0, 128), (64, 128, 128),
    (192, 128, 128), (0, 64, 0), (128, 64, 0), (0, 192, 0), (128, 192, 0),
    (0, 64, 128), (128, 64, 12)
]
NUM_CLASSES = len(colors)

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[45]: # Load dataset
def load_data(path, split=0.1):
    images = sorted(glob(os.path.join(PATH, "image/*")))
    masks = sorted(glob(os.path.join(PATH, "label/*")))
    total_size = len(images)
    valid_size = int(split * total_size)
    test_size = int(split * total_size)

    train_x, valid_x = train_test_split(images, test_size=valid_size,
    ↪random_state=42)
    train_y, valid_y = train_test_split(masks, test_size=valid_size,
    ↪random_state=42)
    train_x, test_x = train_test_split(train_x, test_size=test_size,
    ↪random_state=42)
    train_y, test_y = train_test_split(train_y, test_size=test_size,
    ↪random_state=42)

    return (train_x, train_y), (valid_x, valid_y), (test_x, test_y)

# Read and preprocess image
def read_image(path):
    if isinstance(path, bytes):
        path = path.decode()
    x = cv2.imread(path, cv2.IMREAD_COLOR)
    x = cv2.cvtColor(x, cv2.COLOR_BGR2RGB)
    x = cv2.resize(x, (WIDTH, HEIGHT))
    x = x / 255.0
    return x.astype(np.float32)

# Read and preprocess mask
def read_mask(path):
    if isinstance(path, bytes):
        path = path.decode()
    x = cv2.imread(path, cv2.IMREAD_GRAYSCALE)
    x = cv2.resize(x, (WIDTH, HEIGHT))
    x = x / 255.0
    x = np.round(x * (NUM_CLASSES - 1)).astype(np.uint8)

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x = tf.keras.utils.to_categorical(x, num_classes=NUM_CLASSES)
return x.astype(np.float32)

# Parse function for TensorFlow dataset
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([HEIGHT, WIDTH, 3])
    y.set_shape([HEIGHT, WIDTH, NUM_CLASSES])
    return x, y

# Create TensorFlow dataset
def tf_dataset(x, y, batch=8):
    dataset = tf.data.Dataset.from_tensor_slices((x, y))
    dataset = dataset.map(tf_parse)
    dataset = dataset.batch(batch)
    dataset = dataset.repeat()
    return dataset

# Parse mask for visualization
def mask_parse(mask):
    mask = np.argmax(mask, axis=-1) # Convert one-hot encoded mask to
    ↪single-channel class mask
    rgb_mask = np.zeros((mask.shape[0], mask.shape[1], 3), dtype=np.uint8) #
    ↪Create an empty color mask
    for i, color in enumerate(colors):
        rgb_mask[mask == i] = color # Map class values to predefined colors
    return rgb_mask

# Build the model
def build_model():
    inputs = Input(shape=(HEIGHT, WIDTH, 3), name="input_image")
    encoder = MobileNetV2(input_tensor=inputs, weights="imagenet",
    ↪include_top=False, alpha=0.35)
    skip_connection_names = ["input_image", "block_1_expand_relu",
    ↪"block_3_expand_relu", "block_6_expand_relu"]
    encoder_output = encoder.get_layer("block_13_expand_relu").output
    f = [16, 32, 48, 64]
    x = encoder_output

    for i in range(1, len(skip_connection_names) + 1, 1):
        x_skip = encoder.get_layer(skip_connection_names[-i]).output
        x = UpSampling2D((2, 2))(x)

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height_diff = K.int_shape(x)[1] - K.int_shape(x_skip)[1]
width_diff = K.int_shape(x)[2] - K.int_shape(x_skip)[2]
if height_diff != 0 or width_diff != 0:
    x = Cropping2D(((height_diff // 2, height_diff - height_diff // 2),
                    (width_diff // 2, width_diff - width_diff // 2)))(x)
x = Concatenate()([x, x_skip])
x = Conv2D(f[-i], (3, 3), padding="same")(x)
x = BatchNormalization()(x)
x = Activation("relu")(x)
x = Conv2D(f[-i], (3, 3), padding="same")(x)
x = BatchNormalization()(x)
x = Activation("relu")(x)

x = Conv2D(NUM_CLASSES, (1, 1), padding="same")(x)
x = Activation("softmax")(x)
model = Model(inputs, x)
return model

```

```

[46]: # Load dataset
(train_x, train_y), (valid_x, valid_y), (test_x, test_y) = load_data(PATH)
print("Training data: ", len(train_x))
print("Validation data: ", len(valid_x))
print("Testing data: ", len(test_x))

```

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Training data: 1477
Validation data: 184
Testing data: 184

```

```

[47]: # Define custom metrics
def dice_coefficient(y_true, y_pred):
    y_true_f = tf.keras.backend.flatten(y_true)
    y_pred_f = tf.keras.backend.flatten(y_pred)
    intersection = tf.keras.backend.sum(y_true_f * y_pred_f)
    return (2. * intersection) / (tf.keras.backend.sum(y_true_f) + tf.keras.
    ↪ backend.sum(y_pred_f))

def iou(y_true, y_pred):
    y_true_f = tf.keras.backend.flatten(y_true)
    y_pred_f = tf.keras.backend.flatten(y_pred)
    intersection = tf.keras.backend.sum(y_true_f * y_pred_f)
    union = tf.keras.backend.sum(y_true_f) + tf.keras.backend.sum(y_pred_f) - ↵
    ↪ intersection
    return intersection / union

```

```

[48]: # Create and compile the model
model = build_model()
model.summary()

```

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loss = tf.keras.losses.CategoricalCrossentropy()
opt = tf.keras.optimizers.Nadam(LR)
metrics = ['accuracy']
model.compile(loss=loss, optimizer=opt, metrics=['accuracy', dice_coefficient,
↪iou])

```

/tmp/ipykernel_5823/2080664991.py:68: UserWarning: `input_shape` is undefined or non-square, or `rows` is not in [96, 128, 160, 192, 224]. Weights for input shape (224, 224) will be loaded as the default.

