

# Feed Forward Neural Networks

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# Background Content

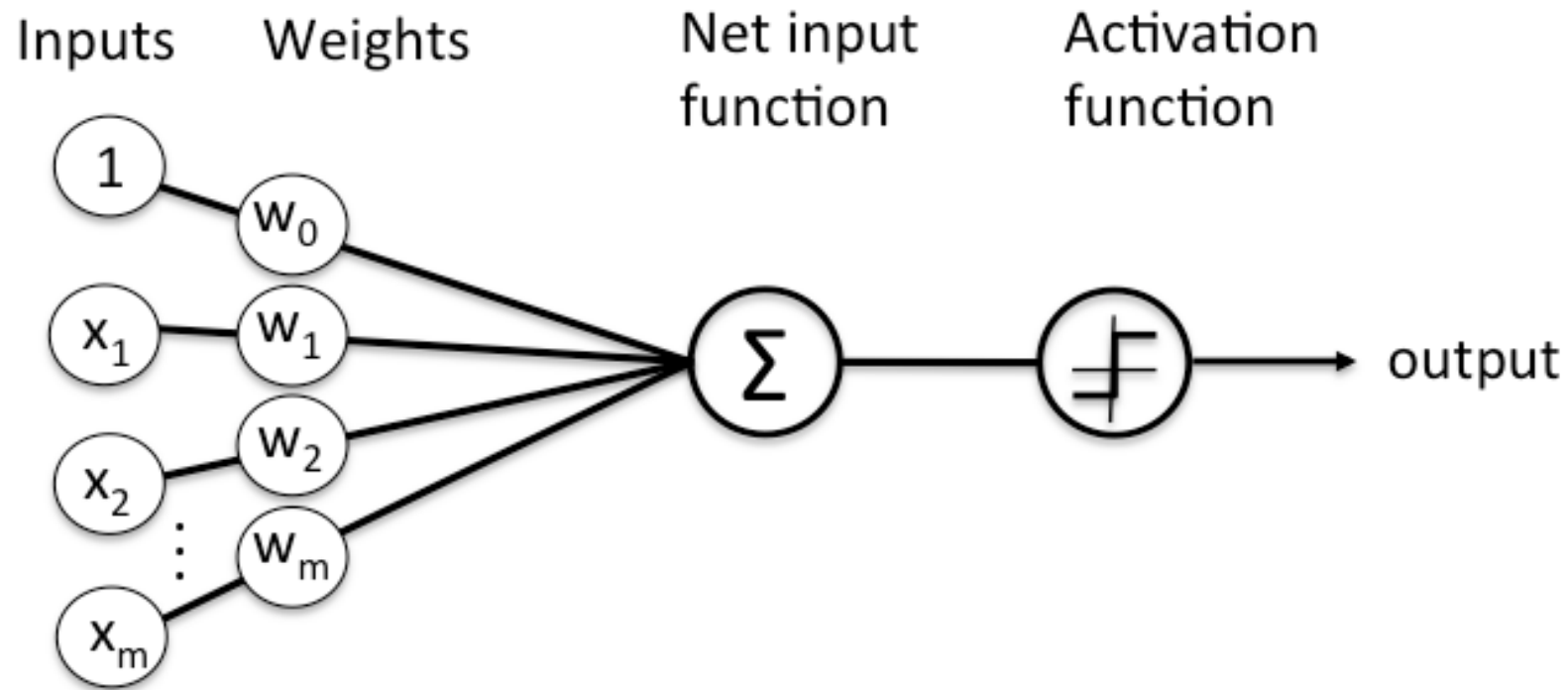
# What is a neural network and what is a node?

