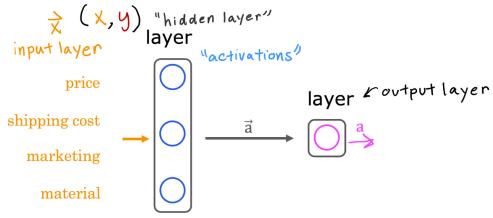
1. 1 point



Which of these are terms used to refer to components of an artificial neural network? (hint: three of these are correct)

- neurons
- layers
- axon
- activation function

 $\textbf{2.} \quad \text{True/False? Neural networks take inspiration from, but do not very accurately mimic, how neurons in a biological brain learn.}$



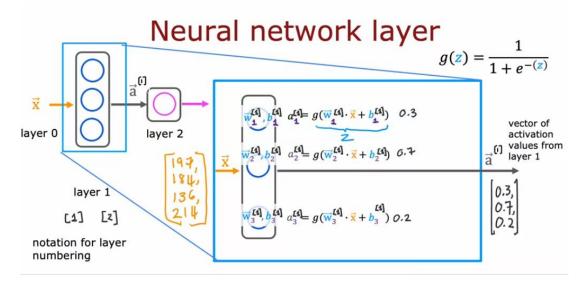
- True
- False

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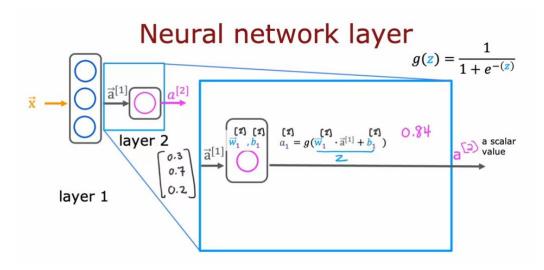
Windows'u Etkinleştir

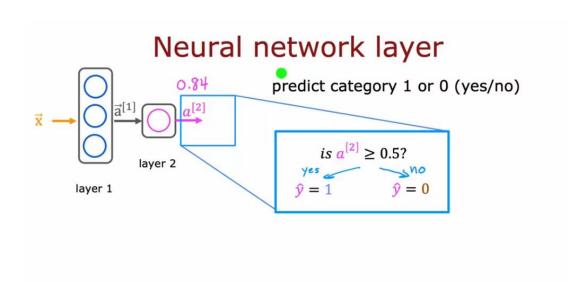
Windows'u etkinleştirmek için Ayarlar'a gidin.

\$\text{\$\text{\$\submitter{A}\text{}}}\$ I, \$\text{\$\text{\$\submitter{A}\text{}}}\$ and \$\text{\$

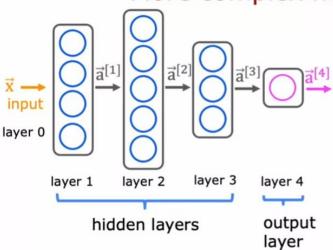


A üstü parantez sayı hangi katmana ait olduğunu gösterir.

