

Introduction to Deep Learning



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General Information

- Teaching assistants
 - Jyoti Kumari
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Course structure

- Introduction to big data problem & representation learning
- Overview of linear algebra and probability ✓
- Basics of feature engineering ←
- Neural network ←
- Introduction to open-source tools ✓
- Deep learning network ✓
- Regularization }
● Optimization }
- Advanced topics ✓
- Practical applications ✓

Evaluation policy

- Real-time quiz (4 Nos) - 20%
- Assignments (2 Nos) - 20% ←
- Midsem - 25% → Quiz - Real time 1
- Endsem - 35% → Assg - 24hrs

Project

- Assignments can be done in a group
- A group can have maximum of 3 students
- You need to create a video and upload on youtube
- Topic details will be provided later

Books

- Deep Learning - Ian Goodfellow, Yoshua Bengio, Aaron Courville ✓
- The Elements of Statistical Learning - Jerome H Friedman, Robert Tibshirani, Trevor Hastie
- Reinforcement Learning: An Introduction - Richard S Sutton, Andrew G Barto
- Neural Network and Learning Machines - Simon S. Haykin
- Neural Networks and Deep Learning - Charu Agarwal

Introduction

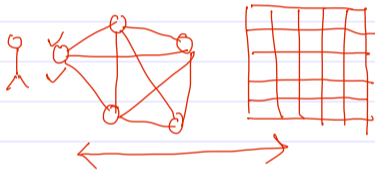
Problem space

- Problems — a matter or situation regarded as unwelcome or harmful and needing to be dealt with and overcome
- Target is to solve the same on a computer

Sorting
①

TSP

chess ②



Object identification \checkmark
 \hookrightarrow Car ③

C prog
 \hookrightarrow recursion ④

undecidable

Problem space

- Problems — *a matter or situation regarded as unwelcome or harmful and needing to be dealt with and overcome*
- Target is to solve the same on a **computer**
- Problems can be **intellectually challenging** for human being but relatively **straight forward** for a computer
 - Travelling salesman problem, chess
- Problems can be **easy** for common people but **difficult** for computer (even expressing it in a formal way)
 - Identifying an object, car (say), in a picture

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- Primary focus will be in the *second category* problems

Problem Solving Strategies for Big Data

- Need to **solve** problems efficiently and accurately when the input data is huge (\sim GB, TB order)
 - Finding a deterministic algorithm is **difficult**
 - Need to find out features ✓
 - Requires significant effort for model building ✓
 - Need to have domain knowledge ✓
 - Statistical inference is found to be suitable ✓
 - Feature selection is not crucial ✓
 - Model will learn from past data ✓
-

Applications: Computer vision

- 2d to 3d conversion
- Street view generation
- Image classifications
- Image segmentation



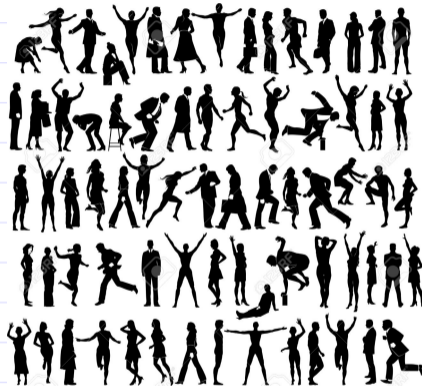
2D



3D







Applications: Activity Recognition

- Recognize activities like walking, running, cooking, etc. from still image or video data



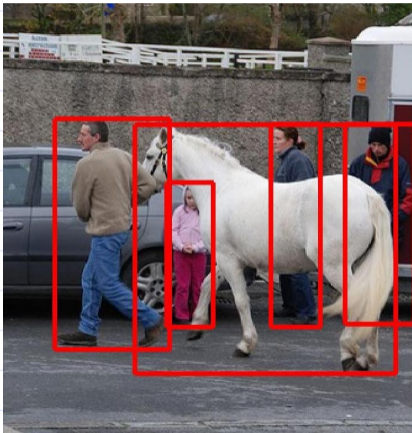
Applications: Image Captioning

- Automated caption generation for a given image

Describes without errors	Describes with minor errors	Somewhat related to the image	Unrelated to the image
 <p>A person riding a motorcycle on a dirt road.</p>	 <p>Two dogs play in the grass.</p>	 <p>A skateboarder does a trick on a ramp.</p>	 <p>A dog is jumping to catch a frisbee.</p>
 <p>A group of young people playing a game of frisbee.</p>	 <p>Two hockey players are fighting over the puck.</p>	 <p>A little girl in a pink hat is blowing bubbles.</p>	 <p>A refrigerator filled with lots of food and drinks.</p>
 <p>A herd of elephants walking across a dry grass field.</p>	 <p>A close up of a cat laying on a couch.</p>	 <p>A red motorcycle parked on the side of the road.</p>	 <p>A yellow school bus parked in a parking lot.</p>

