



2018 Asia University AI Summer Program

(Day1-July3) B: Introduction to AI

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Content

- What is Intelligence
- What is Artificial intelligence
- Machine learning
- Deep learning
- AI applications



(1)WHAT IS INTELLIGENCE





Which animals are smart?

10 Most Intelligent Animals In The World

1. Chimpanzees



2. Elephants



Dolphins



Parrots



Sheep



Rats



Dogs



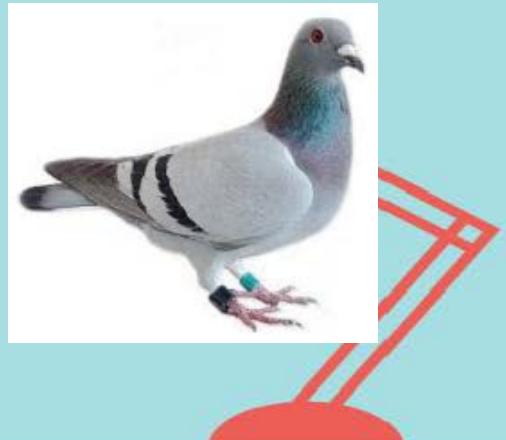
Octopus



Crows



Pigeons



Animal Intelligence

1. Chimpanzees



2. Elephants



Dolphins



Parrots



Sheep

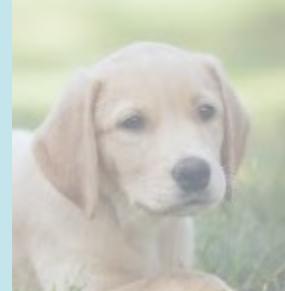


Why we think they are smart?

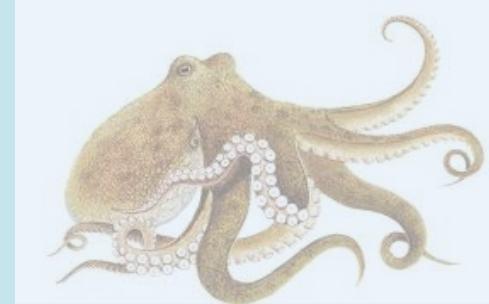
Rats



Dogs



Octopus



Crows



Pigeons



The crow and the pitcher in Aesop's Fables



In a spell of dry weather, when the Birds could find very little to drink, a thirsty Crow found a pitcher with a little water in it. But the pitcher was high and had a narrow neck, and no matter how he tried, the Crow could not reach the water. The poor thing felt as if he must die of thirst. Then an idea came to him. Picking up some small pebbles, he dropped them into the pitcher one by one. With each pebble the water rose a little higher until at last it was near enough so he could drink.

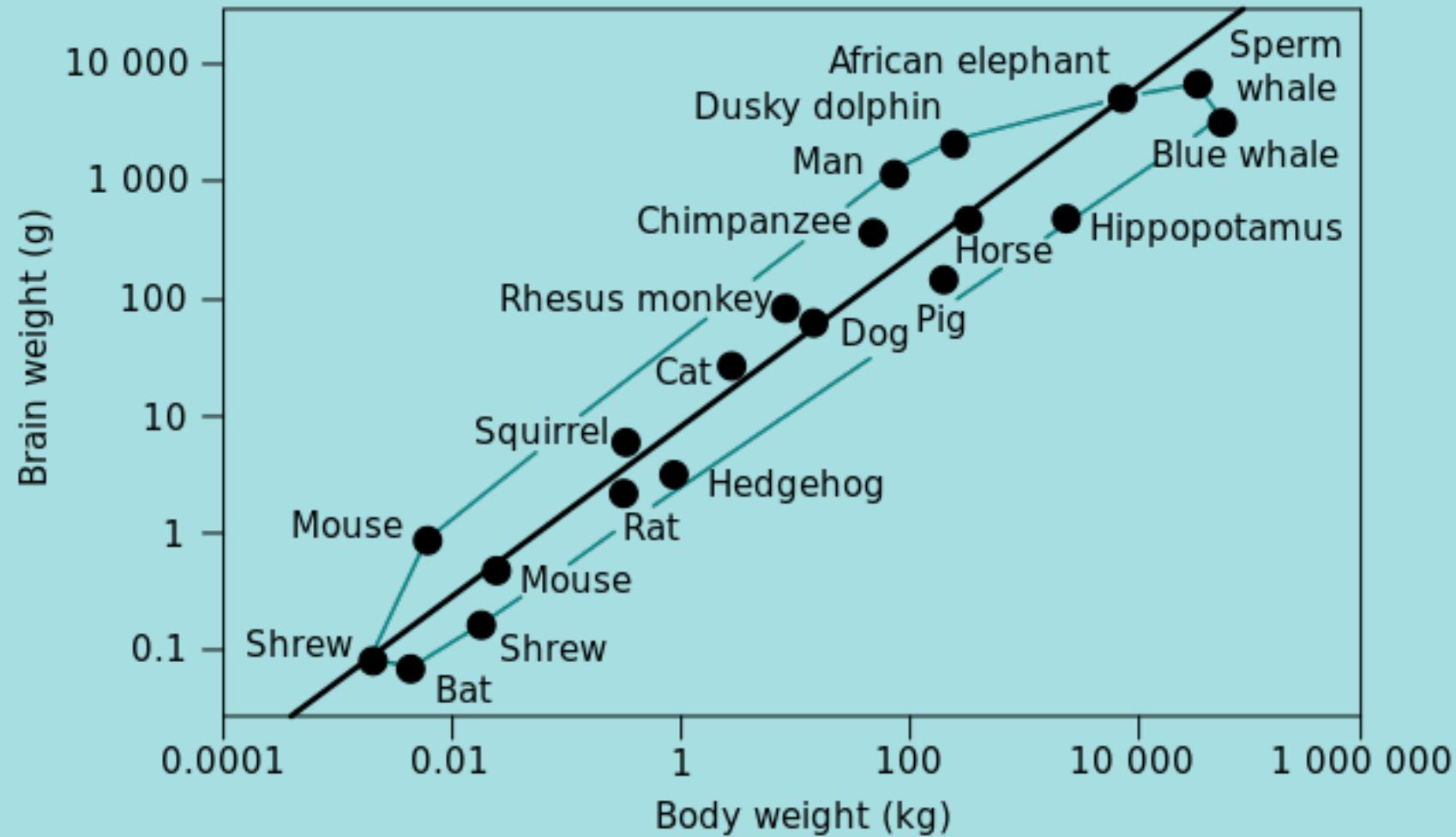


Metrics of intelligence

- Brain-to-body mass ratio
- Number of neurons
- Encephalization quotient

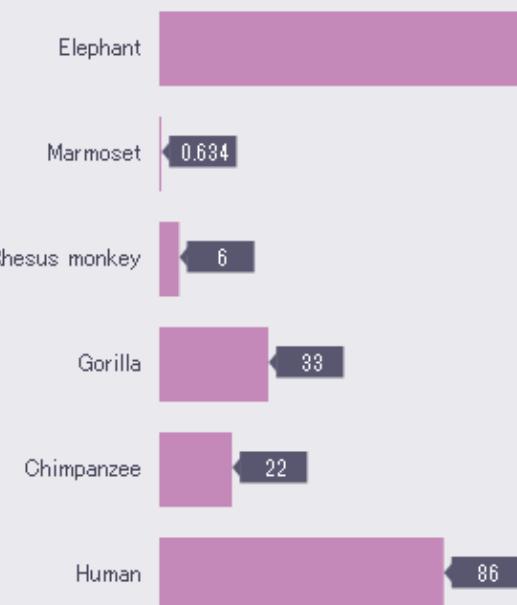


Brain-to-body mass ratio



Number of neurons

Brain neurons (billions)



