Neural Networks

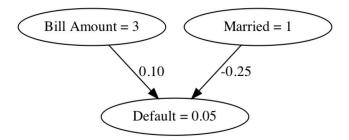
May 17, 2021

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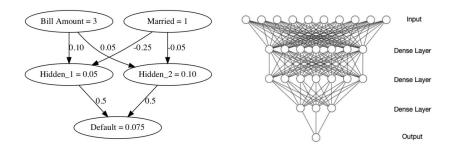
1 Dense layers

1.1 [note-1] The linear regression model

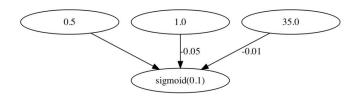


1.2 [note-2] What is a neural network?

• A dense layer applies weights to all nodes from the previous layer.



1.3 [code-1] A simple dense layer



```
[28]: import tensorflow as tf
[29]: # Define inputs (features)
    inputs = tf.constant([[1, 35]], tf.float32)

[30]: # Define weights
    weights = tf.Variable([[-0.05], [-0.01]])

[31]: # Define the bias
    bias = tf.Variable([0.5])

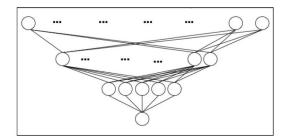
[32]: # Multiply inputs (features) by the weights
    product = tf.matmul(inputs, weights)

    product

[32]: <tf.Tensor: shape=(1, 1), dtype=float32, numpy=array([[-0.4]], dtype=float32)>

[21]: # Define dense layer
    dense = tf.keras.activations.sigmoid(product + bias)
    dense
```

1.4 [code-2] Defining a complete model



```
[22]: import pandas as pd
      import tensorflow as tf
      data = pd.read_csv("../Datasets/1. Borrower features.csv", header=None)
[24]: # Define input (features) layer
      inputs = tf.constant(data, tf.float32)
      inputs.shape
[24]: TensorShape([100, 10])
[25]: # Define first dense layer
      dense1 = tf.keras.layers.Dense(10, activation='sigmoid')(inputs)
      dense1.shape
[25]: TensorShape([100, 10])
[26]: # Define second dense layer
      dense2 = tf.keras.layers.Dense(5, activation='sigmoid')(dense1)
      dense2.shape
[26]: TensorShape([100, 5])
[27]: # Define output (predictions) layer
      outputs = tf.keras.layers.Dense(1, activation='sigmoid')(dense2)
```

[27]: TensorShape([100, 1])

outputs.shape

- 1.5 [note-3] High-level versus low-level approach
 - High-level approach:

- high-level API operations
- example:

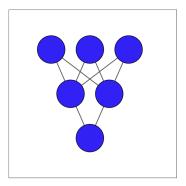
```
dense = tensorflow.keras.layers.Dense(10, activation='sigmoid')(inputs)
```

- Low-level approach:
 - linear-algebraic operations
 - example:

```
prod = tensorflow.matmul(inputs, weights)
dense = tensorflow.keras.activations.sigmoid(prod + bias)
```

1.6 [task-1] The linear algebra of dense layers

▶ Task diagram



▶ Package pre-loading

```
[1]: import numpy as np from tensorflow import Variable, ones, matmul, keras
```

▶ Data pre-loading

```
[2]: borrower_features = np.array([[2., 2., 43.]], dtype=np.float32)
```

▶ Task practice 1/2

```
[3]: # From previous step
bias1 = Variable(1.0)
weights1 = Variable(ones((3, 2)))
product1 = matmul(borrower_features, weights1)
dense1 = keras.activations.sigmoid(product1 + bias1)

# Initialize bias2 and weights2
bias2 = Variable(1.0)
weights2 = Variable(ones((2, 1)))

# Perform matrix multiplication of dense1 and weights2
product2 = matmul(dense1, weights2)

# Apply activation to product2 + bias2 and print the prediction
prediction = keras.activations.sigmoid(product2 + bias2)
print('\n prediction: {}'.format(prediction.numpy()[0, 0]))
print('\n actual: 1')
```

prediction: 0.9525741338729858

actual: 1

► Task practice 2/2

```
[4]: # From previous step
bias1 = Variable(1.0)
weights1 = Variable(ones((3, 2)))
product1 = matmul(borrower_features, weights1)
dense1 = keras.activations.sigmoid(product1 + bias1)

# Initialize bias2 and weights2
bias2 = Variable(1.0)
weights2 = Variable(ones((2, 1)))

# Perform matrix multiplication of dense1 and weights2
product2 = matmul(dense1, weights2)

# Apply activation to product2 + bias2 and print the prediction
prediction = keras.activations.sigmoid(product2 + bias2)
print('\n prediction: {}'.format(prediction.numpy()[0, 0]))
print('\n actual: 1')
```

prediction: 0.9525741338729858

actual: 1

1.7 [task-2] The low-level approach with multiple examples

▶ Task instruction

$$products1 = \begin{bmatrix} 3 & 3 & 23 \\ 2 & 1 & 24 \\ 1 & 1 & 49 \\ 1 & 1 & 49 \\ 2 & 1 & 29 \end{bmatrix} \begin{bmatrix} -0.6 & 0.6 \\ 0.8 & -0.3 \\ -0.09 & -0.08 \end{bmatrix}$$