Applications: Object Identification

- Identify objects in still image or in video stream



Applications: Automated Car

- Self driving car



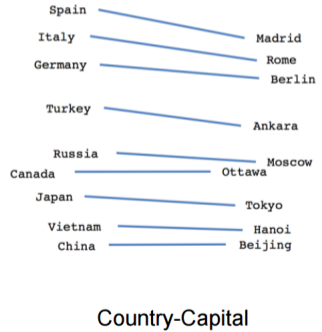
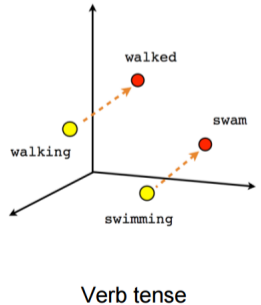
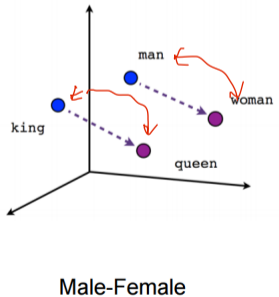
Applications: Drones & Robots

- Managing movement of robot or drones



Applications: Natural Language Processing

- Recommender system
- Sentiment analysis
- Question answering
- Information extraction from website
- Automated email reply




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Applications: Speech processing

- Conversion of speech into text
- Generation of particular voice for a given text



Other possible applications

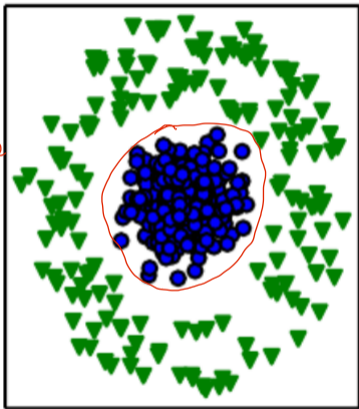
- Language translation
 - Weather prediction
 - Genomics
 - Drug discovery
 - Particle physics
 - Surveillance
 - Cryptography and many more.
- 

Issue of Representation

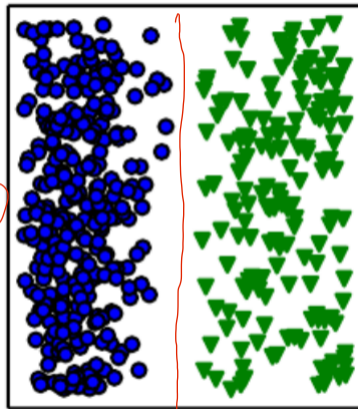
- Representation of data in an efficient/structured manner is crucial for solving problems more effectively
 - Searching of a set of elements in a given list (sorted/unsorted)
 - Arithmetic operations on Arabic and Roman numerals
 - Primality test of n when n is represented as $11111 \dots 111$ (n -number of one)
- Structured representation can help in predicting future values

Choice of Representation

Cartesian coordinates



Polar coordinates



Manages
→ Color
→ Hard

Learning representation/feature

- Traditional approaches

- Pattern recognition ✓
 - Input, output of the problem

ML

- End to end learning | DL

- System automatically learns internal representation |

AI-ML Tasks

- Heavily depends on features ✓
- Requires good domain knowledge ✓
- Feature extraction is not easy job ←
 - Identify a car ✓✓
 - How to describe wheel ✓ ○ ○ |
 - Shadow/brightness ✓✓
 - Obscuring element ✓✓

Representation Learning

- Learned representation often result in **better** performance compared to hand design
- Allows the system to rapidly **adapt** to new task
- Need to discover a good set of **features**
- Manual design of features is nearly **impossible**

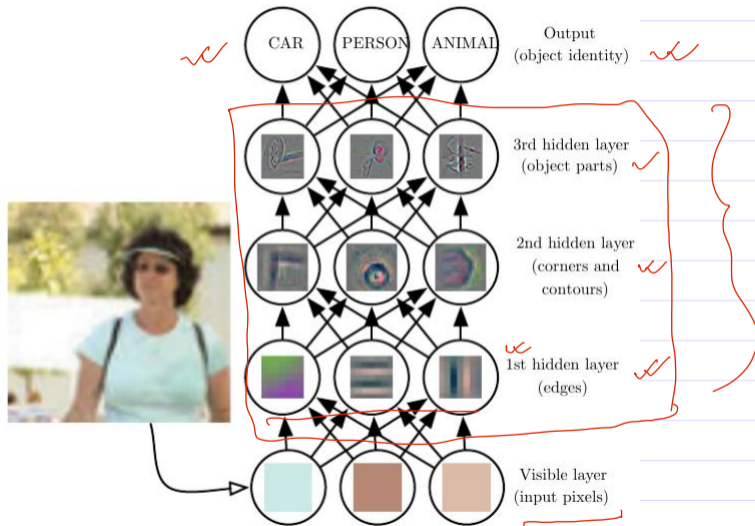
Design of Features

- Goal is to separate out **variation factors**
- These factors are separate **sources of influence**
- It may exist as unobserved object or unobserved forces that **affect observable quantity**
 - **Speech** - Factors are age, sex, accent, etc
 - **Image** - Position, color, brightness, etc.