Cerebral cortex neurons (billions)

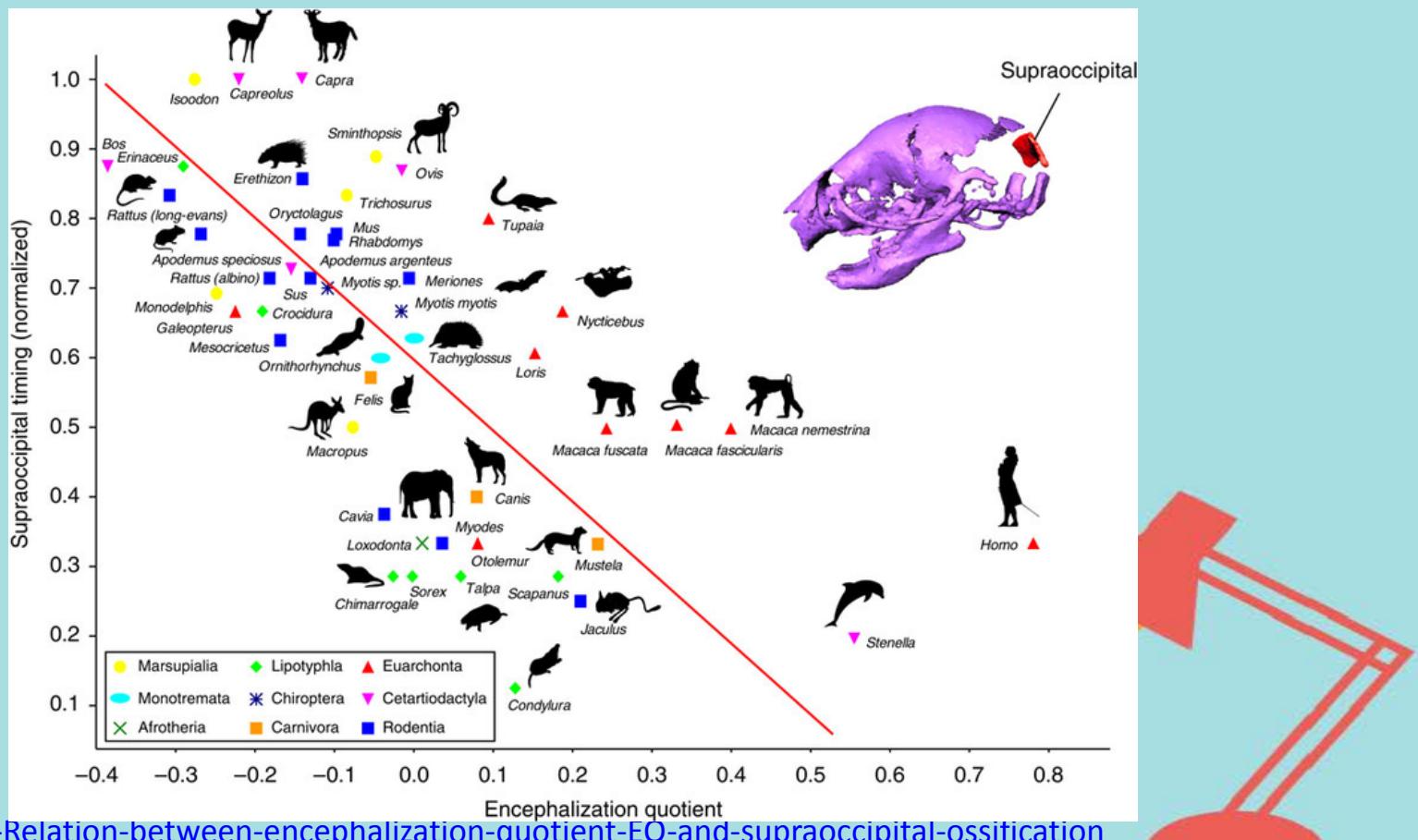


Sources: Suzana Herculano-Houzel; Marino, L. Brain Behav Evol 1998;51:230–238

Sources: Suzana Herculano-Houzel; Marino, L. Brain Behav Evol 1998;51:230–238

Encephalization quotient

- Encephalization quotient (EQ) is a relative brain size measure that is defined as the ratio between actual brain mass and predicted brain mass for an animal of a given size, which may approximate intelligence level or cognition of the species.



https://en.wikipedia.org/wiki/Encephalization_quotient

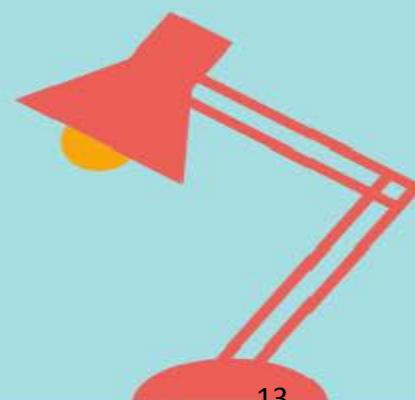
https://www.researchgate.net/figure/261407553_fig6_Figure-3-Relation-between-encephalization-quotient-EQ-and-supraoccipital-ossification

(2) INTRODUCTION TO AI



Concept of Artificial intelligence

- Artificial intelligence (AI) is intelligence demonstrated by machines
- Natural intelligence (NI) displayed by humans and other animals.



Intelligence

- Intelligence consists of the following skills:
- 1. the ability to reason
- 2. the ability to acquire and apply knowledge
- 3. the ability to manipulate and communicate ideas

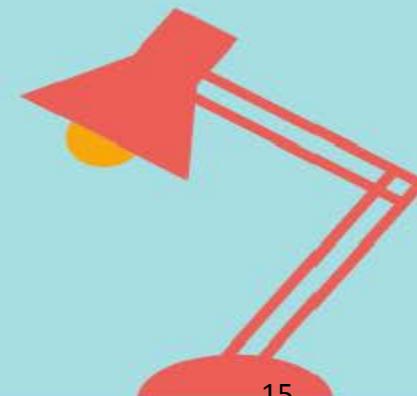
<http://slideplayer.com/slide/272574/>



History of artificial intelligence

https://en.wikipedia.org/wiki/History_of_artificial_intelligence

- The First Boom of AI :
 - 1952–1956
 - Failure of machine translation
- The Second Wave of AI
 - 1980–1987
 - In 1981, Japanese began the Fifth generation computer project.
 - Failure of expert systems
- The Third Wave of AI
 - 2011–present
 - Deep learning and big data



Artificial Intelligence

- Systems that can demonstrate human-like reasoning capability
- AlphaGo Zero: Learning from scratch

Article

Mastering the game of Go without human knowledge

David Silver , Julian Schrittwieser, Karen Simonyan, Ioannis Antonoglou, Aja Huang, Arthur Guez, Thomas Hubert, Lucas Baker, Matthew Lai, Adrian Bolton, Yutian Chen, Timothy Lillicrap, Fan Hui, Laurent Sifre, George van den Driessche, Thore Graepel & Demis Hassabis

Nature 550, 354–359 (19 October 2017)
doi:10.1038/nature24270
[Download Citation](#)

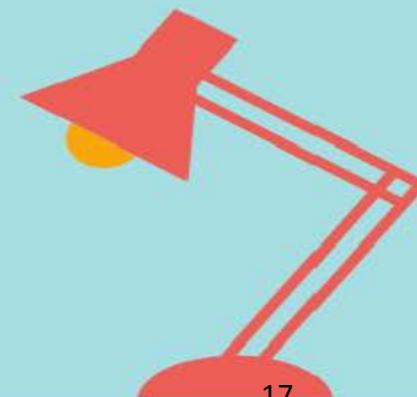
Received: 07 April 2017
Accepted: 13 September 2017
Published online: 18 October 2017



The Prolog programming language

<https://en.wikipedia.org/wiki/Prolog>

- Rules
 - Head :- Body.
 - animal(X) :- cat(X).
- Facts
 - cat(tom).
- Queries
 - ?- cat(tom).
 - ?- animal(X).



