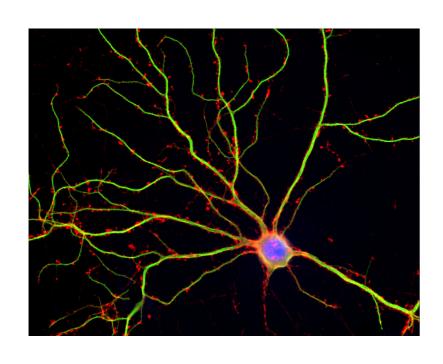
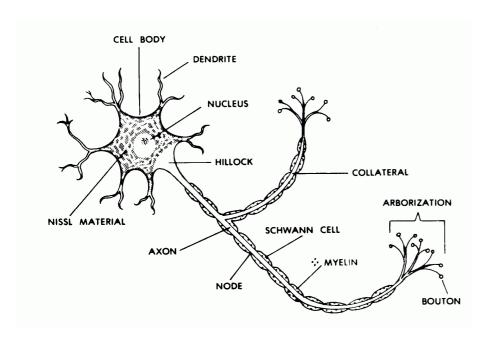
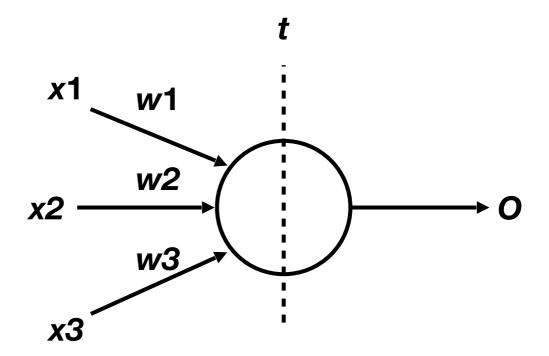
HUDK 4051: ANAIYTICS: PROCESS & THORY

Today

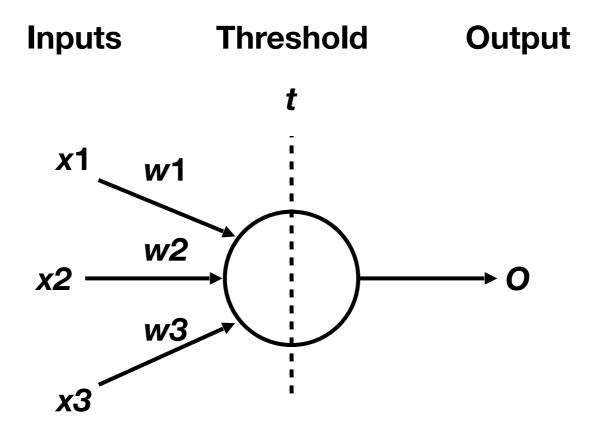
- Artificial Neural Networks
 - Perceptron
 - Sigmoid Function
 - Back propagation
- Work time

