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# A beginner's guide to neural networks for social scientists

# Overview

- Neural Networks are a very powerful type of ML model.
- But their power is that they can be used for very complex problems.
- For simpler problems, simpler models are usually better.
- The simple neural networks we cover here are stepping stones to more complex types of NNs, such as:
  - Convolutional Neural Networks which are suited for processing images.
  - Recurrent Neural Network: Suited for handling time sequences of any length
  - Transformers: which are the core of GPT (ie: ChatGPT) and BERT (ie: Google Translate)

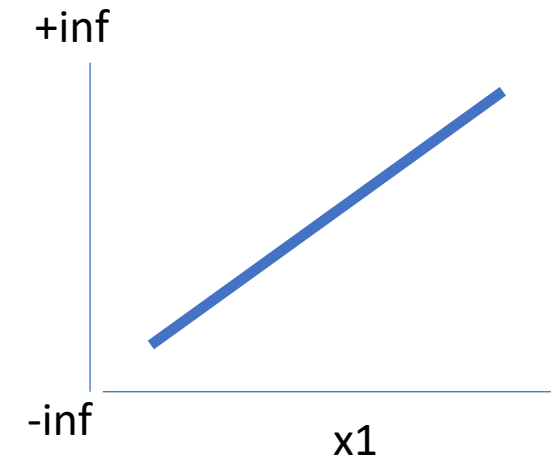
# Revision: What is ML

- Data about cases go in
- Predictions about those cases come out
- Regression = The prediction is a continuous number (ie: annual income)
- Classification = The prediction is a category, or the probability that the given case belongs to certain categories (ie: sub-species of Iris flower).