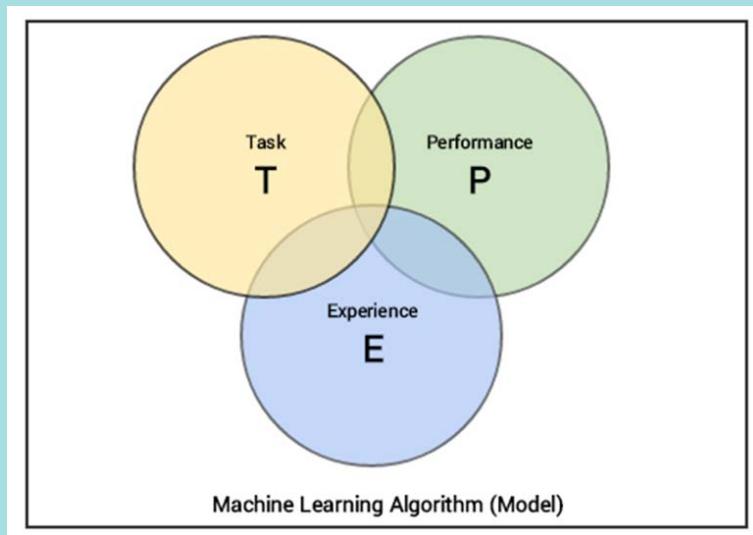
(3) INTRODUCTION TO MACHINE LEARNING



Introduction to Machine Learning

Definition

The art of making machines intelligent without explicit programming



Machine learning is a field which consists of learning algorithms or techniques which

- execute some task **T** (*Regression/Classification*)
- Improve their performance **P** (*model performance*)
- with experience **E** (*Data*)



Introduction contd...

Building Machine learning models is a 3 stage process

- Representation (selection of an algorithm/parameters)
- Evaluation (objective function)
- Optimization (finding the optimal parameters)

Types of Machine Learning techniques

Supervised Machine Learning

Training is done using labelled data

Algorithm learns the mapping function from the input to the output. $Y = f(X)$

Examples

Regression - used to predict continuous values

Classification - used to predict categorical values

Unsupervised Machine Learning

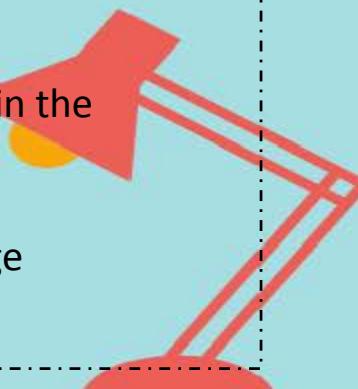
Training is done using unlabelled data

Algorithms are left to their own devices to discover and present the interesting structure in the data.

Examples

Clustering – used to discover the inherent groupings in the data

Association - used to discover rules that describe large portions of the data



Supervised Learning

Supervised M/L – Regression

Definition

Regression is the technique that determines the relationship between one or more independent variables and a dependent variable

Few other naming styles

Dependent	: Independent
Target	: Input
Criterion	: Predictor

Linear

- Simple Linear Regression

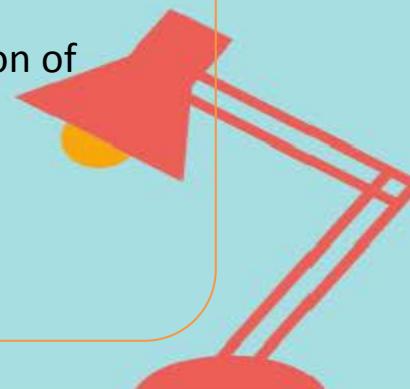
Only one independent variable

- Multiple Linear Regression

Two or more independent variable

Non Linear Regression

Dependent on non-linear transformation of independent variables

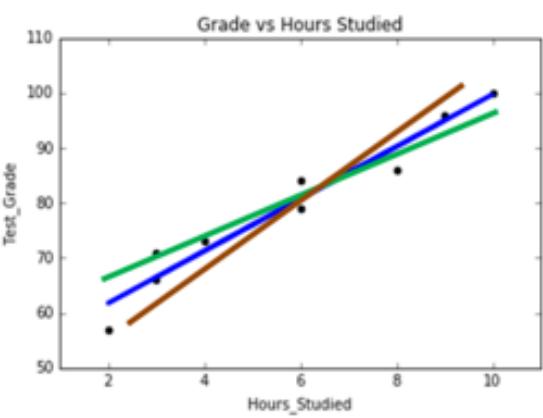


Simple Linear Regression

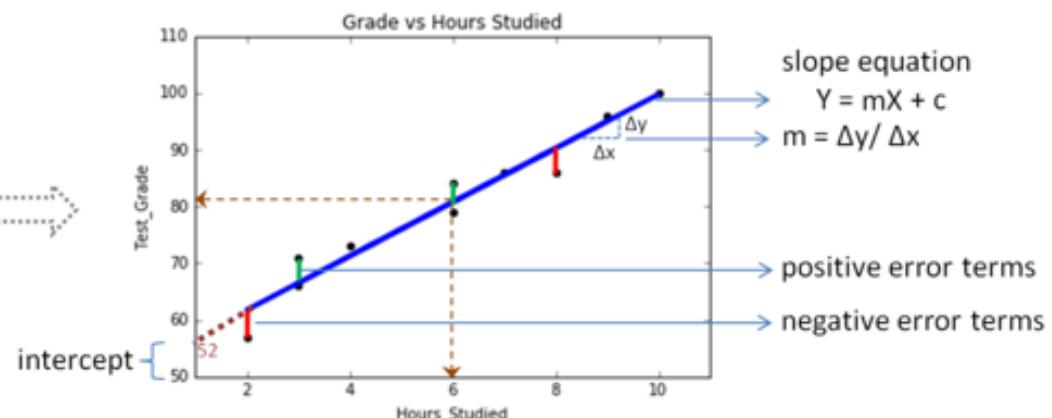
In Simple Linear Regression, we fit the best line between the dependent variable and the independent variable given as $y = mx + c$

m = Co-efficient of x (i.e. change in y divided by change in x)

C = intercept (represents the variability in y , unexplained by x)



Different possible slopes



Best slope that gives minimal sum of squared error

Few points to wander-
How is the best line calculated?

- OLS

Why not use just correlation?
- Intercept



Multiple Linear Regression

Real life scenarios will definitely have many more independent variables than just 1

To model an equation that studies the relationship between the dependent variable and multiple independent variables

The representation equation can be extended to

$$y = m_0 + m_1 x_1 + m_2 x_2 + m_3 x_3 + \dots + m_n x_n$$

Where

- m_0 is the intercept
- m_1 is the coefficient of variable x_1
- m_2 is the coefficient of variable x_2

and so on.....



