

Multilayer Perceptron

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0.1 Sigmoid Function

```
[1]: #import libraries
import numpy as np
import matplotlib.pyplot as plt
```

```
[2]: # sigmoid activation function
def sigmoid(x):
    return 1/(1 + np.exp(-x))
```

0.2 Implementation of Multilayer Perceptron

```
[3]: #define weights and bias as w and b, x is input array, lastly 'a' being the node
x = np.array([0.5, 1])
w1 = np.array([[0.1, 0.2, 0.15], [0.6, 0.2, 0.8]])
b1 = np.array([0.1, 0.2, 0.05])
```

```
[4]: #calculate the dot product and sum with bias
a1 = np.dot(x, w1) + b1
a1
```

```
[4]: array([0.75 , 0.5  , 0.925])
```

```
[5]: #calculate the activation function for a1
z1 = sigmoid(a1)
z1
```

```
[5]: array([0.6791787 , 0.62245933, 0.71605979])
```

```
[6]: #define the weights for second node
w2 = np.array([[0.1, 0.2], [0.05, 0.5], [0.1, 0.2]])
w2
```

```
[6]: array([[0.1 , 0.2 ],
          [0.05, 0.5 ],
          [0.1 , 0.2 ]])
```

```
[7]: #define bias for second node
```

```
b2 = np.array([0.4, 0.5])  
b2
```

```
[7]: array([0.4, 0.5])
```

```
[8]: #calculate the dot product and sum with bias
```

```
a2 = np.dot(z1, w2) + b2  
a2
```

```
[8]: array([0.57064682, 1.09027736])
```

```
[9]: #calculate the activation function for a2
```

```
z2 = sigmoid(a2)  
z2
```

```
[9]: array([0.63891241, 0.74843395])
```

```
[10]: #define weight and bias for second node
```

```
w3 = np.array([[0.1, 0.2], [0.9, 0.8]])  
b3 = np.array([0.9, 0.9])
```

```
[11]: #calculate the activation function
```

```
a3 = np.dot(z2, w3) + b3  
a3
```

```
[11]: array([1.63748179, 1.62652964])
```

```
[12]: #calculate the activation function for a3
```

```
z3 = sigmoid(a3)  
z3
```

```
[12]: array([0.83719199, 0.83569368])
```