```

encoder = MobileNetV2(input_tensor=inputs, weights="imagenet",
include_top=False, alpha=0.35)

```

Model: "functional_4"

Layer (type)	Output Shape	Param #	Connected to
input_image (InputLayer)	(None, 504, 672, 3)	0	-
Conv1 (Conv2D)	(None, 252, 336, 16)	432	input_image[0][0]
bn_Conv1 (BatchNormalizatio...	(None, 252, 336, 16)	64	Conv1[0][0]
Conv1_relu (ReLU)	(None, 252, 336, 16)	0	bn_Conv1[0][0]
expanded_conv_dept... (DepthwiseConv2D)	(None, 252, 336, 16)	144	Conv1_relu[0][0]
expanded_conv_dept... (BatchNormalizatio...	(None, 252, 336, 16)	64	expanded_conv_de...
expanded_conv_dept... (ReLU)	(None, 252, 336, 16)	0	expanded_conv_de...
expanded_conv_proj... (Conv2D)	(None, 252, 336, 8)	128	expanded_conv_de...
expanded_conv_proj... (BatchNormalizatio...	(None, 252, 336, 8)	32	expanded_conv_pr...
block_1_expand (Conv2D)	(None, 252, 336, 48)	384	expanded_conv_pr...

block_1_expand_BN (BatchNormalizatio...	(None, 252, 336, 48)	192	block_1_expand[0...
block_1_expand_relu (ReLU)	(None, 252, 336, 48)	0	block_1_expand_B...
block_1_pad (ZeroPadding2D)	(None, 253, 337, 48)	0	block_1_expand_r...
block_1_depthwise (DepthwiseConv2D)	(None, 126, 168, 48)	432	block_1_pad[0][0]
block_1_depthwise_... (BatchNormalizatio...	(None, 126, 168, 48)	192	block_1_depthwis...
block_1_depthwise_... (ReLU)	(None, 126, 168, 48)	0	block_1_depthwis...
block_1_project (Conv2D)	(None, 126, 168, 8)	384	block_1_depthwis...
block_1_project_BN (BatchNormalizatio...	(None, 126, 168, 8)	32	block_1_project[...
block_2_expand (Conv2D)	(None, 126, 168, 48)	384	block_1_project_...
block_2_expand_BN (BatchNormalizatio...	(None, 126, 168, 48)	192	block_2_expand[0...
block_2_expand_relu (ReLU)	(None, 126, 168, 48)	0	block_2_expand_B...
block_2_depthwise (DepthwiseConv2D)	(None, 126, 168, 48)	432	block_2_expand_r...
block_2_depthwise_... (BatchNormalizatio...	(None, 126, 168, 48)	192	block_2_depthwis...
block_2_depthwise_... (ReLU)	(None, 126, 168, 48)	0	block_2_depthwis...
block_2_project (Conv2D)	(None, 126, 168, 8)	384	block_2_depthwis...
block_2_project_BN (BatchNormalizatio...	(None, 126, 168, 8)	32	block_2_project[...

block_2_add (Add)	(None, 126, 168, 8)	0	block_1_project_ block_2_project_...
block_3_expand (Conv2D)	(None, 126, 168, 48)	384	block_2_add[0][0]
block_3_expand_BN (BatchNormalizatio...	(None, 126, 168, 48)	192	block_3_expand[0...
block_3_expand_relu (ReLU)	(None, 126, 168, 48)	0	block_3_expand_B...
block_3_pad (ZeroPadding2D)	(None, 127, 169, 48)	0	block_3_expand_r...
block_3_depthwise (DepthwiseConv2D)	(None, 63, 84, 48)	432	block_3_pad[0][0]
block_3_depthwise_... (BatchNormalizatio...	(None, 63, 84, 48)	192	block_3_depthwis...
block_3_depthwise_... (ReLU)	(None, 63, 84, 48)	0	block_3_depthwis...
block_3_project (Conv2D)	(None, 63, 84, 16)	768	block_3_depthwis...
block_3_project_BN (BatchNormalizatio...	(None, 63, 84, 16)	64	block_3_project[...
block_4_expand (Conv2D)	(None, 63, 84, 96)	1,536	block_3_project_...
block_4_expand_BN (BatchNormalizatio...	(None, 63, 84, 96)	384	block_4_expand[0...
block_4_expand_relu (ReLU)	(None, 63, 84, 96)	0	block_4_expand_B...
block_4_depthwise (DepthwiseConv2D)	(None, 63, 84, 96)	864	block_4_expand_r...
block_4_depthwise_... (BatchNormalizatio...	(None, 63, 84, 96)	384	block_4_depthwis...
block_4_depthwise_... (ReLU)	(None, 63, 84, 96)	0	block_4_depthwis...

block_4_project (Conv2D)	(None, 63, 84, 16)	1,536	block_4_depthwis...
block_4_project_BN (BatchNormalizatio...	(None, 63, 84, 16)	64	block_4_project[...
block_4_add (Add)	(None, 63, 84, 16)	0	block_3_project_... block_4_project_...
block_5_expand (Conv2D)	(None, 63, 84, 96)	1,536	block_4_add[0][0]
block_5_expand_BN (BatchNormalizatio...	(None, 63, 84, 96)	384	block_5_expand[0...
block_5_expand_relu (ReLU)	(None, 63, 84, 96)	0	block_5_expand_B...
block_5_depthwise (DepthwiseConv2D)	(None, 63, 84, 96)	864	block_5_expand_r...
block_5_depthwise_... (BatchNormalizatio...	(None, 63, 84, 96)	384	block_5_depthwis...
block_5_depthwise_... (ReLU)	(None, 63, 84, 96)	0	block_5_depthwis...
