

Scientific Computing

April 23, 2025

Announcements

→ Homework 6 due Friday, May 2, 11:59pm

→ Final exam is take-home only
assigned Fri, May 2
due Fri, May 9

Today

→ Coding NN and Batching

→ Loss Functions

Office Hours:

Mon + Fri

9:30am - 10:30am

Cudahy 307

Think of each layer as a vector of values.

$$\begin{bmatrix} 0.4 \\ -0.7 \end{bmatrix}$$

$$\begin{bmatrix} 0.7 \\ -0.39 \\ 0.24 \\ -0.58 \end{bmatrix}$$

$$\begin{bmatrix} -0.724 \\ -0.037 \\ 0.598 \end{bmatrix}$$

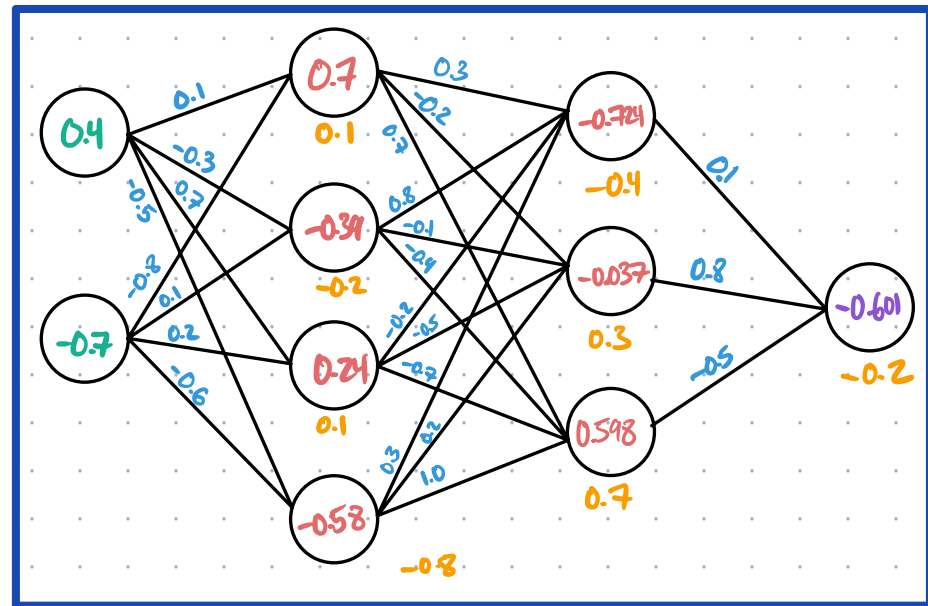
$$[-0.601]$$

How is each layer computed from the previous one?
Make matrices for the weights between each layer and vectors for the biases.

$$M_1 = \begin{bmatrix} 0.1 & -0.8 \\ -0.3 & 0.1 \\ 0.7 & 0.2 \\ -0.5 & -0.6 \end{bmatrix}$$

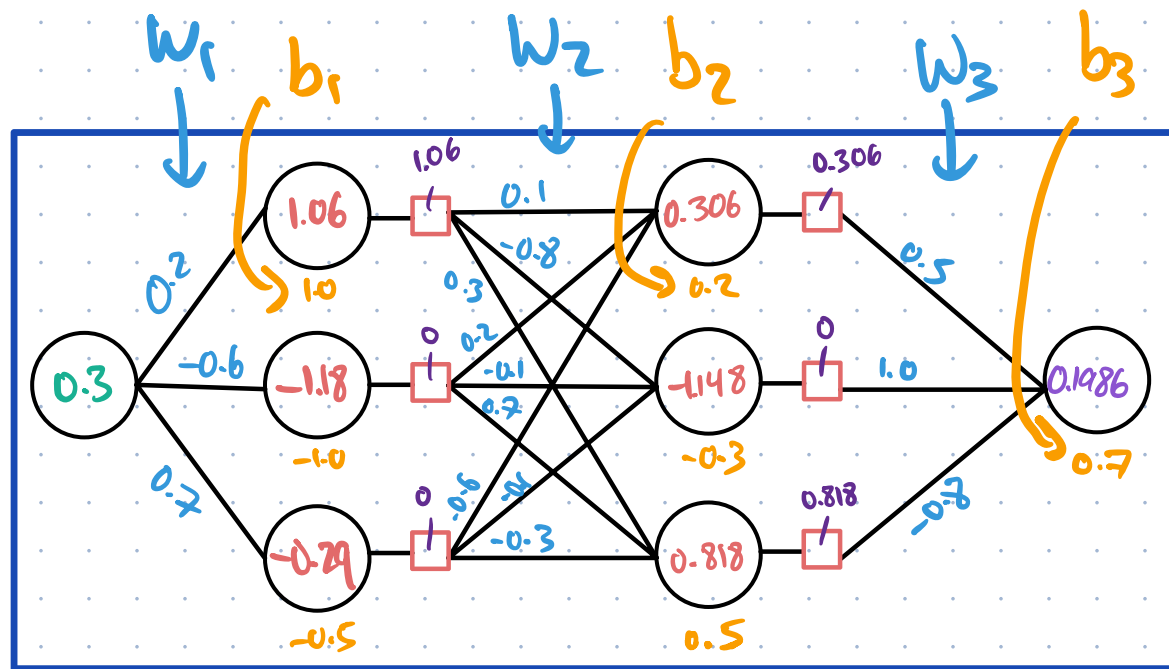
$$b_1 = \begin{bmatrix} 0.1 \\ -0.2 \\ 0.1 \\ -0.8 \end{bmatrix}$$