- Set of algorithms (modelled after human brain) designed to recognize patterns
  - Input data is numbers
  - Output data is class (ex: spam vs not spam) or continuous range of numbers (ex: 0.1, 0.153, etc.)
- Neural network typically comprised of several “layers” of “nodes”
- Hierarchy
  - Node = neuron = perceptron
    - Where computation/math happens
    - Algorithm
    - Computer imitation of human brain’s neuron
  - Layer = group of neurons organized together
  - Neural network = bunch of layers of nodes organized and connected in various ways

# More on what a node is

- Each node has 4 parts
  - Input values
  - Weights and bias
    - Weights = array of numbers that helps control magnitude and sign of output based on input
    - Bias = number that controls when activation function is activated
      - **Similar to “b” in  $y = mx + b$  -> a number that helps you to fit data better/shift output up+down**
  - Net sum
  - Activation function “F”
    - Function (ex: sigmoid, tanh, etc) that introduces non linearity to data
      - **Allows neural network to better simulate brain**
    - Without activation function, neural network would just be 1 big linear classifier
- How does node work? How is output calculated from input array of numbers?
  - $\text{Output} = \text{Activation\_Function} \{(\text{Input} \times \text{Weights}) + \text{Bias}\}$

# Visualization of node (aka perceptron)



Picture from: <https://wiki.pathmind.com/neural-network>

# Why do we need weights and biases pt1

- Imagine we want to use neuron to predict car price (in \$) based on:
  - Year car was made
  - # Miles driven
- We would expect
  - Higher year (more recent car) -> Higher price -> weight for year is (+)
  - Higher # miles driven (more used car) -> Lower price -> weight for price is (-)

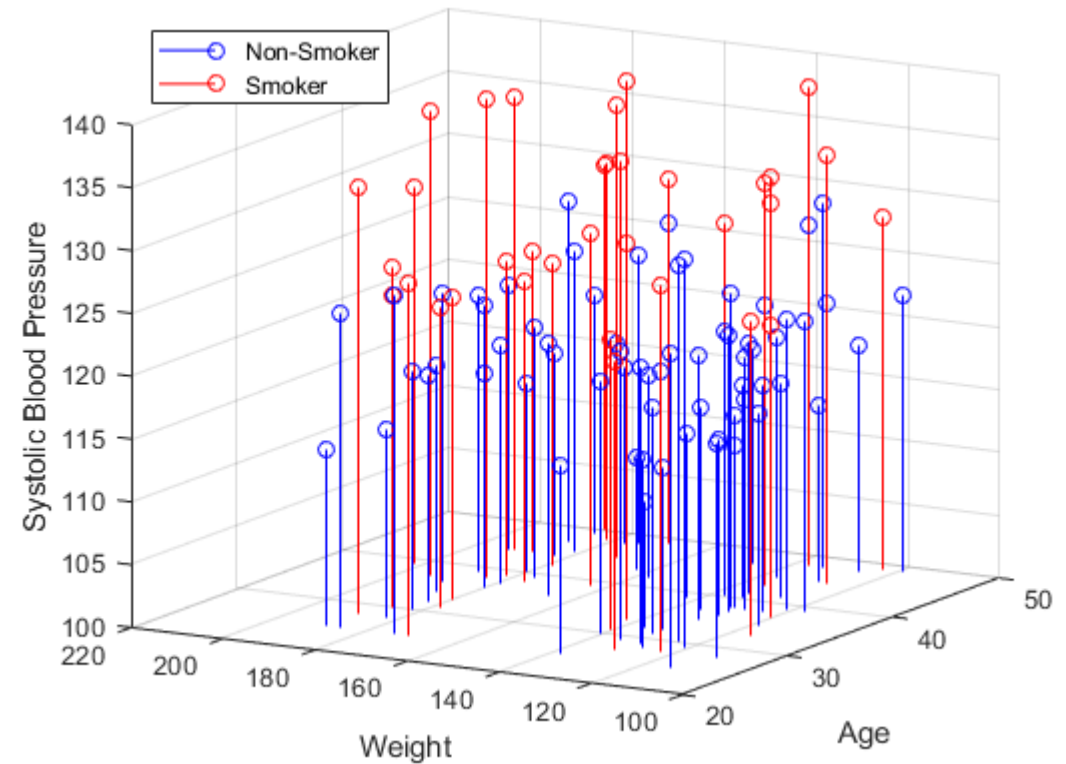
<http://webcache.googleusercontent.com/search?q=cache:vj72OfznF58J:https://medium.com/fintechexplained/neural-networks-bias-and-weights-10b53e6285da&hl=en&gl=us&strip=1&vwsrc=0>

