6.034 Exam 3 Cheat Sheet

Neural Nets

Useful Information

- 1) Neural Nets are numerical classifiers with binary (0/1) output
- 2) The neuron is a primitive circuit element
- 3) Forward propagation computes the overall output of a neural net

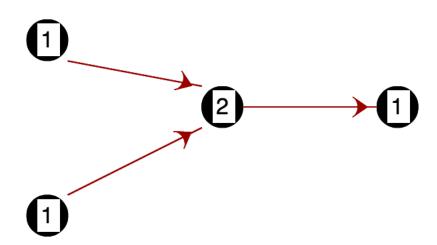
(Input Layer) \rightarrow (Logic Function Layers) \rightarrow Output (0/1)

A single neuron can draw one line and shade above or below it

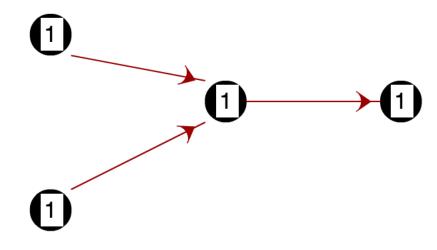
Primitive Logic Functions Computable by a Single Neuron

Note: used in the **logic** layer

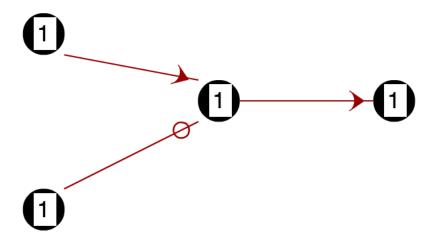
• AND(x,y)



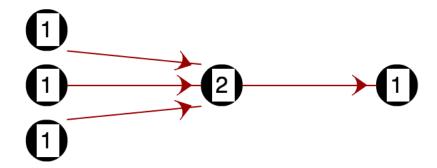
-OR(x,y)



- NOT(x, _)



- * note, the circle on the line means that the weight is -1)
 - "MAJORITY(x1, x2, x3, x4, ...)" (3 input example)



- note, doubling the weight of the bottom input (x3 for instance) makes this gate act like OR(AND(x1, x2), x3)

Helper Functions

Stairstep T(x) =

$$\begin{cases} 1 & \text{if } x \ge T \\ 0 & \text{if } x < T \end{cases}$$

SigmoidS, M(x) =

$$\frac{1}{1 + e^{-S(x-m)}}$$

 $Performance = Accuracy(out*,out) = \tfrac{1}{2}(out*-out)^2$

• * means $\mathbf{desired}$ output

Quick Formulas For Backward Propagation

$$W'_{A \to B} = W_{A \to B} + \Delta W_{A \to B}$$
$$\Delta W_{A \to B} = r \cdot out_A \cdot \delta_B$$

$$\delta_B = \begin{cases} out_B(1-out_B)(out*-out) & \text{if neuron B is in final (output) layer} \\ out_B(1-out_B) \sum_{outgoingC_i} W_{B \to C_i} \delta_{C_i} & \text{if neuron B is not in final (output) layer} \end{cases}$$

Backwards Propagation Steps

- 1. Computing output of each neuron using forward Propagation and $Stairstep_T$ function
- 2. Compute δ_B for final layer
- 3. Compute δ_B for earlier layers
- 4. Compute updates for weights
- 5. Update all weights

Miscellaneous Notes

• You can never classify all points correctly if you have a + data point and a - data point (contraditory) right on top of each other

Overfitting - too strict with regards to the data it's trying to model

Underfitting - too simple with regards to the data it's trying to model

Support Vector Machines

Useful Information

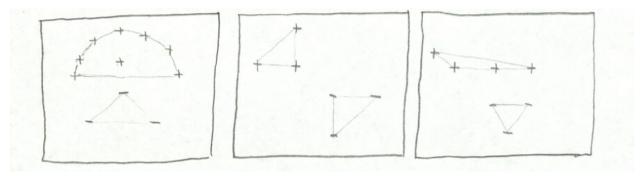
- like Neural Nets, classifies numerical data into two classes: + and -
- draws the decision boundary line that separates the training data with the widest possible margin

Boundaries

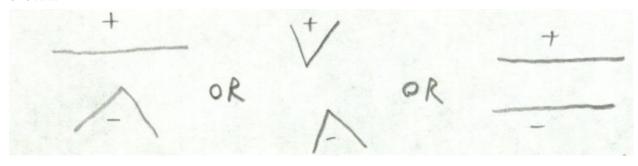
- 1-D just a point
- 2-D some sort of line or curve
- 3-D some sort of plane

How to Draw SVM Boundaries (2D)

- 1. Draw the *convex hulls* for the + and training points.
- a convex hull is the shape you get when you wrap a rubber band around the points and let it contract



- 2. Look at the regions where the convex hulls are closest. $\bf 3$ Cases:



3. The corresponding boundaries look like this:

