

Neural Network Basics

1. 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 / 1 point

- ☐ $\sigma(W \mathbf{x})$
- ☐ $P(y = \hat{y} | \mathbf{x})$
- ☒ $P(y = 1 | \mathbf{x})$
- ☐ $\sigma(W \mathbf{x} + b)$

 Expand

✓ Correct

Yes. We want the output \hat{y} to tell us the probability that $y = 1$ given x .

1. In logistic regression given the input \mathbf{x} , and parameters $w \in \mathbb{R}^{n_x}, b \in \mathbb{R}$, how do we generate the output \hat{y} ?

1 / 1 point

- ☐ $\sigma(W \mathbf{x})$
- ☐ $\tanh(W \mathbf{x} + b)$
- ☐ $W \mathbf{x} + b$
- ☒ $\sigma(W \mathbf{x} + b)$.

 Expand

✓ Correct

Right, in logistic regression we use a linear function $W \mathbf{x} + b$ followed by the sigmoid function σ , to get an output y , referred to as \hat{y} , such that $0 < \hat{y} < 1$.

2. Suppose that $\hat{y} = 0.9$ and $y = 1$. What is the value of the "Logistic Loss"? Choose the best option.

1 / 1 point

- ☐ $+\infty$
- ☐ $\mathcal{L}(\hat{y}, y) = -(\hat{y} \log y + (1 - \hat{y}) \log(1 - y))$
- ☒ 0.105
- ☐ 0.005

 Expand

✓ Correct

Yes. Since $\mathcal{L}(\hat{y}, y) = -(y \log \hat{y} + (1 - y) \log(1 - \hat{y}))$, for the given values we get $\mathcal{L}(\hat{y}, y) = -(1 \log 0.9 + 0 \log 0.1)$

2. Suppose that $\hat{y} = 0.5$ and $y = 0$. What is the value of the "Logistic Loss"? Choose the best option.

1 / 1 point

- ☐ $\mathcal{L}(\hat{y}, y) = -(y \log \hat{y} + (1 - y) \log(1 - \hat{y}))$
- ☒ 0.693
- ☐ $+\infty$
- ☐ 0.5

 Expand

✓ Correct

Yes. Given the values of \hat{y} and y we get $\mathcal{L}(0.5, 0) = -(0 \log 0.5 + 1 \log(0.5)) \approx 0.693$.

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 `x`?

1 / 1 point

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

 Expand

✓ Correct

3. Consider the Numpy array x :

1 / 1 point

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

What is the shape of x ?

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

 Expand

 **Correct**

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 / 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 ?

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

 Expand

 **Correct**

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

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

1 / 1 point

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

```
b = np.random.randn(1, 4) # b.shape = (1, 4)
```

```
c = a + b
```

What will be the shape of c ?

- ☐ c.shape = (1, 4)
- ☐ c.shape = (3, 1)
- ☐ The computation cannot happen because it is not possible to broadcast more than one dimension.
- ☒ c.shape = (3, 4)

 Expand

 **Correct**

Yes. Broadcasting is used, so row b is copied 3 times so it can be summed to each row of a .

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

1 / 1 point

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

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

```
c = a + b
```

What will be the shape of c ?

- ☐ c.shape = (2, 3, 3)
- ☒ The computation cannot happen because it is not possible to broadcast more than one dimension
- ☐ c.shape = (2, 1)
- ☐ c.shape = (3, 3)

 Expand

 **Correct**

Yes. It is not possible to broadcast together a and b . In this case there is no way to generate copies of one of the arrays to match the size of the other.

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

1 / 1 point

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

$b = \text{np.random.randn}(1, 3) \# b.\text{shape} = (1, 3)$

$c = a * b$

What will be the shape of c ?

- ☐ The computation cannot happen because it is not possible to broadcast more than one dimension.
- ☒ $c.\text{shape} = (4, 3)$
- ☐ $c.\text{shape} = (1, 3)$
- ☐ The computation cannot happen because the sizes don't match.

 Expand

 Correct

Yes. Broadcasting is invoked, so row b is multiplied element-wise with each row of a to create c .

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

1 / 1 point

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

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

$c = a * b$

What will be the shape of c ?

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

 Expand

 Correct

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

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

1 / 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 ?

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

[Expand](#)

✓ 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. If we decide to use row vectors \mathbf{x}_j for the features and

1 / 1 point

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

What is the dimension of X ?

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

[Expand](#)

✓ Correct

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.

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

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

 Expand

 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 :

```
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 = (12288, 45)`
- ☐ `c.shape = (12288, 150)`
- ☐ `c.shape = (150, 150)`

 Expand

 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 "

7. Consider the following array:

1 / 1 point

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

What is the result of `np.dot(a, a)`?

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

Expand

✓ Correct

Yes, recall that `*` indicates the element wise multiplication and that `np.dot()` is the matrix multiplication. Thus $\begin{pmatrix} (2)(2) + (1)(1) & (2)(1) + (1)(3) \\ (1)(2) + (3)(1) & (1)(1) + (3)(3) \end{pmatrix}$.

8. Consider the following code snippet:

1 / 1 point

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

Expand

✓ Correct

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

8. Consider the following code snippet:

1 / 1 point

```
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.T`

☐ `c = a + b`

☒ `c = a + b.T`

☐ `c = a.T + b`

 Expand

 Correct

9. Consider the following arrays:

1 / 1 point

```
a = np.array([[1, 1], [1, -1]])
```

```
b = np.array([[2], [3]])
```

```
c = a + b
```

Which of the following arrays is stored in `c`?

$\begin{pmatrix} 4 & 4 \\ 5 & 2 \end{pmatrix}$

☐ $\begin{pmatrix} 3 & 4 \\ 3 & 2 \end{pmatrix}$

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

☒ $\begin{pmatrix} 3 & 3 \\ 4 & 2 \end{pmatrix}$

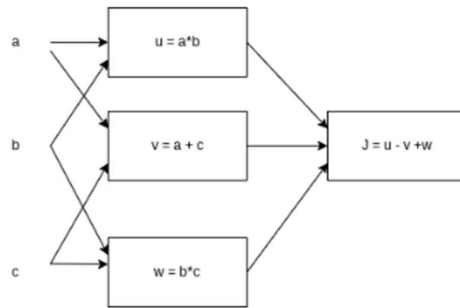
 Expand

 Correct

Yes. The array `b` is a column vector. This is copied two times and added to the array `a` to construct the array `c`.

10. Consider the following computational graph.

1 / 1 point



What is the output of J ?

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

[Expand](#)

✓ **Correct**

Yes.

$$J = u - v + w = ab - (a + c) + bc = ab - a + bc - c = a(b - 1) + c(b - 1) = (a + c)(b - 1)$$