$$\begin{bmatrix} 0.1 & -0.3 & \dots \\ -0.8 & 0.1 & \dots \end{bmatrix}$$



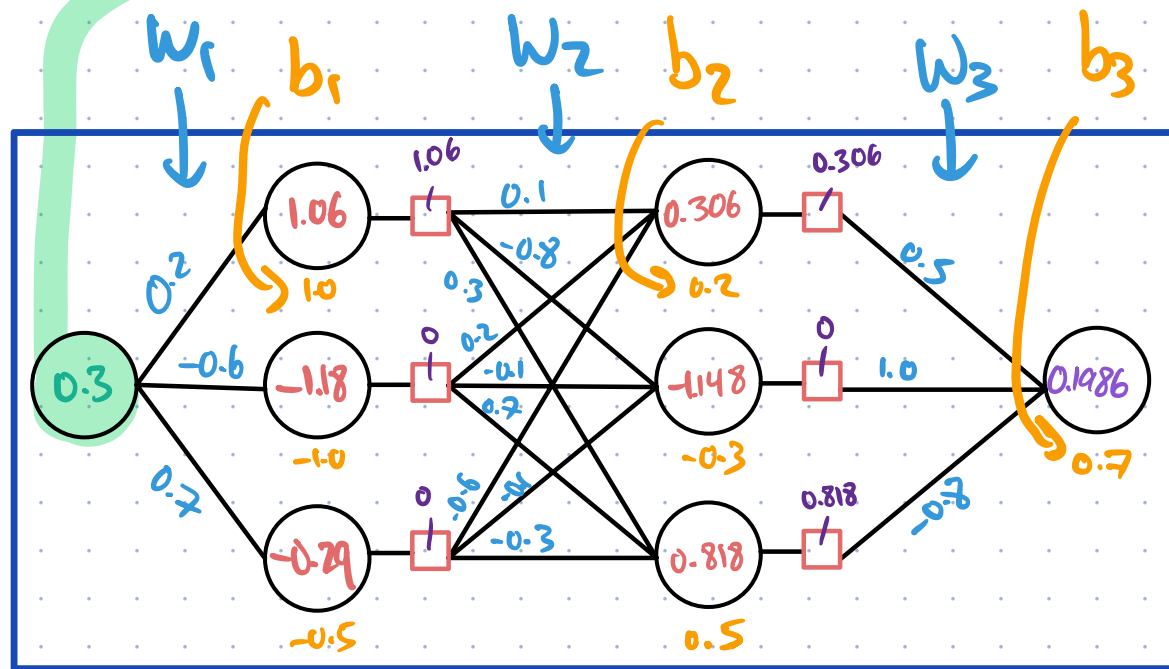
So this neural network, with activation function σ for each layer, is the function:

$$f(x) = \sigma(W_3 \sigma(W_2 \sigma(W_1[x] + b_1) + b_2) + b_3)$$



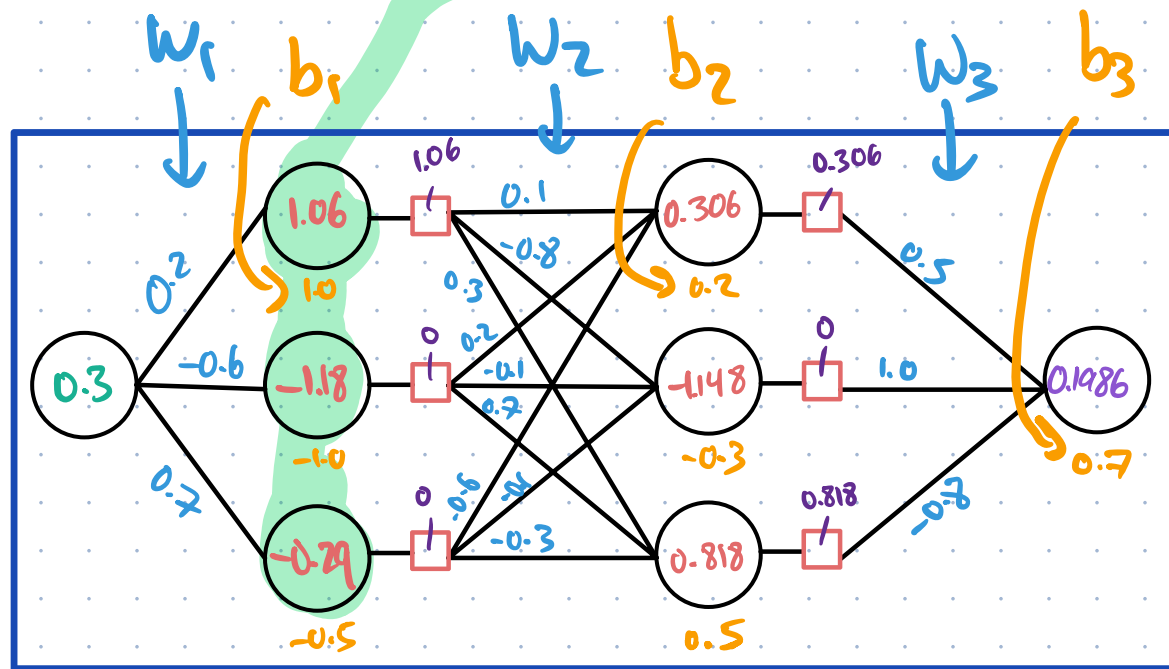
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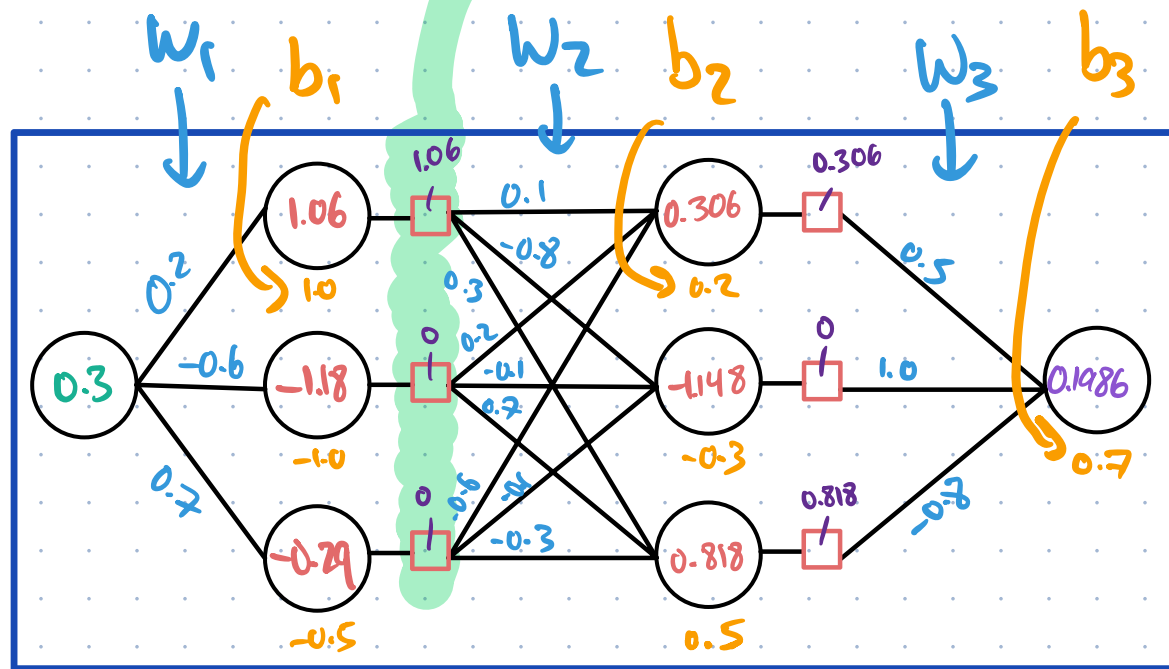
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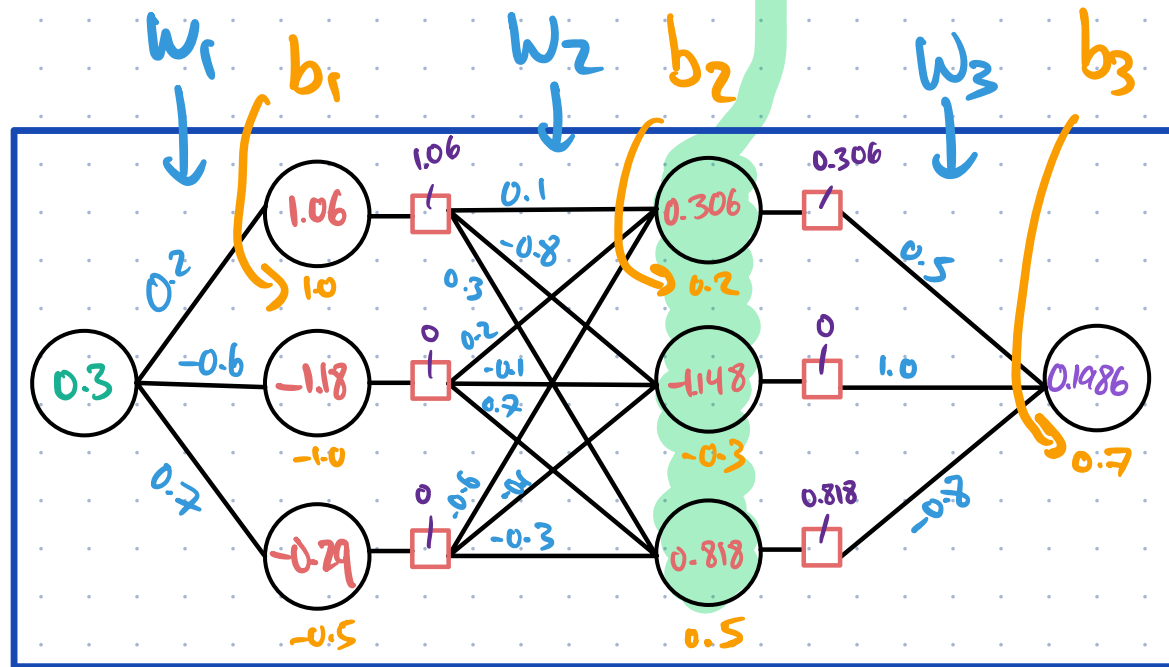
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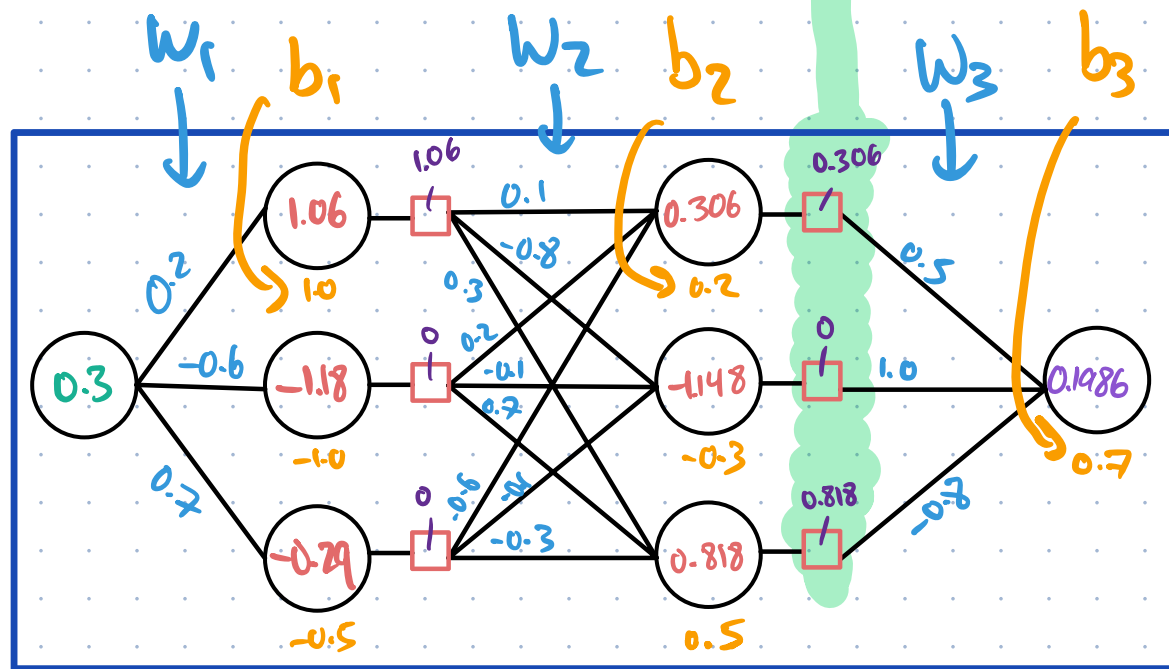
