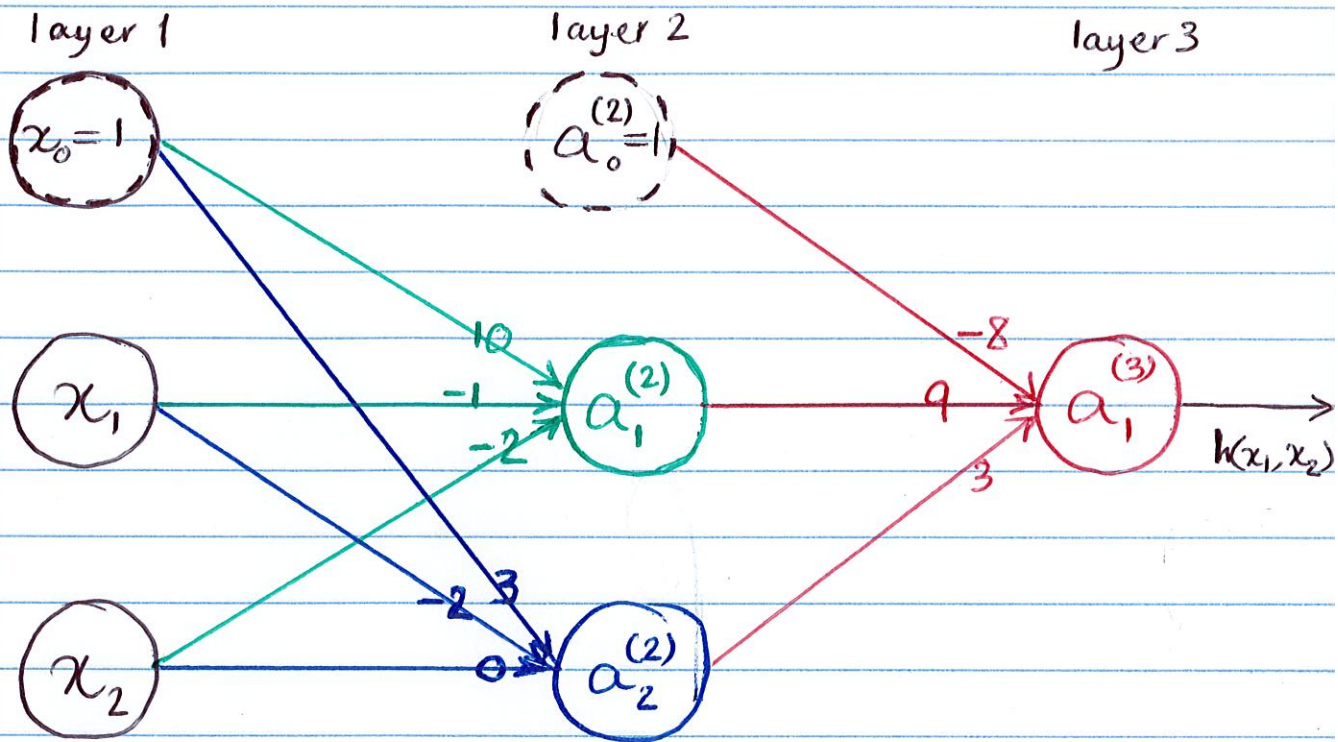


# forward propagation for neural networks — example 1 <sup>①</sup>

Given the following neural network:



where  $\Theta^{(1)} = \begin{bmatrix} \theta_{10}^{(1)} = 10 & \theta_{11}^{(1)} = -1 & \theta_{12}^{(1)} = -2 \\ \theta_{20}^{(1)} = 3 & \theta_{21}^{(1)} = -2 & \theta_{22}^{(1)} = 0 \end{bmatrix}$   $2 \times 3$

and  $\Theta^{(2)} = \begin{bmatrix} \theta_{10}^{(2)} = -8 & \theta_{11}^{(2)} = 9 & \theta_{12}^{(2)} = 3 \end{bmatrix}$   $1 \times 3$

What is  $h(x_1, x_2)$  when  $\begin{bmatrix} x_0=1 \\ x_1=2 \\ x_2=3 \end{bmatrix}$  ?

