

# Week 3 Notes

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# Big Data

# Introduction

- Every object generates data
- Why is this trend emerging now? Because now we have the technology to digitize all existing knowledge, whereas this thing did not exist earlier
- Information is coming from all different kinds of sources, which did not exist before
- A byproduct of our increasing use of technology
- 2020, data volume would be 40 zabytes
  - 40 zabytes = grains of sand on earth \* 75
  - Data processing in last two years = data proc in last 3000 years

## Analogies

Copernicus collected astronomical data – data for the visible world

Microscope opened up the invisible world

Atomic world was opened up by electron microscope

Supervisible world - Big data is a microscope

We collect lots of data and we use powerful algorithms to parse through this data

# Emerging Big Data

## Before

1. We thought of things
2. Wrote them down = knowledge

## Now

1. Bunch of information- not knowledge
2. Play around with it until it becomes knowledge

Every action that we do creates data – that data is mostly meaningless in the raw form until someone contextualizes it and gives meaning to it

# Why is Big Data Happening Now?

1. Development of internet
  - a. Rapidly increasing the amount of data available
2. Ubiquity of small scale devices that can act as crowdsourced sensors
3. Accelerating data storage capacity and computing power at low cost
  - a. Moore's law
  - b. Development of cloud computing, e.g., AWS
  - c. Development of distributed platforms e.g., Hadoop, Apache Spark
4. Development of ML algorithms to process this data
  - a. One has led to the other Big data spurred ML, which in turn led to need for more data

# Data Case: Social Media

## What data could be collected?

- Interests
- Likes/posts/messages/friends
- Photographs
- Relationship status
- Birthday
- Work and educational history, etc.

## Who could use this data?

- Finding potential matches – new feature coming up
- Advertisements
- Cambridge Analytica – Political parties around the world to spread targeted political campaigns
  - American and Indian voters, etc.

# Data Case: Vending Machine

What data could be collected?

- What products sell the most
- Payment information about people (if they use electronic payment)
- Time/days when purchases peak or fade
- Locations of vending machines where people buy these products
- Locations of products inside the machines which were picked up

Who could use this data?

- Could help them decide where to place physical stores/distribution centers, etc. so as to maximize their profits
- Could also help in the design of better vending machines

## Case Study: Flu Prediction

Old method – Accumulating info submitted by doctors about patient visits

2 weeks to reach CDC

Google uses search results of “flu” to predict outbreak in real-time

Issues:

- Flu mentioned several times in the news...ppl interested in this..drove up Google Search

## Unique Case Studies

Big data has allowed us to uncover fairness issues

Big data has been used for disaster response

- Earthquake in Haiti

Big data for uprisings in Tunisia

Target example...Our buying patterns can tell more about our pregnancies than our actual families

# Concerns with Big Data

Anything that's going to change the world, by definition should have the ability to change it for the worse

Data revolution also leading to issues

- Somebody is listening in
- Big brother
- My device knowing me

Huge implications for how we interact with machines

# Deep Learning

# Deep Learning

## ► Classification:

- 1. Classification Problems
- 2. Decision Trees
- 3. Random Forest
- 4. GRT
- 5. Neural Network



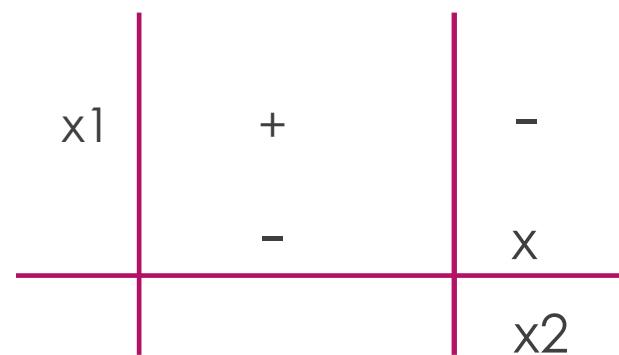
Diff adjust (1920-2010)



