This is a companion notebook for the book <u>Deep Learning with Python, Second Edition</u>. For readability, it only contains runnable code blocks and section titles, and omits everything else in the book: text paragraphs, figures, and pseudocode.

If you want to be able to follow what's going on, I recommend reading the notebook side by side with your copy of the book.

This notebook was generated for TensorFlow 2.6.

## Advanced deep learning for computer vision

## Three essential computer vision tasks

## An image segmentation example

```
| luget http://mm.robots.ox.ac.uk/~vgg/data/pets/data/amoutations.tar.gz | luget http://mm.robots.ox.ac.uk/~vgg/data/pets/data/amoutations.tar.gz | lar - vf amoutations.tar.gz | lar - vf
```

```
In [13]:
    import os
    input_dir = "images/"
    target_dir * "annotations/trimaps/"
    input_img_paths = sorted(
        [os.path.join(input_dir, fname)
            for fname in os.listdir(input_dir)
            if fname.endswith(".jpg")])
    target_paths = sorted(
            [os.path.join(target_dir, fname)
            for fname in os.listdir(target_dir)
            if fname.endswith(".png") and not fname.startswith(".")])

In [14]:
    import matplotlib.pyplot as plt
    from tensorflow.keras.utils import load_img, img_to_array
    plt.axis("off")
    plt.imshow(load_img(input_img_paths[9]))
```

Out[14]:



```
def display_target(target_array):
    normalized_array = (target_array.astype("uint8") - 1) * 127
    plt.axis("off")
    plt.imshow(normalized_array[:, :, 0])

img = img_to_array(load_img(target_paths[9], color_mode="grayscale"))
display_target(img)
```



```
from tensorflow import keras
from tensorflow.keras import layers
    def get_model(img_size, num_classes):
   inputs = keras.Input(shape=img_size + (3,))
   x = layers.Rescaling(1./255)(inputs)
           x = layers.Conv2D(64, 3, strides=2, activation="relu", padding="same")(x)
x = layers.Conv2D(64, 3, activation="relu", padding="same")(x)
x = layers.Conv2D(128, 3, strides=2, activation="relu", padding="same")(x)
x = layers.Conv2D(128, 3, activation="relu", padding="same")(x)
x = layers.Conv2D(256, 3, strides=2, padding="same", activation="relu")(x)
x = layers.Conv2D(256, 3, activation="relu", padding="same")(x)
           x = layers.Conv2DTranspose(256, 3, activation="relu", padding="same")(x)
x = layers.Conv2DTranspose(286, 3, activation="relu", padding="same", strides=2)(x)
x = layers.Conv2DTranspose(128, 3, activation="relu", padding="same")(x)
x = layers.Conv2DTranspose(128, 3, activation="relu", padding="same")(x)
x = layers.Conv2DTranspose(64, 3, activation="relu", padding="same")(x)
x = layers.Conv2DTranspose(64, 3, activation="relu", padding="same", strides=2)(x)
           outputs = layers.Conv2D(num\_classes, 3, activation="softmax", padding="same")(x) \\
            model = keras.Model(inputs, outputs)
return model
    model = get_model(img_size=img_size, num_classes=3)
model.summary()
 Layer (type)
                                                     Output Shape
[(None, 200, 200, 3)]
  input_1 (InputLayer)
  rescaling (Rescaling) (None, 200, 200, 3)
 conv2d (Conv2D) (None, 100, 100, 64)
conv2d_1 (Conv2D) (None, 100, 100, 64)
conv2d_2 (Conv2D) (None, 50, 50, 128)
conv2d_3 (Conv2D) (None, 50, 50, 128)
                                                                                                         73856
                                                                                                        147584
 conv2d_4 (Conv2D) (None, 25, 25, 256)
conv2d_5 (Conv2D) (None, 25, 25, 256)
                                                                                                         295168
                                                                                                         590080
 conv2d_transpose (Conv2DTra (None, 25, 25, 256) nspose)
                                                                                                         590080
  conv2d_transpose_1 (Conv2DT (None, 50, 50, 256)
                                                                                                          590080
  conv2d_transpose_2 (Conv2DT (None, 50, 50, 128)
                                                                                                          295040
   conv2d_transpose_3 (Conv2DT (None, 100, 100, 128)
                                                                                                         147584
  ranspose)
  conv2d_transpose_4 (Conv2DT (None, 100, 100, 64)
ranspose)
                                                                                                          73792
  conv2d_transpose_5 (Conv2DT (None, 200, 200, 64)
ranspose)
  conv2d_6 (Conv2D)
                                                     (None, 200, 200, 3)
Total params: 2,880,643
Trainable params: 2,880,643
Non-trainable params: 0
```

```
In [20]:
    epochs = range(1, len(history.history["loss"]) + 1)
    loss = history.history["loss"]
    val_loss = history.history["val_loss"]
    plt.figure()
    plt.plot(epochs, loss, "bo", label="Training loss")
    plt.plot(epochs, val_loss, "bo", label="validation loss")
    plt.title("Training and validation loss")
    plt.legend()
```

Out[20]:

```
Training and validation loss

12

Validation loss

11

10

09

08

10

15

20

25

30

35

40

45

50
```

```
In [21]:
    from tensorflow.keras.utils import array_to_img
    model = keras.models.load_model("oxford_segmentation.keras")
    i = 4
    test_image = val_input_imgs[i]
    plt.axis("off")
    plt.imshow(array_to_img(test_image))
    mask = model.predict(np.expand_dims(test_image, 0))[0]

def display_mask(pred):
    mask = np.argmax(pred, axis=-1)
    mask = 127
    plt.axis("off")
    plt.imshow(mask)

display_mask(mask)
```

1/1 [-----] - 1s 732ms/step

