

✔ 축하합니다! 통과하셨습니다!

받은 학점 100% 최신 제출물 학점 100% 통과 점수: 80% 이상

다음 항목으로 이동

1. What does a neuron compute?

1 / 1점

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

↗ 더 보기

✔ 맞습니다

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 / 1점

- ☐  $\mathcal{L}^{(i)}(\hat{y}^{(i)}, y^{(i)}) = \max(0, y^{(i)} - \hat{y}^{(i)})$
- ☒  $\mathcal{L}^{(i)}(\hat{y}^{(i)}, y^{(i)}) = -(y^{(i)} \log(\hat{y}^{(i)}) + (1 - y^{(i)}) \log(1 - \hat{y}^{(i)}))$
- ☐  $\mathcal{L}^{(i)}(\hat{y}^{(i)}, y^{(i)}) = |y^{(i)} - \hat{y}^{(i)}|$
- ☐  $\mathcal{L}^{(i)}(\hat{y}^{(i)}, y^{(i)}) = |y^{(i)} - \hat{y}^{(i)}|^2$

↗ 더 보기

✔ 맞습니다

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

3. Consider the Numpy array  $x$ :

1 / 1점

$x = np.array([[[[1], [2]], [[3], [4]]]])$

What is the shape of  $x$ ?

- ☐ (4,)
- ☐ (1, 2, 2)
- ☐ (2, 2)
- ☒ (2, 2, 1)

↗ 더 보기

✔ 맞습니다

Yes. This array has two rows and in each row it has 2 arrays of 1x1.

4. Consider the following random arrays  $a$  and  $b$ , and  $c$ :

1 / 1점

$a = np.random.randn(2, 3) \# a.shape = (2, 3)$

$b = np.random.randn(2, 1) \# b.shape = (2, 1)$

$c = a + b$

What will be the shape of  $c$ ?

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

↗ 더 보기

✔ 맞습니다

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 / 1점

$a = np.random.randn(1, 3) \# a.shape = (1, 3)$

$b = np.random.randn(3, 3) \# b.shape = (3, 3)$

$c = a * b$

What will be the shape of  $c$ ?

- ☐ The computation cannot happen because the sizes don't match.
- ☒  $c.shape = (3, 3)$
- ☐  $c.shape = (1, 3)$
- ☐ The computation cannot happen because it is not possible to broadcast more than one dimension.

↗ 더 보기

✔ 맞습니다

Yes. Broadcasting allows row  $a$  to be multiplied element-wise with each row of  $b$  to form  $c$ .

6.

1/1점

Suppose you have  $n_x$  input features per example. If we decide to use row vectors  $\mathbf{x}_j$  for the features and

$$X = \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \\ \vdots \\ \mathbf{x}_m \end{bmatrix},$$

What is the dimension of  $X$ ?

- ☐  $(n_x, n_x)$
- ☐  $(1, n_x)$
- ☒  $(m, n_x)$
- ☐  $(n \times m)$

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↗ 더 보기

✔ 맞습니다

Yes. Each  $\mathbf{x}_j$  has dimension  $1 \times n_x$ ,  $X$  is built stacking all rows together into a  $m \times n_x$  array.

7. Consider the following array:

1/1점

$a = np.array([[2, 1], [1, 3]])$

What is the result of  $a * a$ ?

- ☐  $\begin{pmatrix} 2 & 6 \\ 5 & 10 \end{pmatrix}$
- ☐  $\begin{pmatrix} 5 & 5 \\ 5 & 10 \end{pmatrix}$
- ☐ The computation cannot happen because the sizes don't match. It's going to be an "Error"!
- ☒  $\begin{pmatrix} 4 & 1 \\ 1 & 9 \end{pmatrix}$

↗ 더 보기

✔ 맞습니다

Yes, recall that  $*$  indicates element-wise multiplication.

8. Consider the following code snippet:

1/1점

$a.shape = (3, 4)$

$b.shape = (4, 1)$

for  $i$  in range(3):

for  $j$  in range(4):

$c[i][j] = a[i][j] * b[j]$

How do you vectorize this?

- ☐  $c = a * b$
- ☐  $c = np.dot(a, b)$
- ☐  $c = a.T * b$
- ☒  $c = a * b.T$

↗ 더 보기

✔ 맞습니다

Yes,  $b.T$  gives a column vector with shape  $(1, 4)$ . The result of  $c$  is equivalent to broadcasting  $a * b.T$ .

9. Consider the following code:

1/1점

```
a = np.random.randn(3, 3)
```

```
b = np.random.randn(3, 1)
```

```
c = a * b
```

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

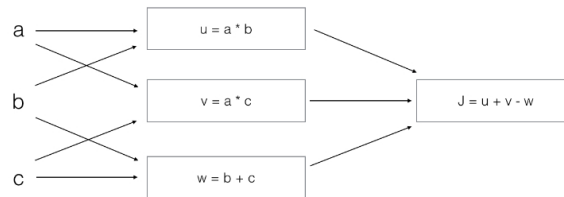
- ☐ It will lead to an error since you cannot use  $*$  to operate on these two matrices. You need to instead use  $\text{np.dot}(a,b)$
- ☐ This will multiply a  $3 \times 3$  matrix  $a$  with a  $3 \times 1$  vector, thus resulting in a  $3 \times 1$  vector. That is,  $c.\text{shape} = (3, 1)$ .
-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

더 보기

✔ 맞습니다

10. Consider the following computation graph.

1/1점



What is the output  $J$ ?

- ☐  $J = a * b + b * c + a * c$
- ☒  $J = (a - 1) * (b + c)$
- ☐  $J = (c - 1) * (b + a)$
- ☐  $J = (b - 1) * (c + a)$

더 보기

✔ 맞습니다

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