

Introduction to Deep Learning



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General Information

- Teaching assistants
 - Niraj Kumar
 - Aakash
- **Course webpage:** `www.iitp.ac.in/~arijit/`, **then follow** Teaching

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

- Mid-sem - 20%
- Project - 40%
- End-sem - 40%
- **75% attendance is compulsory**

Project

- **Group wise project**
- **A group can have maximum of 3 students**
- **Final presentation of project will be held before your end semester**

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**

Acknowledgement

- Deep Learning Book by Ian Goodfellow, Yoshua Bengio, Aaron Courville
- Presentations by Yann LeCun, Geoff Hinton, Yoshua Bengio
- Various websites for images
- **Dr. Jacob Minz (Synopsis)**
- **IIT KGP Batch of 2001**
 - Joydeep Acharya (Hitachi)
 - Sanjeev Kumar (Liv.AI)
 - Mithun Dasgupta (Microsoft)
 - Amit Kumar (Avnera)
 - Mrinmoy Ghosh (Facebook)
 - Animesh Datta (Qualcomm)
 - Bhaskar Saha (PARC)
 - Banit Agrawal (Facebook)

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**

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- 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



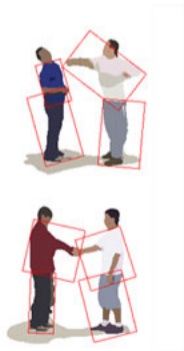
2D



3D

Applications: Activity Recognition

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



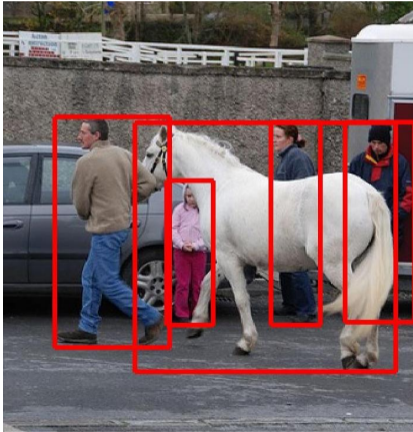
Applications: Image Captioning

- Automated caption generation for a given image

Describes without errors	Describes with minor errors	Somewhat related to the image	Unrelated to the image
 <p data-bbox="369 448 560 484">A person riding a motorcycle on a dirt road.</p>	 <p data-bbox="624 448 833 469">Two dogs play in the grass.</p>	 <p data-bbox="906 448 1115 484">A skateboarder does a trick on a ramp.</p>	 <p data-bbox="1188 448 1397 484">A dog is jumping to catch a frisbee.</p>
 <p data-bbox="369 673 560 709">A group of young people playing a game of frisbee.</p>	 <p data-bbox="624 673 833 709">Two hockey players are fighting over the puck.</p>	 <p data-bbox="906 673 1115 709">A little girl in a pink hat is blowing bubbles.</p>	 <p data-bbox="1188 673 1397 709">A refrigerator filled with lots of food and drinks.</p>
 <p data-bbox="369 903 560 939">A herd of elephants walking across a dry grass field.</p>	 <p data-bbox="624 903 833 939">A close up of a cat laying on a couch.</p>	 <p data-bbox="906 903 1115 939">A red motorcycle parked on the side of the road.</p>	 <p data-bbox="1188 903 1397 939">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



Image source: Internet

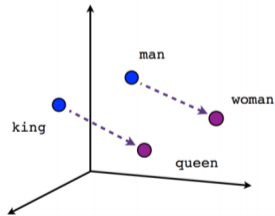
Applications: Drones & Robots

- Managing movement of robot or drones

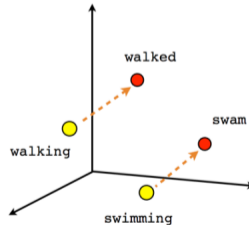


Applications: Natural Language Processing

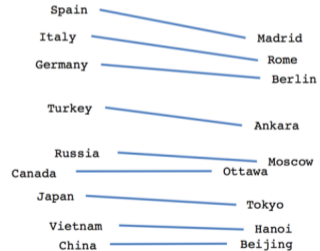
- Recommender system
- Sentiment analysis
- Question answering
- Information extraction from website
- Automated email reply