1960-1970

- Lots of temporary data
- Lots of computer power

# Initial Problems



XOR Example

Neural networks

- Multiple layer perceptron (MLP)
- ↔ NN
- Building block

# Activation Functions

Non-Linear Activation Functions

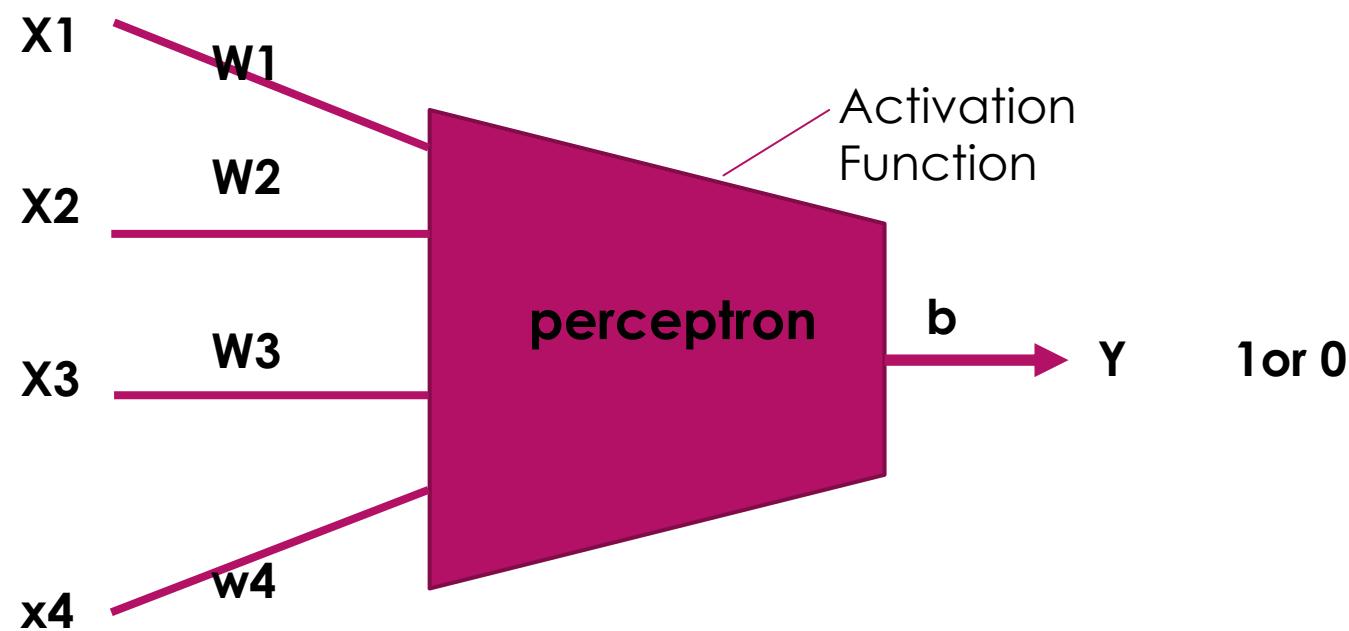
- 1.Sigmoid Function
- 2. ReLU ( Rectified Linear Unit)
- 3. Leaky ReLU
- 4.Tan H

$$\left\{ \begin{array}{l} w_1x_1 + w_2x_2 + w_3x_3 + w_4x_4 \geq b \\ w_1x_1 + w_2x_2 + w_3x_3 + w_4x_4 < b \end{array} \right.$$

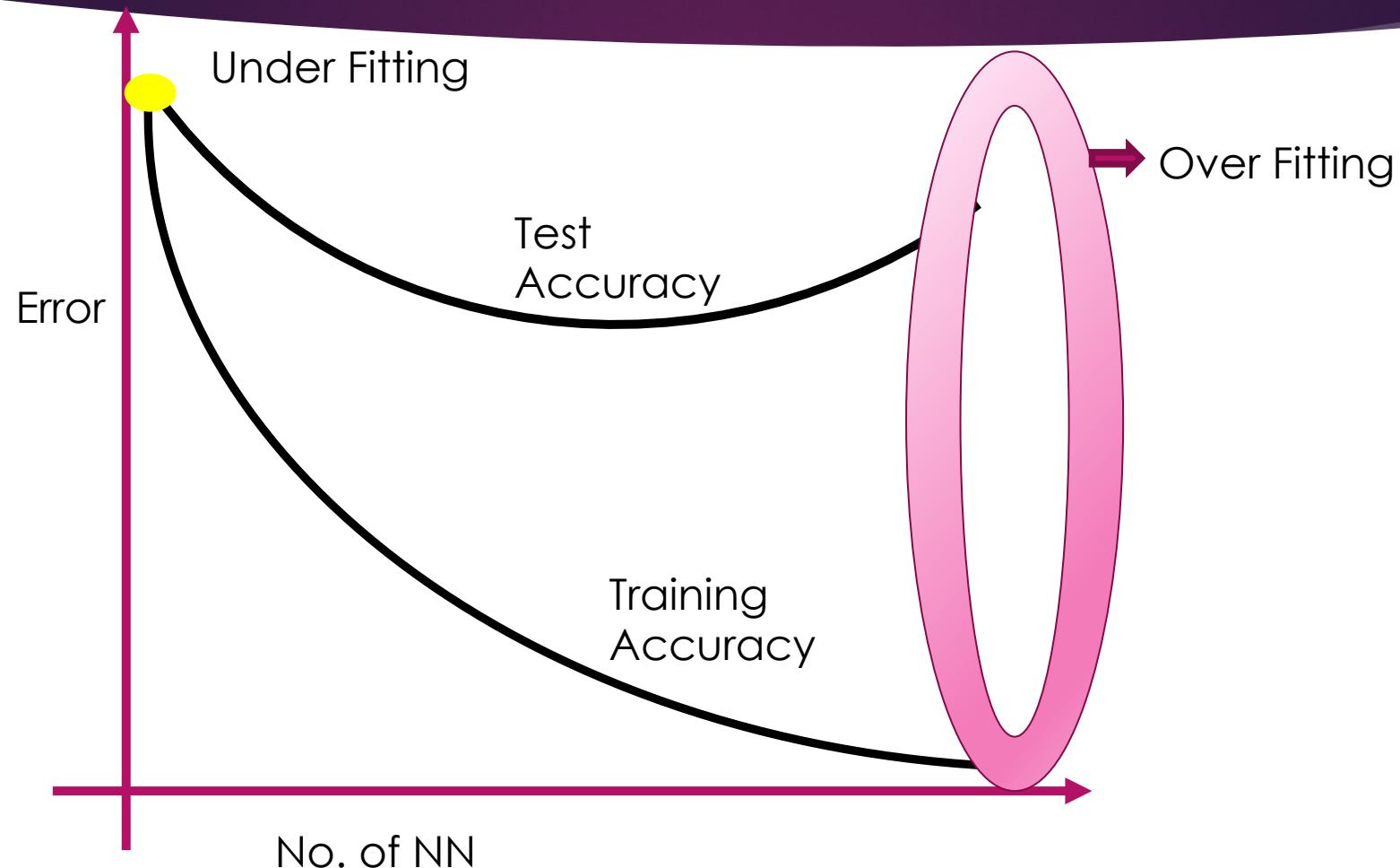
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$$\frac{1}{H * e^{-(w_1x_1 + w_2x_2 + w_3x_3 + w_4x_4 + b)}}$$

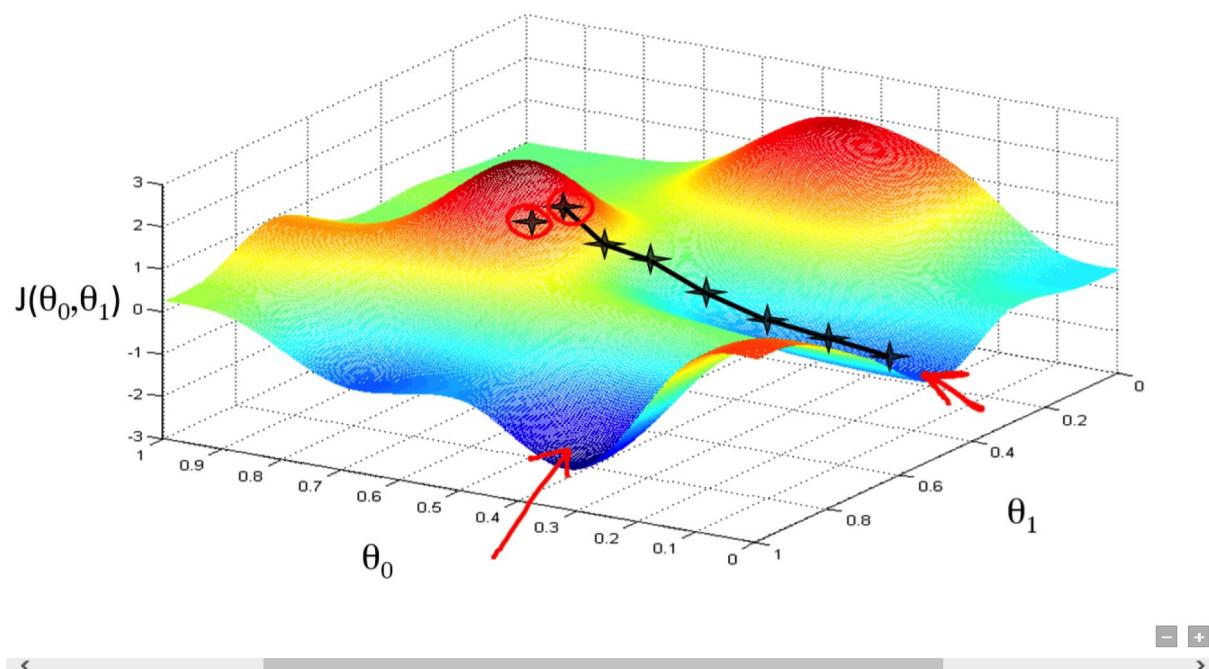
# Perceptron



# Bias variance trade off



# Gradient Descent



- Deep network are trained using gradient descent
- Back propagation is often how you do gradient descent with neural networks