

## Overview:

- Large language models (LLM)  
built on problem of "next word predictor"
- Some perspective on this from brute force  
next letter predictor  $\rightarrow$  language like output
- next word predictor w/ large texts unreasonably  
hard for brute force frequency analysis.

Problem: how to reasonably do next word predictor?  
Answer = neural nets (lots of them)

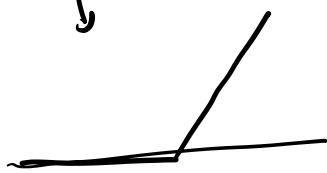
Neural nets give a solution to the following ill posed problem  
given inputs, we believe there should exist  
 $x_0, x_1, \dots$  a function  $f$   
taking these to certain desired outputs  
 $y_0, y_1, \dots$

$f(x_0) = y_0$   $f(x_1) = y_1$  ... is one reasonable way.  
and we believe if we can plug  $\xi$  check values for  $f$ ,  
we'll know if it's right.

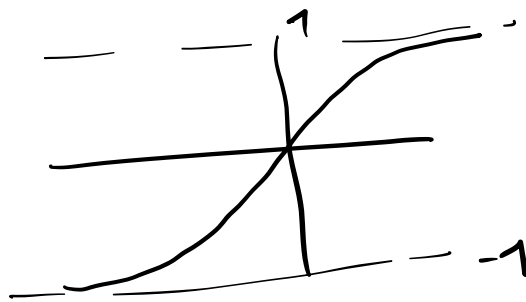
The design/architecture of a net is highly variable,  
depends on problem at hand.

ReLU Activation / similar for visual data (handwriting)

Sigmoid Activation for boolean / logic / classifier

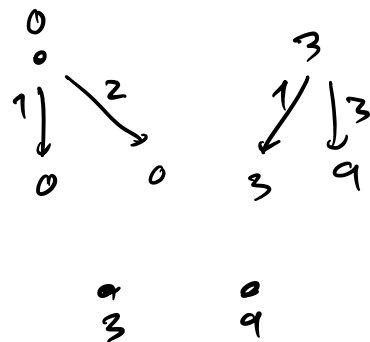
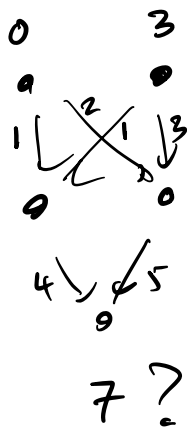
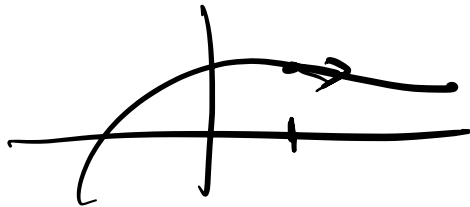


ReLU



Neural nets have the flavor of linear approximation /  
regression, but actually very nonlinear.

Gradient descent:  
 method of approximating solns to eqns  
 $f(x) = 0$



$w_1 = 4 \quad \swarrow \quad \searrow \quad 5 = w_2$

$12 + 45 = 67$

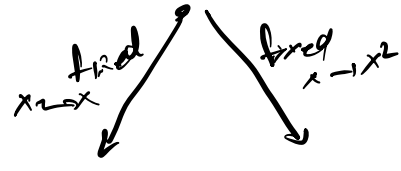
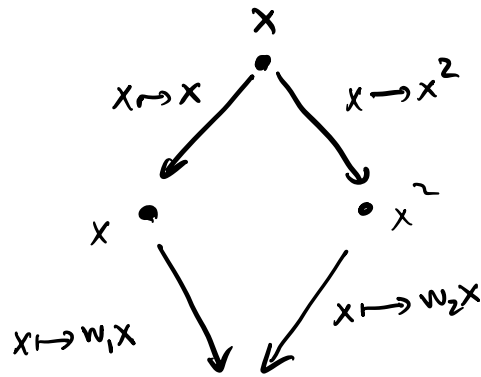
"  
 $3w_1 + 5w_2$

$w_1 \rightarrow w_1 - 3/10$

$w_2 \rightarrow w_2 - 5/10$

$$f(x) = \sin x$$

$$\sin 30 = \frac{1}{2}$$



$$\sin(5) = .087$$

$$\sin(40) = 0.64$$

$$\sin(20) = 0.34$$