Male-Female



Verb tense



Country-Capital

Applications: Speech processing

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



Other possible applications

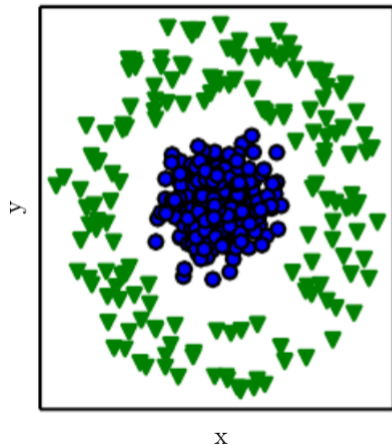
- Language translation
- Weather prediction
- Genomics
- Drug discovery
- Particle physics
- Surveillance
- Cryptography and many more.

Issue of Representation

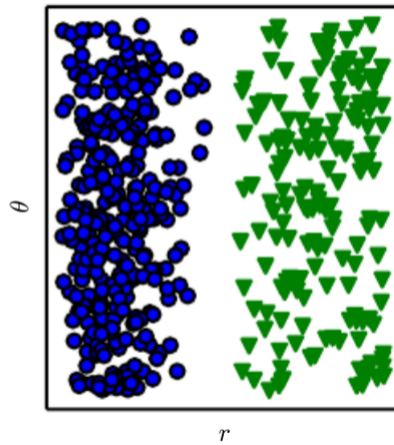
- Representation of data in an efficient/structured manner is **crucial** for solving problems more effectively
 - Searching of a set of elements in a given list (sorted/unsorted)
 - Arithmetic operations on Arabic and Roman numerals
 - Primality test of n when n is represented as $11111 \dots 111$ (n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

Choice of Representation

Cartesian coordinates



Polar coordinates



Learning representation/feature

- **Traditional approaches**
 - **Pattern recognition**
 - **Input, output of the problem**
- **End to end learning**
 - **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**

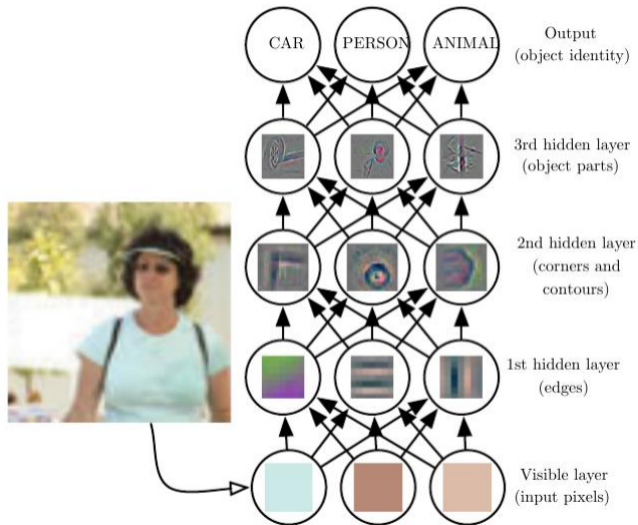
Design of Features

- 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.