# Why do we need weights and biases pt2

- Recall neuron made of multiple parts
  - Input values: Assume it is a 1 by 2 array = [year, # miles driven]
  - Weights array: Let =  
[weight for year,  
weight for miles driven]
  - If we just used input values and weights array, price would be
    - Output = Price =  $\text{year} * \text{weight\_for\_year} + \text{miles\_driven} * \text{weight\_for\_miles\_driven}$
    - Let  $\text{weight\_for\_year} = 2$
    - Let  $\text{weight\_for\_miles\_driven} = -1$
    - Assume year = 2008, and miles driven = 50,000
    - Price would =  $2008 * 2 - 1 * 50000 = 4016 - 50000 = -45,984$  dollars
  - As you can see, using weights and bias alone doesn't make sense (data might not fit very well or make sense) -> introduce bias to make better predictions of output based on input
  - Bias: Let bias = 50,008
  - Now, output = price =  $\text{year} * \text{weight\_for\_year} + \text{miles\_driven} * \text{weight\_for\_miles\_driven} + \text{bias}$ 
    - Output = price =  $2008 * 2 - 1 * 50000 + 50008 = \$4,024$  -> makes more sense (to have positive price rather than a negative price)
  - Activation function "F" –ignore for now, look at next slide to understand why we need activation functions

# Why do we need activation functions pt 1

- Recall: activation functions introduce non-linearity to data
- If we tried to classify non smokers vs smokers in picture to right with linear classifier (i.e a flat/linear 3-D plane), we wouldn't be able to since the data is NON-LINEAR
- BUT if we used a curved 3-D plane, we WOULD be able to differentiate non-smokers and smokers



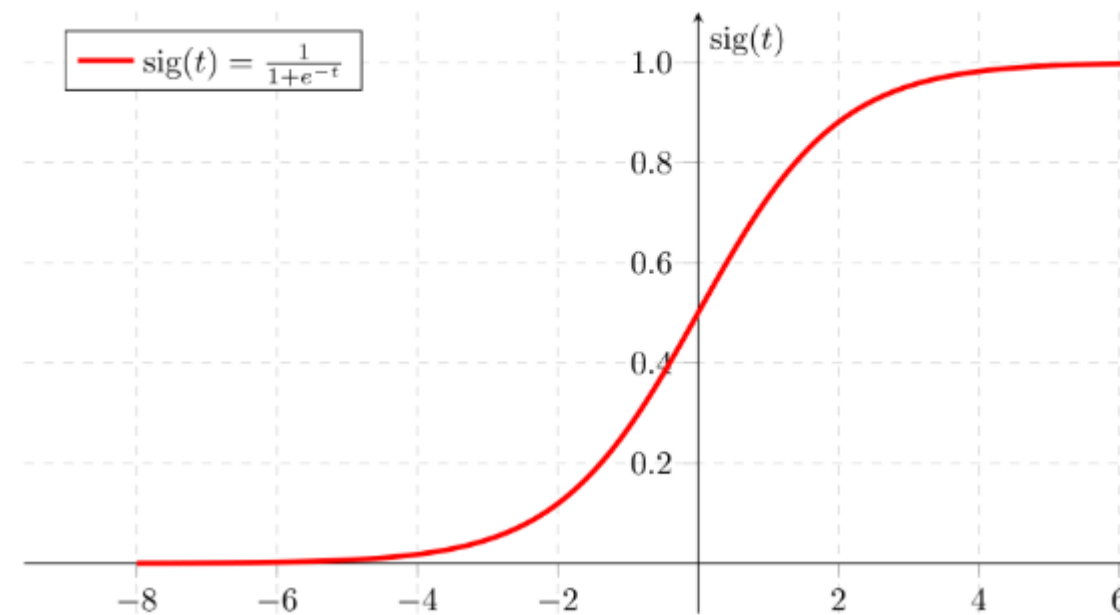
<https://towardsdatascience.com/everything-you-need-to-know-about-activation-functions-in-deep-learning-models-84ba9f82c253>



# Why do we need activation functions pt2

- If a neuron only had inputs, weights, and biases, the output would be:
  - Output = Weight\*Input + Bias =  $wx+b$
  - Model has degree of 1 -> identical to linear classifier -> can't classify complex nonlinear data
- BUT if we made the model non-linear by adding activation function at end (output = activation\_function( $wx+b$ )), the model would not have degree 1 and would be able to classify non-linear data