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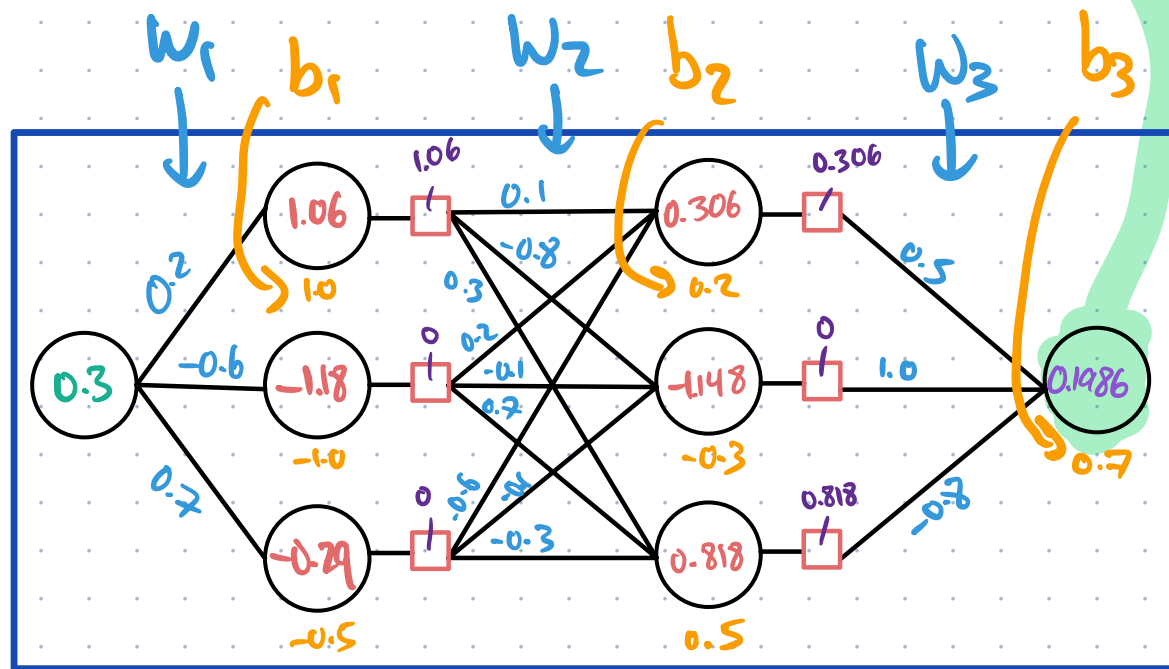
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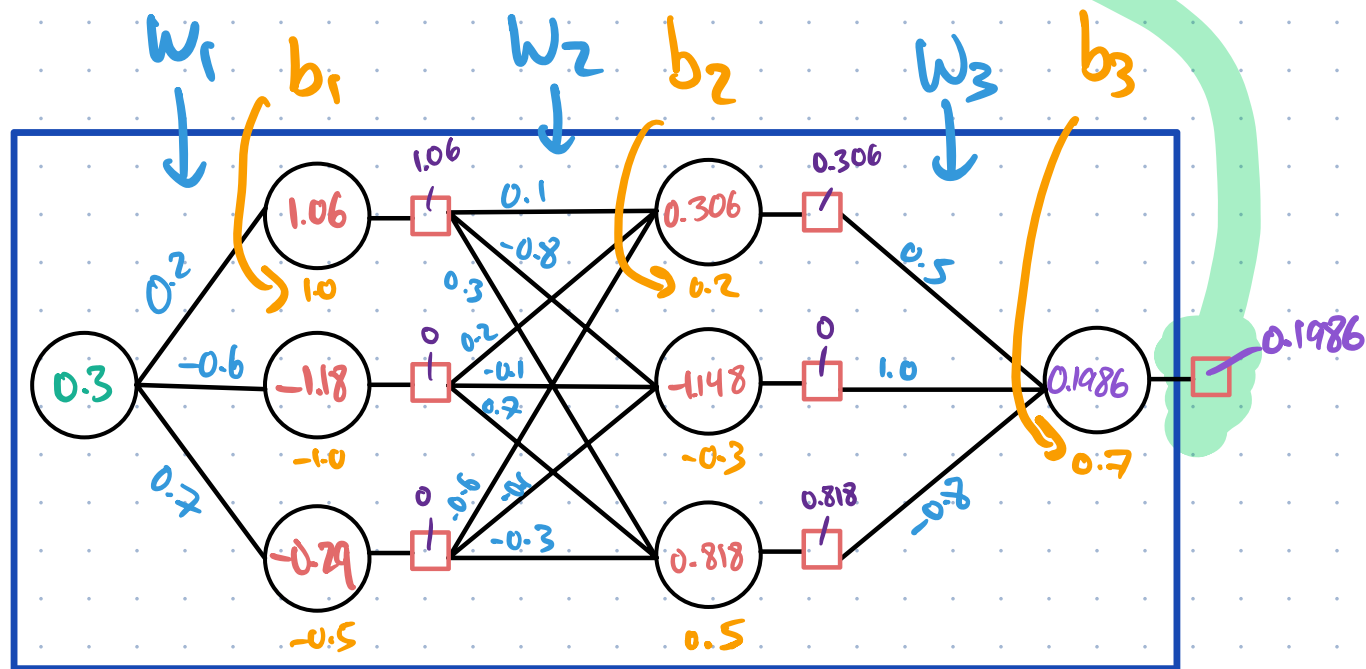
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* Sometimes the output layer has a different activation function, or no AF.