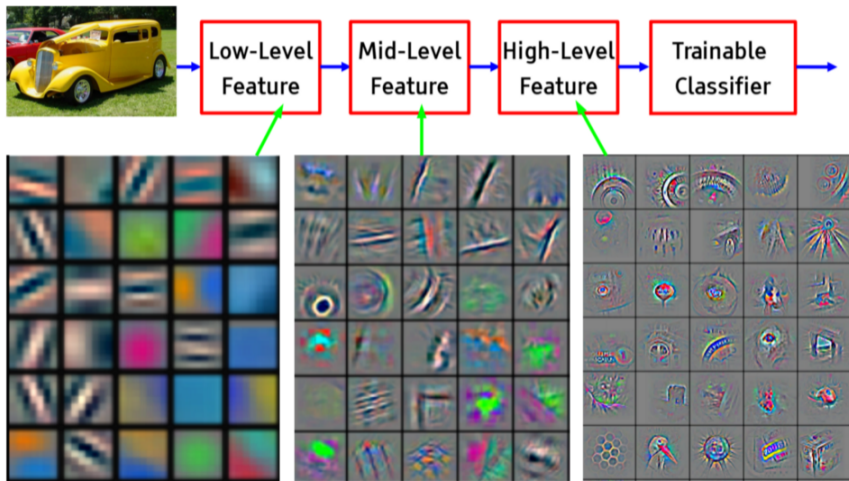
Deep Learning

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

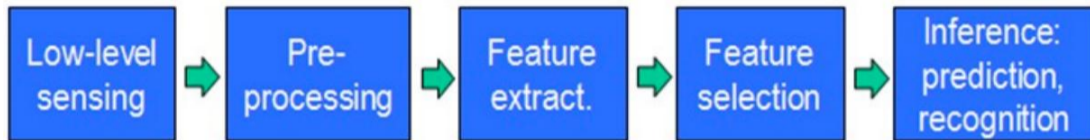
Simple to Complex Features



Simple to Complex Features



Conventional Machine Learning

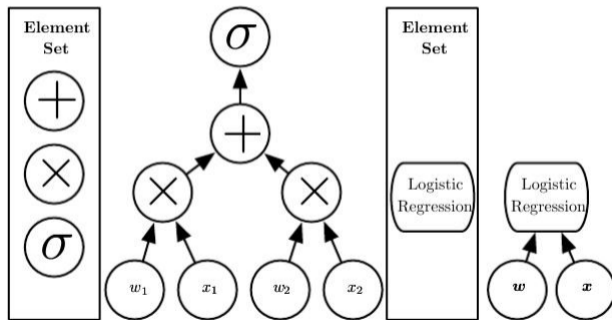


Deep Learning Model

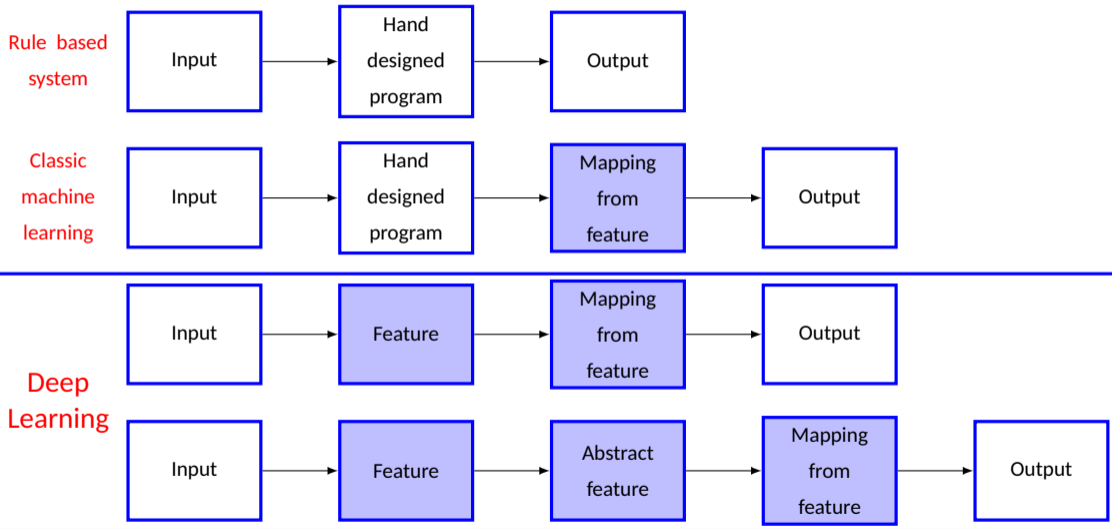
- **Feed-forward deep network or multilayer perceptron**
- **Mathematical functions that map input to output**
- **Composed of simpler functions**
- **Each layer provides a new representation**
- **Learning right representation**