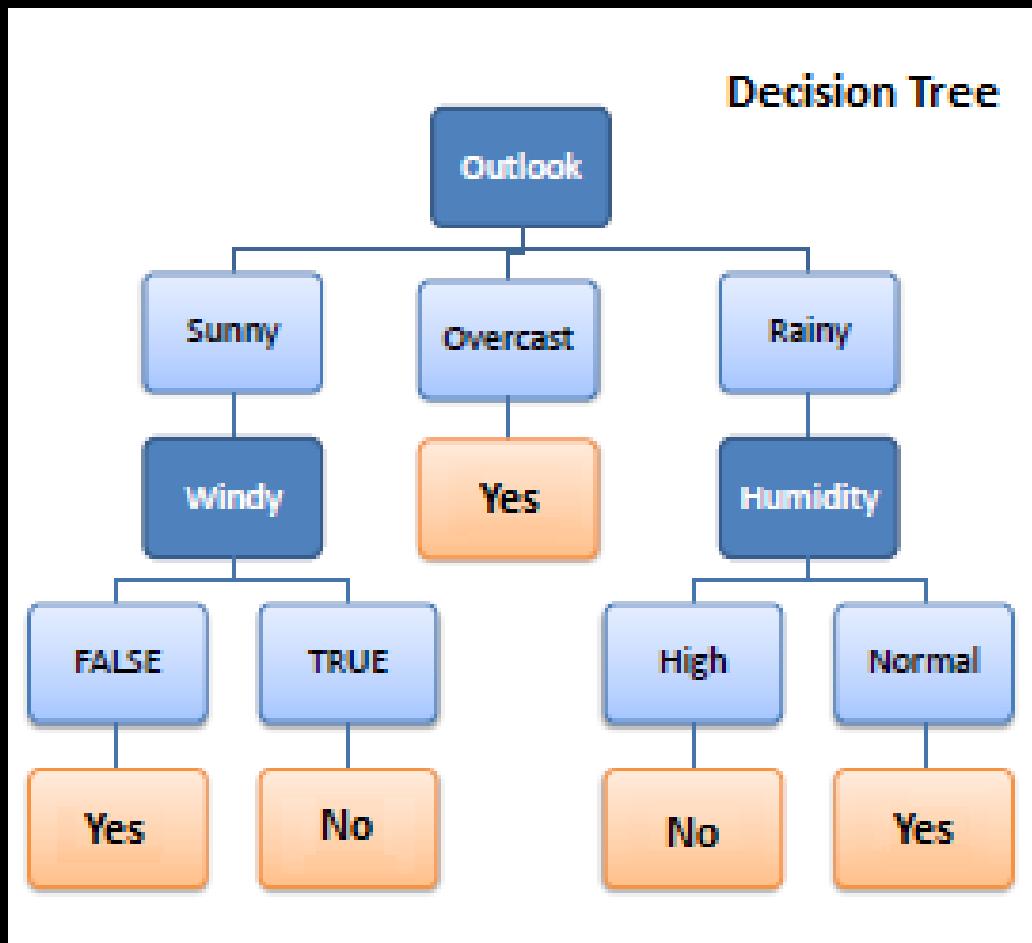
Supervised ML - Decision Trees

Definition

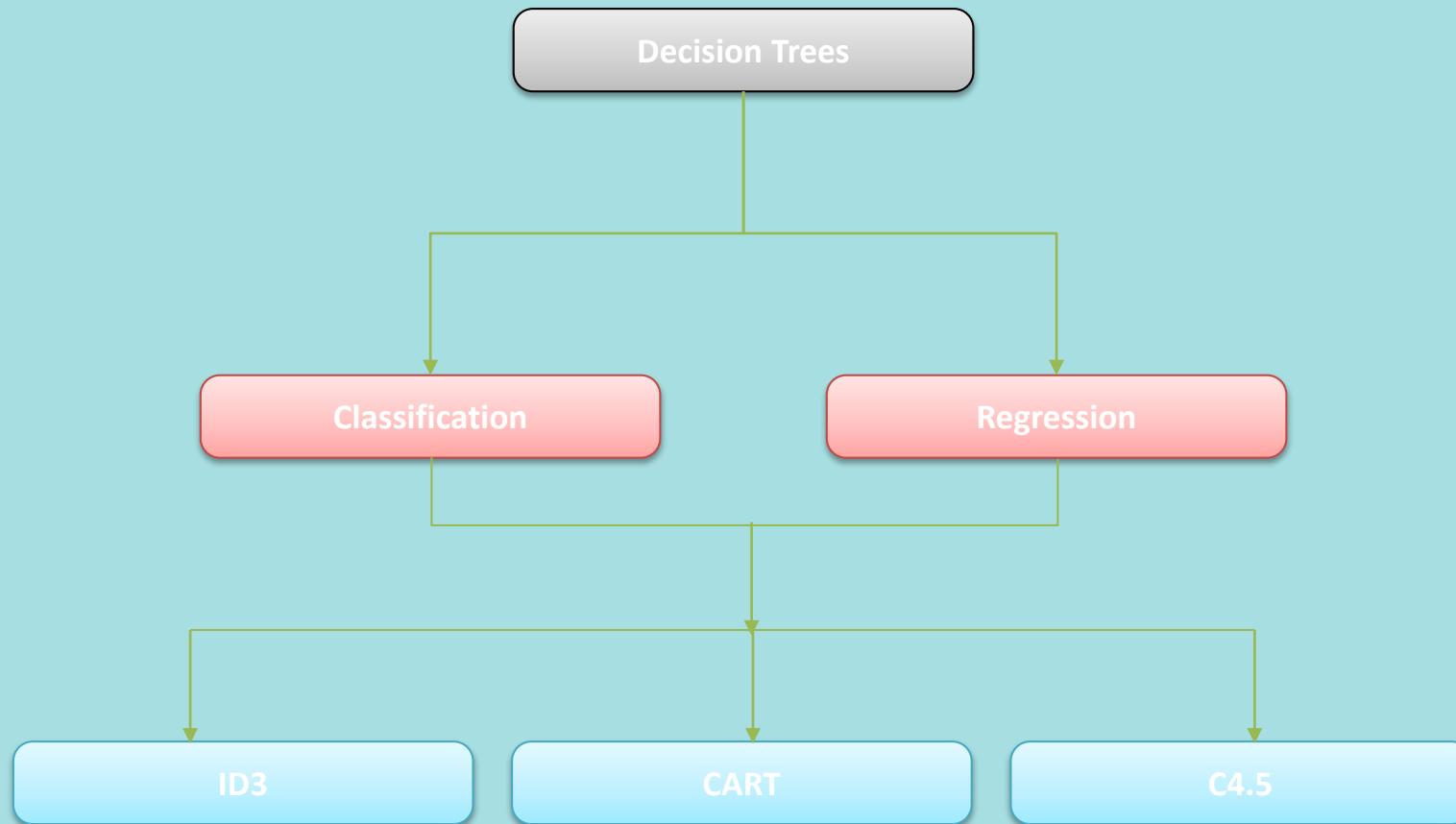
Decision Trees are a class of Supervised Learning Algorithms which can be used for predicting categorical or continuous variables

How does it work?

- It works by breaking data from the root node into smaller and smaller subsets while incrementally building associated decision tree.
- The final result is a tree with a root node, decision nodes and leaf nodes.
- Decision nodes create a rule and leaf nodes deliver a result



Types of Decision Trees



And a few more...



Building Decision Trees

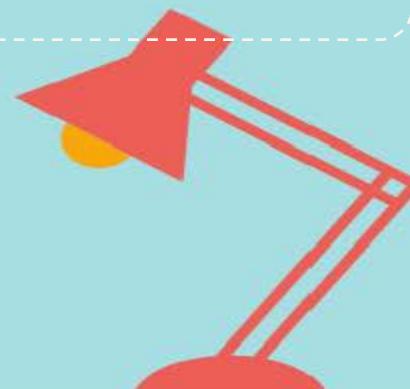
Pseudocode

- 1) Select Root Node
- 2) Partition Data into respective groups
- 3) Create a Decision node
- 4) Partition Data into respective groups

Repeat until the node size > threshold or Features = empty

Few questions?