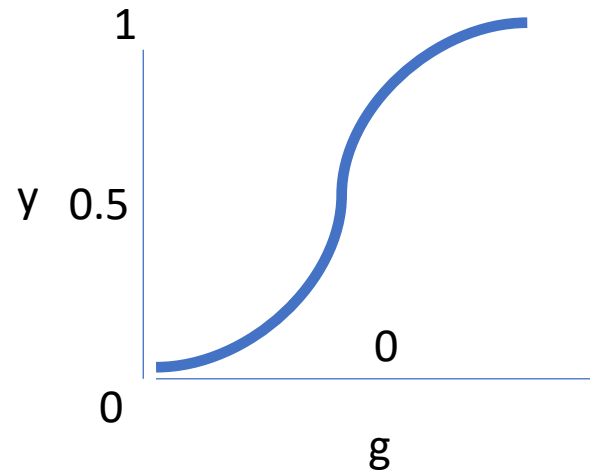
$$\hat{y} = f(X)$$

# Revision: Logistic Regression

$$g = x_1 \cdot w_1 + x_2 \cdot w_2 + \dots x_n \cdot w_n + \beta$$

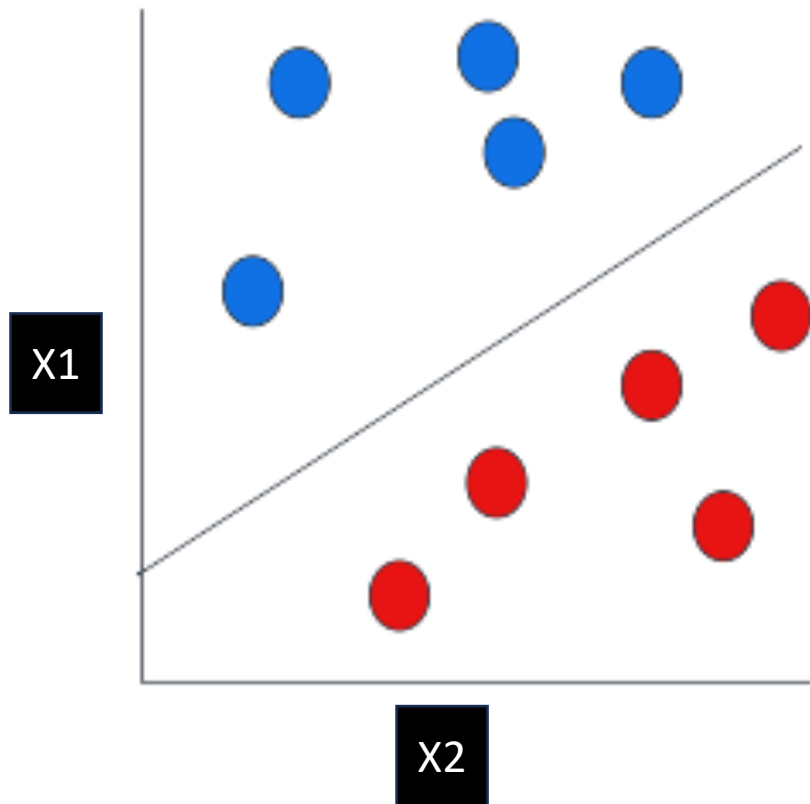


$$\hat{y} = \frac{1}{1 + e^{-(g-1)}}$$

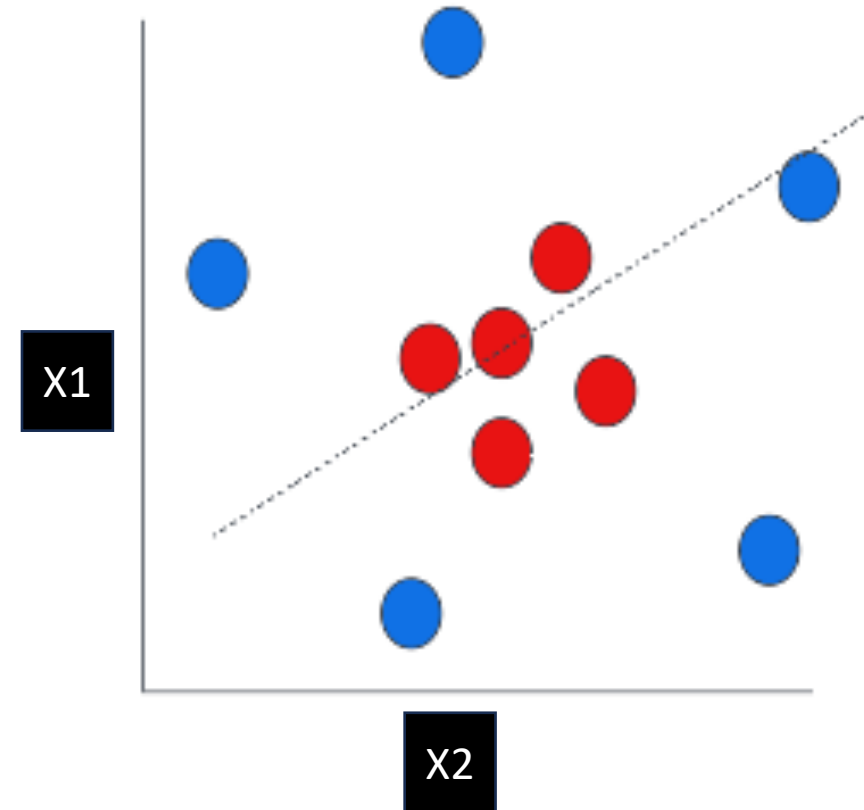