block_5_project (Conv2D)	(None, 63, 84, 16)	1,536	block_5_depthwis...
block_5_project_BN (BatchNormalizatio...	(None, 63, 84, 16)	64	block_5_project[...
block_5_add (Add)	(None, 63, 84, 16)	0	block_4_add[0][0... block_5_project_...
block_6_expand (Conv2D)	(None, 63, 84, 96)	1,536	block_5_add[0][0]
block_6_expand_BN (BatchNormalizatio...	(None, 63, 84, 96)	384	block_6_expand[0...
block_6_expand_relu (ReLU)	(None, 63, 84, 96)	0	block_6_expand_B...
block_6_pad (ZeroPadding2D)	(None, 65, 85, 96)	0	block_6_expand_r...

block_6_depthwise (DepthwiseConv2D)	(None, 32, 42, 96)	864	block_6_pad[0][0]
block_6_depthwise_...	(None, 32, 42, (BatchNormalizatio... 96)	384	block_6_depthwis...
block_6_depthwise_...	(None, 32, 42, (ReLU) 96)	0	block_6_depthwis...
block_6_project (Conv2D)	(None, 32, 42, 24)	2,304	block_6_depthwis...
block_6_project_BN (BatchNormalizatio...	(None, 32, 42, 24)	96	block_6_project[...
block_7_expand (Conv2D)	(None, 32, 42, 144)	3,456	block_6_project_...
block_7_expand_BN (BatchNormalizatio...	(None, 32, 42, 144)	576	block_7_expand[0...
block_7_expand_relu (ReLU)	(None, 32, 42, 144)	0	block_7_expand_B...
block_7_depthwise (DepthwiseConv2D)	(None, 32, 42, 144)	1,296	block_7_expand_r...
block_7_depthwise_...	(None, 32, 42, (BatchNormalizatio... 144)	576	block_7_depthwis...
block_7_depthwise_...	(None, 32, 42, (ReLU) 144)	0	block_7_depthwis...
block_7_project (Conv2D)	(None, 32, 42, 24)	3,456	block_7_depthwis...
block_7_project_BN (BatchNormalizatio...	(None, 32, 42, 24)	96	block_7_project[...
block_7_add (Add)	(None, 32, 42, 24)	0	block_6_project_... block_7_project_...
block_8_expand (Conv2D)	(None, 32, 42, 144)	3,456	block_7_add[0][0]
block_8_expand_BN (BatchNormalizatio...	(None, 32, 42, 144)	576	block_8_expand[0...

block_8_expand_relu (ReLU)	(None, 32, 42, 144)	0	block_8_expand_B...
block_8_depthwise (DepthwiseConv2D)	(None, 32, 42, 144)	1,296	block_8_expand_r...
block_8_depthwise_... (BatchNormalizatio...	(None, 32, 42, 144)	576	block_8_depthwis...
block_8_depthwise_... (ReLU)	(None, 32, 42, 144)	0	block_8_depthwis...
block_8_project (Conv2D)	(None, 32, 42, 24)	3,456	block_8_depthwis...
block_8_project_BN (BatchNormalizatio...	(None, 32, 42, 24)	96	block_8_project[...
block_8_add (Add)	(None, 32, 42, 24)	0	block_7_add[0][0... block_8_project_...
block_9_expand (Conv2D)	(None, 32, 42, 144)	3,456	block_8_add[0][0]
block_9_expand_BN (BatchNormalizatio...	(None, 32, 42, 144)	576	block_9_expand[0...
block_9_expand_relu (ReLU)	(None, 32, 42, 144)	0	block_9_expand_B...
block_9_depthwise (DepthwiseConv2D)	(None, 32, 42, 144)	1,296	block_9_expand_r...
block_9_depthwise_... (BatchNormalizatio...	(None, 32, 42, 144)	576	block_9_depthwis...
block_9_depthwise_... (ReLU)	(None, 32, 42, 144)	0	block_9_depthwis...
block_9_project (Conv2D)	(None, 32, 42, 24)	3,456	block_9_depthwis...
block_9_project_BN (BatchNormalizatio...	(None, 32, 42, 24)	96	block_9_project[...
block_9_add (Add)	(None, 32, 42, 24)	0	block_8_add[0][0... block_9_project_...

block_10_expand (Conv2D)	(None, 32, 42, 144)	3,456	block_9_add[0][0]
block_10_expand_BN (BatchNormalizatio...	(None, 32, 42, 144)	576	block_10_expand[...
block_10_expand_re... (ReLU)	(None, 32, 42, 144)	0	block_10_expand_...
block_10_depthwise (DepthwiseConv2D)	(None, 32, 42, 144)	1,296	block_10_expand_...
block_10_depthwise... (BatchNormalizatio...	(None, 32, 42, 144)	576	block_10_depthwi...
block_10_depthwise... (ReLU)	(None, 32, 42, 144)	0	block_10_depthwi...
block_10_project (Conv2D)	(None, 32, 42, 32)	4,608	block_10_depthwi...
block_10_project_BN (BatchNormalizatio...	(None, 32, 42, 32)	128	block_10_project...
block_11_expand (Conv2D)	(None, 32, 42, 192)	6,144	block_10_project...
block_11_expand_BN (BatchNormalizatio...	(None, 32, 42, 192)	768	block_11_expand[...
block_11_expand_re... (ReLU)	(None, 32, 42, 192)	0	block_11_expand_...
block_11_depthwise (DepthwiseConv2D)	(None, 32, 42, 192)	1,728	block_11_expand_...
block_11_depthwise... (BatchNormalizatio...	(None, 32, 42, 192)	768	block_11_depthwi...