- **How** to select the Root Node?
- **How** are the decision nodes ordered/chosen?
- **When** does the branching stop?
- **How** does the tree treat continuous variables?
- **How** different is the process for classification and Regression?



Entropy

What is it?

Entropy (measures the homogeneity of the sample)

$$-p\log_2 p - q\log_2 q$$

Where

p = Probability of event happening

q = probability of event not happening

How is it done?

Consider the problem of

Predicting whether the Employee is a Techie or a Non-Techie?

Techie	Non-techie
46	15

p	0.75
q	0.25

Entropy(Emp) : Entropy(Techie,Non-Techie)

$$\text{Entropy}(46,15) = -0.75\log_2(0.75) - 0.25\log_2(0.25) = 0.8$$

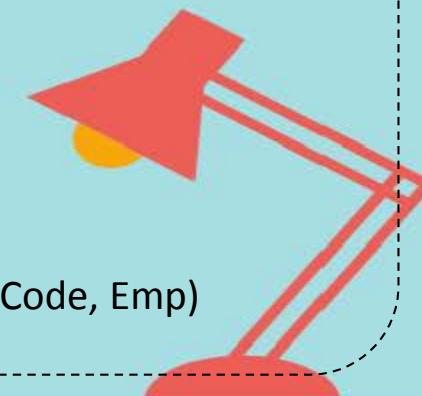
Dress codes	Techie	Non-Techie	Total
Formals	6	8	14
Casuals	31	4	35
Business Casuals	9	3	12
			61

Entropy(Employee,Dress-code)

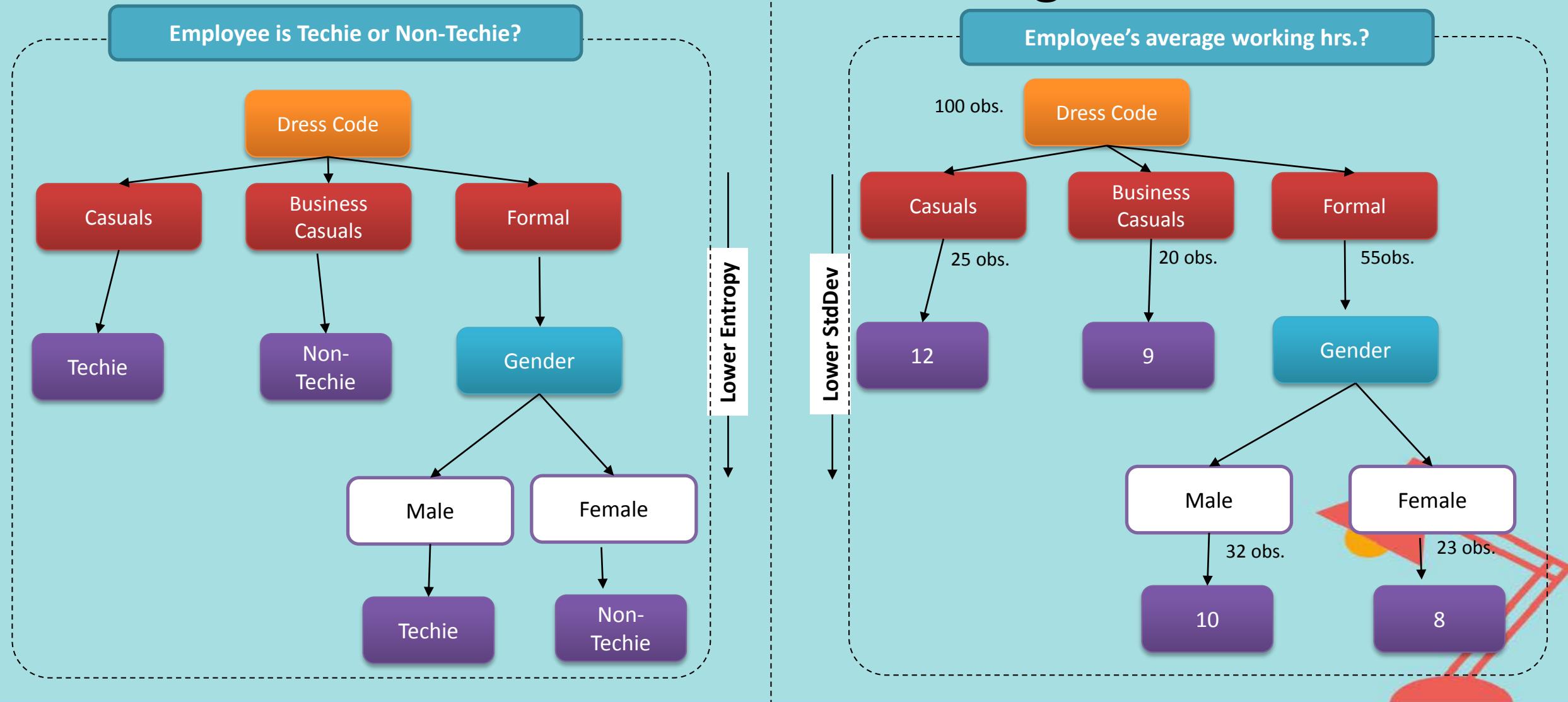
$$\text{Prob}(\text{Formals}) * \text{Entropy}_{\text{formals}} + \text{Prob}(\text{Casuals}) * \text{Entropy}_{\text{casuals}} + \text{Prob}(\text{BCasuals}) * \text{Entropy}_{\text{Bcasuals}}$$

$$(14/61) * \text{Entropy}(6,8) + (35/61) * \text{Entropy}(31,4) + (12/61) * \text{Entropy}(9,3) = 0.68$$

$$\text{Gain}(\text{Emp}) = \text{Entropy}(\text{Emp}) - \text{Entropy}(\text{DressCode}, \text{Emp}) = 0.80 - 0.69 = 0.11$$



Decision Trees - Classification & Regression



Unsupervised Learning

Clustering

The key objective in clustering is to identify distinct groups/clusters)based on some notion of similarity within a given dataset.

Types of clustering

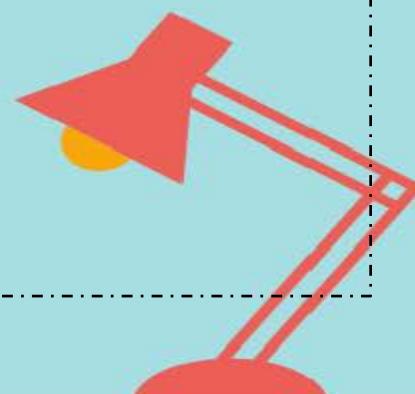
- Agglomerative (hierarchical)
- Divisive (k – means)

K-Means clustering

- Start with random point initialization of the required number of centers. ('K' in K-means stands for the number of clusters)
- Assign each data point to the 'center' closest to it. (distance metric := normal Euclidian distance)
- Recalculate centers by averaging the dimensions of the points belonging to the cluster.
- Repeat with new centers until we reach a point where the assignments become stable.