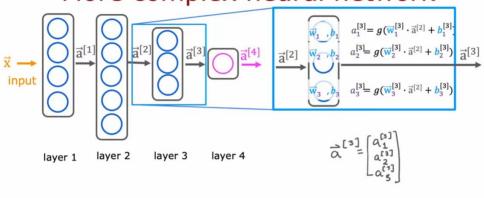


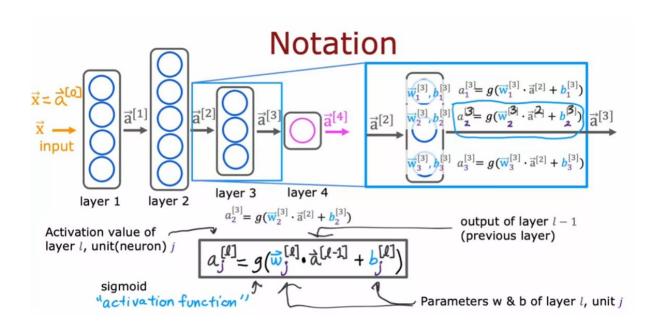


More complex neural network

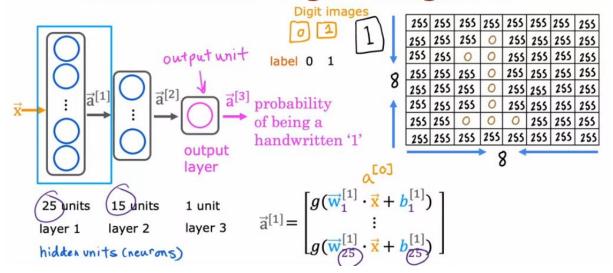


More complex neural network

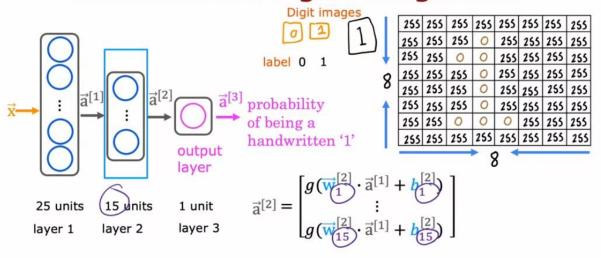




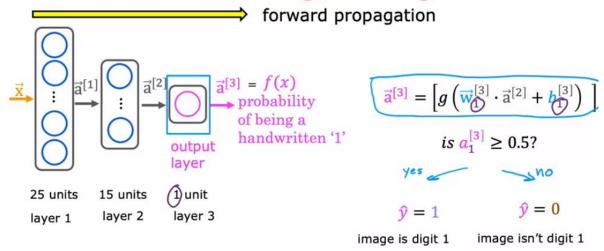
Handwritten digit recognition



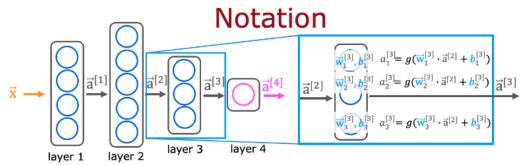
Handwritten digit recognition



Handwritten digit recognition



1.



$$a_j^{[l]} = g(\overrightarrow{\mathbf{w}}_j^{[l]} \cdot \overrightarrow{\mathbf{a}}^{[l-1]} + b_j^{[l]})$$

For a neural network, what is the expression for calculating the activation of the third neuron in layer 2? Note, this is different from the question that you saw in the lecture video.

$$\bigcap a_3^{[2]} = g(\vec{w}_3^{[2]} \cdot \vec{a}^{[2]} + b_3^{[2]})$$

$$\bigcap a_3^{[2]} = g(\vec{w}_2^{[3]} \cdot \vec{a}^{[2]} + b_2^{[3]})$$

$$\bigcirc \ a_3^{[2]} = g(\vec{w}_2^{[3]} \cdot \vec{a}^{[1]} + b_2^{[3]})$$

Handwritten digit recognition

 $\vec{a}^{[1]} \stackrel{\vec{a}^{[2]}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}}{\stackrel{\vec{a}^{[3]}}}{\stackrel{\vec{a}^{[3]}}}}{\stackrel{\vec{$

image is digit 1 image isn't digit 1

1 point

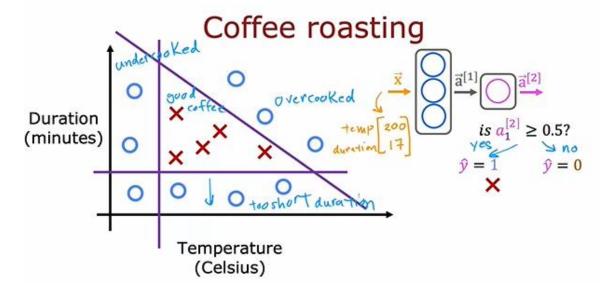
For the handwriting recognition task discussed in lecture, what is the output $a_1^{[3]}$?

- A vector of several numbers, each of which is either exactly 0 or 1
- A vector of several numbers that take values between 0 and 1
- A number that is either exactly 0 or 1, comprising the network's prediction
- The estimated probability that the input image is of a number 1, a number that ranges from 0 to 1.

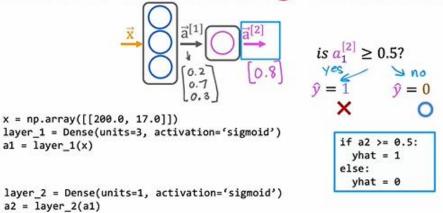
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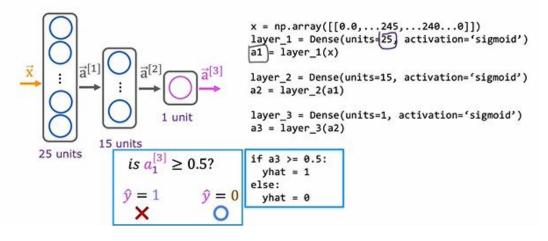
Inference in Code