block_11_depthwise... (ReLU)	(None, 32, 42, 192)	0	block_11_depthwi...
block_11_project (Conv2D)	(None, 32, 42, 32)	6,144	block_11_depthwi...
block_11_project_BN (BatchNormalizatio...	(None, 32, 42, 32)	128	block_11_project...

block_11_add (Add)	(None, 32, 42, 32)	0	block_10_project... block_11_project...
block_12_expand (Conv2D)	(None, 32, 42, 192)	6,144	block_11_add[0] [...]
block_12_expand_BN (BatchNormalizatio...	(None, 32, 42, 192)	768	block_12_expand[...]
block_12_expand_re... (ReLU)	(None, 32, 42, 192)	0	block_12_expand_...
block_12_depthwise (DepthwiseConv2D)	(None, 32, 42, 192)	1,728	block_12_expand_...
block_12_depthwise... (BatchNormalizatio...	(None, 32, 42, 192)	768	block_12_depthwi...
block_12_depthwise... (ReLU)	(None, 32, 42, 192)	0	block_12_depthwi...
block_12_project (Conv2D)	(None, 32, 42, 32)	6,144	block_12_depthwi...
block_12_project_BN (BatchNormalizatio...	(None, 32, 42, 32)	128	block_12_project...
block_12_add (Add)	(None, 32, 42, 32)	0	block_11_add[0] [...] block_12_project...
block_13_expand (Conv2D)	(None, 32, 42, 192)	6,144	block_12_add[0] [...]
block_13_expand_BN (BatchNormalizatio...	(None, 32, 42, 192)	768	block_13_expand[...]
block_13_expand_re... (ReLU)	(None, 32, 42, 192)	0	block_13_expand_...
up_sampling2d_16 (UpSampling2D)	(None, 64, 84, 192)	0	block_13_expand_...
cropping2d_4 (Cropping2D)	(None, 63, 84, 192)	0	up_sampling2d_16...
concatenate_16 (Concatenate)	(None, 63, 84, 288)	0	cropping2d_4[0] [...] block_6_expand_r...

conv2d_36 (Conv2D)	(None, 63, 84, 64)	165,952	concatenate_16[0...
batch_normalizatio... (BatchNormalizatio...	(None, 63, 84, 64)	256	conv2d_36[0][0]
activation_36 (Activation)	(None, 63, 84, 64)	0	batch_normalizat...
conv2d_37 (Conv2D)	(None, 63, 84, 64)	36,928	activation_36[0]...
batch_normalizatio... (BatchNormalizatio...	(None, 63, 84, 64)	256	conv2d_37[0][0]
activation_37 (Activation)	(None, 63, 84, 64)	0	batch_normalizat...
up_sampling2d_17 (UpSampling2D)	(None, 126, 168, 64)	0	activation_37[0]...
concatenate_17 (Concatenate)	(None, 126, 168, 112)	0	up_sampling2d_17... block_3_expand_r...
conv2d_38 (Conv2D)	(None, 126, 168, 48)	48,432	concatenate_17[0...
batch_normalizatio... (BatchNormalizatio...	(None, 126, 168, 48)	192	conv2d_38[0][0]
activation_38 (Activation)	(None, 126, 168, 48)	0	batch_normalizat...
conv2d_39 (Conv2D)	(None, 126, 168, 48)	20,784	activation_38[0]...
batch_normalizatio... (BatchNormalizatio...	(None, 126, 168, 48)	192	conv2d_39[0][0]
activation_39 (Activation)	(None, 126, 168, 48)	0	batch_normalizat...
up_sampling2d_18 (UpSampling2D)	(None, 252, 336, 48)	0	activation_39[0]...
concatenate_18 (Concatenate)	(None, 252, 336, 96)	0	up_sampling2d_18... block_1_expand_r...

conv2d_40 (Conv2D)	(None, 252, 336, 32)	27,680	concatenate_18[0...
batch_normalizatio... (BatchNormalizatio...	(None, 252, 336, 32)	128	conv2d_40[0][0]
activation_40 (Activation)	(None, 252, 336, 32)	0	batch_normalizat...
conv2d_41 (Conv2D)	(None, 252, 336, 32)	9,248	activation_40[0]...
batch_normalizatio... (BatchNormalizatio...	(None, 252, 336, 32)	128	conv2d_41[0][0]
activation_41 (Activation)	(None, 252, 336, 32)	0	batch_normalizat...
up_sampling2d_19 (UpSampling2D)	(None, 504, 672, 32)	0	activation_41[0]...
concatenate_19 (Concatenate)	(None, 504, 672, 35)	0	up_sampling2d_19... input_image[0][0]
conv2d_42 (Conv2D)	(None, 504, 672, 16)	5,056	concatenate_19[0...
batch_normalizatio... (BatchNormalizatio...	(None, 504, 672, 16)	64	conv2d_42[0][0]
activation_42 (Activation)	(None, 504, 672, 16)	0	batch_normalizat...
conv2d_43 (Conv2D)	(None, 504, 672, 16)	2,320	activation_42[0]...
batch_normalizatio... (BatchNormalizatio...	(None, 504, 672, 16)	64	conv2d_43[0][0]
activation_43 (Activation)	(None, 504, 672, 16)	0	batch_normalizat...
conv2d_44 (Conv2D)	(None, 504, 672, 22)	374	activation_43[0]...
activation_44 (Activation)	(None, 504, 672, 22)	0	conv2d_44[0][0]

Total params: 416,566 (1.59 MB)

Trainable params: 409,382 (1.56 MB)

Non-trainable params: 7,184 (28.06 KB)