Hierarchical clustering

- Start with n clusters ($n = \# \text{ of datapoints}$)
- Combine the 2 closest clusters
- Repeat till only 1 cluster exists



Association Rule Mining

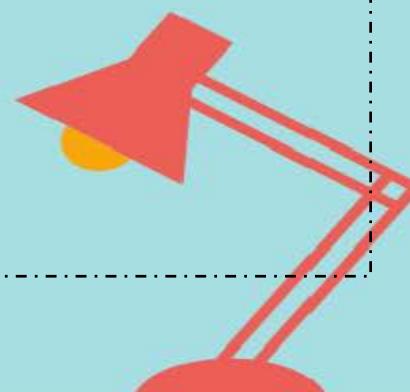
Association rule mining is a procedure to find frequent patterns, correlations, associations, or causal structures from data sets

Algorithm details

- Item set – The list of all transactions; {milk, bread}, {apples, oranges}, {milk}
- Support - # of times an item appears in a dataset
i.e. support(milk, bread) = (# transactions with milk and bread)/(Total # of transactions)
- Confidence – Measure of # of times a rule is found to exist in the dataset
i.e. confidence(milk \rightarrow bread) = support(milk and bread)/ support(milk)

Main Applications

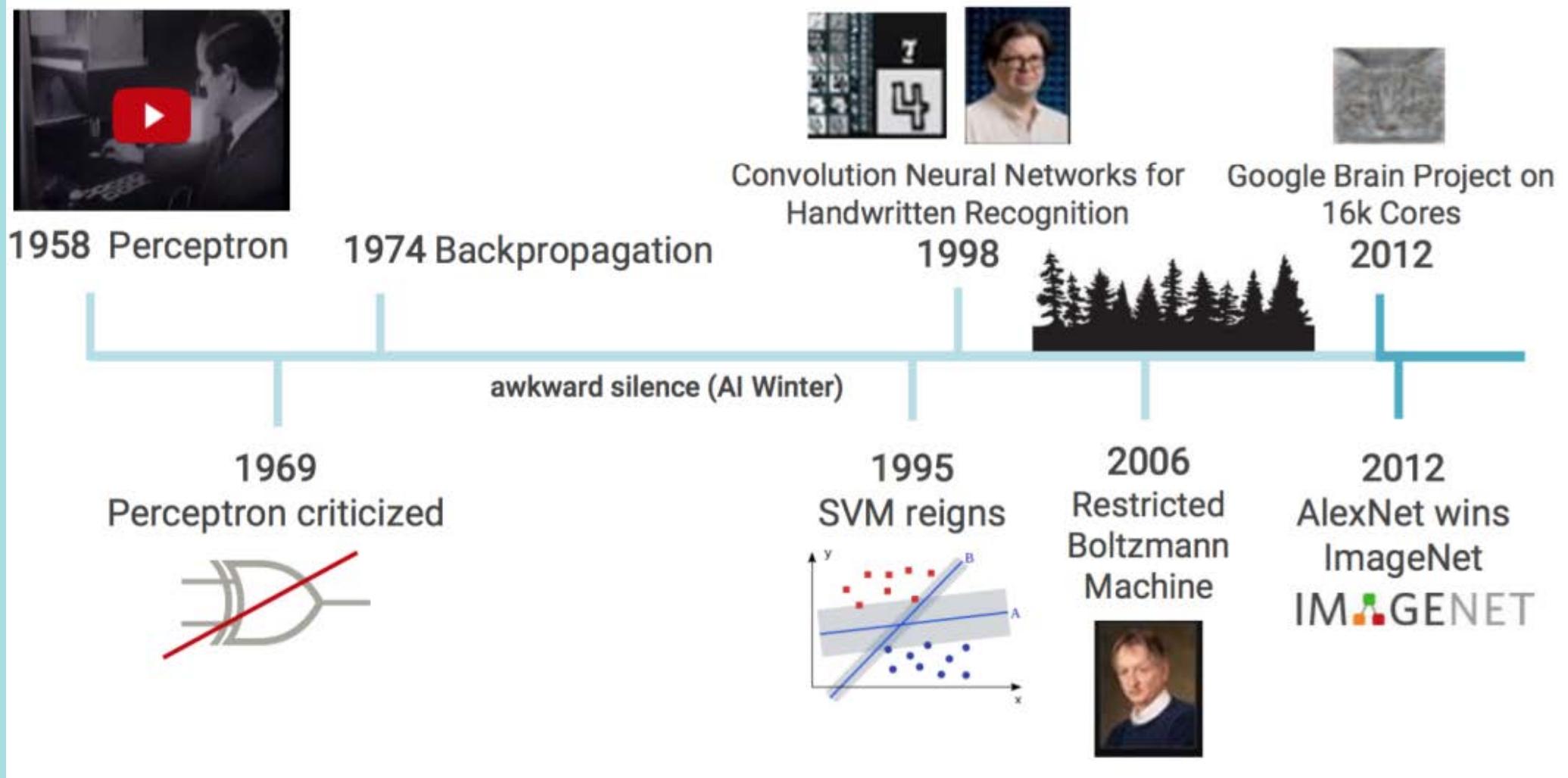
- Cross sell/Up-sell
- Market Basket Analysis



