

Deep Learning



Md. Jalil Piran, PhD

Asst. Professor

Computer Science and Eng.

Sejong University

Spring, 2021

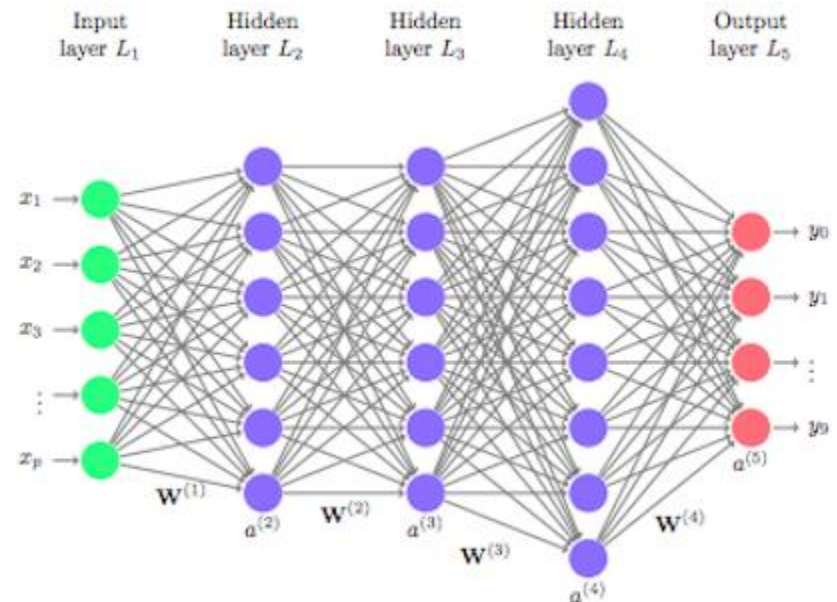
- **Introduction to Deep Learning (DL)**
- **The History of DL**