```
[49]: # Define callbacks
callbacks = [
    ReduceLROnPlateau(monitor='val_loss', factor=0.1, patience=4),
    EarlyStopping(monitor='val_loss', patience=10, restore_best_weights=False)
]

train_steps = len(train_x) // BATCH
valid_steps = len(valid_x) // BATCH
if len(train_x) % BATCH != 0:
    train_steps += 1
if len(valid_x) % BATCH != 0:
    valid_steps += 1

train_dataset = tf_dataset(train_x, train_y, batch=BATCH)
valid_dataset = tf_dataset(valid_x, valid_y, batch=BATCH)

# Train the model and record history
history = model.fit(
    train_dataset,
    validation_data=valid_dataset,
    epochs=EPOCHS,
    steps_per_epoch=train_steps,
    validation_steps=valid_steps,
    callbacks=callbacks
)
```

Epoch 1/5

370/370 0s 329ms/step -

accuracy: 0.4819 - dice_coefficient: 0.1340 - iou: 0.0728 - loss: 2.3275

2024-07-27 10:43:58.658050: I

external/local_xla/xla/stream_executor/cuda/cuda_asm_compiler.cc:393] ptxas

warning : Registers are spilled to local memory in function

'input_and_reduce_fusion_5', 8 bytes spill stores, 8 bytes spill loads

ptxas warning : Registers are spilled to local memory in function

'input_and_reduce_fusion_9', 8 bytes spill stores, 8 bytes spill loads

370/370 160s 369ms/step -

accuracy: 0.4825 - dice_coefficient: 0.1341 - iou: 0.0729 - loss: 2.3265 -
val_accuracy: 0.8273 - val_dice_coefficient: 0.2409 - val_iou: 0.1370 -
val_loss: 1.6164 - learning_rate: 1.0000e-04

Epoch 2/5

370/370 126s 341ms/step -

accuracy: 0.8235 - dice_coefficient: 0.2871 - iou: 0.1680 - loss: 1.4549 -
val_accuracy: 0.8267 - val_dice_coefficient: 0.3766 - val_iou: 0.2321 -
val_loss: 1.2104 - learning_rate: 1.0000e-04

Epoch 3/5

370/370 128s 346ms/step -

accuracy: 0.8370 - dice_coefficient: 0.4115 - iou: 0.2595 - loss: 1.0964 -
val_accuracy: 0.8193 - val_dice_coefficient: 0.4918 - val_iou: 0.3264 -
val_loss: 0.9781 - learning_rate: 1.0000e-04

Epoch 4/5

370/370 126s 342ms/step -

accuracy: 0.8454 - dice_coefficient: 0.5215 - iou: 0.3532 - loss: 0.8578 -
val_accuracy: 0.8153 - val_dice_coefficient: 0.5761 - val_iou: 0.4052 -
val_loss: 0.8474 - learning_rate: 1.0000e-04

Epoch 5/5

370/370 118s 321ms/step -

accuracy: 0.8504 - dice_coefficient: 0.6054 - iou: 0.4346 - loss: 0.7061 -
val_accuracy: 0.8223 - val_dice_coefficient: 0.6497 - val_iou: 0.4820 -
val_loss: 0.7285 - learning_rate: 1.0000e-04

```
[50]: # Evaluate the model
test_dataset = tf_dataset(test_x, test_y, batch=BATCH)
test_steps = (len(test_x) // BATCH)
if len(test_x) % BATCH != 0:
    test_steps += 1
results = model.evaluate(test_dataset, steps=test_steps)

