

# A Gentle Introduction to Deep Learning

Roberto Souza  
Associate Professor  
ESE - Schulich School of Engineering

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# Outline

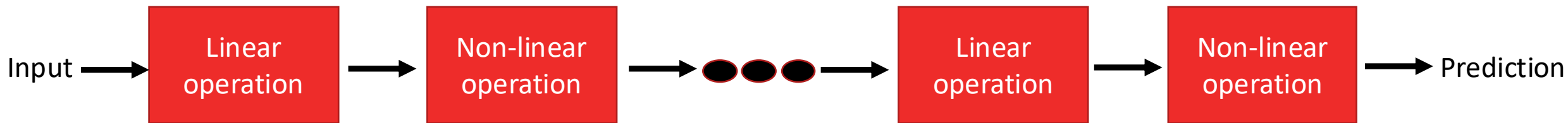
- Learning Goals
- Deep Learning Intuition
- Commonly Used Activations
- Summary

# Learning Goals

- Get an intuitive feel for how the majority of deep learning methods work
- Get familiar with different activation functions used in deep learning

# Deep Learning Intuition

- Alternated stack of linear and non-linear operations
- Non-linear operations that come immediately after a linear operation are called “activations”
- The activation at the end of the network determines if the model is a regression or classification network
- You can have two consecutive non-linear operations
- Two consecutive linear operations often do not make sense