THE HISTORY OF DEEP LEARNING

Deep Learning

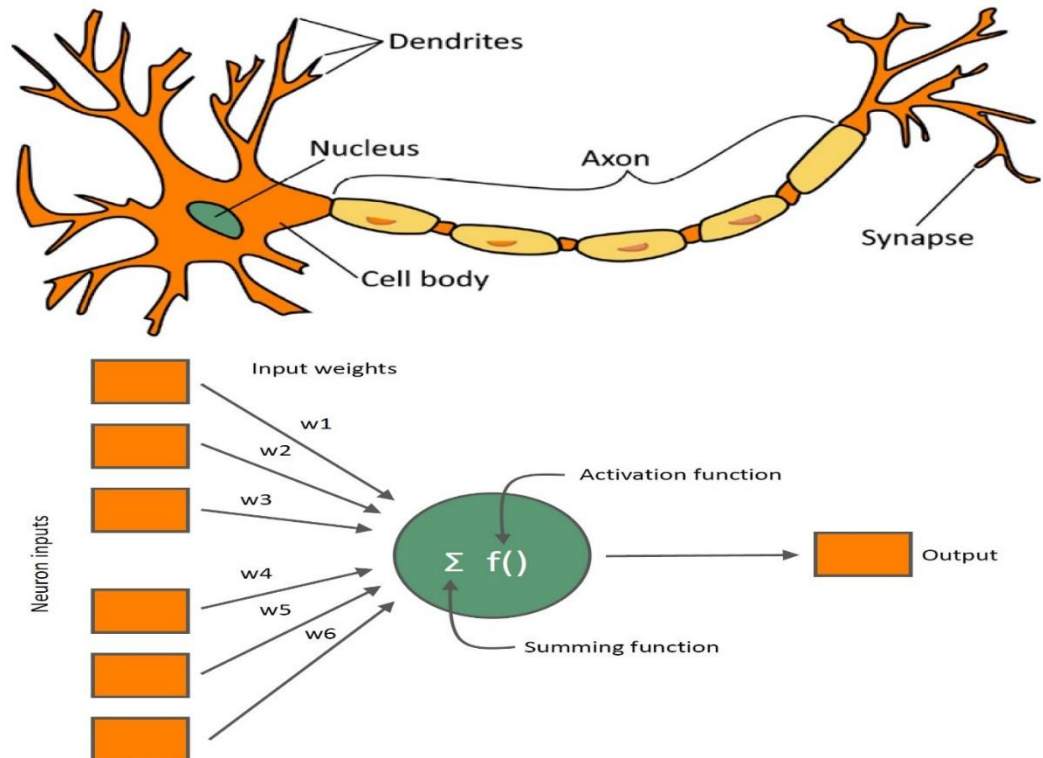
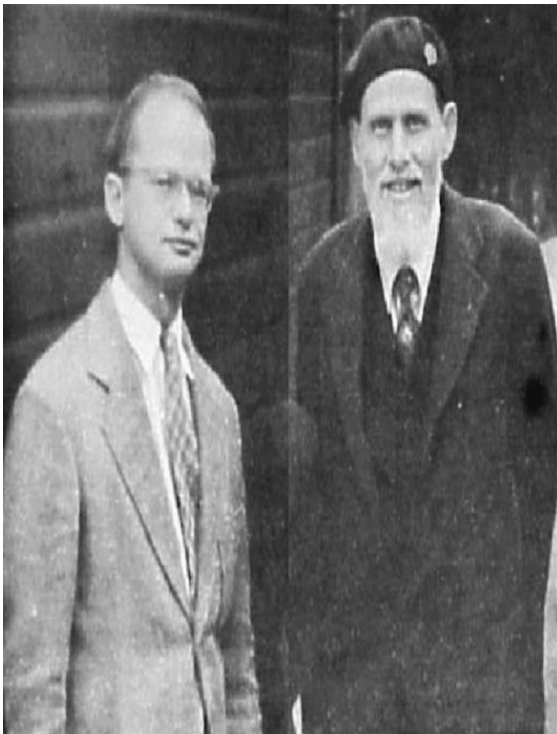
- DL is a branch of ML.
- Used to:
 - Process data
 - Imitate the thinking process
 - Understand human speech
 - Visually recognize objects



Artificial Neurons

1943

- McCulloch and Pitts simplified mathematical model of neurons and introduced **Neural Networks (NNs)** to model the Brain.



Birth of AI



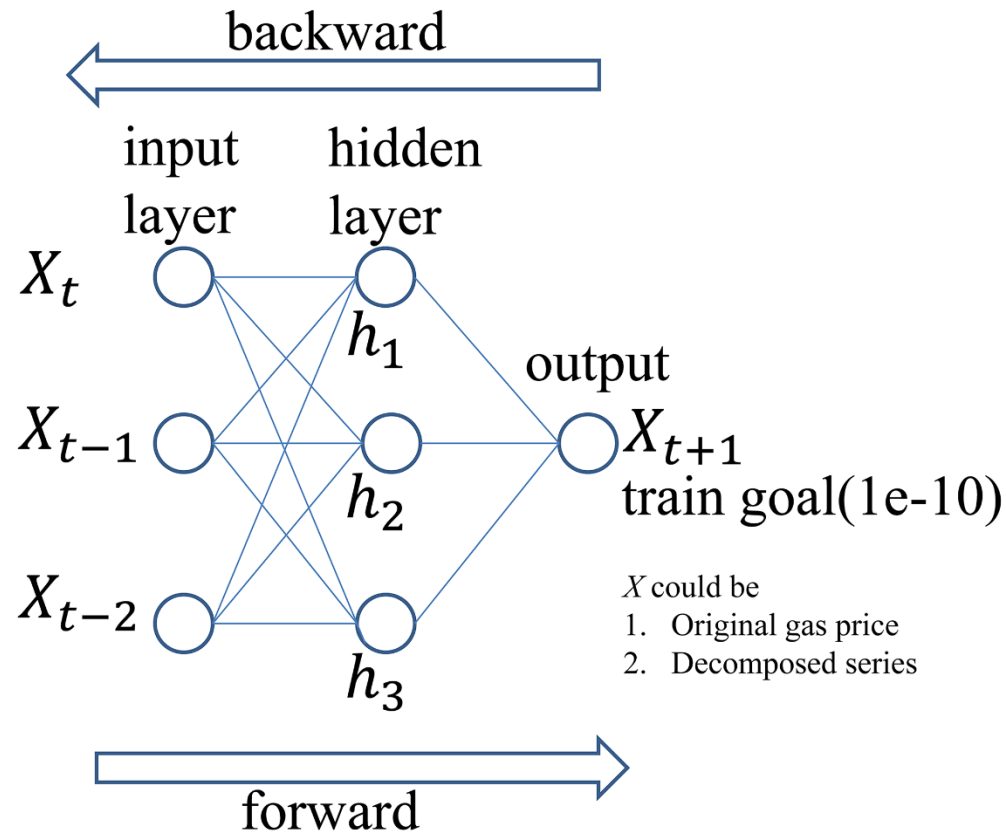
- McCarthy coined the term “**Artificial Intelligence**”.
- The term ‘**Artificial Intelligence**’ made its advent at the “Dartmouth Summer Research Project on Artificial Intelligence” conference in Hanover, New Hampshire.



