

## Question

What are the parameters of logistic regression?

- ☐ W and b, both  $n_x$  dimensional vectors.
- ☐ W and b, both real numbers.
- ☒ W, an  $n_x$  dimensional vector, and b, a real number.
- ☐ W, an identity vector, and b, a real number.

✓ Correct

What is the difference between the cost function and the loss function for logistic regression?

- ☐ The cost function computes the error for a single training example; the loss function is the average of the cost functions of the entire training set.
- ☒ The loss function computes the error for a single training example; the cost function is the average of the loss functions of the entire training set.
- ☐ They are different names for the same function.

✓ **Correct**

---

True or false. A convex function always has multiple local optima.

☐ True

☒ False

✓ **Correct**

One step of \_\_\_\_\_ propagation on a computation graph yields derivative of final output variable.

☐ Forward

☒ Backward

In this class, what does the coding convention `dvar` represent?

- ☐ The derivative of any variable used in the code.
- ☐ The derivative of input variables with respect to various intermediate quantities.
- ☒ The derivative of a final output variable with respect to various intermediate quantities.

✓ **Correct**

In this video, what is the simplified formula for the derivative of the loss with respect to  $z$ ?

- ☐  $a / (1-a)$
- ☒  $a - y$
- ☐  $a(1 - y)$

✓ Correct

1. What does a neuron compute?

1 / 1 point

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



Correct

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

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



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))`
- ☐ `x = img.reshape((1,32*32,*3))`
- ☒ `x = img.reshape((32*32*3,1))`
- ☐ `x = img.reshape((3,32*32))`

✓ Correct

4. Consider the two following random arrays `a` and `b`:

1 / 1 point

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



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

1 / 1 point

```
a = np.random.randn(4,3) # a.shape = (4,3)
```

```
b = np.random.randn(3,2) # b.shape = (3,2)
```

```
c = a * b
```

What will be the shape of  $c$ ?

☐  $c.shape = (3, 3)$

☐  $c.shape = (4, 2)$

☐  $c.shape = (4, 3)$

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

✓ 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  $X = [x^{(1)} x^{(2)} \dots x^{(m)}]$ . What is the dimension of  $X$ ?

1 / 1 point

☐  $(1, m)$

☐  $(m, n_x)$

☐  $(m, 1)$

☒  $(n_x, m)$

1 / 1 point

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

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  $c$ ?

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

✓ 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:

```
# 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 = a.T + b`

☐ `c = a.T + b.T`

☒ `c = a + b.T`

 **Correct**

9. Consider the following code:

1 / 1 point

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

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

```
c = a * b
```

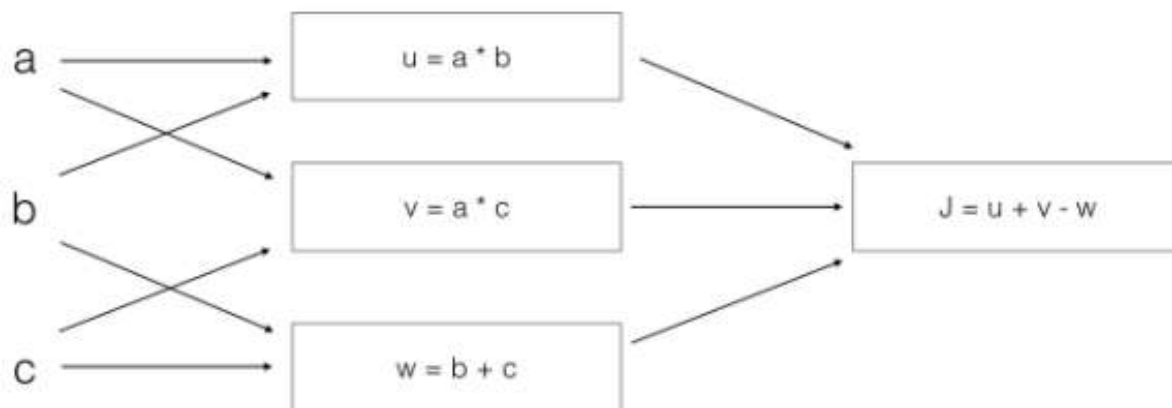
What will be *c*? (If you're not sure, feel free to run this in python to find out).

- ☐ 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 \* invokes a matrix multiplication operation of two 3x3 matrices so *c.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.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).

✓ Correct

10. Consider the following computation graph.

1 / 1 point



What is the output  $J$ ?

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

✓ Correct

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