Ex of activation function (sigmoid)



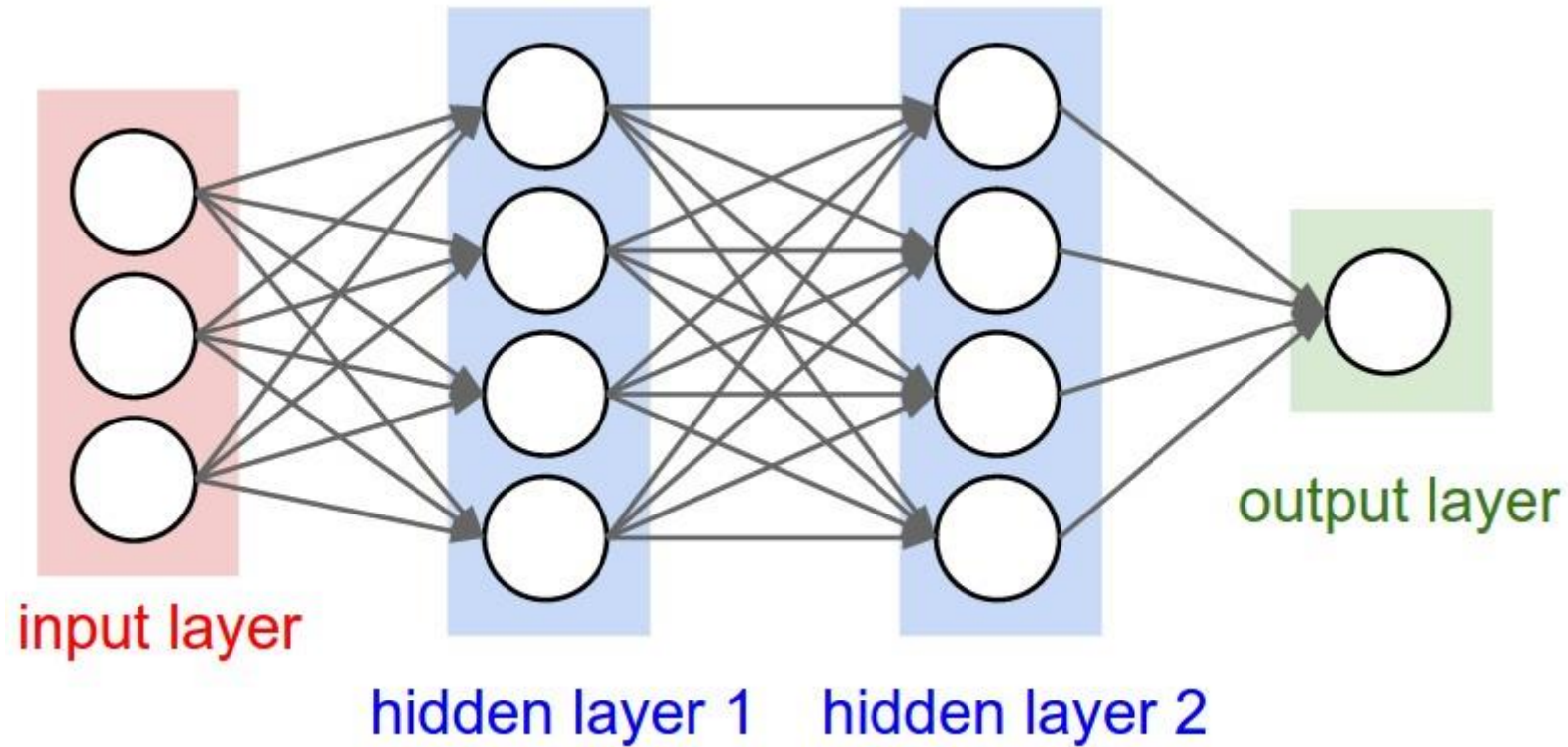
# What is a neural network

- Recall that a neural network = bunch of organized neurons linked together

# What makes a neural network a feed forward neural network?

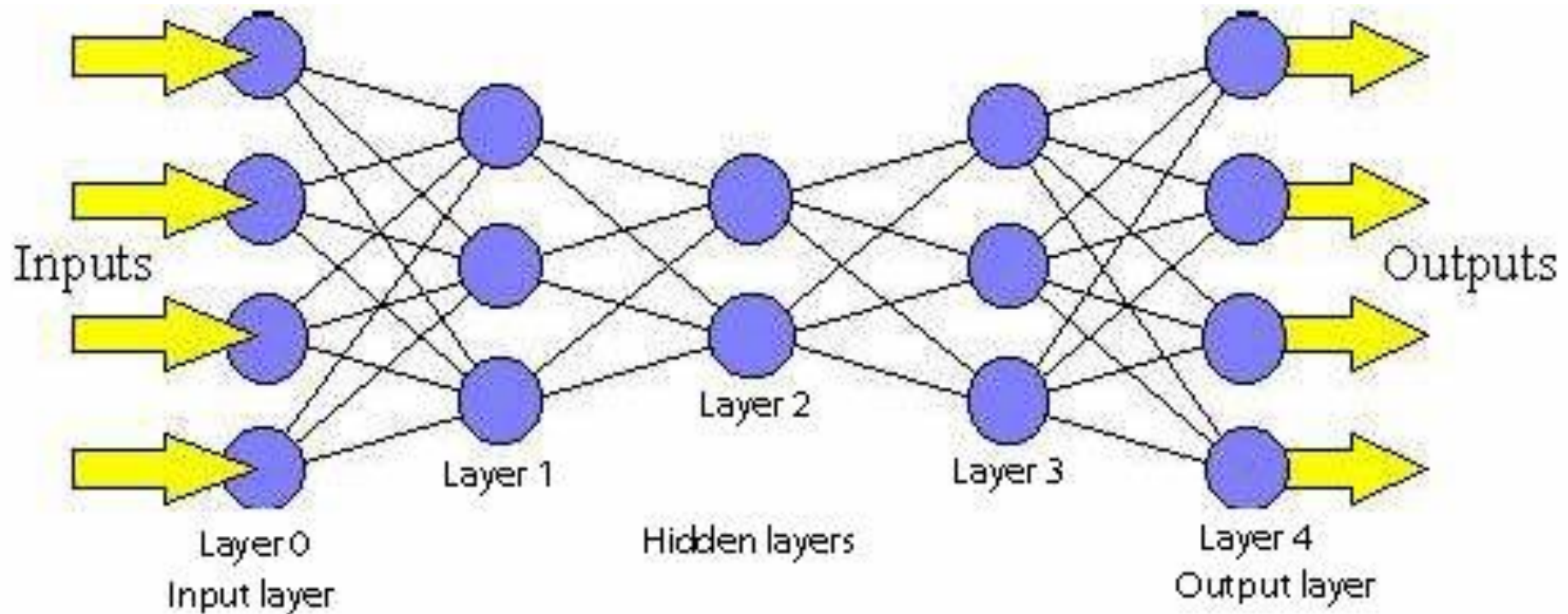
1. Perceptrons (nodes) organized by layer
  1. 1<sup>st</sup> layer – Input
  2. Middle layers – Hidden layers
  3. Last layer – Output
2. Each perceptron in one layer connected to each perceptron in next layer (“info fed forward from one layer to next”)
3. No connection between perceptrons in same layer