Backpropagation

1960

- Backpropagation Model was developed by Henry J. Kelley.



Backpropagation



- Stuart Dreyfus developed **improved** backpropagation model based on the chain rules.
- However, backpropagation was not used widely till 1985 due to its clumsy and inefficient.



Backpropagation



- The Group Method of Data Handling was developed by Alexey Ivankhnenko, as the earliest efforts in developing DL.

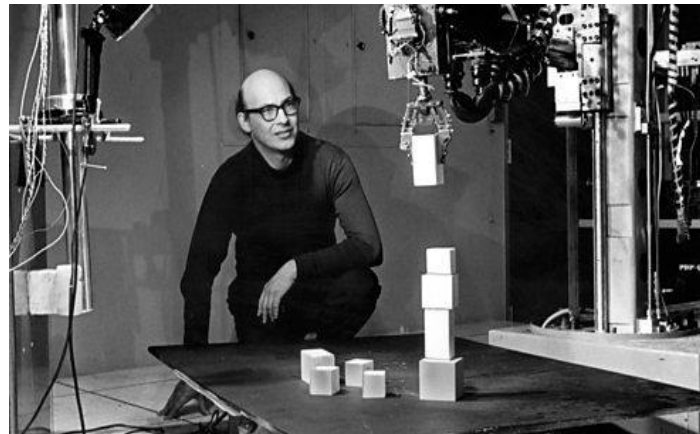


(.q 7001) ojhnhxvsl .T.O

Neural Network

1968

- Marvin Minsky and Dean Edmonds built the first **Artificial Neural Network (ANN)**— a computer-based simulation of the way organic brains work.
- The Stochastic Neural Analog Reinforcement Computer (SNARC) learned from experience and was used to search a *maze*, like a rat in a psychology experiment.



Marvin Minsky at MIT in 1968

The First AI 'Winter'



- In the 1970s, the capabilities of AI programs were limited.
 - AI was subject to critiques and financial setbacks.
 - AI researchers had **failed** to appreciate the difficulty of the problems they faced.
 - Reasons:
 - Limited computer **power**
 - Lack of **data**
 - Commonsense **knowledge** and **reasoning**.
 - Governments and corporations were **losing faith** in AI.
-
- The **Second AI 'winter'** happened during 1987~1993.

A blue-outlined icon of a calendar page with the year 1979 written on it.