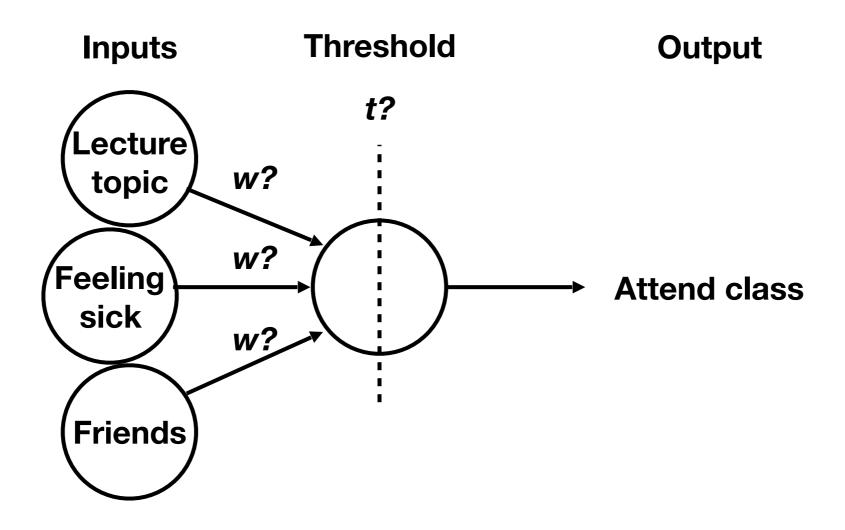


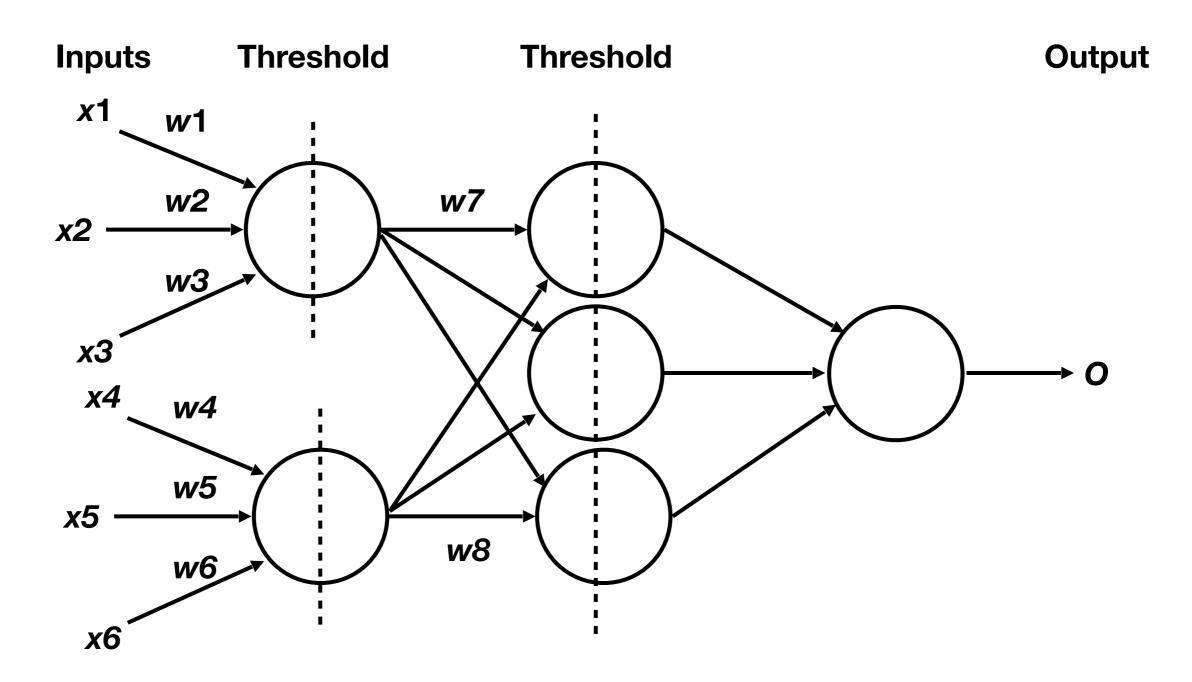


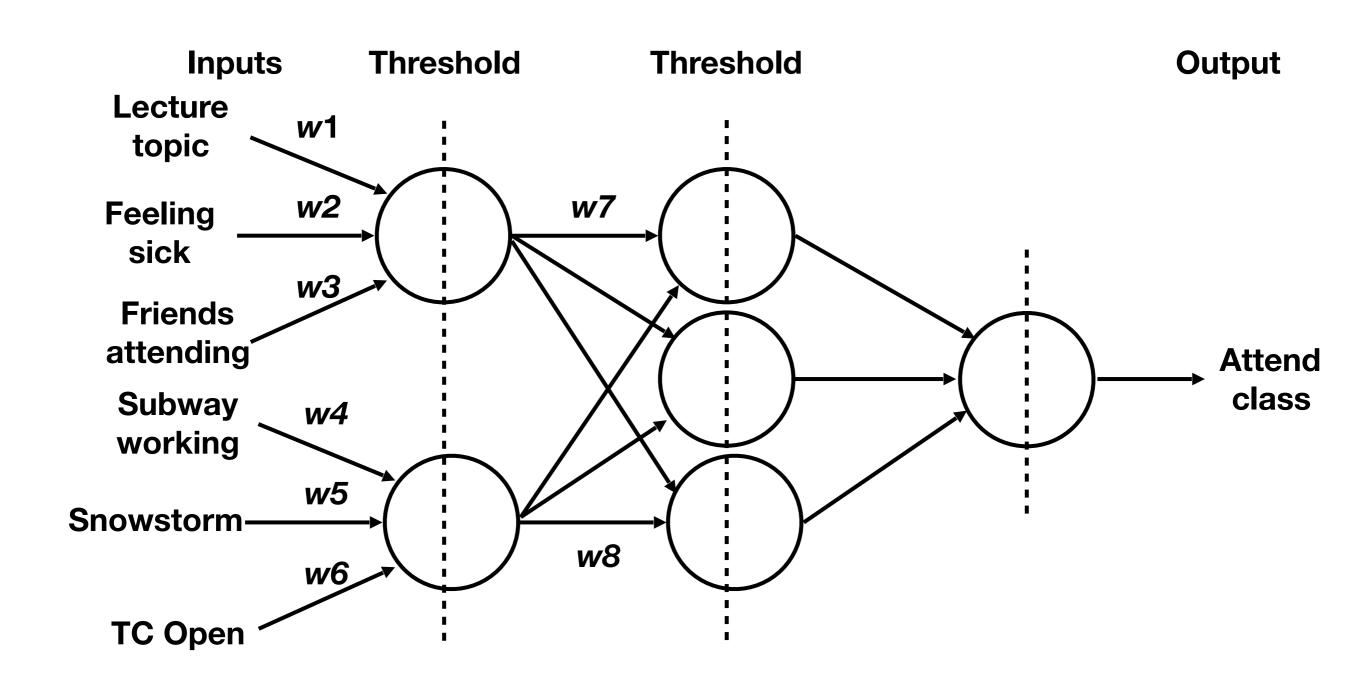


Frank Rosenblatt, 1957









Logic

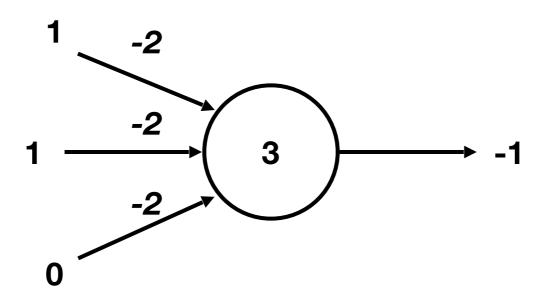
- From the perceptron we can create a NAND gate
- From a NAND gate can create all other logic units (AND, NOR, etc.)
- See Nielson 2016*

Bias

- Another way to describe the threshold
- Negative threshold
- More convenient for notation
- Describes how easy it is to get make the perceptron "fire"

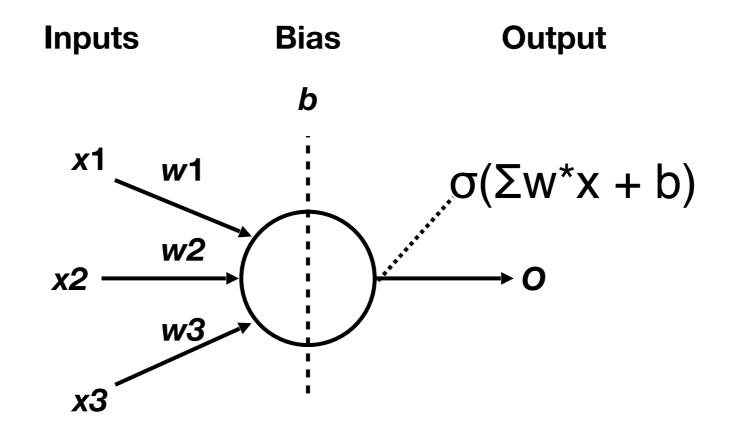
Notation

Inputs Bias Output



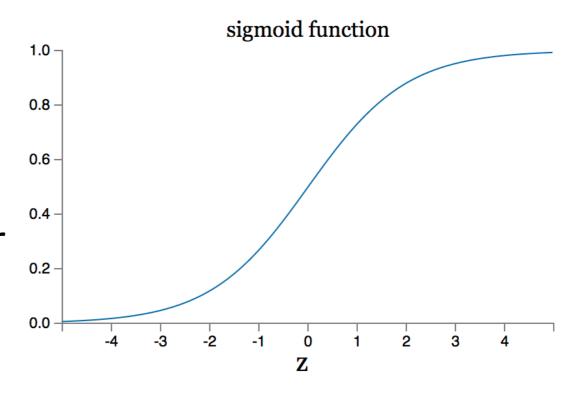
$$(1)^*-2 + (1)^*-2 + (0)^*-2 + 3 = -1$$

- Want to build a learning algorithm
- Could change b or w
- BUT that will cause very large changes
- Network will never "fix"
- Solution: "smooth" the output



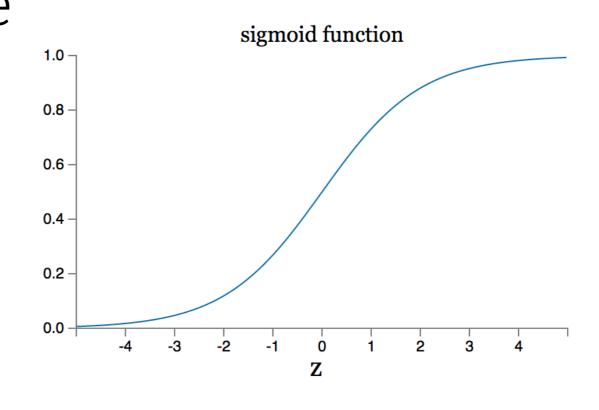
- Sigmoid function
 "smooths" the output
- Makes changing w
 and b less sudden and
 more predictable
- Could use lots of other functions...