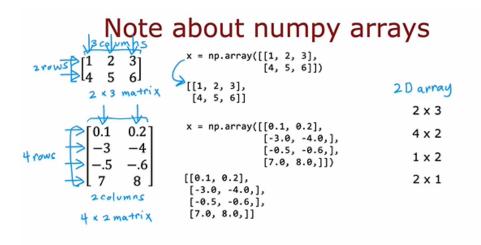
Build the model using TensorFlow



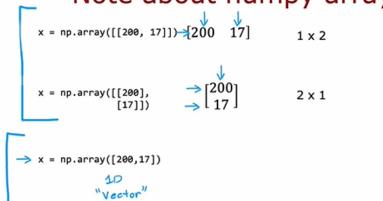
Model for digit classification



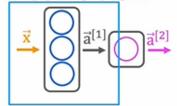
Data in Tensorflow



Note about numpy arrays



Activation vector



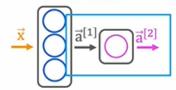
```
x = np.array([[200.0, 17.0]])
layer_1 = Dense(units=3, activation='sigmoid')
a1 = layer_1(x)

> [[0.2, 0.7, 0.3]] 1 x 3 matrix

tf.Tensor([[0.2 0.7 0.3]], shape=(1, 3), dtype=float32)

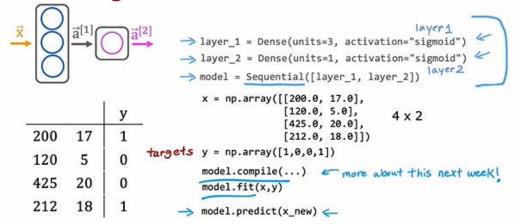
> a1.numpy()
array([[0.2, 0.7, 0.3]], dtype=float32)
```

Activation vector

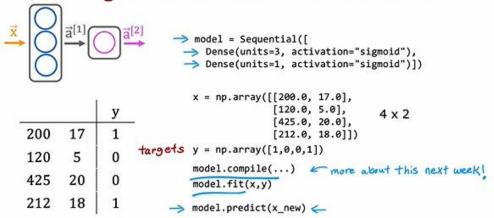


Building a Neural Network

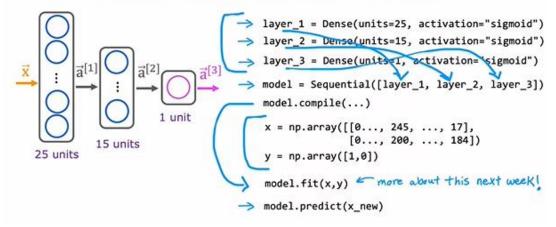
Building a neural network architecture



Building a neural network architecture



Digit classification model



model = Sequential([

Dense(units=25, activation="sigmoid"),

Dense(units=15, activation="sigmoid"),

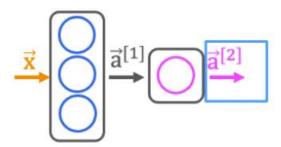
Dense(units=10, activation="sigmoid"),

Dense(units=1, activation="sigmoid")])

This code will define a neural network with how many layers?

- O 3
- 4
- O 25
- 0 5

2.



x = np.array([[200.0, 17.0]])
layer_1 = Dense(units=3, activation='sigmoid')
a1 = layer_1(x)

How do you define the second layer of a neural network that has 4 neurons and a sigmoid activation?

- O Dense(units=[4], activation=['sigmoid'])
- O Dense(units=4)
- Dense(units=4, activation='sigmoid')
- O Dense(layer=2, units=4, activation = 'sigmoid')