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

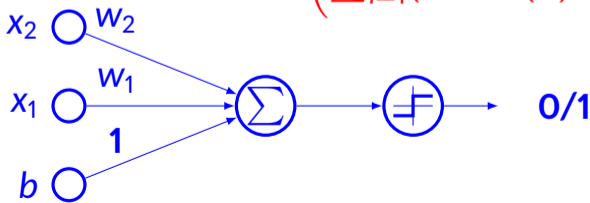
- Has many names and view point
 - Cybernetics (1940-1960)
 - Connectionism (1980-1990) (neural net)
 - Deep learning (2006+)
- More useful as the amount of **data is increased**
- Models have grown in size as **increase** in computing resources
- Solving complex problem with **increasing accuracy**

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

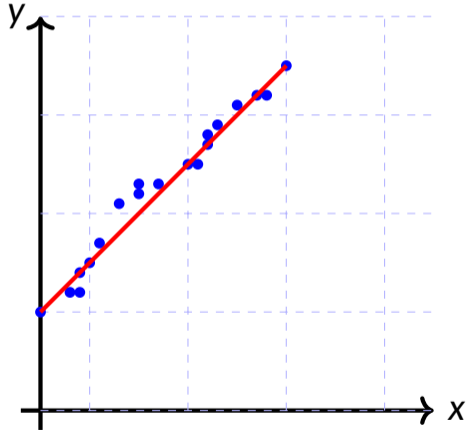
History of basic model

- 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 $y = \text{sign} \left(\sum_{i=1}^N (w_i \times f_i(X) + b) \right)$

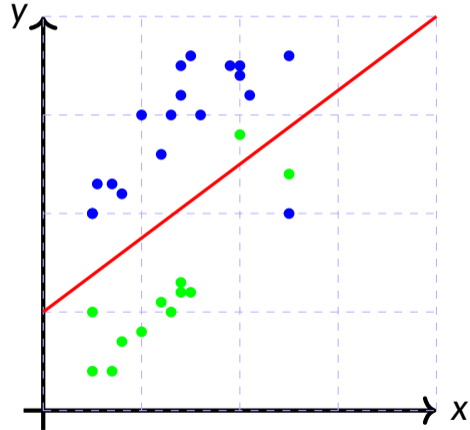


Broad Categories of Problem

- Regression

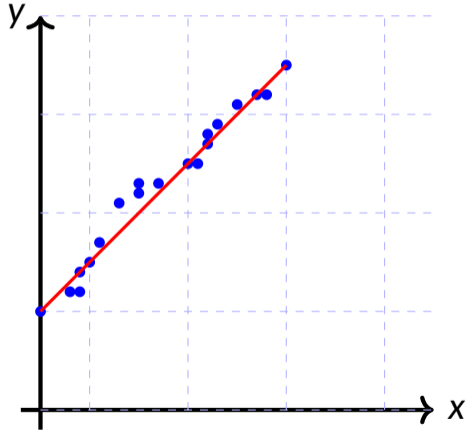


- Classification

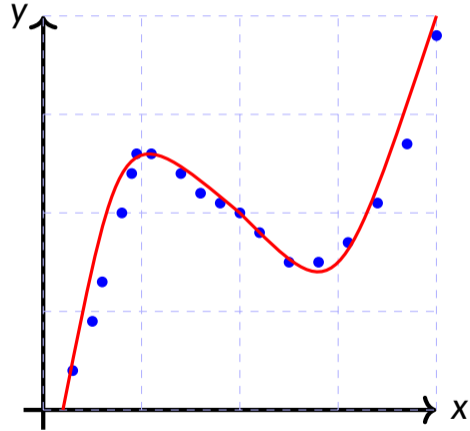


Regression

- Regression (linear)

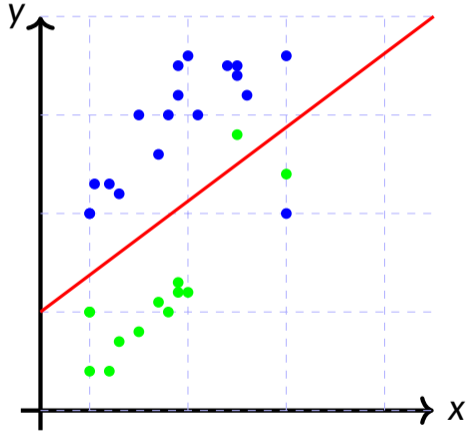


- Regression (Non-linear)

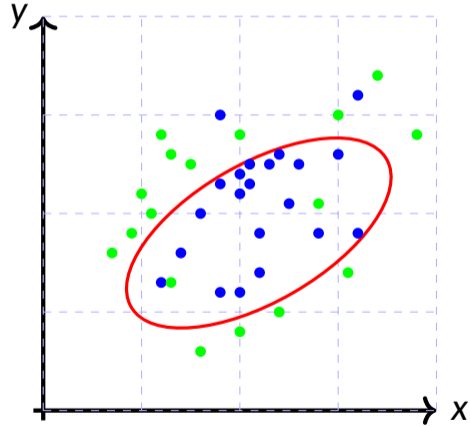


Classification

- Linear

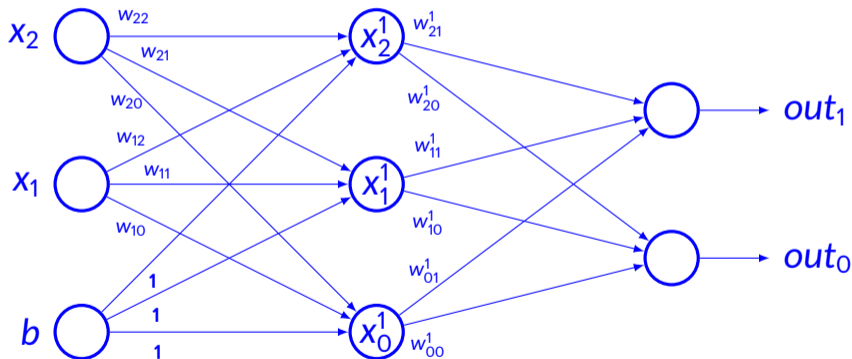


- Non-linear

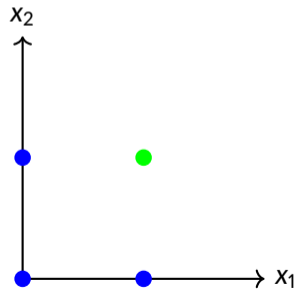
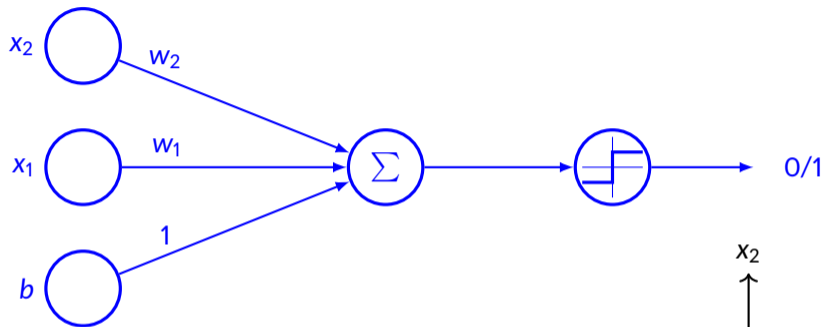


Artificial Neural Network

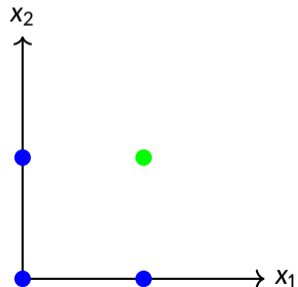
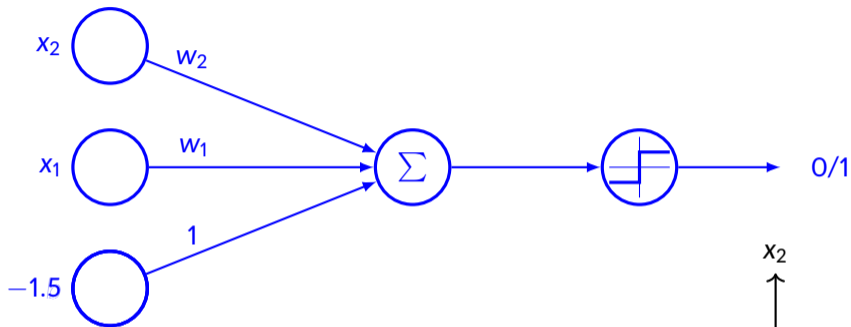
- A simple model