$$\sigma(z) \equiv \frac{1}{1 + e^{-z}}.$$

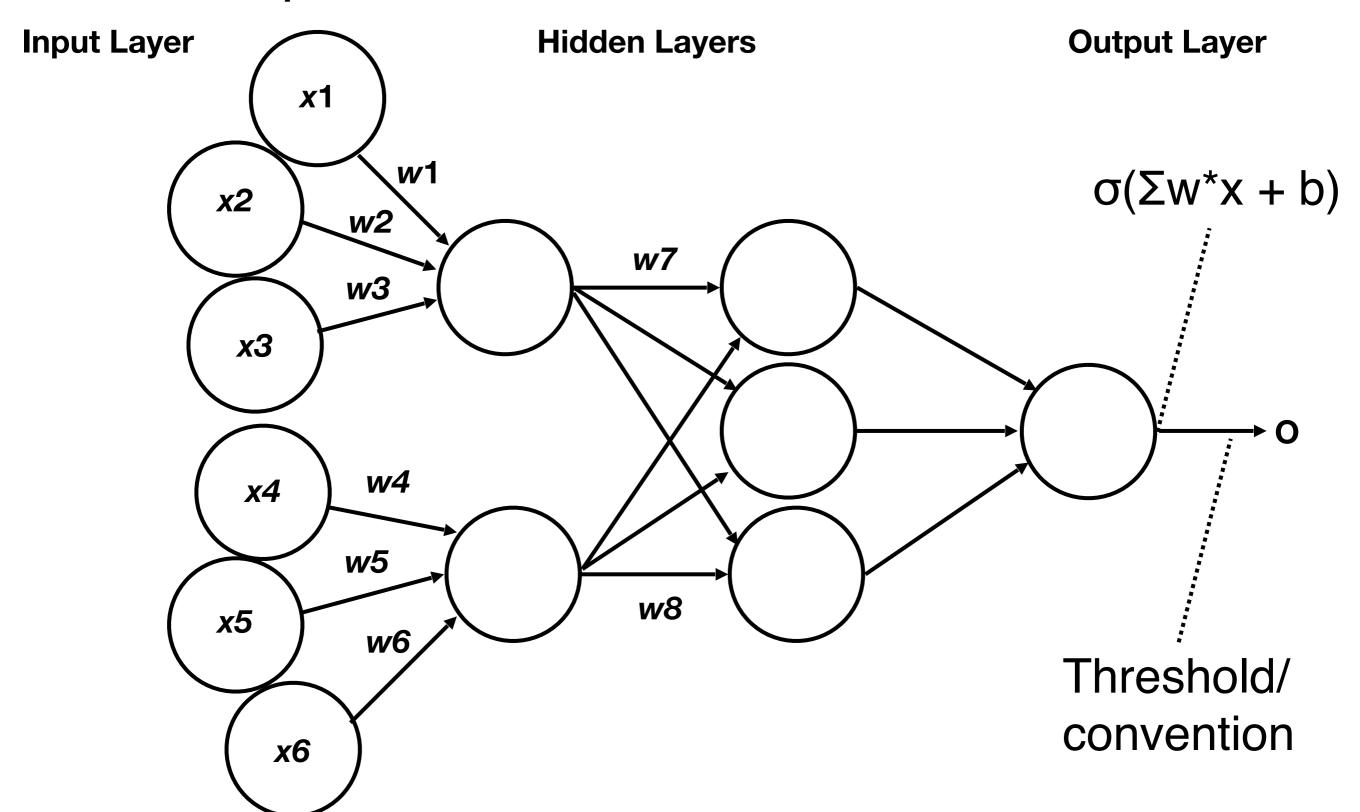


- Perceptrons have 0/1 output
- Sigmoid neurons have
 0 1 output (eg. 0.1,
 0.6778, etc.)
- How to interpret sigmoid neuron output?

$$\sigma(z) \equiv \frac{1}{1 + e^{-z}}.$$



Complete Feedfoward Network



How many Hidden Layers?

- No foolproof method
- The only method is really trial and error
- Heuristics:
 - Theory based starting point?
 - Number of inputs and outputs?

Exercise

Back Propagation

- Need a way to minimize error
- Error is defined by a cost function
- Then we imagine error as a surface that needs to be "searched" for the minimum

Weight