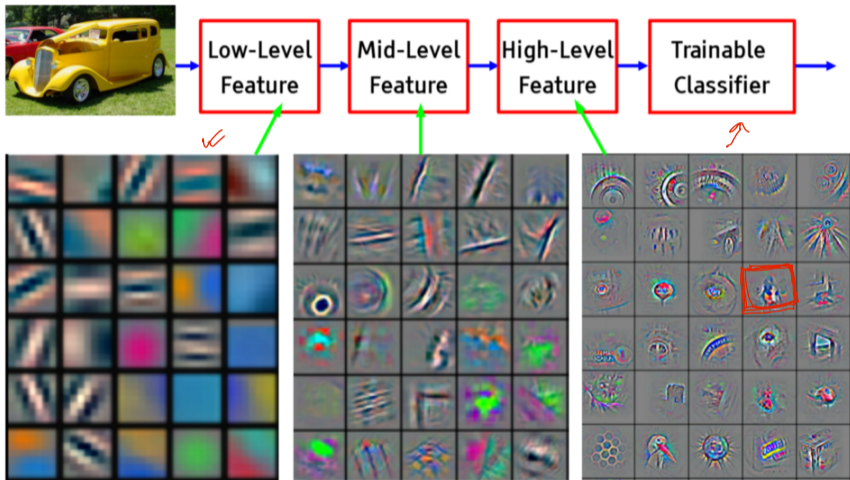
Deep Learning

- Try to address the problem of representation learning
- Representation are expressed in terms of other simpler representation
- Develop complex concept using simpler concept

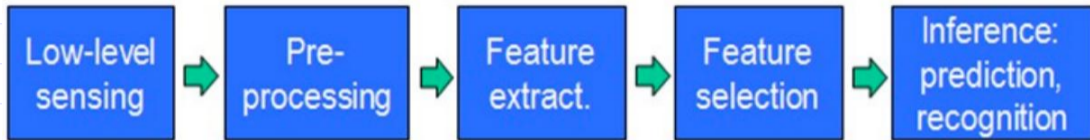
Simple to Complex Features



Simple to Complex Features

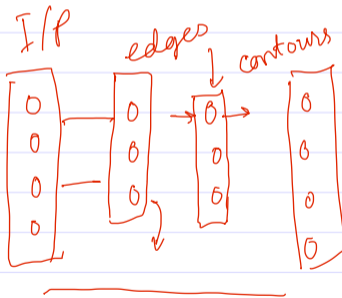


Conventional Machine Learning



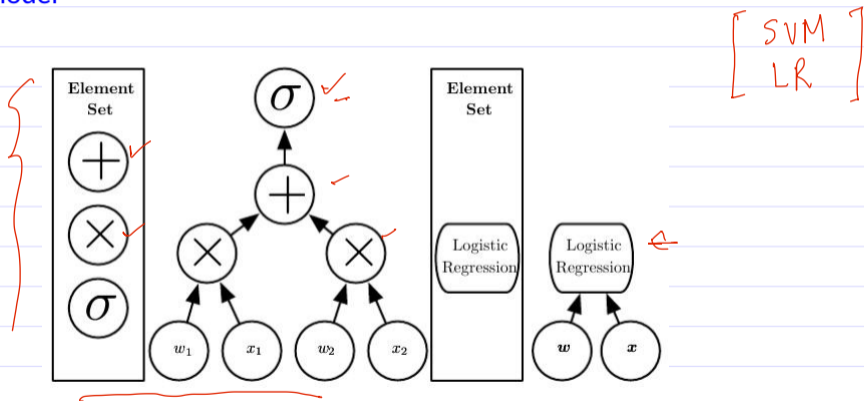
Deep Learning Model

- Feed-forward deep network or multilayer perceptron ✓
- Mathematical functions that map input to output ✓
- Composed of simpler functions ✓
- Each layer provides a new representation ✓
- Learning right representation ✓