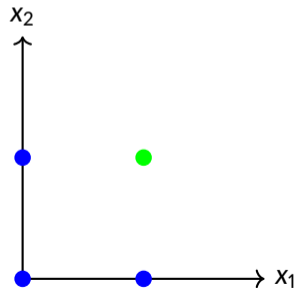
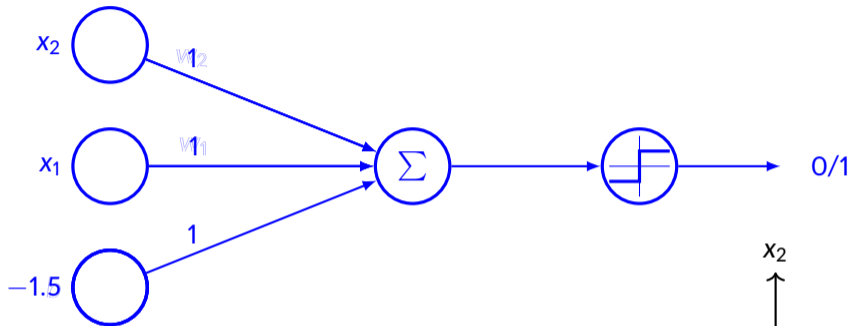
Example NN: AND gate



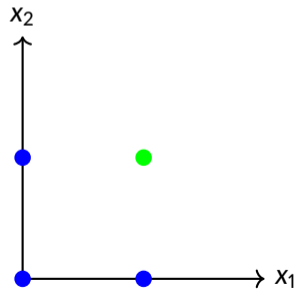
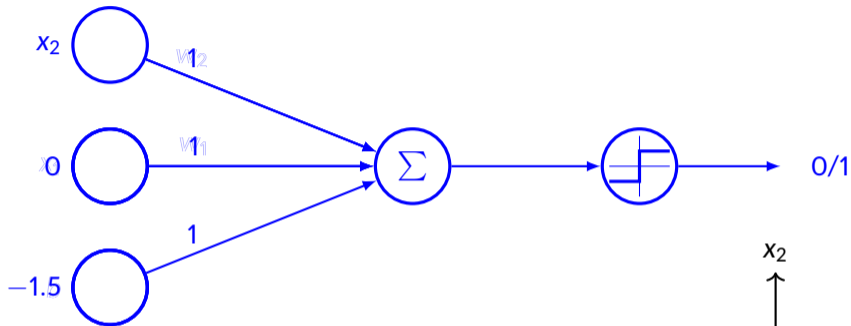
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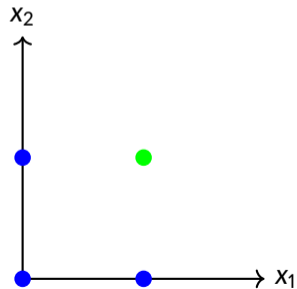
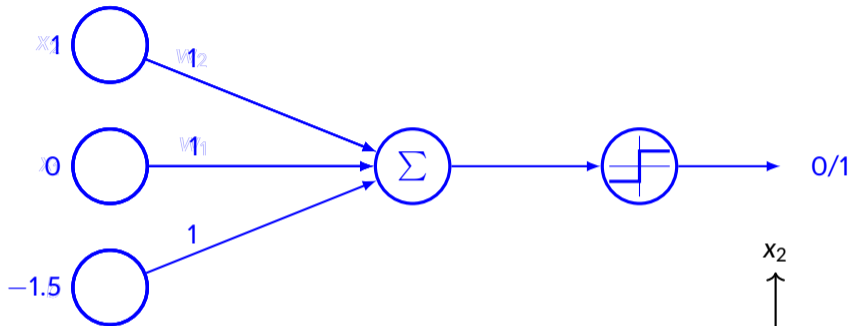
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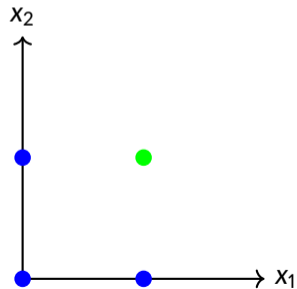
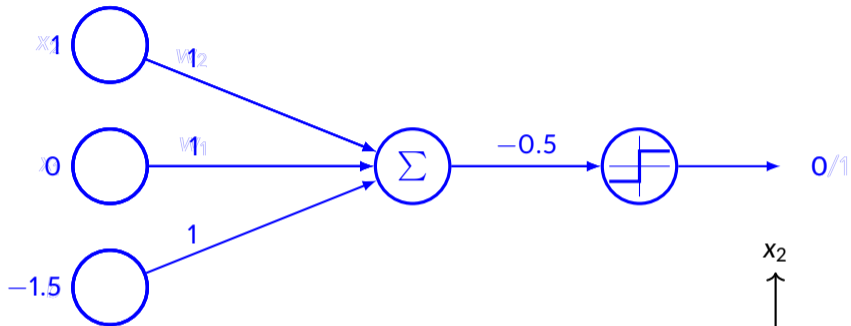
