

In logistic regression given \mathbf{x} and parameters $w \in \mathbb{R}^{n_x}$, $b \in \mathbb{R}$. Which of the following best expresses what we want \hat{y} to tell us?

1 point

☐ $P(y = \hat{y}|\mathbf{x})$

☐ $\sigma(W \mathbf{x})$

☐ $\sigma(W \mathbf{x} + b)$

☐ $P(y = 1|\mathbf{x})$

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

1 point

☐ $L^{(i)}(\hat{y}^{(i)}, y^{(i)}) = -(y^{(i)} \log(\hat{y}^{(i)}) + (1 - y^{(i)}) \log(1 - \hat{y}^{(i)}))$

☐ $L^{(i)}(\hat{y}^{(i)}, y^{(i)}) = |y^{(i)} - \hat{y}^{(i)}|$

☐ $L^{(i)}(\hat{y}^{(i)}, y^{(i)}) = \max(0, y^{(i)} - \hat{y}^{(i)})$

☐ $L^{(i)}(\hat{y}^{(i)}, y^{(i)}) = |y^{(i)} - \hat{y}^{(i)}|^2$

3. Consider the Numpy array x :

1 point

```
 $x = \text{np.array}([[[1], [2]], [[3], [4]]])$ 
```

What is the shape of x ?

☐ $(4,)$

☐ $(2, 2, 1)$

☐ $(1, 2, 2)$

☐ $(2, 2)$

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

1 point

```
 $a = \text{np.random.randn}(2, 3) \# a.\text{shape} = (2, 3)$ 
```

```
 $b = \text{np.random.randn}(2, 1) \# b.\text{shape} = (2, 1)$ 
```

```
 $c = a + b$ 
```

What will be the shape of c ?

☐ The computation cannot happen because the sizes don't match. It's going to be "Error"!

☐ $c.\text{shape} = (2, 3)$

☐ $c.\text{shape} = (2, 1)$

☐ $c.\text{shape} = (3, 2)$

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

1 point

```
 $a = \text{np.random.randn}(4, 3) \# a.\text{shape} = (4, 3)$ 
```

```
 $b = \text{np.random.randn}(3, 2) \# b.\text{shape} = (3, 2)$ 
```

```
 $c = a * b$ 
```

What will be the shape of c ?

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

1 point

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

$$\text{and } X = \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \\ \vdots \\ \mathbf{x}_m \end{bmatrix}.$$

What is the dimension of X ?

- ☐ (n_x, n_x)
- ☐ (m, n_x)
- ☐ (n_x, m)
- ☐ $(1, n_x)$

7. Recall that `np.dot(a, b)` performs a matrix multiplication on a and b , whereas $a * b$ performs an element-wise multiplication.

1 point

Consider the two following random arrays a and b :

`a = np.random.randn(12288, 150) # a.shape = (12288, 150)`

`b = np.random.randn(150, 45) # b.shape = (150, 45)`

`c = np.dot(a, b)`

What is the shape of c ?

- ☐ The computation cannot happen because the sizes don't match. It's going to be "Error"!
- ☐ `c.shape = (150,150)`
- ☐ `c.shape = (12288, 150)`
- ☐ `c.shape = (12288, 45)`

1 point

8. Consider the following code snippet:

`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?

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

9. Consider the following code:

1 point

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

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

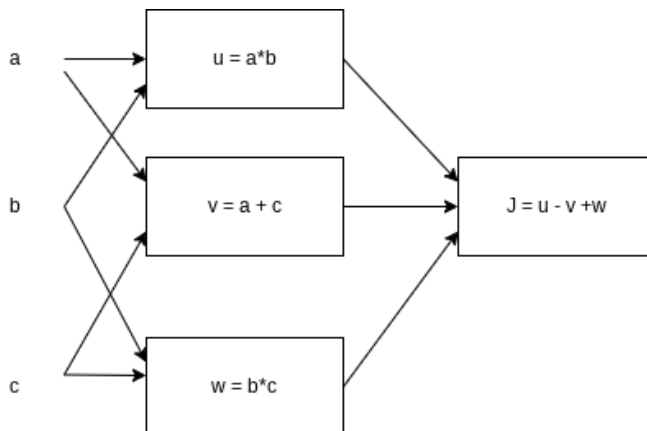
$c = a * b$

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 $np.dot(a, b)$
- ☐ 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 $(3, 3)$
- ☐ This will invoke broadcasting, so b is copied three times to become $(3, 3)$, and $*$ invokes a matrix multiplication operation of two 3×3 matrices so $c.shape$ will be $(3, 3)$
- ☐ This will multiply a 3×3 matrix a with a 3×1 vector, thus resulting in a 3×1 vector. That is, $c.shape = (3, 1)$.

10. Consider the following computational graph.

1 point



What is the output of J ?

- ☐ $(c - 1)(a + c)$
- ☐ $ab + bc + ac$
- ☐ $(a + c)(b - 1)$
- ☐ $(a - 1)(b + c)$