$$W_1 = \begin{bmatrix} 0.2 \\ -0.6 \\ 0.7 \end{bmatrix}$$

$$W_2 = \begin{bmatrix} 0.1 & -0.8 & 0.3 \\ 0.2 & -0.1 & 0.7 \\ -0.6 & -0.1 & -0.3 \end{bmatrix}$$

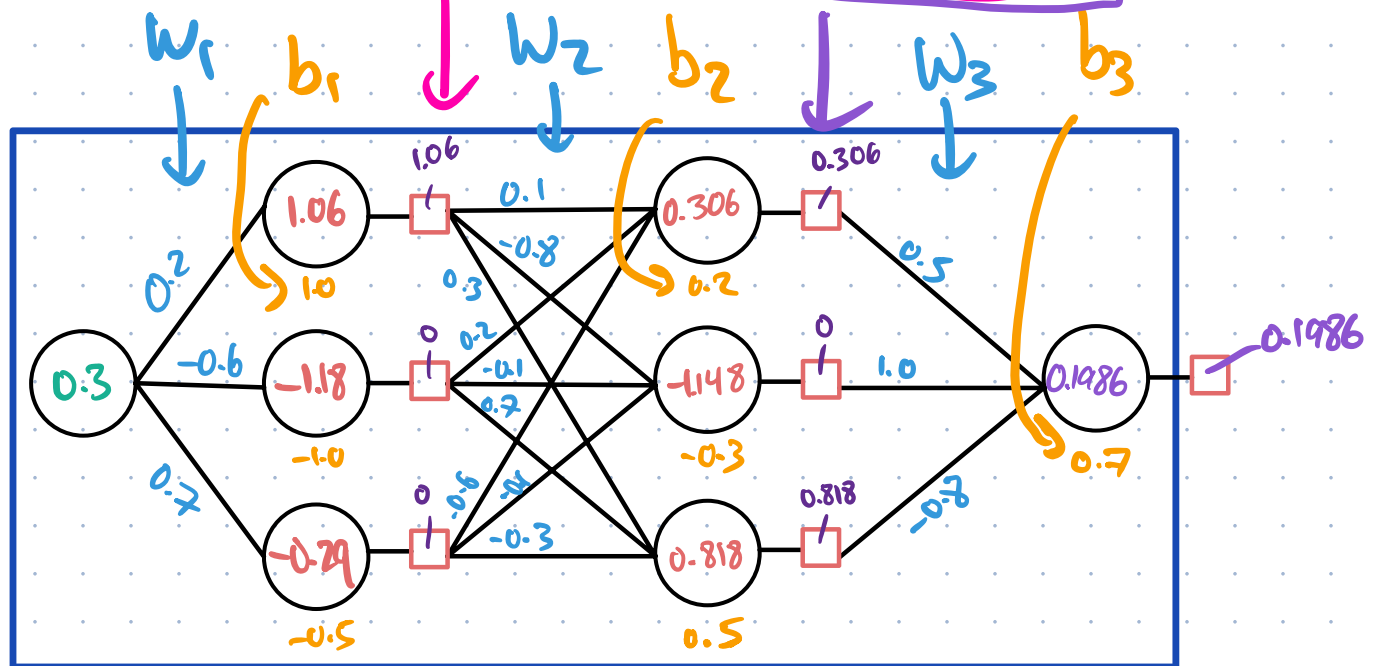
$$W_3 = \begin{bmatrix} 0.5 & 1.0 & -0.8 \end{bmatrix}$$

$$b_1 = \begin{bmatrix} 1.0 \\ -1.0 \\ -0.5 \end{bmatrix}, b_2 = \begin{bmatrix} 0.2 \\ -0.3 \\ 0.5 \end{bmatrix}, b_3 = \begin{bmatrix} 0.7 \end{bmatrix}$$

each row of a W is the weights coming out of one neuron

$$f(0.3) = \max(0, [0.5 \ 1.0 \ -0.8] (\max(0, [0.1 \ -0.8 \ 0.3] (\max(0, [0.2 \ -0.6 \ 0.7] [0.3] + \begin{bmatrix} 1.0 \\ -1.0 \\ -0.5 \end{bmatrix}) + \begin{bmatrix} 0.2 \\ -0.3 \\ 0.5 \end{bmatrix}) + [0.7]))$$

$$= 0.1986$$



Now you understand exactly what a neural network is:
a fancy way to define a function

But not yet:

- * How to find a good one (training)
- * How NNs accomplish different tasks
(what good is a fancy function?)

Next topic : Coding our first NN, and batching the
forward pass

Topic 14 - Implementing our first NN and Batching

* Goal: Write code to perform the forward pass of a NN, with activation functions

The purpose of writing this code is to understand the topic.

If you needed a fast NN for research, you would probably use a Python library like PyTorch

Structure:

- * Object Oriented

- * Objects:

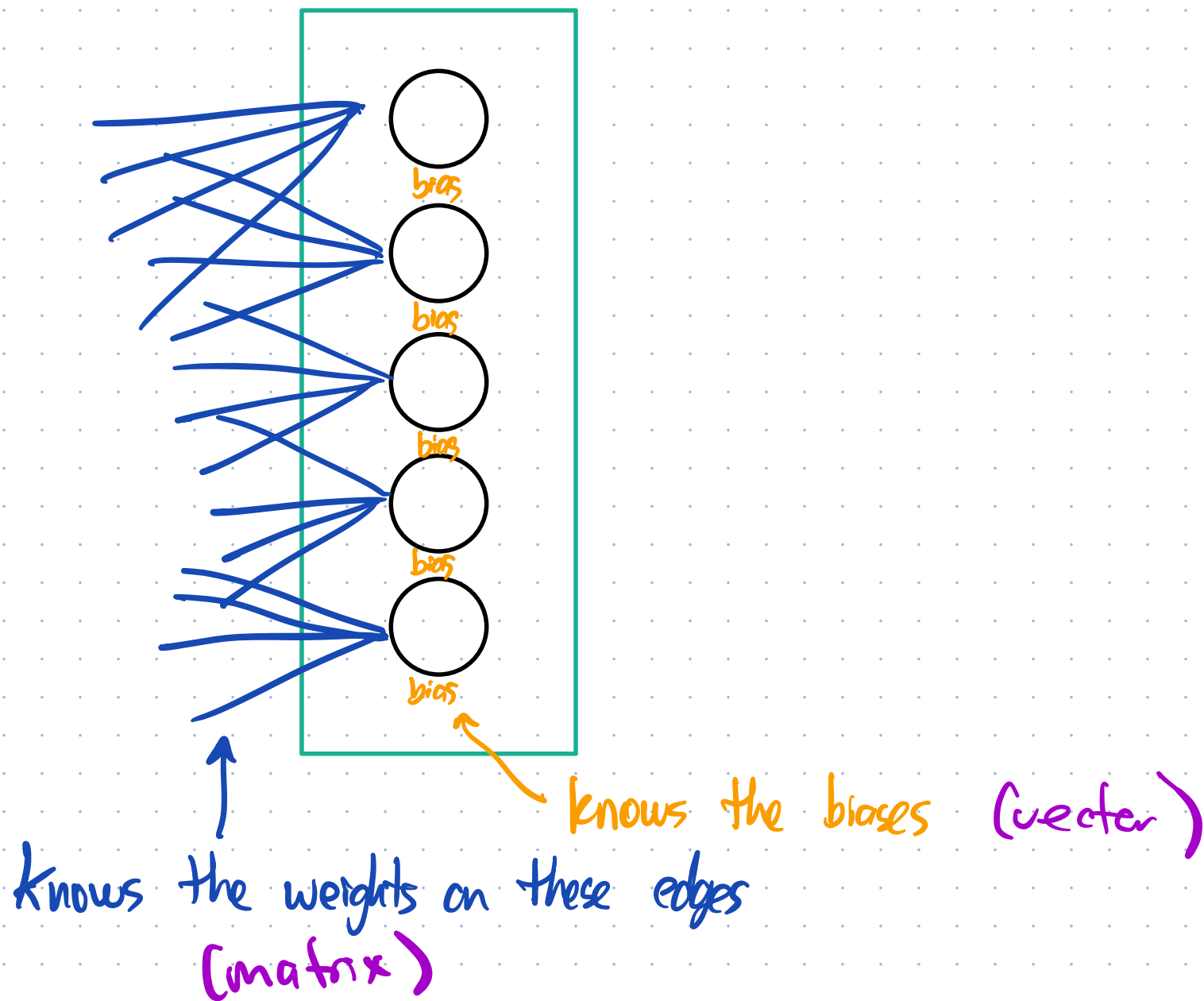
Layer

- knows the weights that feed into it from the previous layer
- knows the biases of its neurons
- can take input data and compute output data

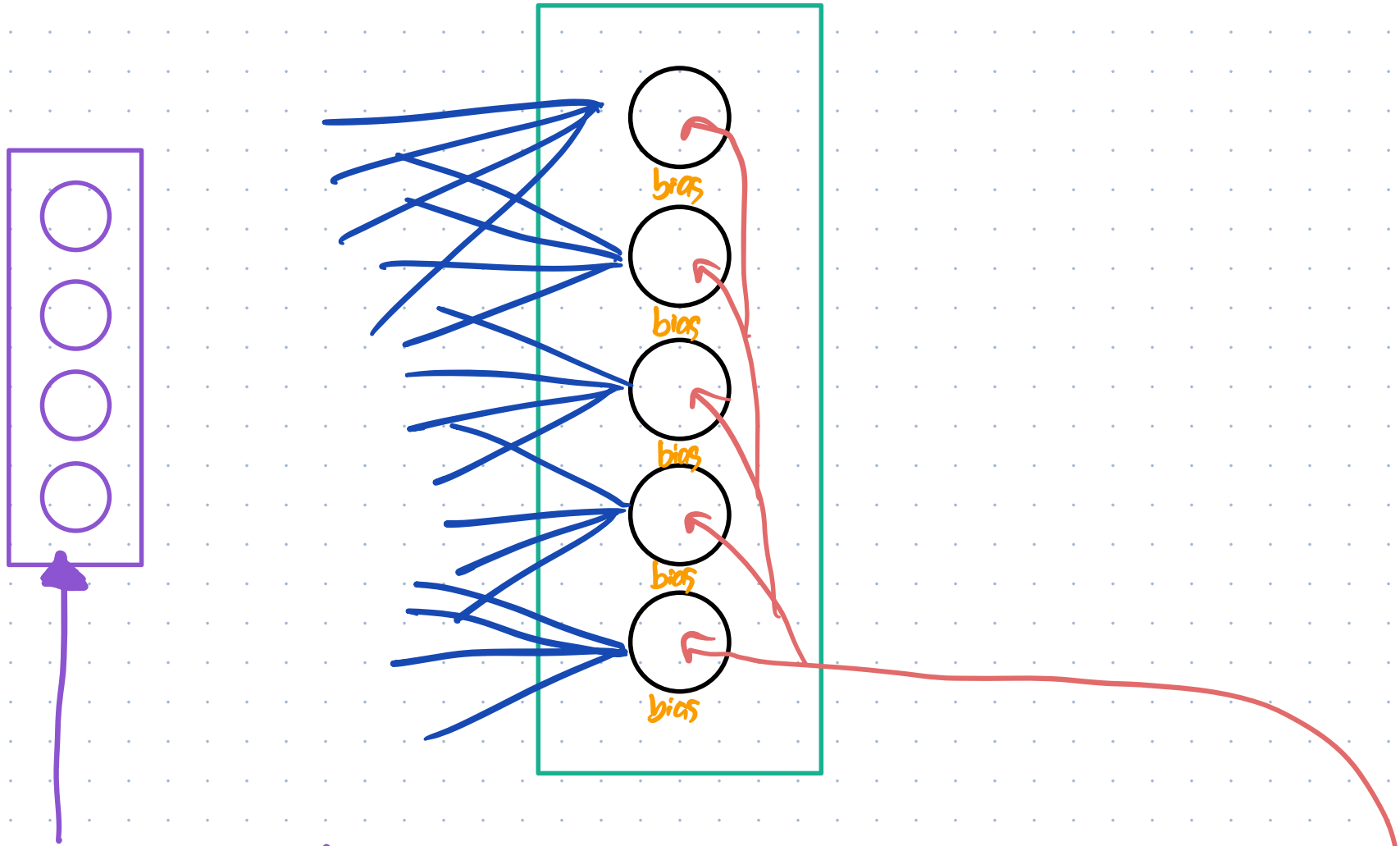
Activation Function

- can take input data and compute output data

Layer:



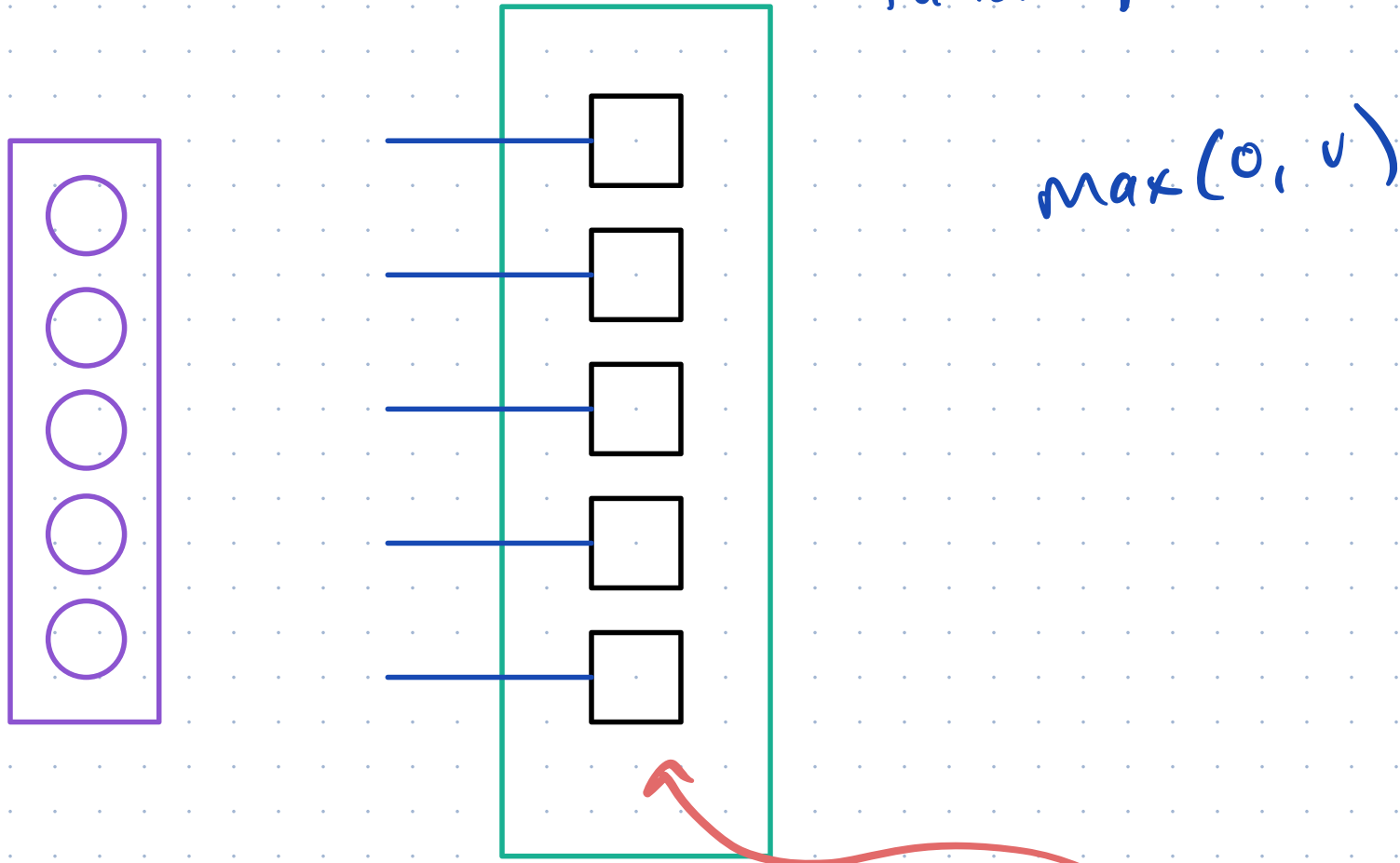
Layer:



input: value of previous layer's neurons
(actually, the activation function of those neurons)

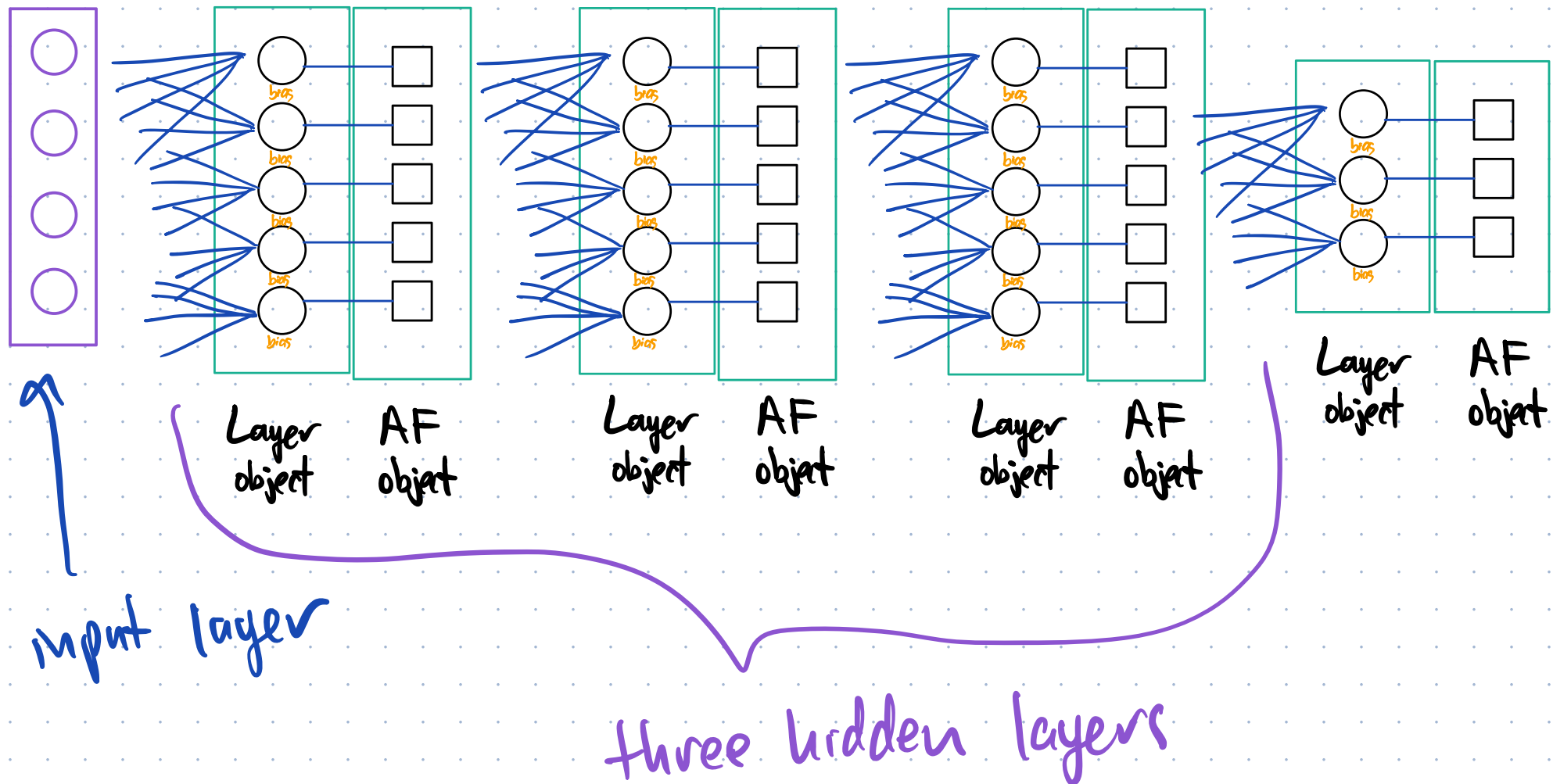
output: value of this layer's neurons (pre-activation)

Activation Function classes (one for each activation function)



inputs: values of neurons we're activating
outputs: activated values

Then we can chain instances of these objects together to make a full NN.