propagating from 1<sup>st</sup> layer to 2<sup>nd</sup> layer: ②

Let  $z_1^{(2)}$  be the value that goes into the 1<sup>st</sup> unit of the 2<sup>nd</sup> layer  
and let  $z_2^{(2)}$  be the value that goes into the 2<sup>nd</sup> unit of the 2<sup>nd</sup> layer

$$\begin{bmatrix} z_1^{(2)} \\ z_2^{(2)} \end{bmatrix} = \Theta^{(1)} * X = \begin{bmatrix} 10 & -1 & -2 \\ 3 & -2 & 0 \end{bmatrix}_{2 \times 3} * \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}_{3 \times 1} = \begin{bmatrix} 10(1) + (-1)2 + (-2)3 \\ (3)1 + (-2)2 + (0)3 \end{bmatrix}_{2 \times 1} = \begin{bmatrix} 2 \\ -1 \end{bmatrix}$$

$$\begin{bmatrix} a_1^{(2)} \\ a_2^{(2)} \end{bmatrix} = \begin{bmatrix} g(z_1^{(2)}) \\ g(z_2^{(2)}) \end{bmatrix} = \begin{bmatrix} \frac{1}{1 + e^{-(2)}} \\ \frac{1}{1 + e^{-(-1)}} \end{bmatrix} = \begin{bmatrix} 0.88 \\ 0.27 \end{bmatrix}_{2 \times 1}$$

Now, we can add  $a_0^{(2)} = 1$  and get  $\begin{bmatrix} a_0^{(2)} = 1 \\ a_1^{(2)} = 0.88 \\ a_2^{(2)} = 0.27 \end{bmatrix}$

So, as  $X = \begin{bmatrix} x_0 = 1 \\ x_1 = 2 \\ x_2 = 3 \end{bmatrix}$  propagates from 1<sup>st</sup> layer to 2<sup>nd</sup> layer

it will become  $a^{(2)} = \begin{bmatrix} a_0^{(2)} = 1 \\ a_1^{(2)} = 0.88 \\ a_2^{(2)} = 0.27 \end{bmatrix}$



propagating from 2<sup>nd</sup> layer to 3<sup>rd</sup> layer: ③

Let  $Z_1^{(3)}$  be the value that goes into the 1<sup>st</sup> unit of the 3<sup>rd</sup> layer

$$\begin{bmatrix} Z_1^{(3)} \end{bmatrix}_{1 \times 1} = \Theta^{(2)} * a^{(2)} = \begin{bmatrix} -8 & 9 & 3 \end{bmatrix}_{1 \times 3} * \begin{bmatrix} 1 \\ 0.88 \\ 0.27 \end{bmatrix}_{3 \times 1} = \begin{bmatrix} (-8)1 + (9)(0.88) + (3)(0.27) \end{bmatrix} = 0.73$$

$$\begin{bmatrix} a_1^{(3)} \end{bmatrix} = \begin{bmatrix} g(Z_1^{(3)}) \end{bmatrix} = \begin{bmatrix} \frac{1}{1 + e^{-(0.73)}} \end{bmatrix} = 0.67$$

This means  $h(x_1=2, x_2=3) = 0.67$

This means  $P(y=1) = 0.67$  for input point  $\begin{bmatrix} 2 = x_1 \\ 3 = x_2 \end{bmatrix}$

This means input point  $\begin{bmatrix} x_1=2 \\ x_2=3 \end{bmatrix}$  should be classified as belonging to class 1.

```

1 function h = probabilities(Theta1, Theta2, X)
2 %PREDICT Predict the label of an input given a trained neural network
3 % p = PREDICT(Theta1, Theta2, X) outputs the predicted label of X given
4 % the trained weights of a neural network (Theta1, Theta2)
5
6 m = size(X, 2);
7
8 a1 = X;
9 a2 = sigmoid(Theta1 * [ones(1, m); a1]);
10 a3 = sigmoid(Theta2 * [ones(1, m); a2]);
11
12 h = a3;
13 % =====
14
15 end
16
17
18
19 function g = sigmoid(z)
20 g = 1 ./ (1 + exp(-z));
21 end

```

## Command Window

New to MATLAB? See resources for [Getting Started](#).

```
>> X = [2; 3]
```

```
X =
```

```
2
3
```

```
>> Theta1 = [10 -1 -2; 3 -2 0]
```

```
Theta1 =
```

```
10 -1 -2
3 -2 0
```

```
>> Theta2 = [-8 9 3]
```

```
Theta2 =
```

```
-8 9 3
```

```
>> probabilities(Theta1, Theta2, X)
```

```
ans =
```

```
0.6757
```