# Get metric names
metrics_names = model.metrics_names

# Print evaluation results
for name, value in zip(metrics_names, results):
    print(f"{name}: {value:.4f}")
```

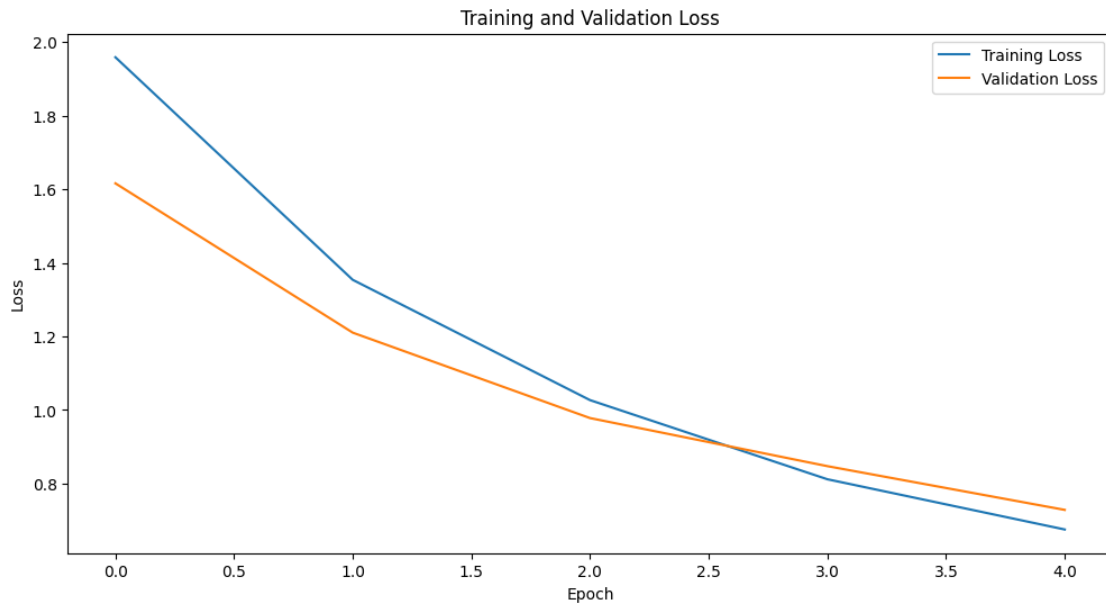
46/46 14s 311ms/step -

accuracy: 0.8167 - dice_coefficient: 0.6445 - iou: 0.4769 - loss: 0.7344
loss: 0.7518
compile_metrics: 0.8114

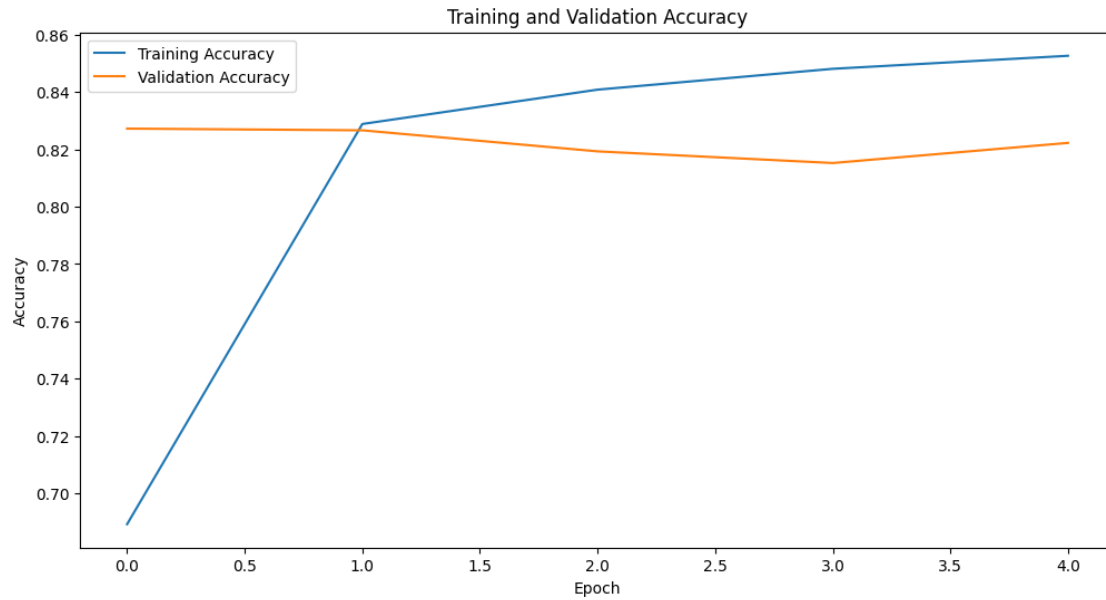
```
[51]: # Plot training and validation loss
plt.figure(figsize=(12, 6))
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
```



```
plt.legend()
plt.title('Training and Validation Loss')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.show()
```



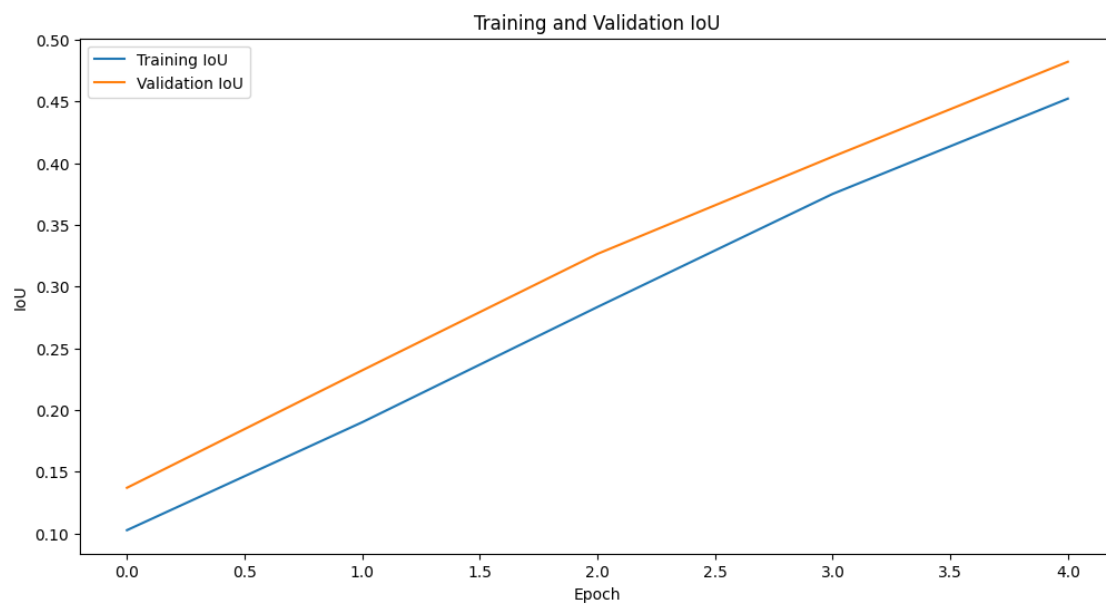
```
[52]: # Plot training and validation accuracy
plt.figure(figsize=(12, 6))
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.legend()
plt.title('Training and Validation Accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.show()
```



```
[53]: # Plot training and validation Dice coefficient
plt.figure(figsize=(12, 6))
plt.plot(history.history['dice_coefficient'], label='Training Dice Coefficient')
plt.plot(history.history['val_dice_coefficient'], label='Validation Dice_
    ↪Coefficient')
plt.legend()
plt.title('Training and Validation Dice Coefficient')
plt.xlabel('Epoch')
plt.ylabel('Dice Coefficient')
plt.show()
```