# Deep Learning Intuition

$C = A \times B \rightarrow$  Equivalent linear operation

$$Y = \overbrace{A \times B} \times X$$

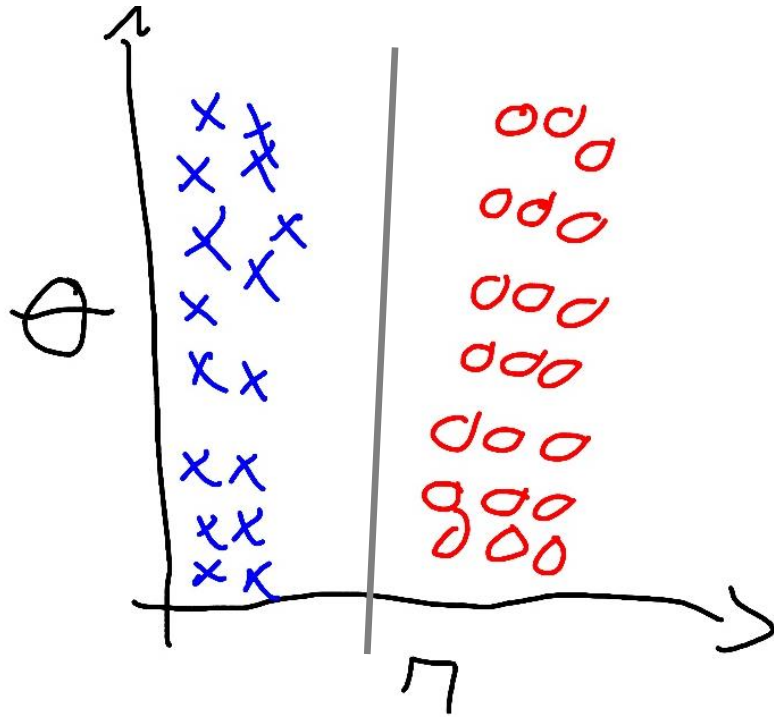
$X \rightarrow$  Input

$Y \rightarrow$  Output

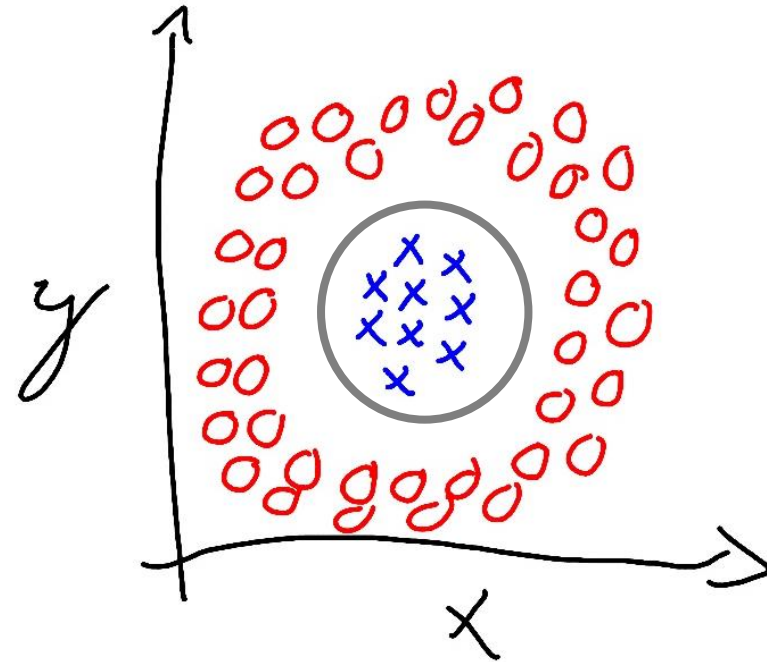
$A \rightarrow$  Linear Operation

$B \rightarrow$  Linear Operation

# Deep Learning Intuition



Linear model



Non-linear models allow you to get more complex decision boundaries.

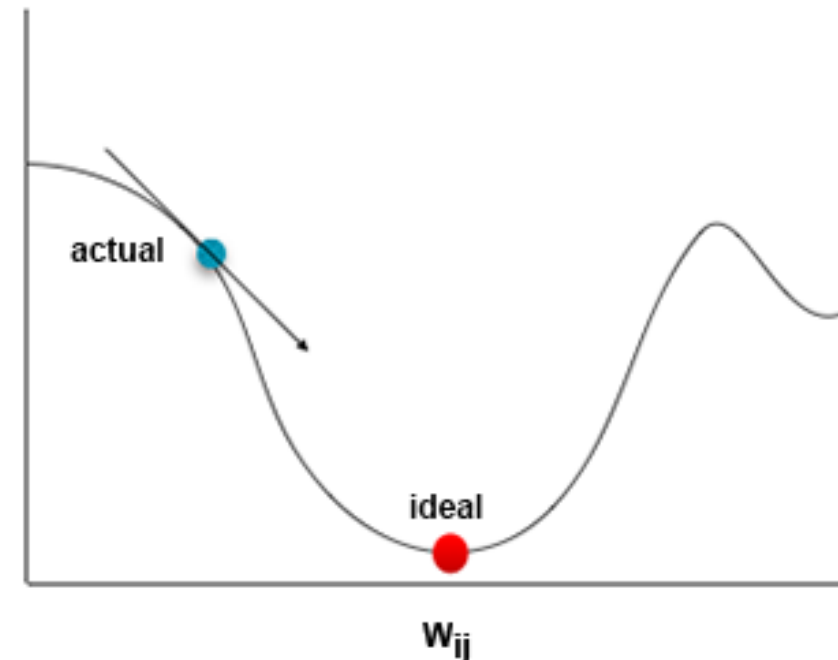
# Deep Learning Intuition

1. Data
2. Model
3. Cost function or loss or objective

Fit the model to the data by minimizing your cost function.

# Deep Learning Intuition

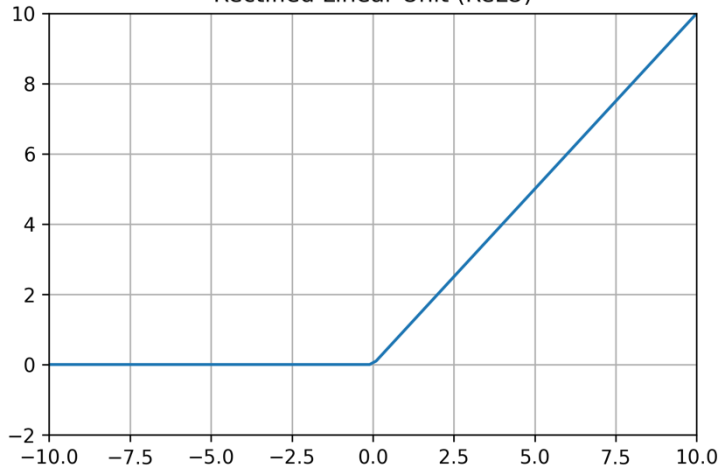
- Gradient descent optimization of the cost function
  - Linear and non-linear operations need to be differentiable
- Compute the gradient across the training set (the whole set or mini-batches)
- Update the model weights by giving a step in the opposite direction (i.e., minimize the cost)
- Compute the average cost function in the train and validation sets after each epoch