# Picture of typical feed forward neural network



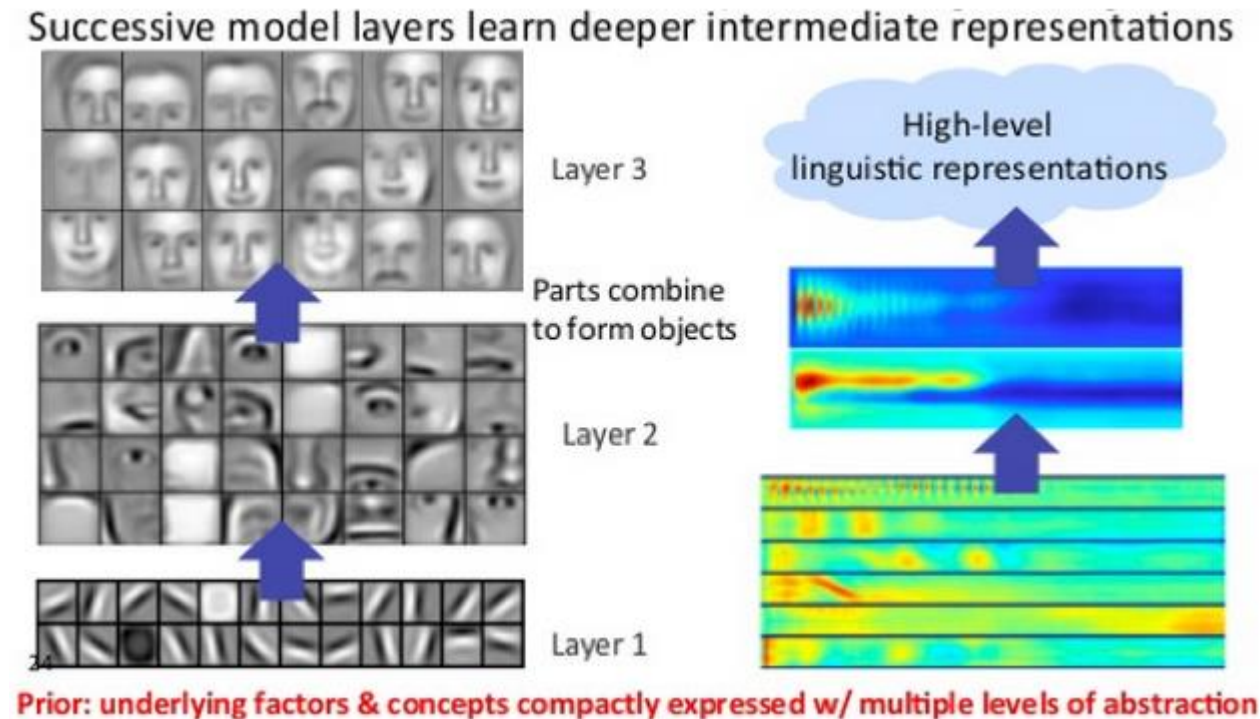
- Picture from: <https://stackoverflow.com/questions/35345191/what-is-a-layer-in-a-neural-network>

# Another example of feed forward neural network



# Why do we need hidden layers?

- We need at least 2 layers (input and output layer), but we can also have hidden layers -> why?
- Adding multiple hidden layers allows us neural network to learn higher level features in later layers from lower level features in earlier layers



# How do neural networks “learn”?

- “learning” – refers to adjusting weights and biases in layers/neurons to achieve better performance
  - Better performance = better predicting outputs based on input
- In order to learn:
  - Need error function to quantify difference between prediction and true/correct result
    - Loss function – quantifies difference for single training example
    - Cost function – average of loss function over entire training set
  - Need to adjust weights and biases based on error, to reduce error
    - Back propagation – algorithm that uses gradients and partial derivatives to adjust weights and biases in order to reduce loss (achieve higher accuracy)

# WAIT - How are layers, neurons, weights, biases, initialized?

- Random Initialization - Generally weights initialized with random numbers and biases initialized with 0
- Other initialization techniques:
  - “He” initialization
  - “Xavier” initialization

For more info: <https://towardsdatascience.com/weight-initialization-techniques-in-neural-networks-26c649eb3b78>



# General Structure of Feed Forward NN Code

- Explain and walk through MNIST code in PyTorch

# Works Cited and links for more details

1. <https://wiki.pathmind.com/neural-network>
2. <https://stackoverflow.com/questions/35345191/what-is-a-layer-in-a-neural-network>
3. <https://cs.stanford.edu/people/eroberts/courses/soco/projects/neural-networks/Architecture/feedforward.html>
4. <http://webcache.googleusercontent.com/search?q=cache:vj72OfznF58J:https://medium.com/fintechexplained/neural-networks-bias-and-weights-10b53e6285da&hl=en&gl=us&strip=1&vwsrc=0>
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7. <https://stats.stackexchange.com/questions/179026/objective-function-cost-function-loss-function-are-they-the-same-thing#:~:text=The%20loss%20function%20computes%20the,of%20the%20entire%20training%20set>
8. <http://neuralnetworksanddeeplearning.com/chap2.html>