```
[54]: # Plot training and validation IoU
plt.figure(figsize=(12, 6))
plt.plot(history.history['iou'], label='Training IoU')
plt.plot(history.history['val_iou'], label='Validation IoU')
plt.legend()
plt.title('Training and Validation IoU')
plt.xlabel('Epoch')
plt.ylabel('IoU')
plt.show()
```



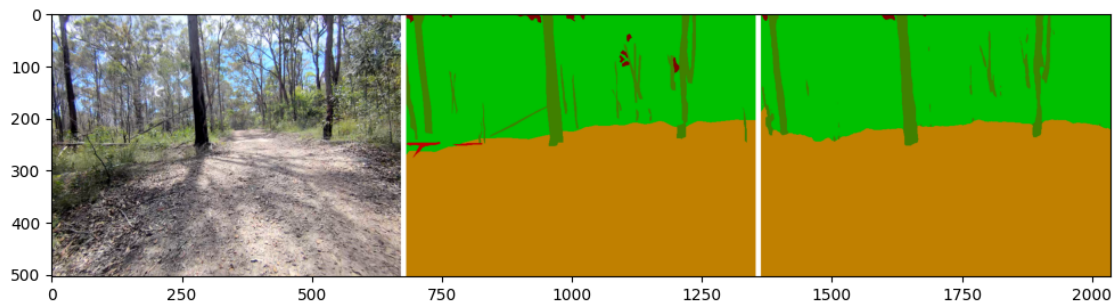
```
[55]: # Visualize prediction results
for i, (x_path, y_path) in enumerate(zip(test_x[:10], test_y[:10])):
    x = read_image(x_path)
    y_pred = model.predict(np.expand_dims(x, axis=0))[0]
    y_original = read_mask(y_path) # Directly read the mask

    h, w, _ = x.shape
    white_line = np.ones((h, 10, 3)) * 255

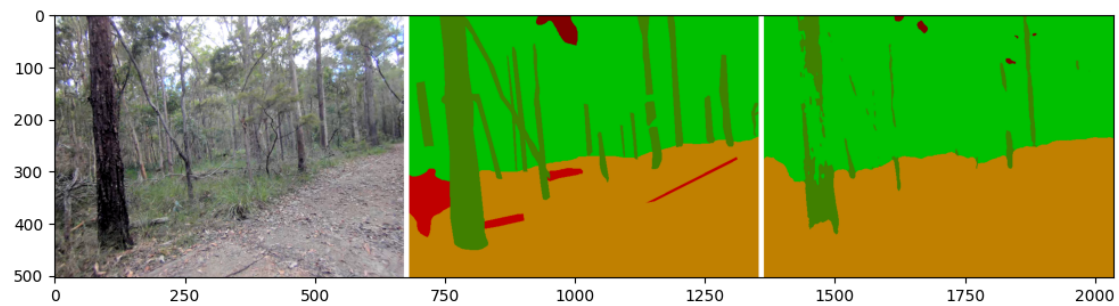
    all_images = [
        x * 255, white_line,
        mask_parse(y_original), white_line, # Parse and display the original
        ↪mask mask_parse(y_pred)
    ]
    image = np.concatenate(all_images, axis=1).astype(np.uint8)

