## **Neural Network Basics**

## LATEST SUBMISSION GRADE

100%

1. What does a neuron compute?

1 / 1 point

- A neuron computes a linear function (z = Wx + b) followed by an activation function
- $\bigcirc$  A neuron computes an activation function followed by a linear function (z = Wx + b)
- A neuron computes the mean of all features before applying the output to an activation function
- A neuron computes a function g that scales the input x linearly (Wx + b)



Correct, we generally say that the output of a neuron is a = g(Wx + b) where g is the activation function (sigmoid, tanh, ReLU, ...).

2. Which of these is the "Logistic Loss"?

1 / 1 point

- $\bigcap \mathcal{L}^{(i)}(\hat{y}^{(i)}, y^{(i)}) = |y^{(i)} \hat{y}^{(i)}|$
- $\bigcap \mathcal{L}^{(i)}(\hat{y}^{(i)}, y^{(i)}) = max(0, y^{(i)} \hat{y}^{(i)})$
- $\bigcirc \ \mathcal{L}^{(i)}(\hat{y}^{(i)}, y^{(i)}) = \mid y^{(i)} \hat{y}^{(i)} \mid^2$

✓ Correct

Correct, this is the logistic loss you've seen in lecture!

3. Suppose img is a (32,32,3) array, representing a 32x32 image with 3 color channels red, green and blue. How do you reshape this into a column vector?

1 / 1 point

- x = img.reshape((32\*32\*3,1))
- x = img.reshape((32\*32,3))
- x = img.reshape((1,32\*32,\*3))

✓ Correct

<ol> <li>Consider the two following random arrays "a" an</li> </ol>	na "b"	:
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2	a = np.random.randn(2, 3) # a.shape = (2, 3) b = np.random.randn(2, 1) # b.shape = (2, 1) c = a + b	

What will be the shape of "c"?

- c.shape = (3, 2)
- c.shape = (2, 1)
- c.shape = (2, 3)
- The computation cannot happen because the sizes don't match. It's going to be "Error"!



Yes! This is broadcasting. b (column vector) is copied 3 times so that it can be summed to each column of a.

5. Consider the two following random arrays "a" and "b":

```
1 a = np.random.randn(4, 3) # a.shape = (4, 3)
2 b = np.random.randn(3, 2) # b.shape = (3, 2)
3 c = a*b
```

What will be the shape of "c"?

- $\bigcirc$  c.shape = (3, 3)
- The computation cannot happen because the sizes don't match. It's going to be "Error"!
- c.shape = (4,2)
- c.shape = (4, 3)



## ✓ Correct

Indeed! In numpy the "\*" operator indicates element-wise multiplication. It is different from "np.dot()". If you would try "c = np.dot(a,b)" you would get c.shape = (4, 2).

6.	Suppose you have $n_x$ input features per example. Recall that $X=[x^{(1)}x^{(2)}x^{(m)}]$ . What is the dimension of X?	1/1 point
	$\bigcirc$ $(m,1)$ $\checkmark$ Correct	
7.	Recall that "np.dot(a,b)" performs a matrix multiplication on a and b, whereas "a*b" performs an element-wise multiplication.	1/1 point
	Consider the two following random arrays "a" and "b":  1	
	What is the shape of c?  c.shape = (12288, 45)  The computation cannot happen because the sizes don't match. It's going to be "Error"!  c.shape = (150,150)  c.shape = (12288, 150)	
	Correct Correct, remember that a np.dot(a, b) has shape (number of rows of a, number of columns of b). The sizes match because: "number of columns of a = 150 = number of rows of b"	

8. Consider the following code snippet:

How do you vectorize this?

- $\bigcirc$  c = a + b.T
- $\bigcirc$  c = a.T + b
- $\bigcirc$  c = a.T + b.T
- $\bigcirc$  c=a+b



9. Consider the following code:

1/1 point

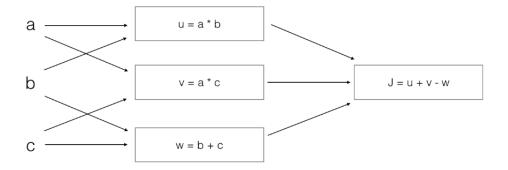
1/1 point

```
1 a = np.random.randn(3, 3)
2 b = np.random.randn(3, 1)
3 c = a*b
```

What will be c? (If you're not sure, feel free to run this in python to find out).

- This will invoke broadcasting, so b is copied three times to become (3,3), and \* is an element-wise product so c.shape will be (3, 3)
- O This will invoke broadcasting, so b is copied three times to become (3, 3), and \* invokes a matrix multiplication operation of two 3x3 matrices so c.shape will be (3, 3)
- This will multiply a 3x3 matrix a with a 3x1 vector, thus resulting in a 3x1 vector. That is, c.shape = (3,1).
- It will lead to an error since you cannot use "\*" to operate on these two matrices. You need to instead use np.dot(a,b)





What is the output J?

$$\int J = (c - 1)*(b + a)$$

$$\int J = a*b + b*c + a*c$$

$$\int J = (b - 1) * (c + a)$$

## ✓ Correct

Yes. 
$$J = u + v - w = a*b + a*c - (b + c) = a*(b + c) - (b + c) = (a - 1)*(b + c).$$