

# **Activation Function**



# **★**Title: Activation Function



### **Summary**

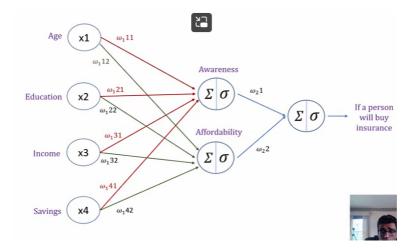
- Activation function and weighted sum of the input is in every neuron
- Activation function helps us to formulate the output to a specific range
- Activation function helps us to make the model non-linear
- ▼ Google Collab Practice ■
- ▼ Further research
- ▼ Resources

<u>Activation Functions | Deep Learning Tutorial 8 (Tensorflow Tutorial, Keras & Python) - YouTube</u>

## M

## Why Activation Function?

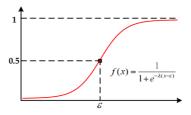
- Helpful in output layer
  - The sigmoid function helps us see if the neuron will fire or not (0 or 1?)
- Helpful in hidden layer
  - To make the neuron output non-linear



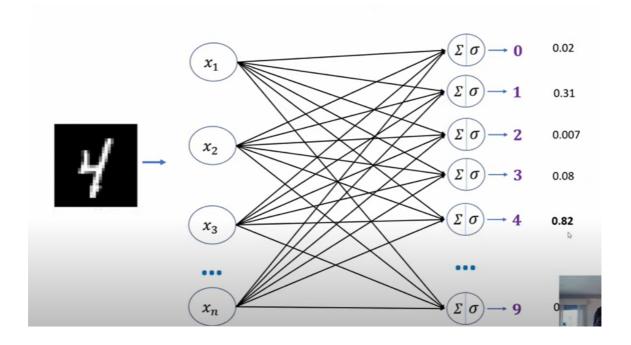
## **Examples of activation functions**

### **Generally in binary classifications:**

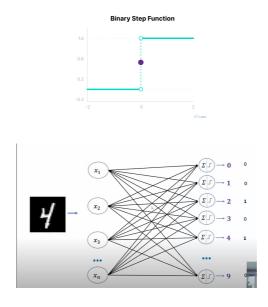
- Juse tanh in all hidden layers, and sigmoid in output layer
- will use **ReLU** is the default choice for hidden layer activation function
- ▼ sigmoid (number between 0 and 1)
  - Better than the step function because you choose the one with higher probability



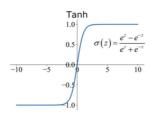
Activation Function 2



- Has the vanishing gradient problem as you get closer to the asymptote the learning starts to get so slow
- ▼ step function (0 or 1)

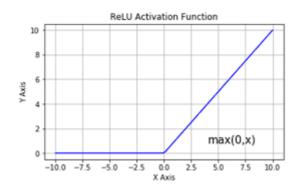


### **▼** tanh



• Has the vanishing gradient problem as you get closer to the asymptote the learning starts to get so slow

#### **▼** ReLU



- solves the **vanishing gradient problem** as its always increasing, but still same problem in decreasing.
- very light-weight function

### ▼ Leaky ReLU

• solves the vanishing gradient as you increase and decrease

Activation Function 4