    fig = plt.figure(figsize=(12, 12))
    a = fig.add_subplot(1, 1, 1)
    imgplot = plt.imshow(image)
    plt.show()
```

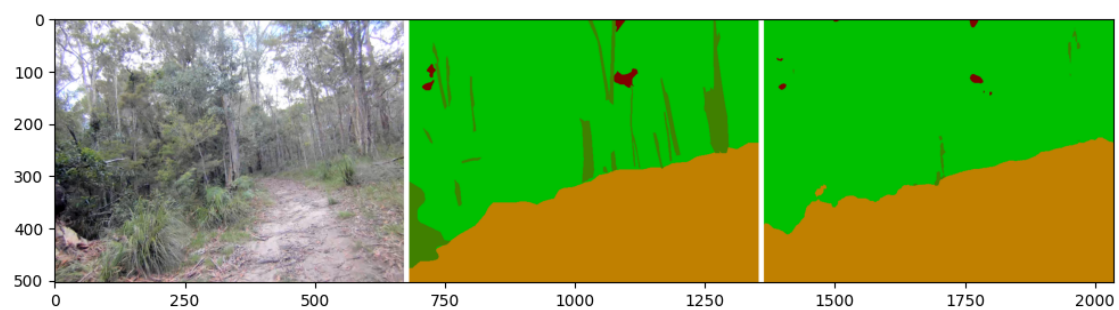
1/1 1s 791ms/step



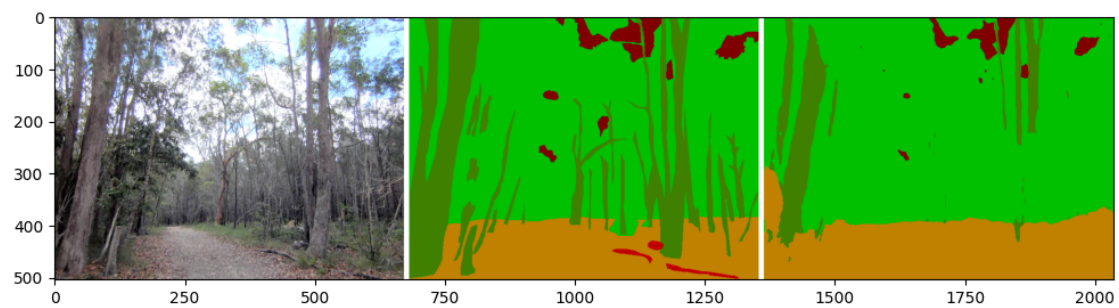
1/1 0s 11ms/step



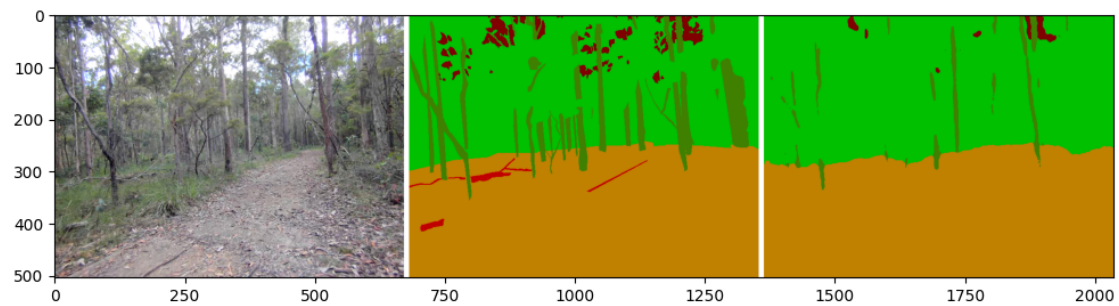
1/1 0s 11ms/step



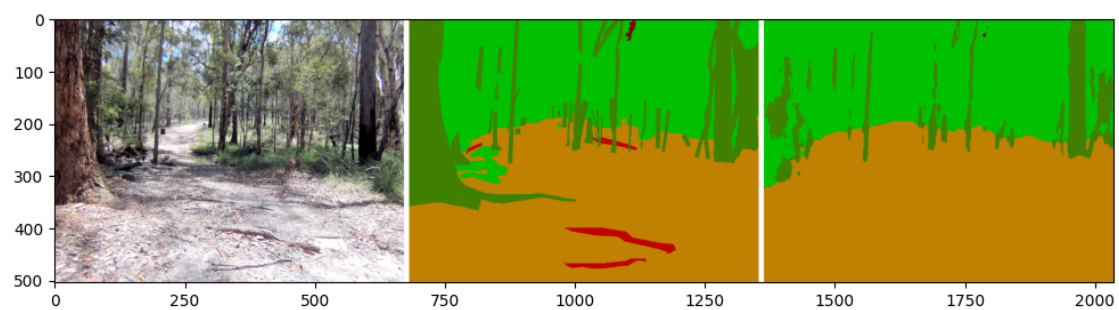
1/1 0s 11ms/step



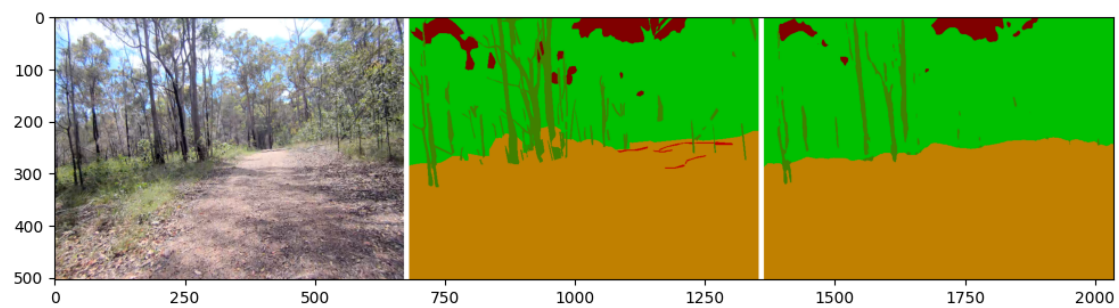
1/1 0s 11ms/step



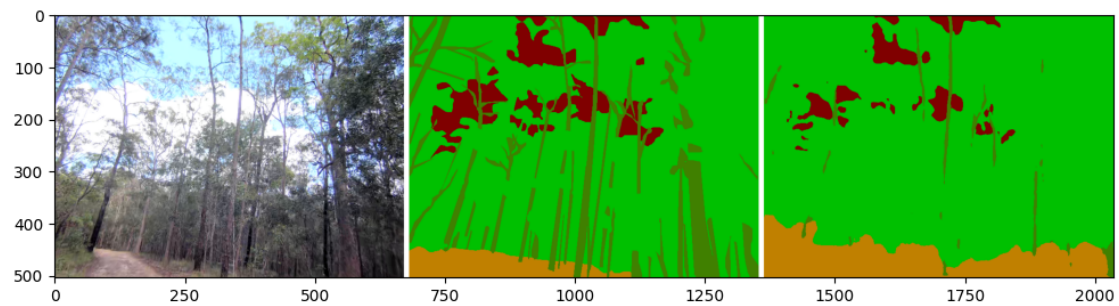
1/1 0s 10ms/step



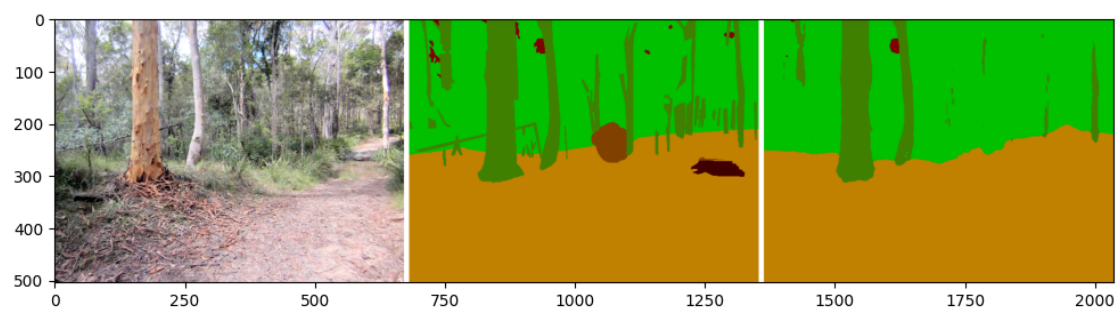
1/1 0s 10ms/step



1/1 0s 10ms/step



1/1 0s 11ms/step



1/1 0s 10ms/step