# Linearly Separable Problems

Linearly Separable



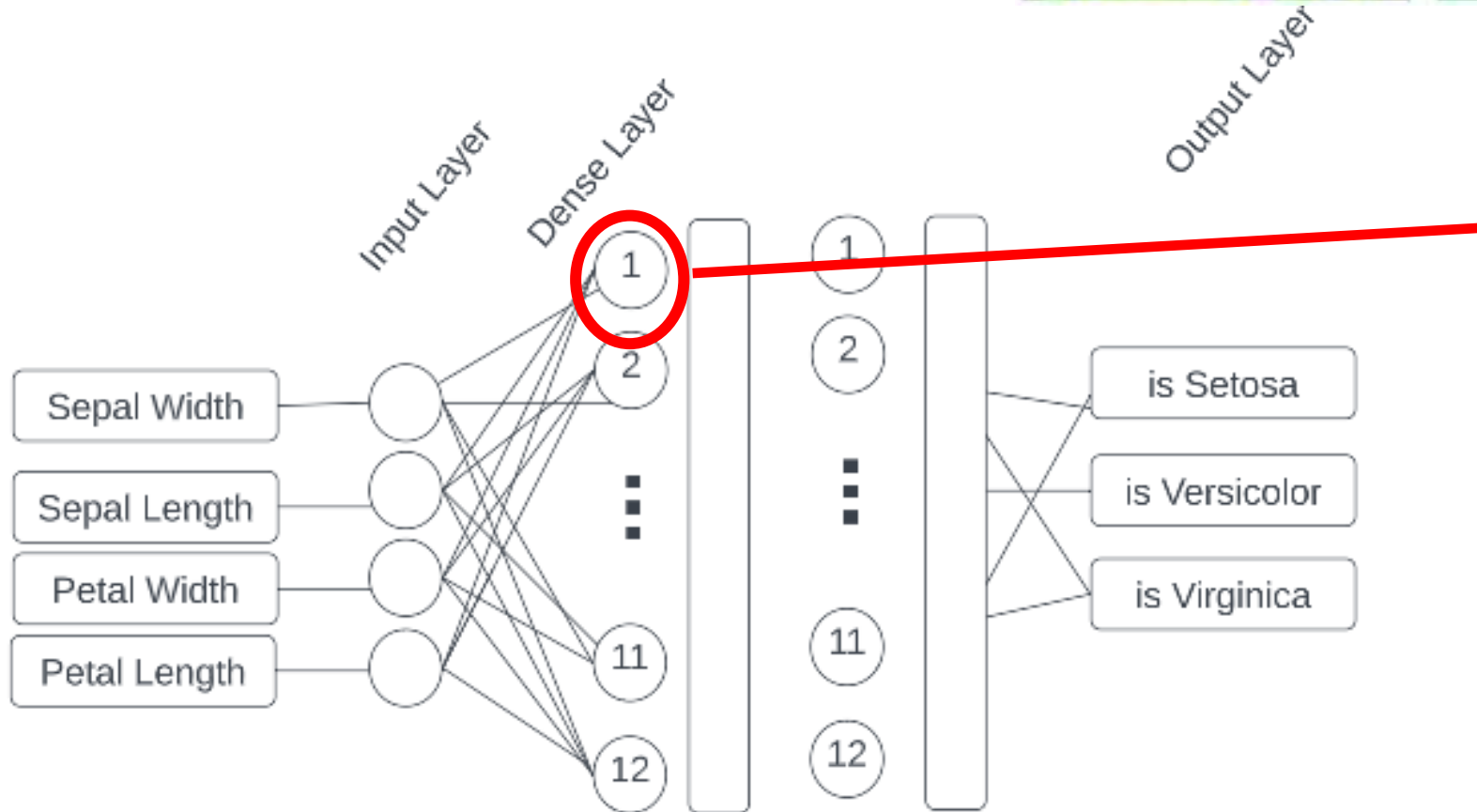
Non-Linearly Separable



# Neural Networks solve non-linear problems!



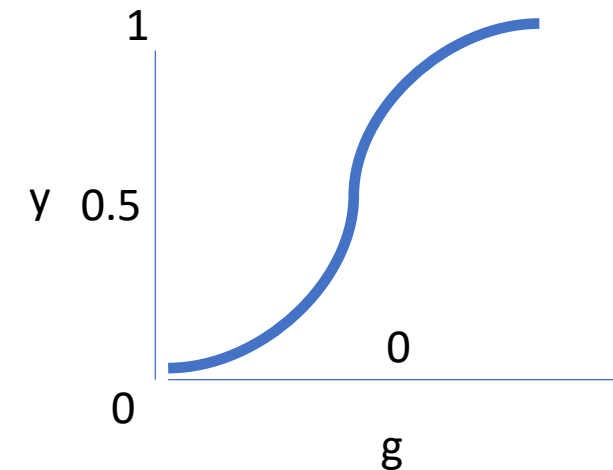
# Classifying Iris Flowers



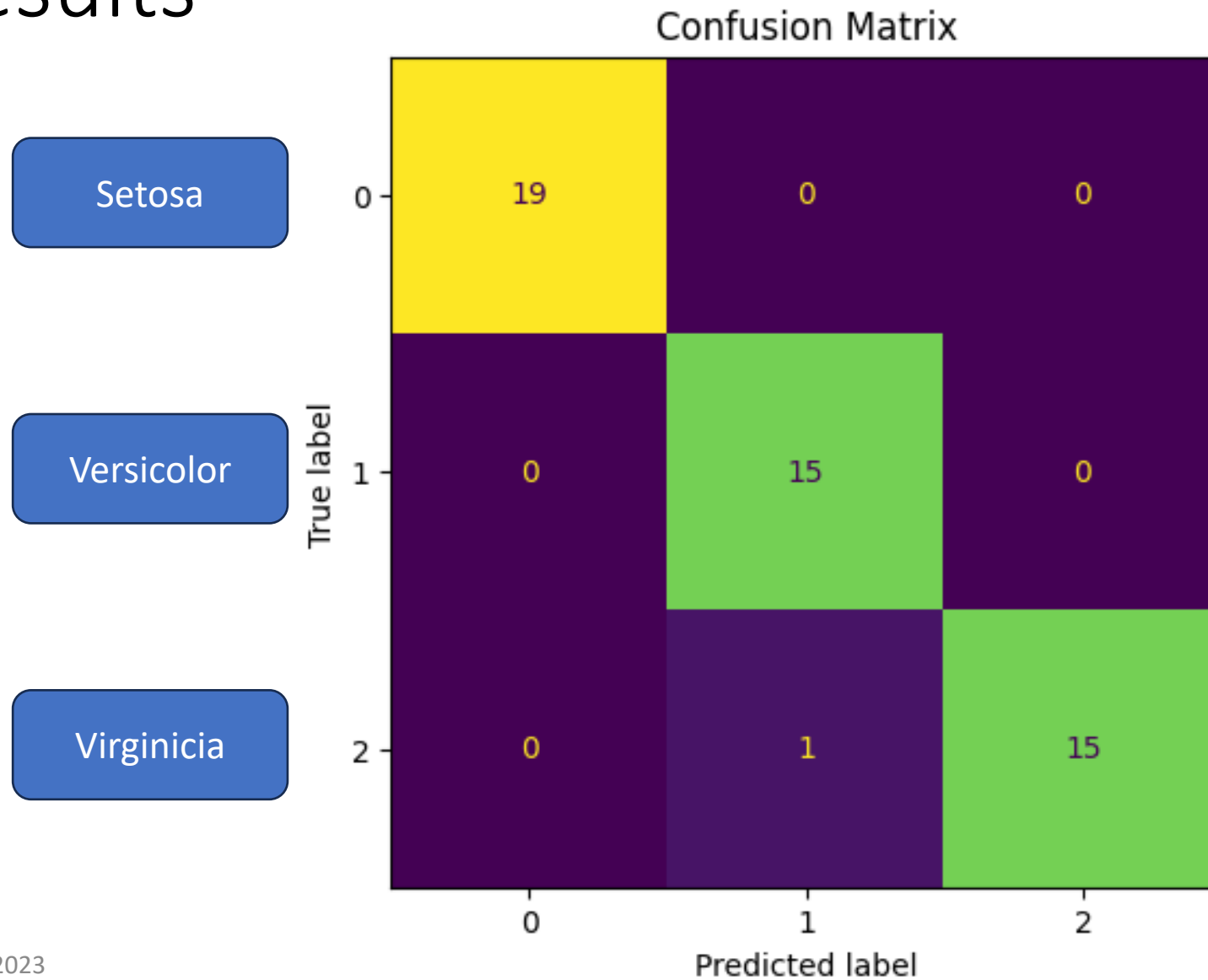