1979

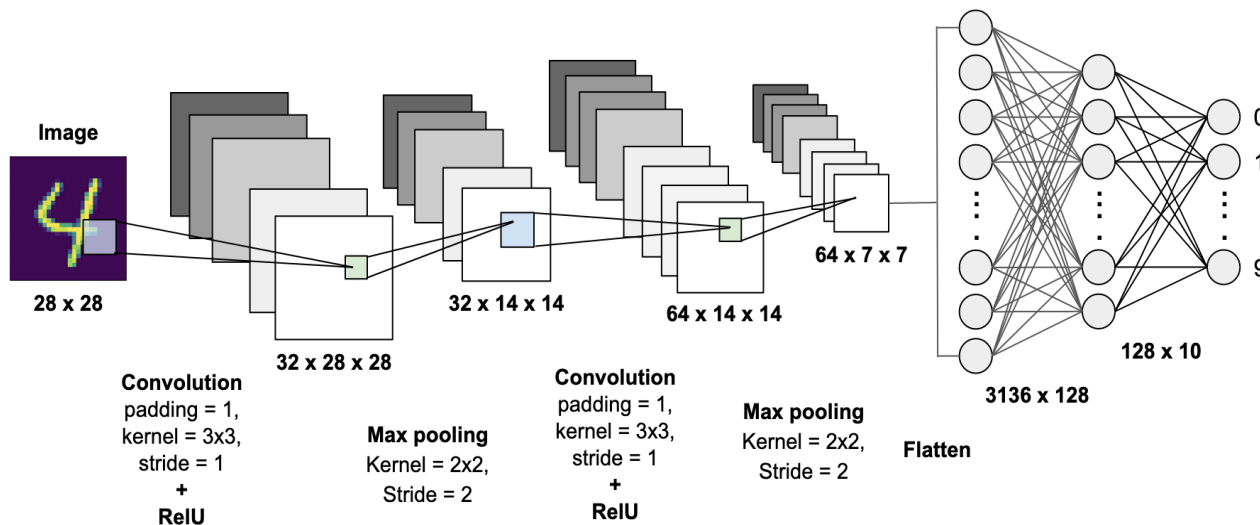
- Kunihiro Fukushima designed ANN with **multiple pooling and convolutional layers**.
- He also developed an ANN, Neocognitron



Handwriting Recognition

1989

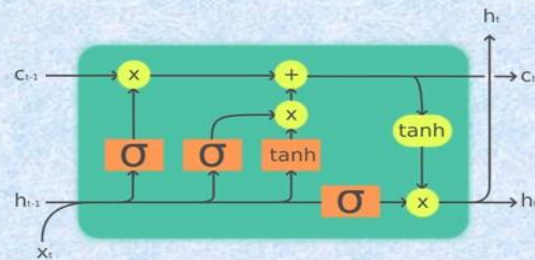
- Yann LeCun invented a machine that can read handwritten digits.



LSTM

1997

- Sepp Hochreiter and Jürgen Schmidhuber publishes a milestone paper on “Long Short-Term Memory” (LSTM).
- LSTM is a type of recurrent neural network architecture which will go on to revolutionize deep learning in decades to come.



Legend:



Pointwise op



Copy



Deep Blue

1997

- **Deep Blue** beat world chess champion Garry Kasparov in the first game of a match.
- Kasparov won the 1996 match, but in 1997 an upgraded Deep Blue then won a second match $3\frac{1}{2}$ games to $2\frac{1}{2}$.



DL Birth




- Hinton coins the term of **Deep Learning**.



Deep Belief Network



- Geoffrey Hinton, Ruslan Salakhutdinov, Osindero and Teh publishes the paper “*A fast learning algorithm for deep belief nets*”
- In which they stacked multiple RBMs together in layers and called them Deep Belief Networks.
- The training process is much more efficient for large amount of data.