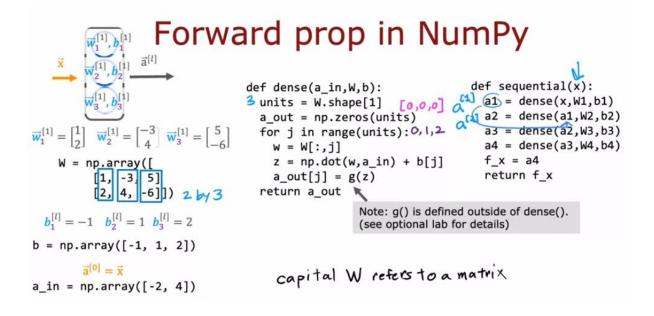
1 point

Forward prop in a single layer

forward prop (coffee roasting model)

```
a_1^{[2]} = g(\vec{\mathbf{w}}_1^{[2]} \cdot \vec{\mathbf{a}}^{[1]} + b_1^{[2]})
                                                          \rightarrow w2_1 = np.array([-7, 8, 9])
                                                          \rightarrow b2_1 = np.array([3])
                                                                                                               W<sub>1</sub> w<sub>2_1</sub>
                                                          \rightarrow z2_1 = np.dot(w2_1,a1)+b2_1
                                                          \rightarrow a2_1 = sigmoid(z2_1)
                                           10 arrays
x = np.array([200, 17])
                                                                               a_3^{[1]} = g(\vec{\mathbf{w}}_3^{[1]} \cdot \vec{\mathbf{x}} + b_3^{[1]})
a_1^{[1]} = g(\vec{\mathbf{w}}_1^{[1]} \cdot \vec{\mathbf{x}} + b_1^{[1]})
                                          a_2^{[1]} = g(\vec{\mathbf{w}}_2^{[1]} \cdot \vec{\mathbf{x}} + b_2^{[1]})
                                          w1_2 = np.array([-3, 4]) w1_3 = np.array([5, -6])
w1_1 = np.array([1, 2])
b1_1 = np.array([-1])
                                          b1_2 = np.array([1])
                                                                                    b1_3 = np.array([2])
z1_1 = np.dot(w1_1,x)+b1_1 z1_2 = np.dot(w1_2,x)+b1_2 z1_3 = np.dot(w1_3,x)+b1_3
                                                                                 a1_3 = sigmoid(z1_3)
                                       Ga1_2 = sigmoid(z1_2)
a1_1 = sigmoid(z1_1)
                                       = np.array([a1_1, a1_2, a1_3])
```

Genearal implemantation of forward propagation



1 point

forward prop (coffee roasting model)

```
a_1^{[2]} = g(\vec{\mathbf{w}}_1^{[2]} \cdot \vec{\mathbf{a}}^{[1]} + b_1^{[2]})
                                                            \rightarrow w2_1 = np.array([-7, 8, 9])
                                                           \rightarrow b2_1 = np.array([3])
                                                           \rightarrow z2_1 = np.dot(w2_1,a1)+b2_1
                                                           \rightarrowa2_1 = sigmoid(z2_1)
x = np.array([200, 17])
                                            10 arrays
a_1^{[1]} = g(\vec{\mathbf{w}}_1^{[1]} \cdot \vec{\mathbf{x}} + b_1^{[1]})
                                           a_2^{[1]} = g(\vec{\mathbf{w}}_2^{[1]} \cdot \vec{\mathbf{x}} + b_2^{[1]})
                                                                                    a_2^{[1]} = g(\vec{\mathbf{w}}_2^{[1]} \cdot \vec{\mathbf{x}} + b_2^{[1]})
w1_1 = np.array([1, 2])
                                          w1_2 = np.array([-3, 4])
                                                                                      w1_3 = np.array([5, -6])
b1_1 = np.array([-1])
                                          b1_2 = np.array([1])
                                                                                      b1_3 = np.array([2])
z1_1 = np.dot(w1_1,x)+b1_1 z1_2 = np.dot(w1_2,x)+b1_2 z1_3 = 

    a1_3 =

a1_1 = sigmoid(z1_1)
                                       a1_2 = sigmoid(z1_2)
                                      = np.array([a1_1, a1_2, a1_3])
```

According to the lecture, how do you calculate the activation of the third neuron in the first layer using NumPy?

O layer_1 = Dense(units=3, activation='sigmoid') a_1 = layer_1(x)

z1_3 = w1_3 * x + b a1_3 = sigmoid(z1_3)

z1_3 = np.dot(w1_3, x) + b1_3
a1_3 = sigmoid(z1_3)

2. 1point

```
Forward prop in NumPy

\vec{x}

\vec{w}_{1}^{(1)}, \vec{b}_{1}^{(1)}

\vec{w}_{2}^{(1)}, \vec{b}_{2}^{(1)}

\vec{w}_{1}^{(1)}, \vec{b}_{1}^{(1)}

\vec{w}_{1}^{(1)} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}

\vec{w}_{2}^{(1)} = \begin{bmatrix} -3 \\ 4 \end{bmatrix}

\vec{w}_{3}^{(1)} = \begin{bmatrix} 5 \\ -6 \end{bmatrix}

We np.array([

\begin{bmatrix} 1 \\ 1 \\ -3 \end{bmatrix}, \begin{bmatrix} 5 \\ 5 \\ 2 \end{bmatrix}, \begin{bmatrix} 4 \\ 4 \end{bmatrix}, \begin{bmatrix} -6 \end{bmatrix}])

\vec{w} = \vec{w}_{1}^{(1)} = \vec{w}_{2}^{(1)} = \vec{w}_{3}^{(1)} = \vec{w}_{3}^
```

According to the lecture, when coding up the numpy array W, where would you place the w parameters for each neuron?

- O In the rows of W.
- In the columns of W.

3.

1 point

Forward prop in NumPy

```
\vec{x} = \begin{bmatrix} \vec{x} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} & \vec{y} \\ \vec{y} & \vec{y} & \vec{y}
```

For the code above in the "dense" function that defines a single layer of neurons, how many times does the code go through the "for loop"? Note that W has 2 rows and 3 columns.

- O 2 times
- O 6 times
- 3 times
- 5 times

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