## Activation Functions

- $g = x_1 \cdot w_1 + x_2 \cdot w_2 + \dots x_n \cdot w_n + \beta$

- $y = \frac{1}{1 + e^{-(g-1)}}$

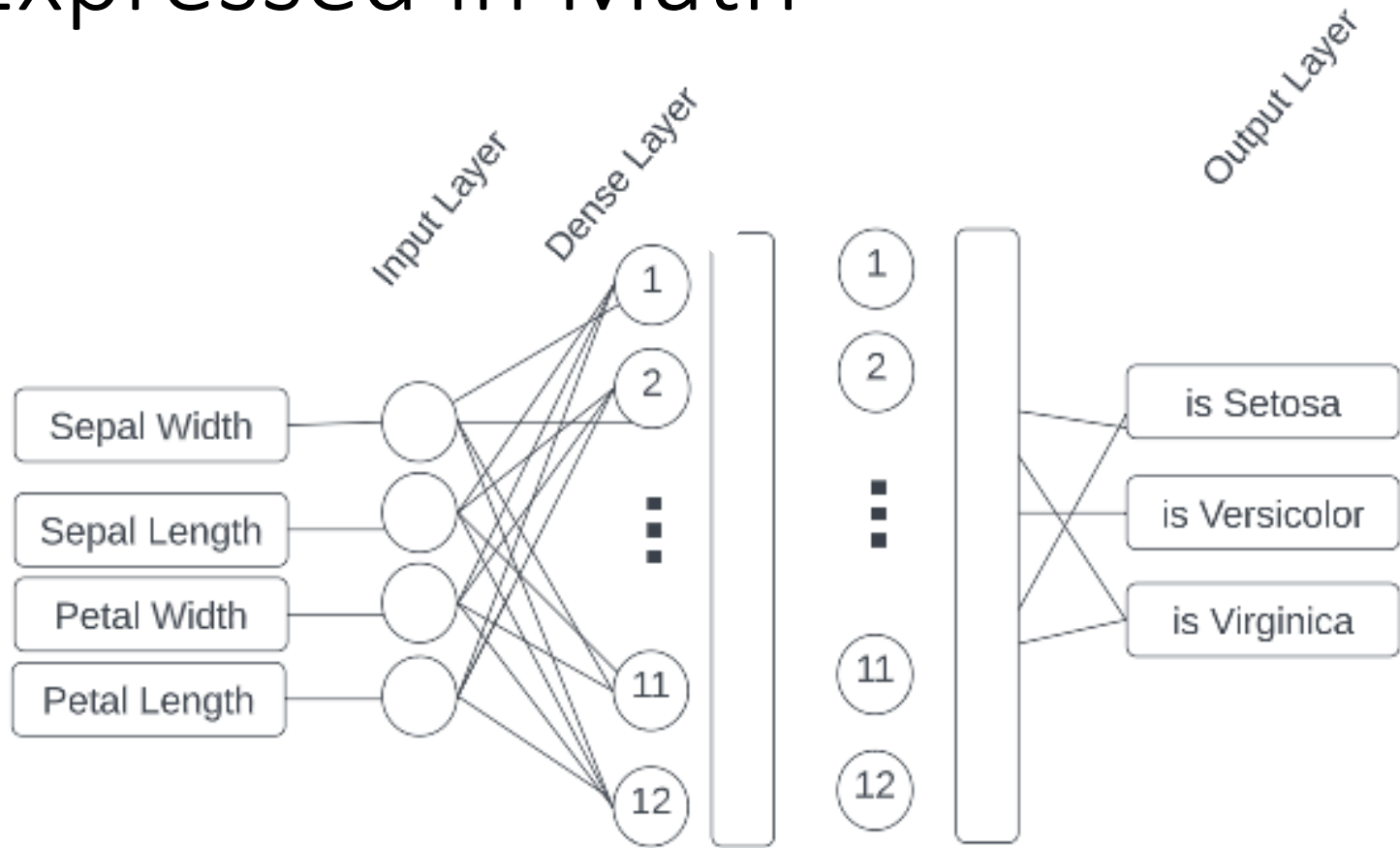


# Results

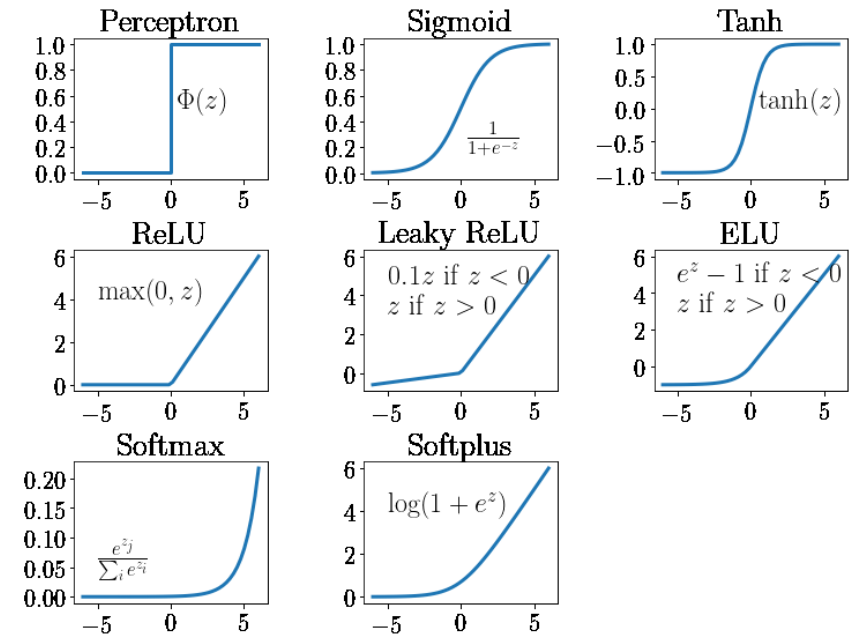