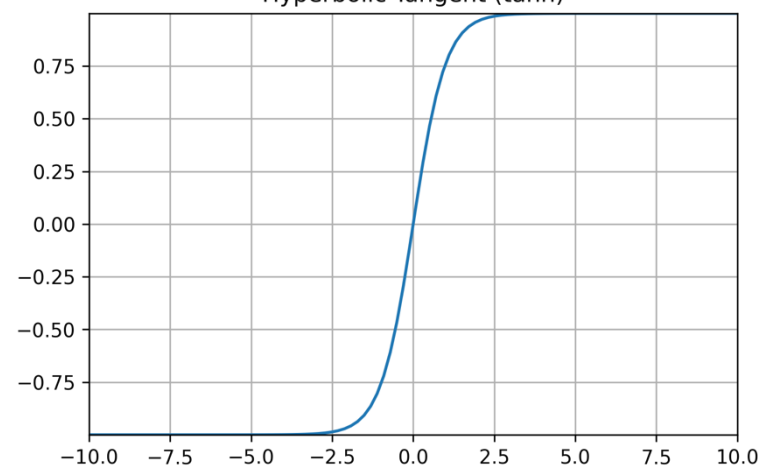
# Activations

Rectified Linear Unit (ReLU)



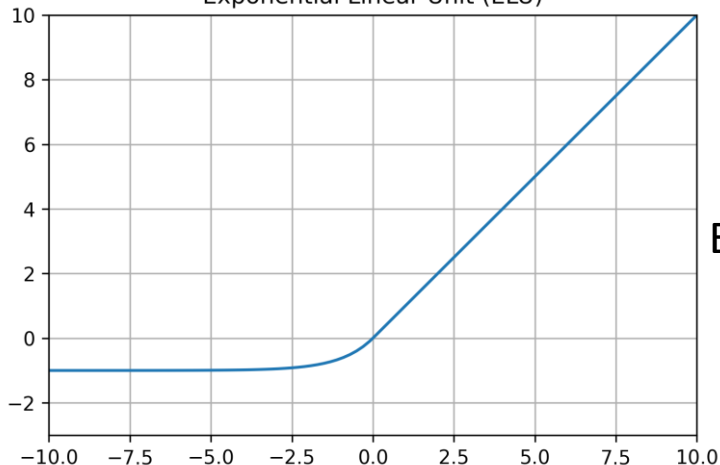
$$\text{ReLU}(x) = \begin{cases} x & x \geq 0 \\ 0 & x < 0 \end{cases}$$

Hyperbolic Tangent (tanh)



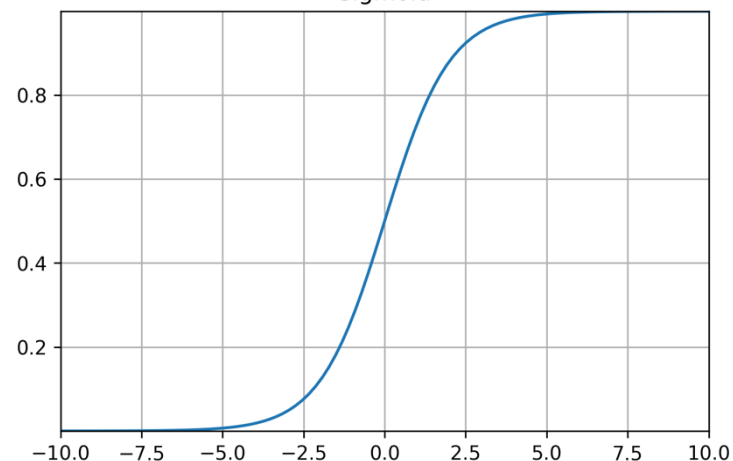
$$\tanh(x) = \frac{e^{2x} - 1}{e^{2x} + 1}$$

Exponential Linear Unit (ELU)