A fast learning algorithm for deep belief nets

Geoffrey Hinton and Simon Osindero
 Department of Computer Science
 University of Toronto
 270 College Road
 Toronto, Canada M5S 3G4
 {hinton, osindero}@cs.toronto.edu

Yee-Whye Teh
 Department of Computer Science
 National University of Singapore
 3 Science Drive 3, Singapore
 tehyw@comp.nus.edu.sg

Abstract

“Complementary priors” to way effects that make densely-connected belief networks. Using complementary priors, we develop a fast, greedy algorithm to construct deep belief networks. We initialize the top two layers with complementary priors. The algorithm initializes a slower network with three layers, and then fine-tunes the weights using the wake-sleep algorithm. The resulting network is a good generative model.

remaining hidden layers form a generative model. The algorithm converts the representations in the hidden layers into observable variables such as the labels. The resulting hybrid model has some attractive properties.

1. There is a fast, greedy learning algorithm that finds a fairly good set of parameters for deep belief networks with millions of parameters and millions of training examples.
2. The learning algorithm is simple and can be applied to labeled data by learning both the label and the data.
3. There is a fine-tuning algorithm that converts a deep belief network into a generative model with a good likelihood.

GPU Revolution



- Andrew NG's group in Stanford starts advocating for the use of GPUs for training DNNs to speed up the training time by many folds.
- This could bring practicality in the field of Deep Learning for training on huge volume of data efficiently.



Watson

2011

- Watson won '**Jeopardy!**'



<https://www.youtube.com/watch?v=P18EdAKuC1U>

GoogleBrain

2012

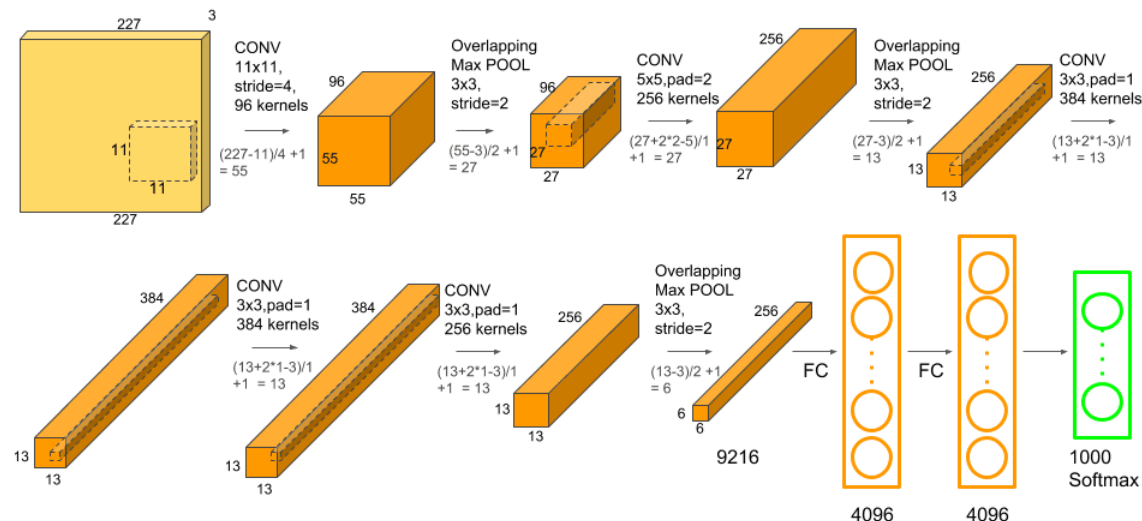
- A DNN created by Jeff Dean, which focused on **pattern detection** in images and videos.



AlexNet



- AlexNet won the ImageNet competition by a large margin in 2012,
- It led to the use of GPUs and **CNNs** in machine learning.
- They also created **ReLU**, which is an activation function that greatly improves efficiency of CNNs.



DeepFace



- A DNN created by Facebook, which they claimed can **recognize** people with the same precision as a human can.