► Package pre-loading

```
[5]: from tensorflow import constant, float32
```

▶ Data pre-loading

▶ Task practice

```
[7]: # Compute the product of borrower_features and weights1
products1 = matmul(borrower_features, weights1)

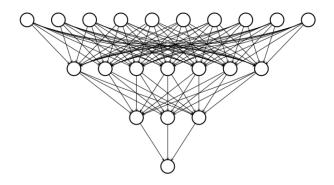
# Apply a sigmoid activation function to products1 + bias1
dense1 = keras.activations.sigmoid(products1 + bias1)

# Print the shapes of borrower_features, weights1, bias1, and dense1
print('\n shape of borrower_features: ', borrower_features.shape)
print('\n shape of weights1: ', weights1.shape)
print('\n shape of bias1: ', bias1.shape)
print('\n shape of dense1: ', dense1.shape)
```

```
shape of borrower_features: (5, 3)
shape of weights1: (3, 2)
shape of bias1: (1,)
shape of dense1: (5, 2)
```

1.8 [task-3] Using the dense layer operation

► Task diagram



▶ Package pre-loading

```
[8]: import pandas as pd
```

▶ Data pre-loading

```
[9]: df = pd.read_csv("../Datasets/1. Borrower features.csv", header=None) borrower_features = constant(df, dtype=float32)
```

► Task practice

```
[10]: # Define the first dense layer
dense1 = keras.layers.Dense(7, activation='sigmoid')(borrower_features)

# Define a dense layer with 3 output nodes
dense2 = keras.layers.Dense(3, activation='sigmoid')(dense1)

# Define a dense layer with 1 output node
predictions = keras.layers.Dense(1, activation='sigmoid')(dense2)

# Print the shapes of dense1, dense2, and predictions
print('\n shape of dense1: ', dense1.shape)
print('\n shape of dense2: ', dense2.shape)
print('\n shape of predictions: ', predictions.shape)
```

```
shape of dense1: (100, 7)
shape of dense2: (100, 3)
shape of predictions: (100, 1)
```

```
[34]: import tensorflow
```

```
[38]: dense = tensorflow.keras.layers.Dense(10, activation='sigmoid')(dense2)
```

```
[39]: dense.shape
[39]: TensorShape([100, 10])
[40]: dense2.shape
[40]: TensorShape([100, 5])
[]:
[]:
[]:
[]:
[]:
[]:
[]:
```

2 Requirements

```
[]: from platform import python_version
     import sklearn
     python_version = ('python=={}'.format(python_version()))
     numpy_version = ('numpy=={}'.format(np.__version__))
     pandas_version = ('pandas=={}'.format(pd.__version__))
     tensorflow_version = ('tensorflow=={}'.format(tf.__version__))
     scikit_learn_version = ('scikit-learn=={}'.format(sklearn.__version__))
     writepath = '../../requirements.txt'
     requirements = []
     packages = [
         numpy_version, pandas_version, tensorflow_version, scikit_learn_version
     ]
     try:
         with open(writepath, 'r+') as file:
             for line in file:
                 requirements.append(line.strip('\n'))
     except:
         with open(writepath, 'w+') as file:
             for line in file:
                 requirements.append(line.strip('\n'))
```

```
with open(writepath, 'a') as file:
    for package in packages:
        if package not in requirements:
            file.write(package + '\n')
max_characters = len(python_version)
for package in packages:
    if max(max_characters, len(package)) > max_characters:
        max_characters = max(max_characters, len(package))
print('#' * (max_characters + 8))
print('#' * 2 + ' ' * (max_characters + 4) + '#' * 2)
print('#' * 2 + ' ' * 2 + python_version + ' ' *
      (max_characters - len(python_version) + 2) + '#' * 2)
for package in packages:
    print('#' * 2 + ' ' * 2 + package + ' ' *
          (\max_{\text{characters}} - \text{len}(\text{package}) + 2) + '#' * 2)
print('#' * 2 + ' ' * (max_characters + 4) + '#' * 2)
print('#' * (max_characters + 8))
```