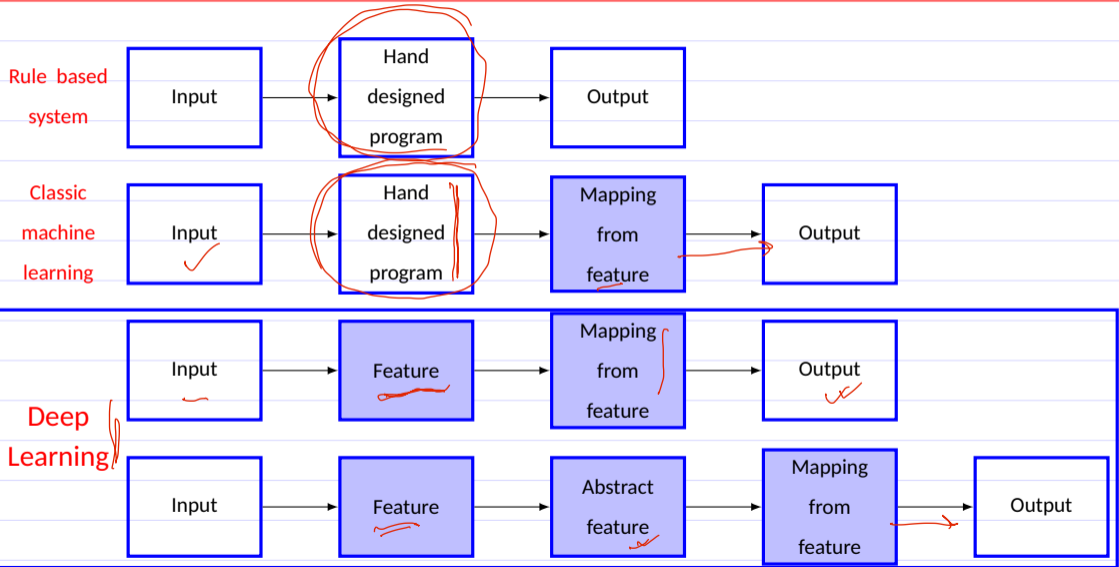


Depth of network

- Number of sequential instruction must be executed to evaluate the architecture
 - Length of the longest path
- Depth of the model



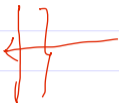
Representation learning



History

- Has many names and view point

- Cybernetics (1940-1960) ✓
- Connectionism (1980-1990) (neural net) ✓
- Deep learning (2006+) ✓



- More useful as the amount of data is increased

- Models have grown in size as increase in computing resources

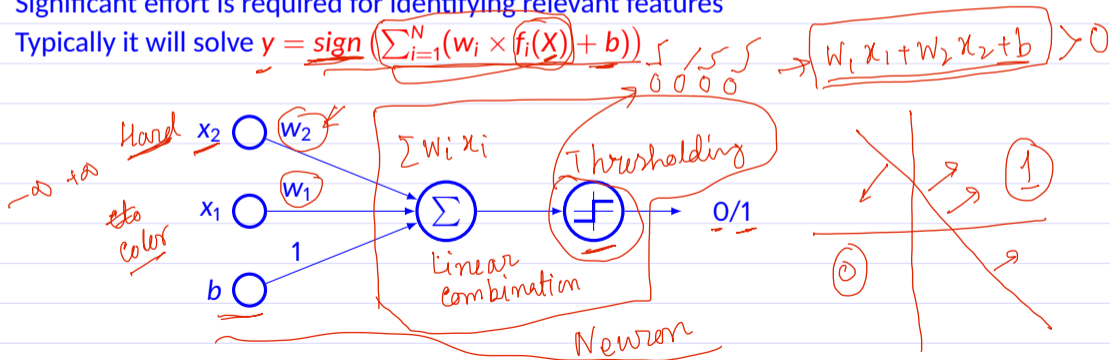
- Solving complex problem with increasing accuracy ✓

Learning Algorithm

- Early learning algorithm
 - How learning happen in brain?
 - Computational model of biological learning
- Neural perspective of DL
 - Brains provide a proof by example
 - Reverse engineer the computational principle behind the brain and duplicate its functionality

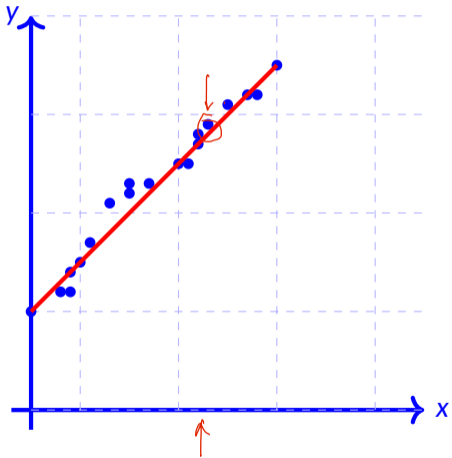
History of basic model

- The first learning machine: the Perceptron
 - Built at Cornell, 1960
- Perceptron was linear classifier on top of simple feature extractor
- Most of the practical applications of ML today use glorified linear classifiers or glorified template matching.
- Significant effort is required for identifying relevant features
- Typically it will solve $y = \text{sign}(\sum_{i=1}^N (w_i \times f_i(X)) + b)$

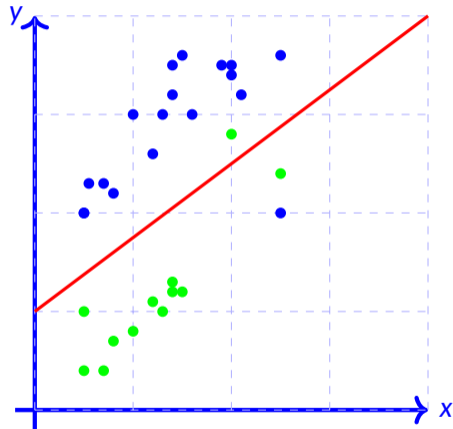


Broad Categories of Problem

• Regression ✓

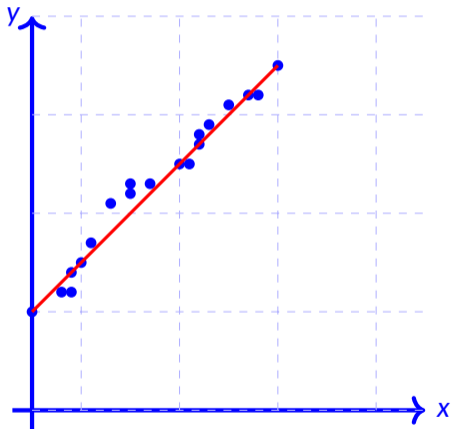


• Classification ✓

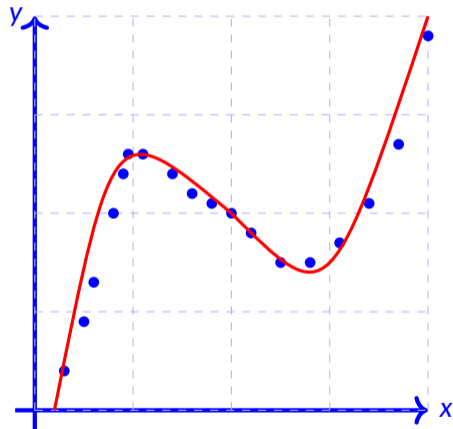


Regression

- Regression (linear)

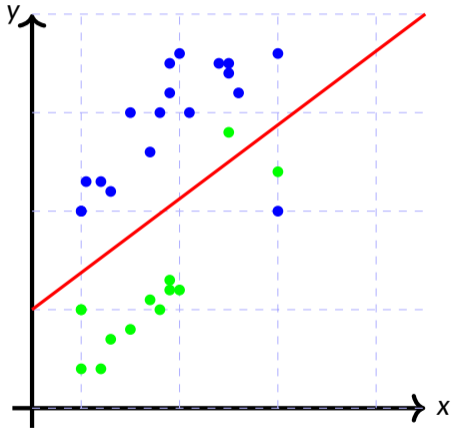


- Regression (Non-linear)

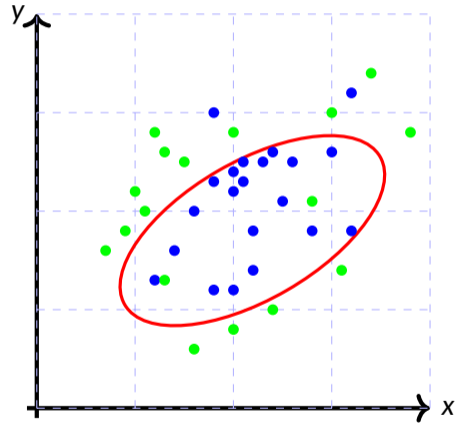


Classification

• Linear

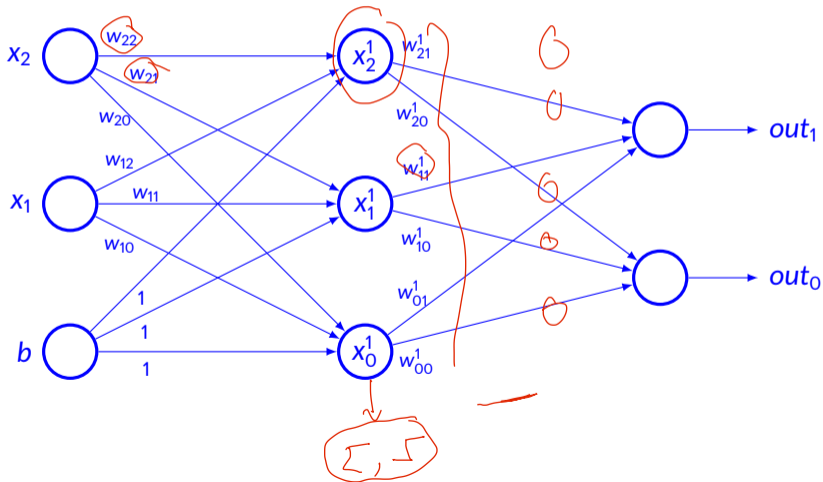


• Non-linear

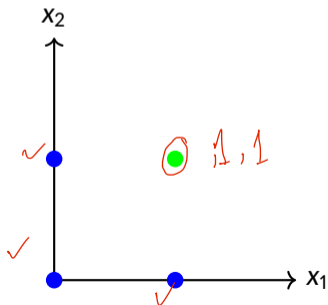
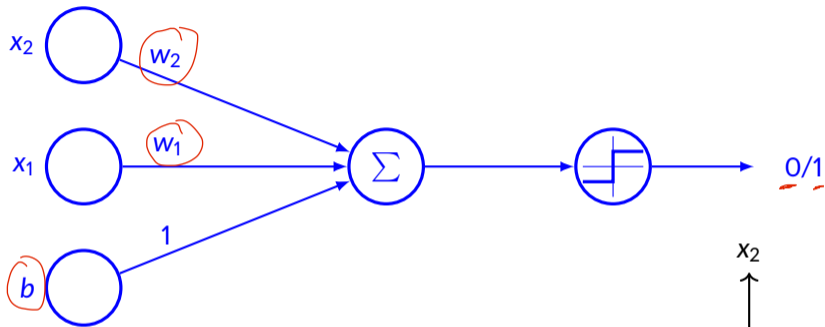


Artificial Neural Network

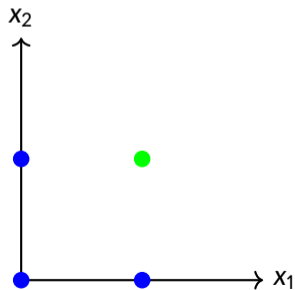
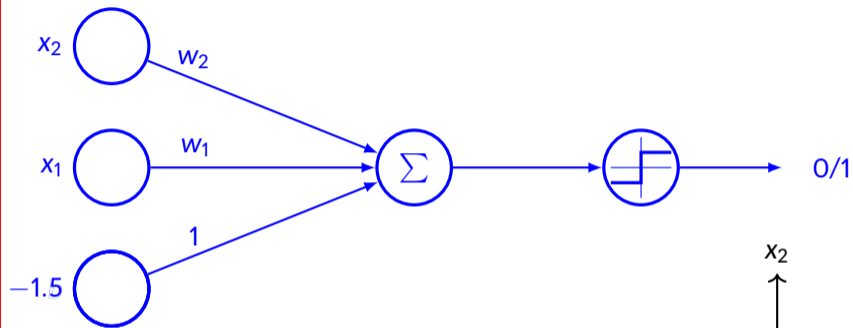
- A simple model



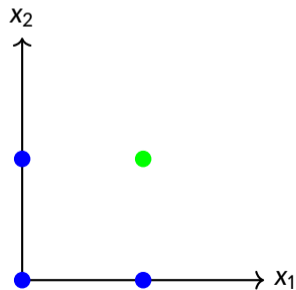
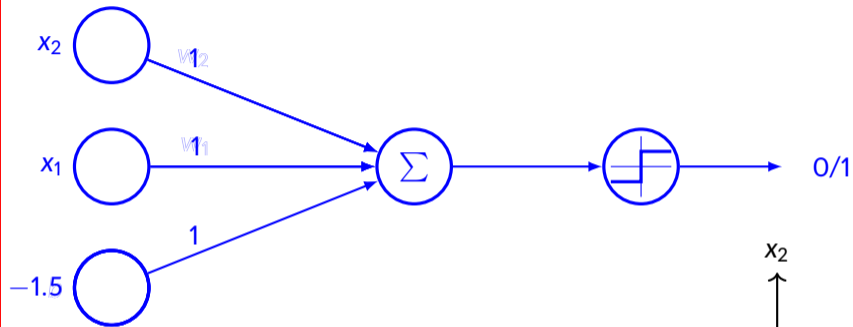
Example NN: AND gate



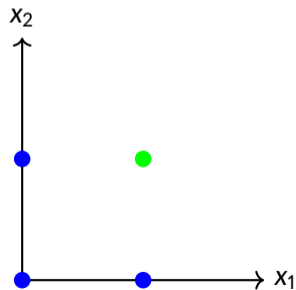
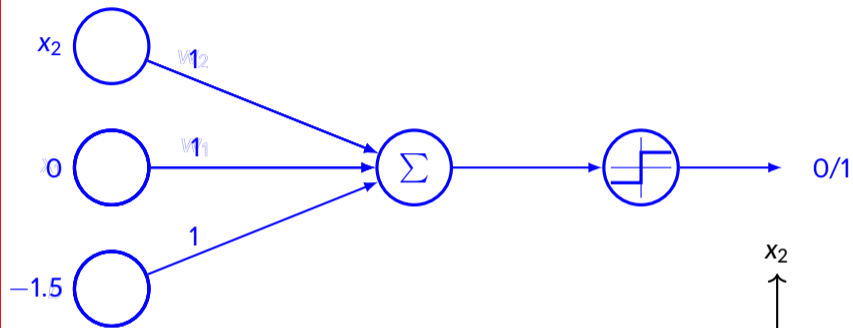
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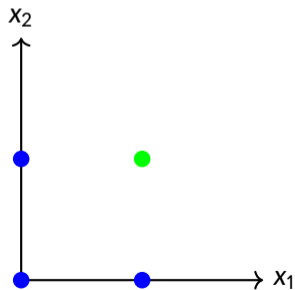
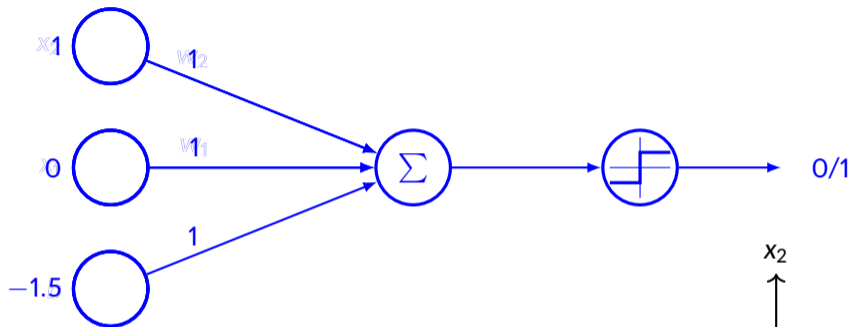
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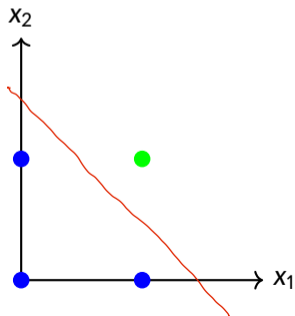
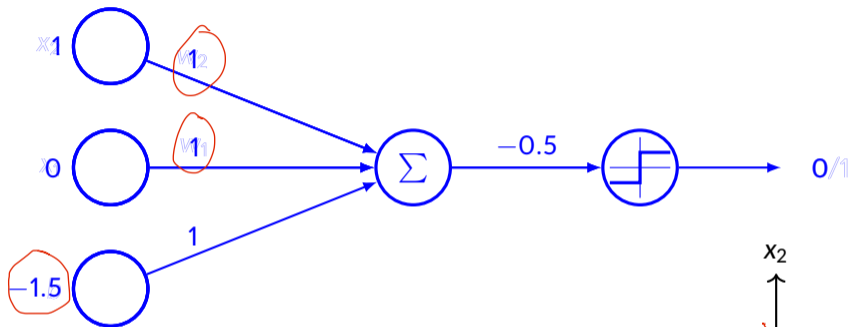
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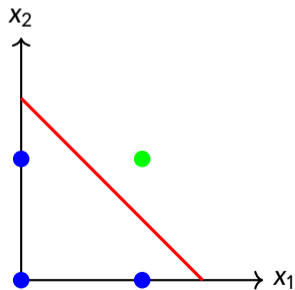
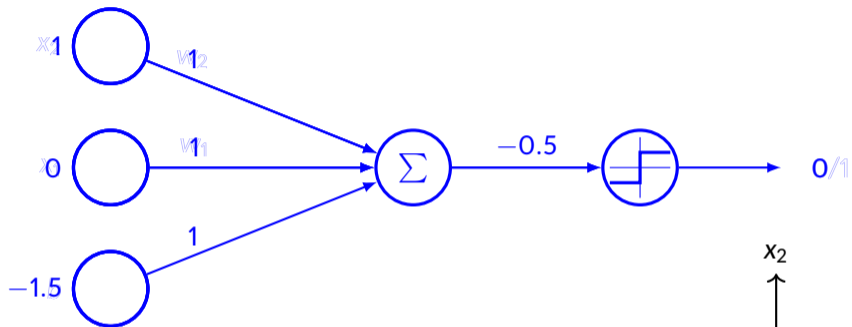
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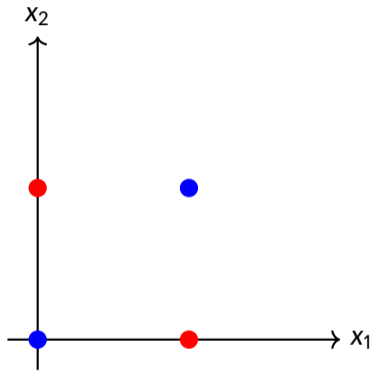
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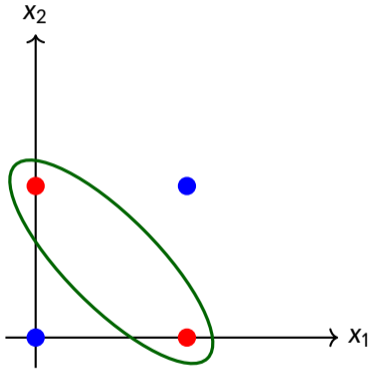
Example NN: AND gate



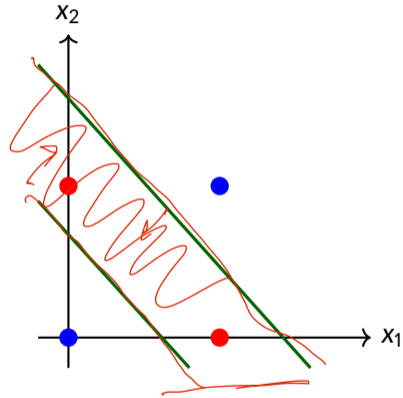
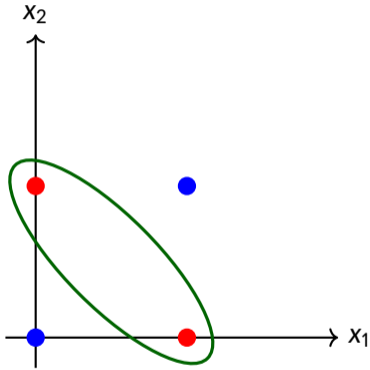
Example NN: XOR gate