GAN



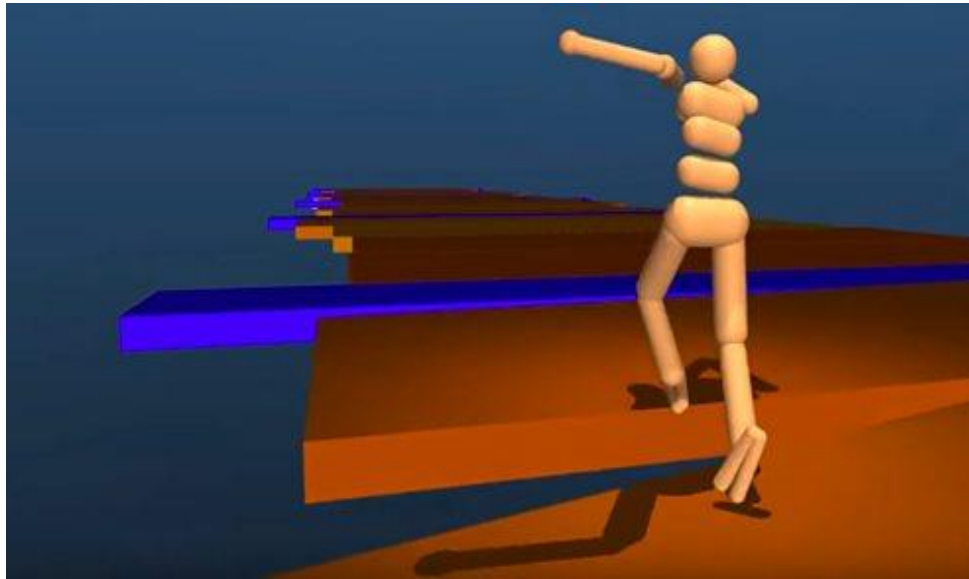
- Ian Goodfellow introduces the Generative Adversarial Networks (GANs)



DeepMind



- **Google DeepMind** gained prominence when it developed a neural network that could learn to **play video games**
- By analyzing the behavior of pixels on a screen.



DeepMind artificial intelligence moving an animated figure

AlphaGo beats Lee Sedol

2016

- Developed by **DeepMind** researchers,
- AlphaGo won its first match against a professional in 2015, beat the world's number two player Lee Sedol in March 2016 and the number one player Ke Jie in 2017.
- AlphaGo's neural network is trained by playing both humans and computers, and uses a **Monte Carlo tree search algorithm** to find moves.



Turing Award



- Yoshua Bengio, Geoffrey Hinton, and Yann LeCun wins Turing Award 2018 for their immense contribution in advancements in area of deep learning and artificial intelligence.



Handol



- **HanDol**, developed by South Korea's NHN Entertainment Corp.,
- Took down the 36-year-old master after 180 stones in Lee's final match,
- Held in his hometown of Sinan, 400 kilometers south of Seoul.