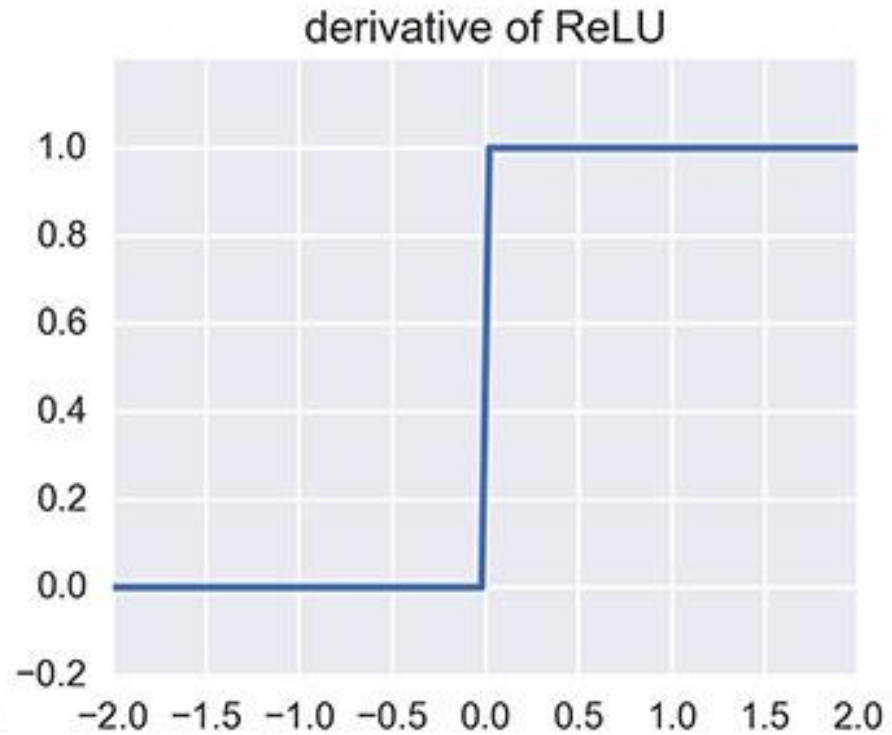
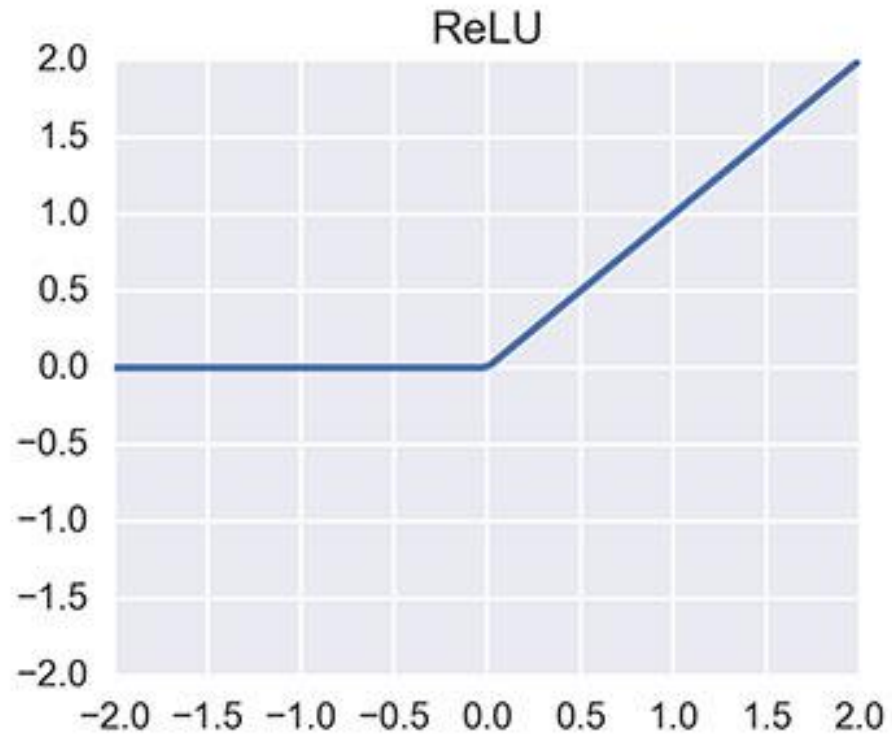
$$\text{ELU}(x) = \begin{cases} x & x \geq 0 \\ \alpha(e^x - 1) & x < 0 \end{cases}$$

Sigmoid



$$\text{sigmoid}(x) = \frac{1}{1 + e^{-x}}$$

# Activations - ReLU



$$\text{ReLU}(x) = \begin{cases} x & x \geq 0 \\ 0 & x < 0 \end{cases}$$

$$\frac{d\text{ReLU}(x)}{dx} = \begin{cases} 1 & x > 0 \\ 0 & x < 0 \end{cases}$$

# Activations - Softmax

$$\text{softmax}(\vec{z}) = \frac{e^{z_i}}{\sum_{j=1}^K e^{z_j}}$$

- Softmax converts a real vector to a vector of categorical probabilities.
- The elements of the output vector are in range (0, 1) and sum to 1.
- Softmax is often used as the activation for the last layer of a classification network -> results are interpreted as a probability distribution.

# Summary

- Deep learning models alternate between differentiable linear and non-linear operations
- Deep learning models are fit (i.e., trained) to the data by minimizing a cost function using gradient descent methods
- There are many potential non-linear operations
- ReLUs are commonly used due to their computational simplicity and simple derivative

# Thank you!