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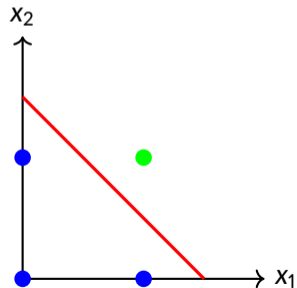
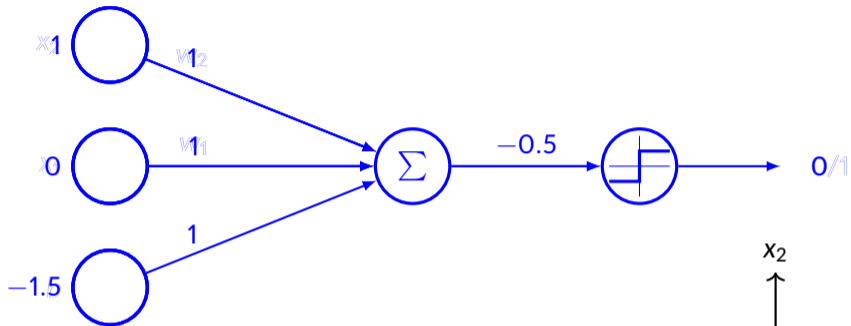
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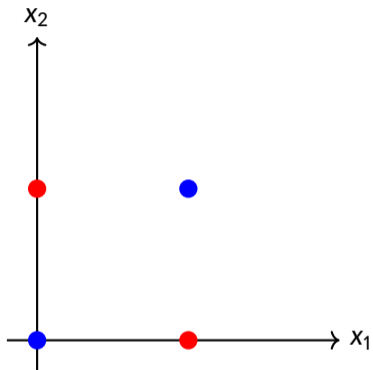
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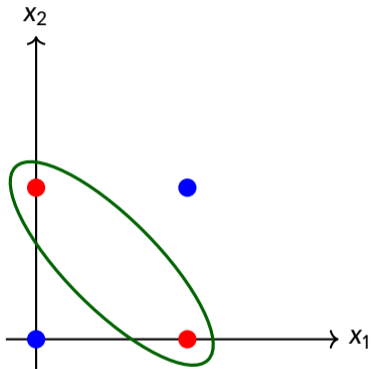
Example NN: AND gate



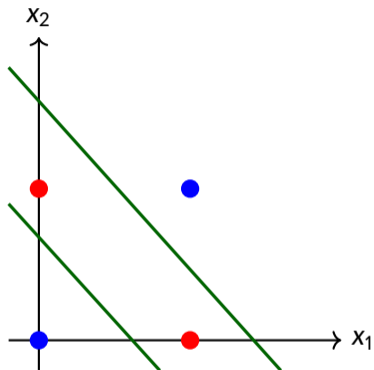
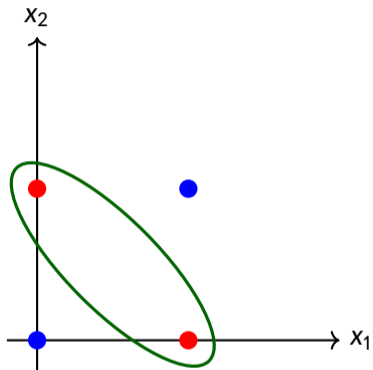
Example NN: XOR gate