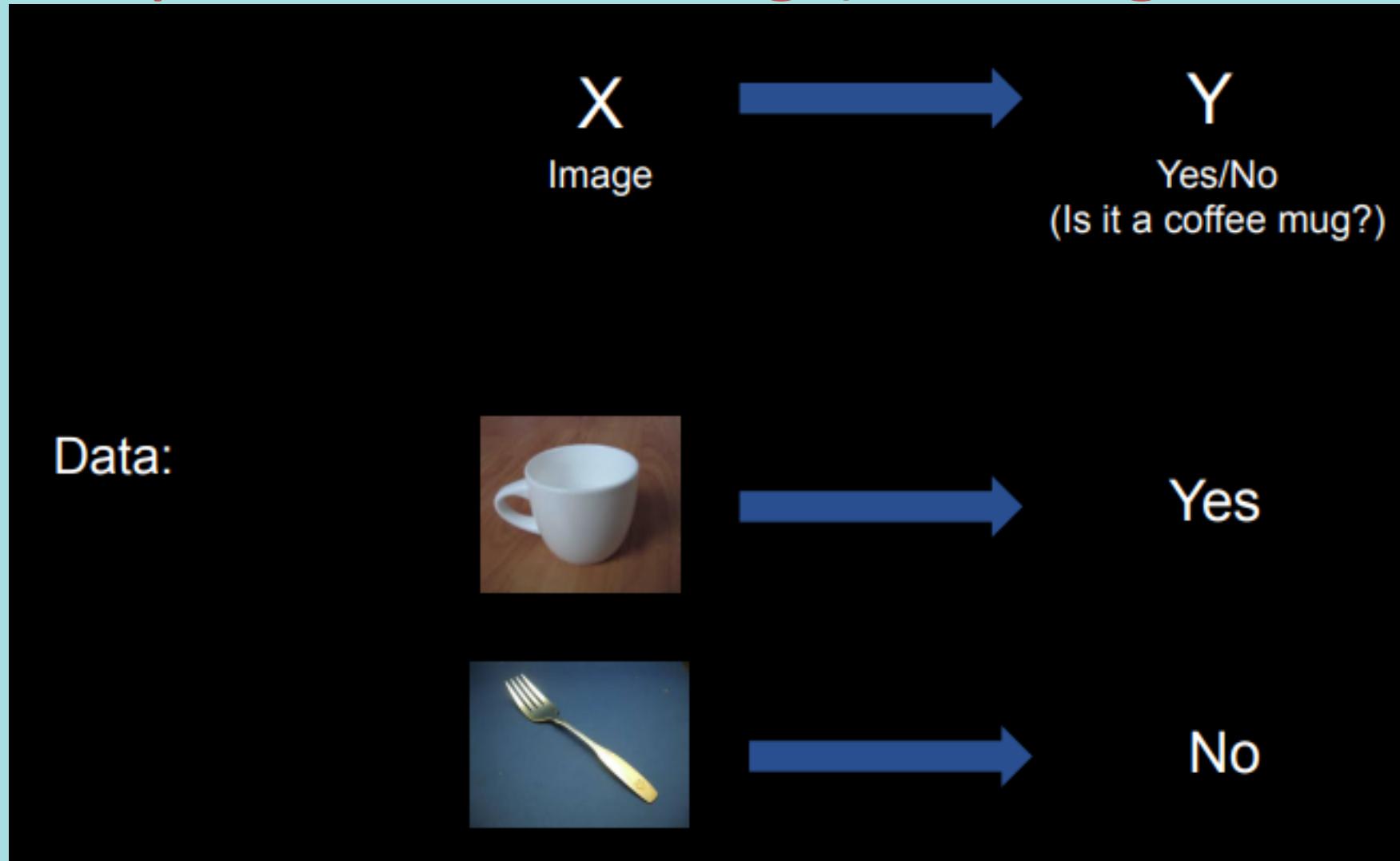
(4) INTRODUCTION TO DEEP LEARNING



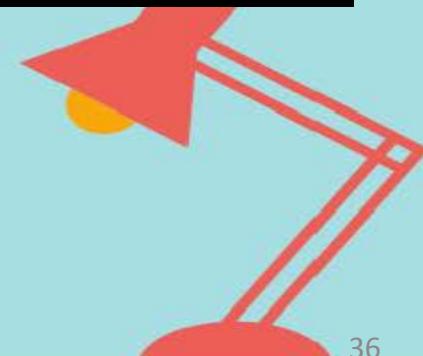
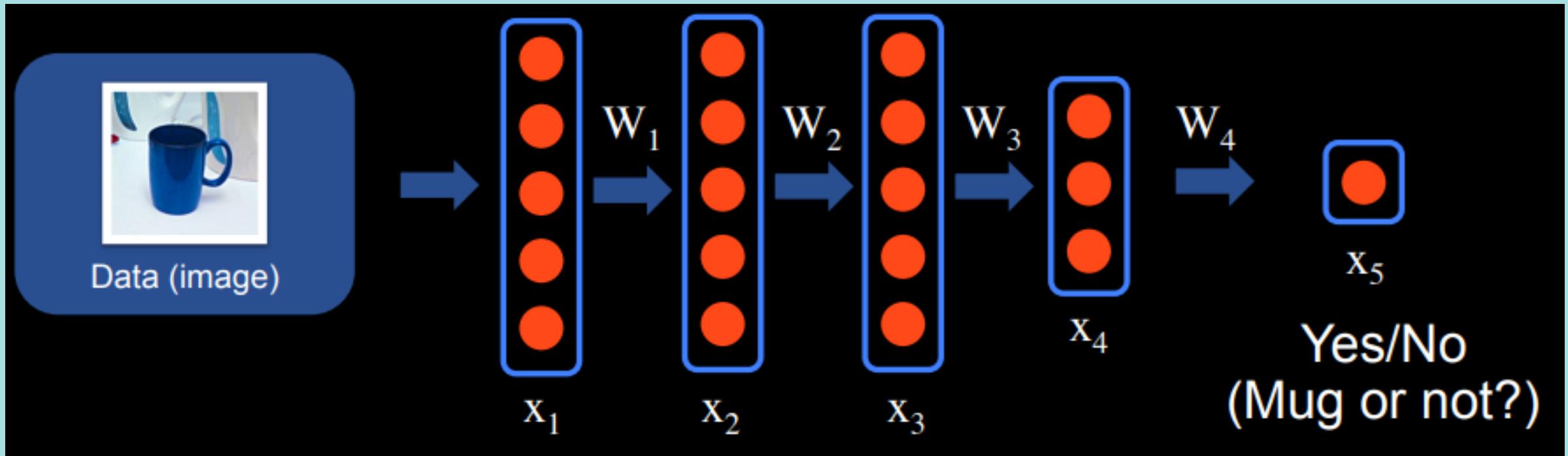
Brief History - Neural Networks



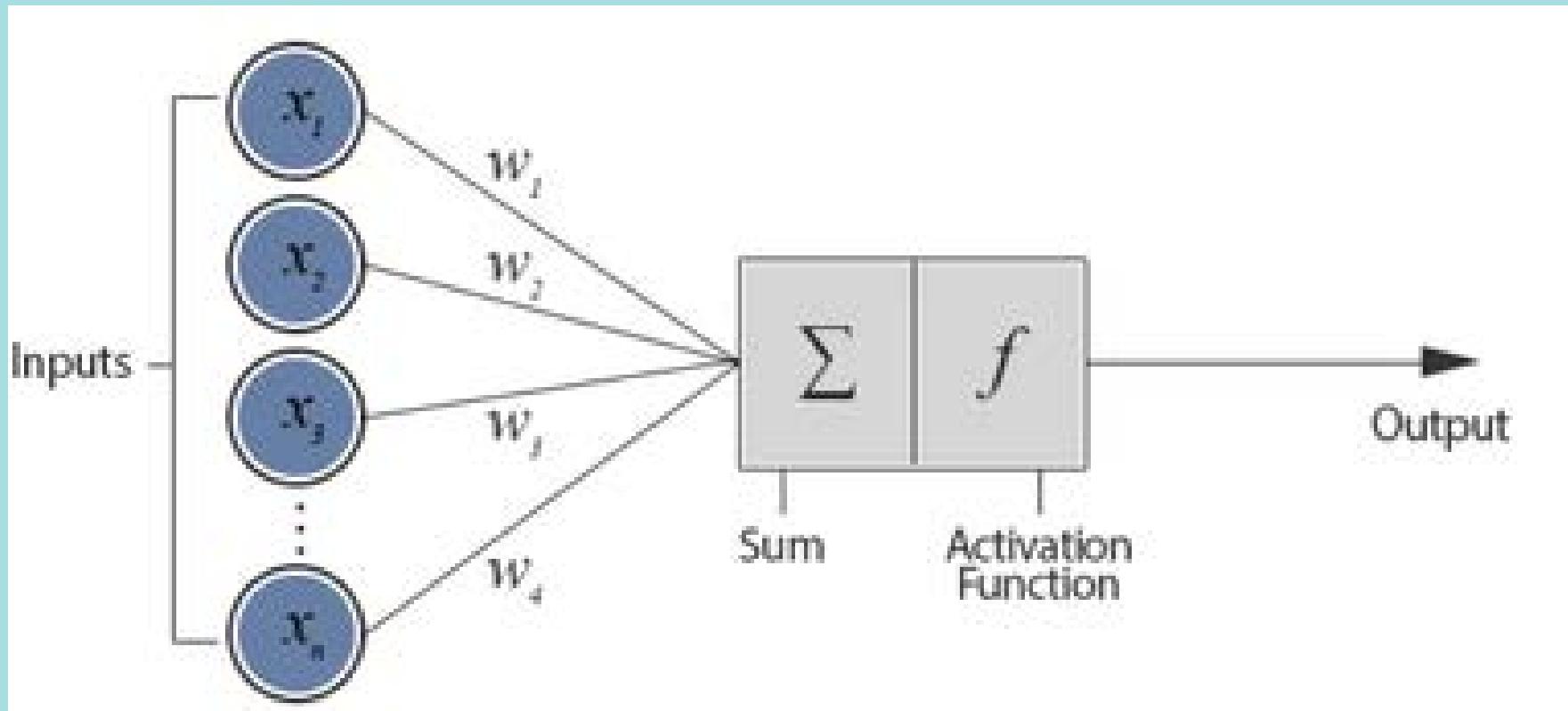
Supervised learning (learning from tagged data)



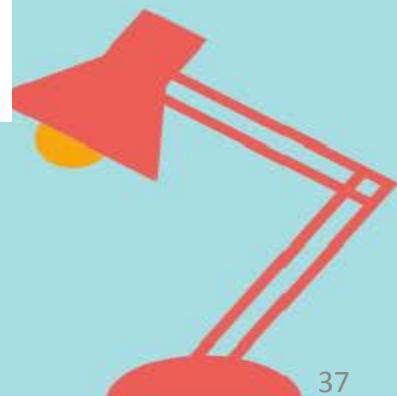
What is a neural network?



