Recap: Perceptron

- learning a linear separator
- · corrects itself on each failed prediction in training

Today: Logistic Regression

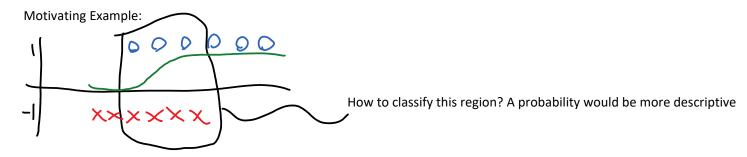
• a probabilistic model - we predict how likely the result is to take on one value over another

Online Learning vs. Batch Learning

- Online Learning takes one example at a time and uses them one by one to train the model
- Batch takes all the data and updates the model using multiple data at once
- Pros of Batch
 - takes less iterations
 - o may be able to see some geometry or trend in all the data
- Pros of Online
 - o model is more adaptable to new data
 - o individual updates are usually simpler, resulting in smaller time complexity
 - o space complexity smaller

Logistic Regression is Classification, but...

- output is discrete valued (-1 or 1) but we output a probability
- P(y=1 | x)
- as opposed to the perceptron algorithm which just gives a 1 or -1



In Logistic Regression we use a logistic curve (green above) to assign probabilities

Sigmoid Function

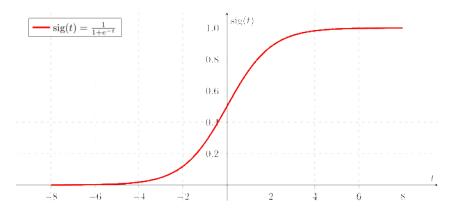
$$\sigma(z) = \frac{e^{z}}{1 + e^{z}} = \frac{1}{1 + e^{-z}}$$

$$-\sin(t) = \frac{1}{1 + e^{-t}}$$

$$0.8$$

$$0.6$$

$$0.2$$



What is
$$\frac{d}{dz} \frac{e^z}{1+e^z}$$
?

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

$$= \frac{1}{|+e^{-2}|} \left(|-|\frac{1}{|+e^{-2}|}\right) = \delta(2)(|-\sigma(2)|)$$

How to make a prediction based on the probability? if P(y=1|x) > 0.5, we predict 1. otherwise, -1

In multiple dimensions, $P(y = 1 | x; w) = \sigma(w^T x) = \frac{1}{1 + e^{-w^T x}}$

When does y=1?

$$\frac{1}{1 + e^{-w^T x}} \ge 0.5$$
$$1 + e^{-w^T x} \le 2$$

$$1 + e^{-w^{T}x} \le 2$$

$$e^{-w^Tx} \le 1$$

 $w^T x \ge 0$ (just like perceptron!)

The decision boundary is linear

We draw papers from an envelope. They're either purple or yellow. Let's say we drew k yellow, n-k purple. What is the probability that maximizes the possibility of this outcome? $\max\left(\theta^k(1-\theta^k)\right)$

turns out it's just k/n which I probably could've told you in middle school