Example NN: XOR gate



Example NN: XOR gate



Distributed representation

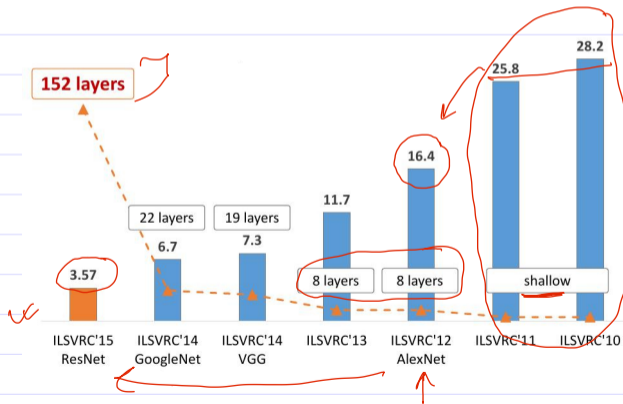
- Each input should be represented by **many** features
- Each feature should be involved in the representation of **many** possible inputs
- Example: car, flower, birds — red, green, blue
 - 9 neurons
 - For each combination of color and object
- Distributed neurons ✓
 - 3 Neurons for color ✓
 - 3 Neurons for object ←
 - Total 6 neurons

9



Popularization of Neural Network

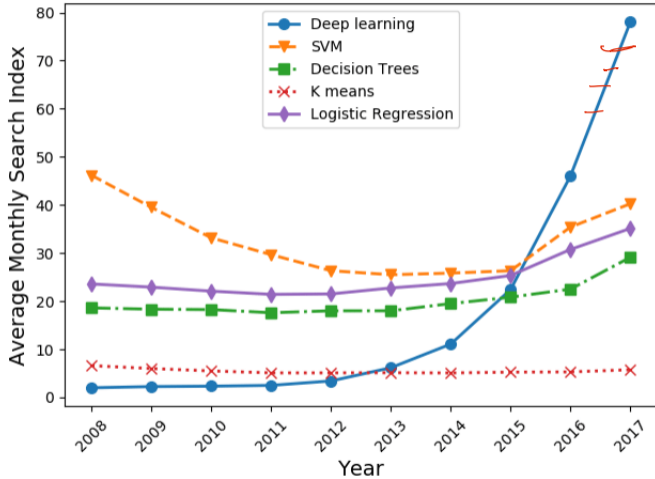
- Most of the theory of neural network was developed in the 1980s
- Started gaining popularity around 2012
 - Geoffrey Hinton and Alex Krizhevsky winning the ImageNet competition where they beat the nearest competitor by a **huge margin** (2012)



Popularity

- Increase data size ✓
 - Computing resources are available
 - Accepting performance 5000 labeled example per category
 - 10 million for human performance
- Increasing model size ✓
- Increasing accuracy, complexity, real world impact ✓
- Used by many companies
 - Google, Microsoft, Facebook, IBM, Baidu, Apple, Adobe, Nvidia, NEC, etc. ↑
- Availability of good commercial & open-source tools
 - Theano, Torch, DistBelief, Caffe, TensorFlow, Keras, etc. ↓

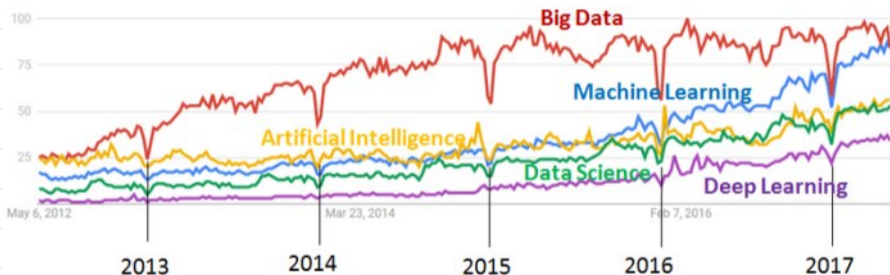
DL Trend



Search trend in Google

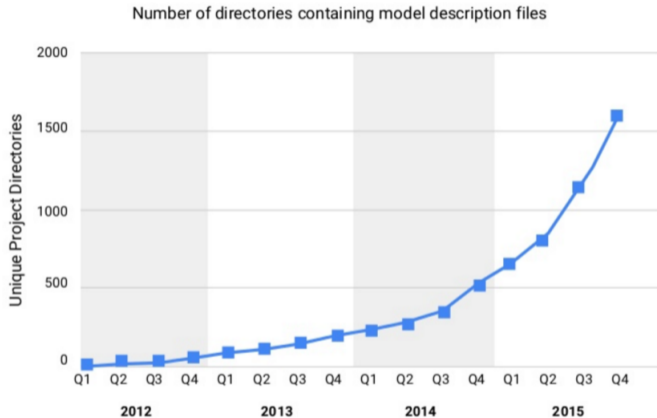
Google Trends, May 2012 - April 2017, Worldwide

Big Data, Machine Learning, Artificial Intelligence, Data Science, Deep Learning



CS551

AI/DL in Google



Across many products/areas

- Apps
- Maps
- Photos
- Gmail
- Speech
- Android
- YouTube
- Translation
- Robotics Research
- Image Understanding
- Natural Language Understanding
- Drug Discovery



Artificial Intelligence is the New Electricity — Andrew Ng

Artificial Intelligence is the New Electricity — Andrew Ng

Thank you!