Example NN: XOR gate



Example NN: XOR gate

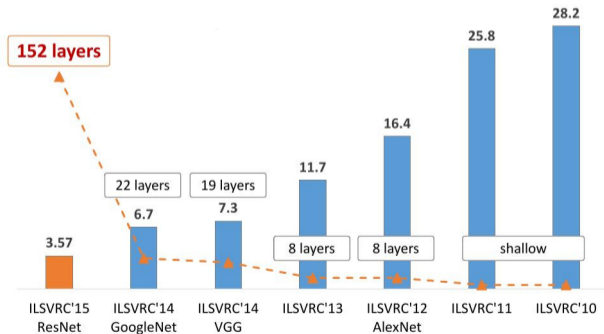


Distributed representation

- 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

Popularization of Neural Network

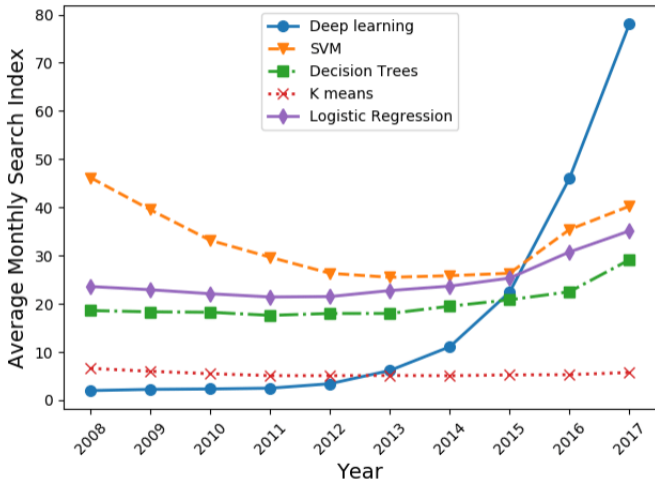
- Most of the theory of neural network was developed in the 1980s
- Started gaining popularity around 2012
 - Geoffrey Hinton and Alex Krizhevsky winning the ImageNet competition where they beat the nearest competitor by a **huge margin** (2012)



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.

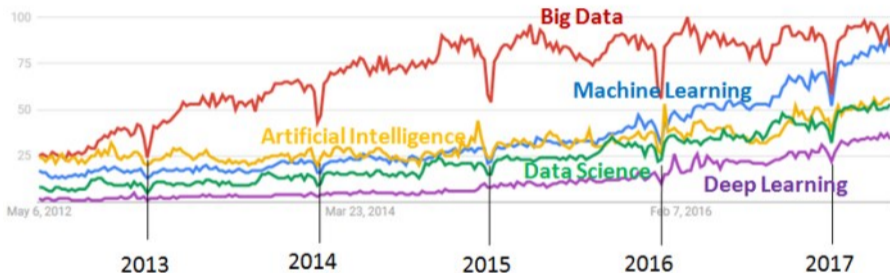
DL Trend



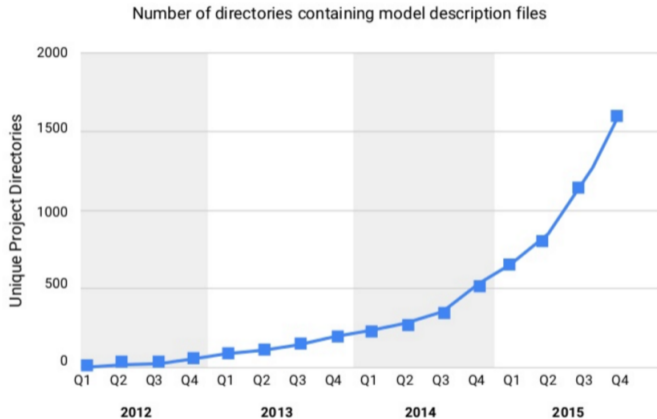
Search trend in Google

Google Trends, May 2012 - April 2017, Worldwide

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



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

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Thank you!