Coding time:

First: the "numpy" Python library

* Jupyter notebook demo

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Next: Coding our first layers together
from scratch.

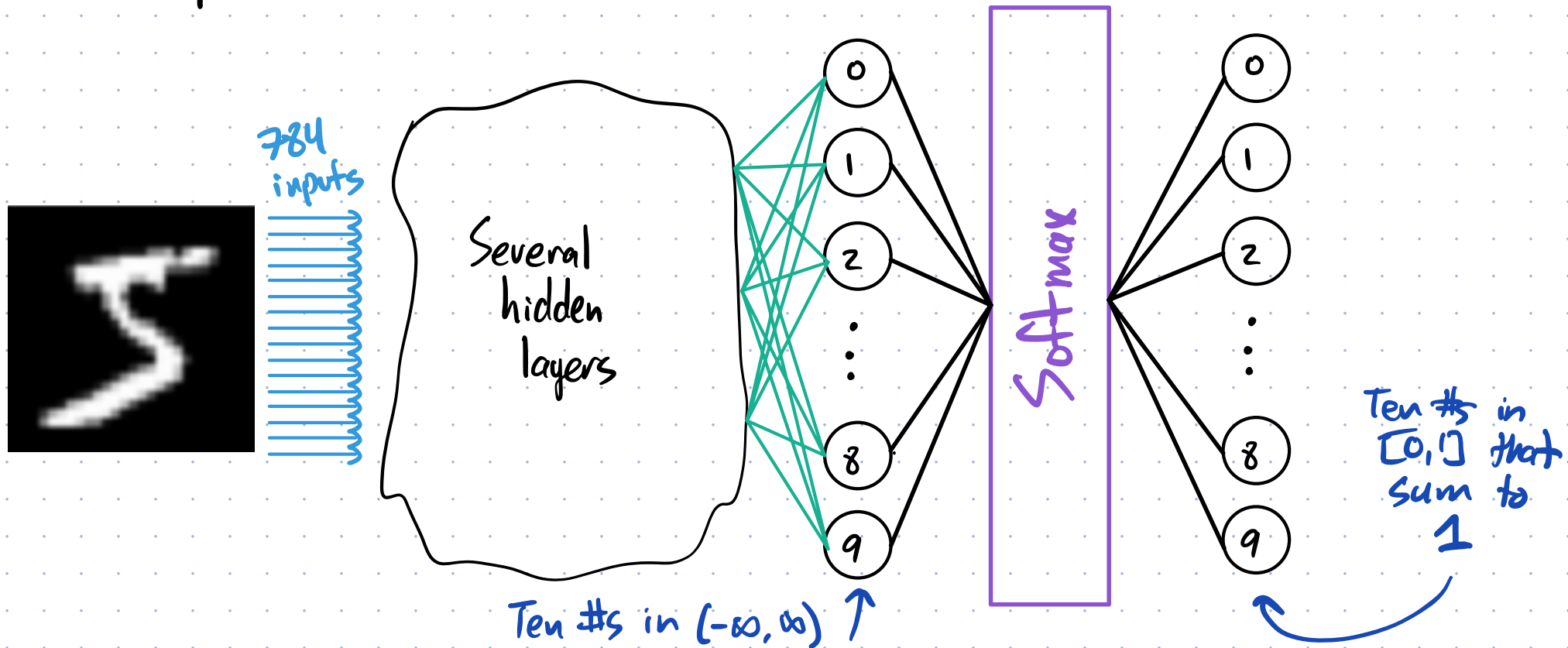
(very heavy inspiration from the
"Neural Networks from Scratch"
book!)

New Activation Function: Softmax

- * Turns a vector of $\#s$ into a probability distribution

(a vector of $\#s$ in $[0,1]$ that sums to 1)

- * Useful for the output layer in a classification problem.



Unlike our other activation functions, Softmax works on the whole vector at once, not one value at a time individually.

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_k \end{bmatrix} \xrightarrow{\text{softmax}} \begin{bmatrix} e^{x_1}/s \\ e^{x_2}/s \\ e^{x_3}/s \\ \vdots \\ e^{x_k}/s \end{bmatrix} \quad \text{where} \quad s = \sum_{i=1}^k e^{x_i}$$

Obviously these add up to 1 because the denominator is their sum.

Obviously > 0 because $e^x > 0$.

Ex:

0	-0.8
1	-0.7
2	0.3
3	0.1
4	0.8
5	0.5
6	0.2
7	-0.3
8	0
9	0.6

softmax

$$\sum_{i=0}^9 e^{x_i} \approx 12.06$$

0.037
0.041
0.111
0.091
0.184
0.136
0.101
0.061
0.082
0.150

* 18.4% chance
the digit is
a 4

Numeric Stability (because e^x gets big!)

$$\text{let } m = \max(\vec{x})$$

Then do $\frac{e^{x_i - m}}{\sum_{j=1}^k e^{x_j - m}}$ *all components ≤ 0*

$$\frac{e^{x_i - m}}{\sum_{j=1}^k e^{x_j - m}} = \frac{\frac{e^{x_i}}{e^m}}{\sum_{j=1}^k \frac{e^{x_j}}{e^m}} = \frac{\cancel{\frac{1}{e^m}} \cdot e^{x_i}}{\cancel{\frac{1}{e^m}} \cdot \sum_{j=1}^k e^{x_j}}$$

* Let's add this as a new Activation Function class