# Expressed in Math



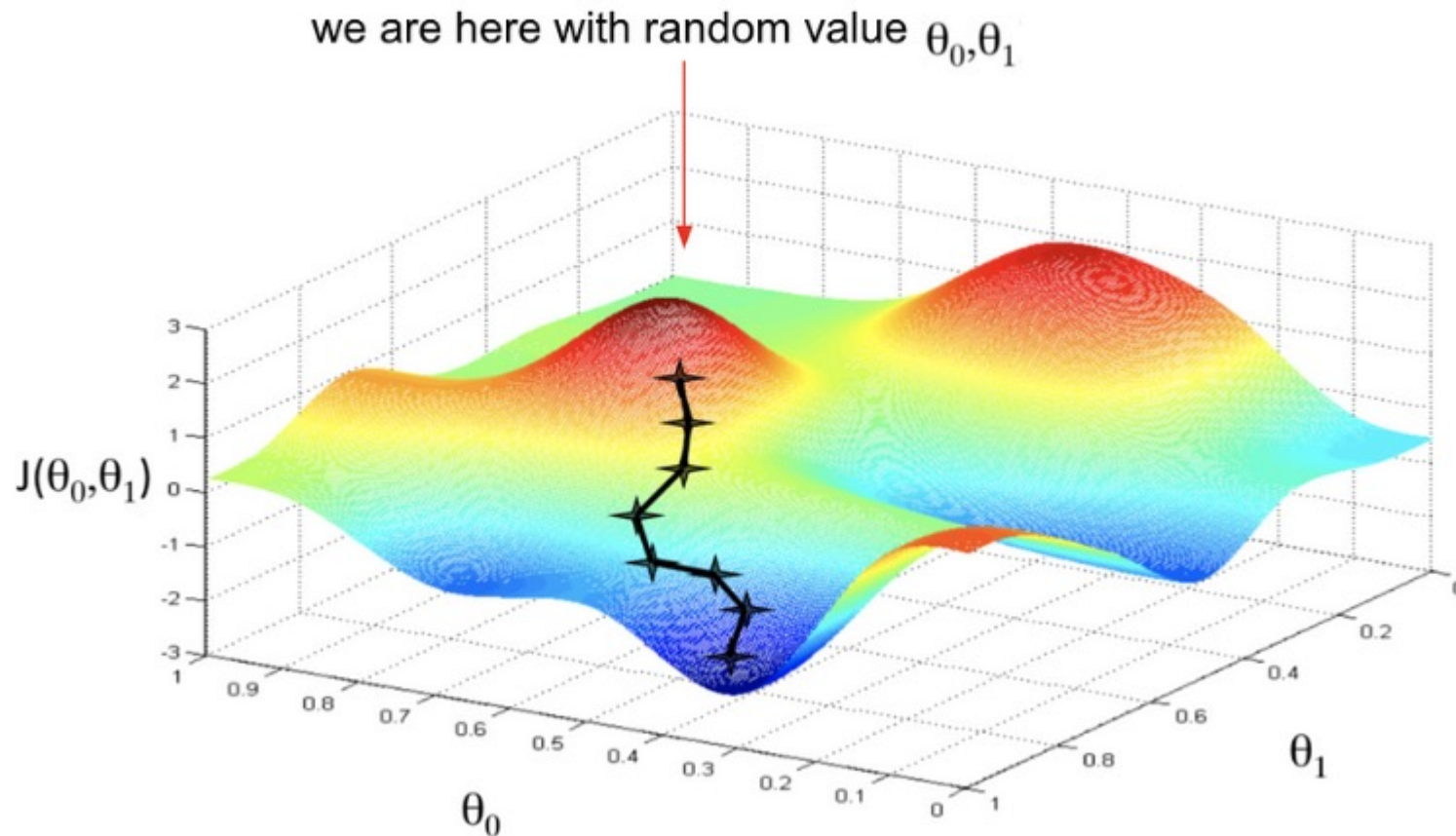
## Activation Functions



$$\hat{y} = \alpha(W_L \cdot X_{L-1} + \beta_L)$$

# How do we train neural networks?

- Trial and Error (almost) = Gradient Descent



# Neural Networks solve non-linear problems!



# How can I test ANNs myself?



# What have we learned?

- What types of problems neural networks are best at: complex problems, with many feature interactions
- The mathematics of Feed Forward Neural Networks
- A step towards understanding more complex neural networks