State-of-the-art

Computer Vision

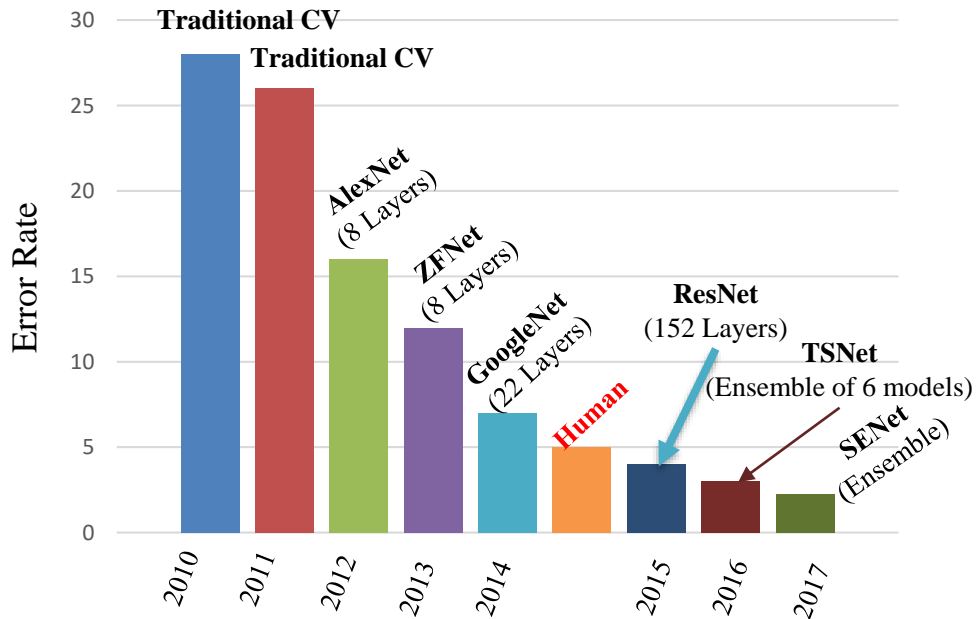
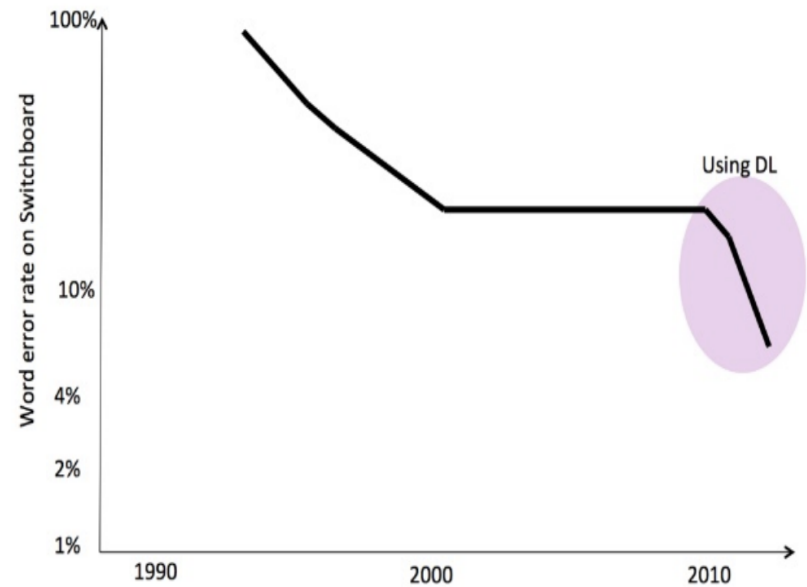


Image classifiers have surpassed human level accuracy.

Speech Recognition



DL improved the accuracy of speech recognition significantly!

State-of-the-art

Game	AI Application	Year of Takeover
Go	Alpha Go	2016
Jeopardy	Watson	2010
Chess	Deep Blue	1997
Checkers	Chinook	1994
Backgammon	BKG	1979

State-of-the-art



Advantages of DL

- Has best-in-class performance on problems that significantly outperforms other solutions in multiple domains, e.g.
 - Speech,
 - Language,
 - Vision,
 - playing games like Go
 - etc.
- Reduces the need for feature engineering, one of the most time-consuming parts of ML practice.
- Can be adapted to new problems relatively easily
- Feature engineering can be automatically executed inside DL model
- Can solve complex problems
- flexible to be adapted to new challenge in the future (or transfer learning can be easily applied)
- High automation.
- Deep learning library (Tensorflow, keras, or MATLAB...) can help users build a deep learning model in seconds (without the need of deep understanding)

Drawbacks of DL

- Requires very large amount of data in order to outperform the other techniques
- Requires high-performance and expensive hardware, e.g. GPUs
- Needs intensive training
- Overfitting if applied into uncomplicated problems
- It's a blackbox, not straightforward to understand inside each layer
- There is no standard theory to select the best learning tool.
- Determining the topology/flavor/training method/hyperparameters for deep learning is a black art with no theory to guide you.
- What is learned is not easy to comprehend. Other classifiers (e.g. decision trees, logistic regression etc) make it much easier to understand what's going on.

Assignment

- What is the technique beyond DL?



