Introduction

In the tutorial we learned about the building blocks of neural networks: *linear units*. We saw that a model of just one linear unit will fit a linear function to a dataset (equivalent to linear regression). In this exercise, you'll build a linear model and get some practice working with models in Keras.

Before you get started, run the code cell below to set everything up.

The *Red Wine Quality* dataset consists of physiochemical measurements from about 1600 Portuguese red wines. Also included is a quality rating for each wine from blind taste-tests.

First, run the next cell to display the first few rows of this dataset.

```
import pandas as pd

red_wine = pd.read_csv('../input/dl-course-data/red-wine.csv')
red_wine.head()
```

```
Out[2]:
                                                                 free
                                                                          total
                fixed volatile citric residual
                                                  chlorides
                                                               sulfur
                                                                         sulfur
                                                                                density
                                                                                           pH sulphates alcohol qualit
              acidity
                        acidity
                                 acid
                                          sugar
                                                             dioxide dioxide
           0
                  7.4
                           0.70
                                  0.00
                                             1.9
                                                      0.076
                                                                 11.0
                                                                          34.0
                                                                                  0.9978 3.51
                                                                                                      0.56
                                                                                                                 9.4
           1
                  7.8
                           88.0
                                  0.00
                                             2.6
                                                      0.098
                                                                 25.0
                                                                          67.0
                                                                                 0.9968 3.20
                                                                                                      0.68
                                                                                                                 9.8
           2
                  7.8
                          0.76
                                 0.04
                                             2.3
                                                      0.092
                                                                 15.0
                                                                                 0.9970 3.26
                                                                                                      0.65
                                                                                                                 9.8
                                                                          54.0
           3
                 11.2
                          0.28
                                 0.56
                                             1.9
                                                      0.075
                                                                 17.0
                                                                          60.0
                                                                                 0.9980 3.16
                                                                                                      0.58
                                                                                                                 9.8
           4
                          0.70
                                 0.00
                                                                                                                 9.4
                  7.4
                                             1.9
                                                      0.076
                                                                 11.0
                                                                          34.0
                                                                                 0.9978 3.51
                                                                                                      0.56
```

You can get the number of rows and columns of a dataframe (or a Numpy array) with the shape attribute.

```
In [3]: red_wine.shape # (rows, columns)
```

```
Out[3]: (1599, 12)
```

1) Input shape

How well can we predict a wine's perceived quality from the physiochemical measurements?

The target is 'quality', and the remaining columns are the features. How would you set the input_shape parameter for a Keras model on this task?

```
In [4]: # YOUR CODE HERE
  input_shape = [len(red_wine.columns)-1]

# Check your answer
q_1.check()
```

Correct

```
In [5]:
# Lines below will give you a hint or solution code
#q_1.hint()
#q_1.solution()
```

2) Define a linear model

Now define a linear model appropriate for this task. Pay attention to how many inputs and outputs the model should have.

```
from tensorflow import keras
from tensorflow.keras import layers

# YOUR CODE HERE
model = keras.Sequential([layers.Dense(units=1, input_shape=input_shape)])

# Check your answer
q_2.check()
```

2022-12-18 05:12:50.107668: I tensorflow/core/common_runtime/process_util.cc:146] Creating new thread pool with default inter op setting: 2. Tune using inter_op_parallelism_threads for best performance.

Correct

```
In [7]:
# Lines below will give you a hint or solution code
#q_2.hint()
#q_2.solution()
```

3) Look at the weights

Internally, Keras represents the weights of a neural network with **tensors**. Tensors are basically TensorFlow's version of a Numpy array with a few differences that make them better suited to deep learning. One of the most important is that tensors are compatible with GPU and TPU) accelerators. TPUs, in fact, are designed specifically for tensor computations.

A model's weights are kept in its weights attribute as a list of tensors. Get the weights of the model you defined above. (If you want, you could display the weights with something like: print("Weights\n{}\n\nBias\n{}\".format(w, b))).

```
In [8]: # YOUR CODE HERE
w, b = model.weights

# Check your answer
q_3.check()
```

Correct: Do you see how there's one weight for each input (and a bias)? Notice though that there doesn't seem to be any pattern to the values the weights have. Before the model is trained, the weights are set to random numbers (and the bias to 0.0). A neural network learns by finding better values for its weights.

```
In [9]:
# Lines below will give you a hint or solution code
#q_3.hint()
#q_3.solution()
```

(By the way, Keras represents weights as tensors, but also uses tensors to represent data. When you set the input_shape argument, you are telling Keras the dimensions of the array it should expect for each example in the training data. Setting input_shape=[3] would create a network accepting vectors of length 3, like [0.2, 0.4, 0.6].)

Optional: Plot the output of an untrained linear model

The kinds of problems we'll work on through Lesson 5 will be *regression* problems, where the goal is to predict some numeric target. Regression problems are like "curve-fitting" problems: we're trying to find a curve that best fits the data. Let's take a look at the "curve" produced by a linear model. (You've probably guessed that it's a line!)

We mentioned that before training a model's weights are set randomly. Run the cell below a few times to see the different lines produced with a random initialization. (There's no coding for this exercise -- it's just a demonstration.)

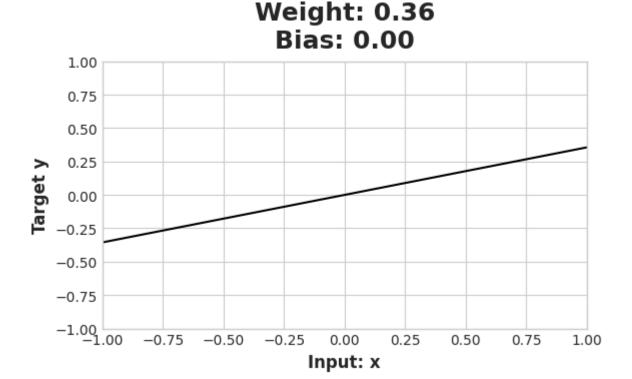
```
import tensorflow as tf
import matplotlib.pyplot as plt
model = keras.Sequential([
```

```
layers.Dense(1, input_shape=[1]),
])

x = tf.linspace(-1.0, 1.0, 100)
y = model.predict(x)

plt.figure(dpi=100)
plt.plot(x, y, 'k')
plt.xlim(-1, 1)
plt.ylim(-1, 1)
plt.ylim(-1, 1)
plt.xlabel("Input: x")
plt.ylabel("Target y")
w, b = model.weights # you could also use model.get_weights() here
plt.title("Weight: {:0.2f}\nBias: {:0.2f}".format(w[0][0], b[0]))
plt.show()
```

2022-12-18 05:12:50.430050: I tensorflow/compiler/mlir_graph_optimization_pass.cc:1 85] None of the MLIR Optimization Passes are enabled (registered 2)



Keep Going

Add hidden layers and make your models deep in Lesson 2.

Have questions or comments? Visit the course discussion forum to chat with other learners.