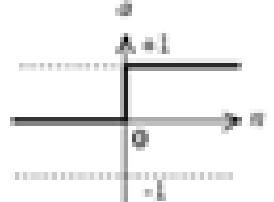
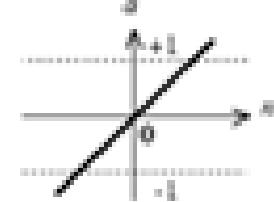
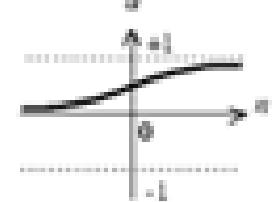
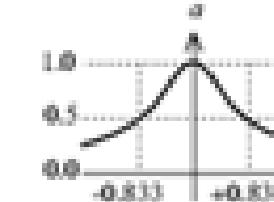
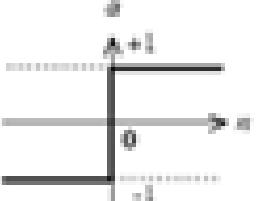
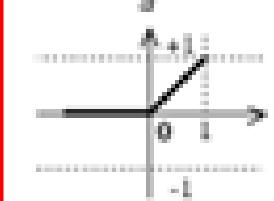
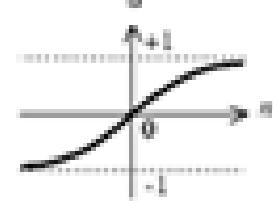
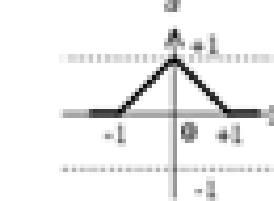
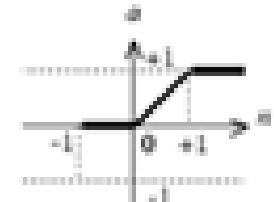
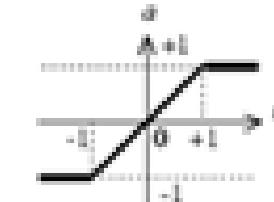
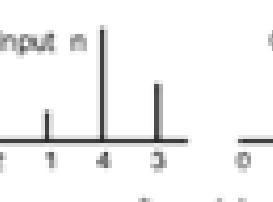
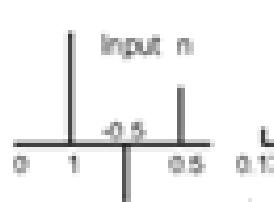
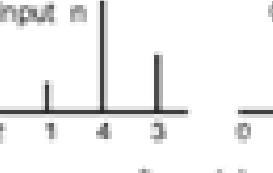
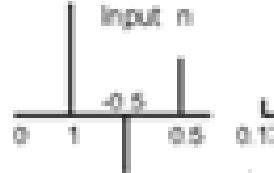
Perceptron



$$\text{Output} = f\left(\sum_{k=1}^n x_k \cdot w_k\right)$$



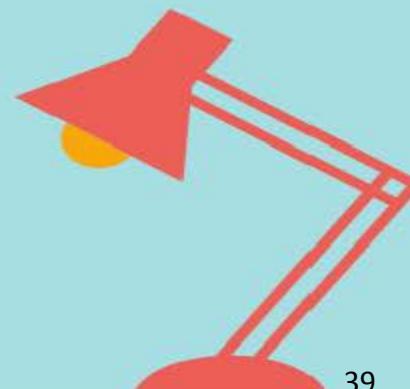
Activation functions

 $a = \text{hardlim}(n)$	 $a = \text{purelin}(n)$	 $a = \text{logsig}(n)$	 $a = \text{radbas}(n)$
Hard-Limit TF  $a = \text{hardlim}(n)$	Linear TF  $a = \text{purelin}(n)$	Log-Sigmoid TF  $a = \text{logsig}(n)$	Radial Basis TF  $a = \text{radbas}(n)$
Symmetric Hard-Limit TF  $a = \text{satlin}(n)$	Positive Linear TF  $a = \text{purelin}(n)$	Tan-Sigmoid TF  $a = \text{tansig}(n)$	Triangular Basis TF  $a = \text{tribas}(n)$
Satlin TF  $a = \text{satlin}(n)$	Satlins TF  $a = \text{satlins}(n)$	 $a = \text{softmax}(n)$	 $a = \text{softmax}(n)$
Compet TF	Softmax TF		



Deep Neural Networks

- Dramatic improvements in accuracy and runtime in multiple domains
 - Image Detection
 - Face Recognition
 - Gesture Recognition
 - Video Search & Analytics
 - Speech Recognition & Translation
 - Recommendation Engines
 - Indexing & Search



Images Classification - Neural Networks

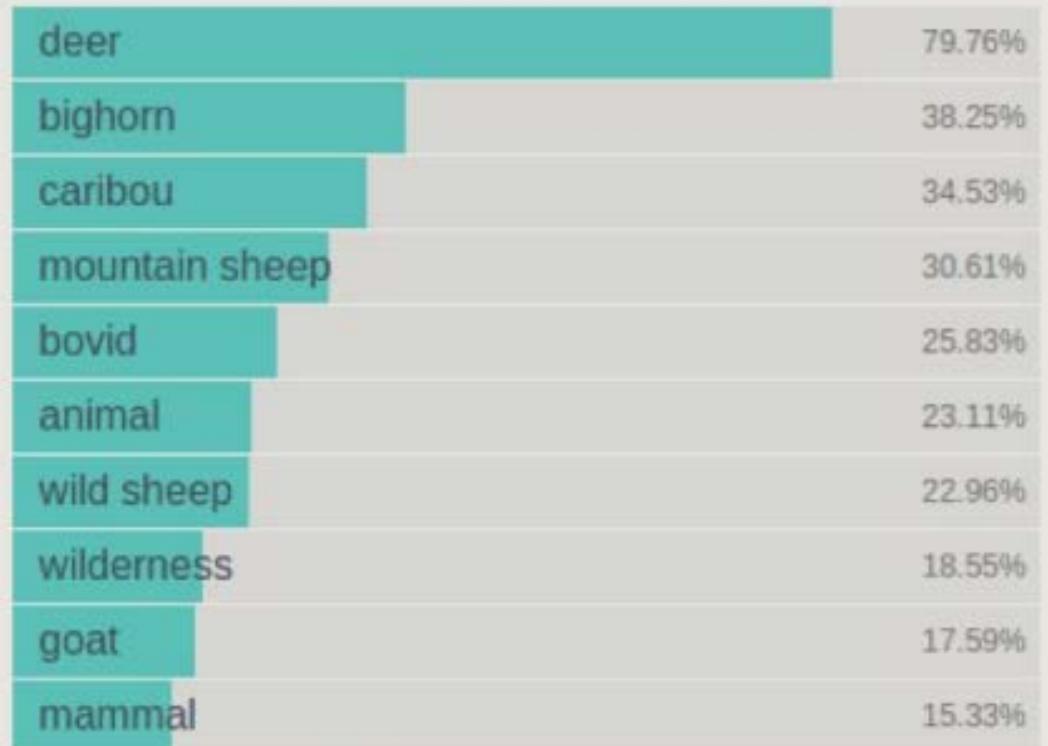
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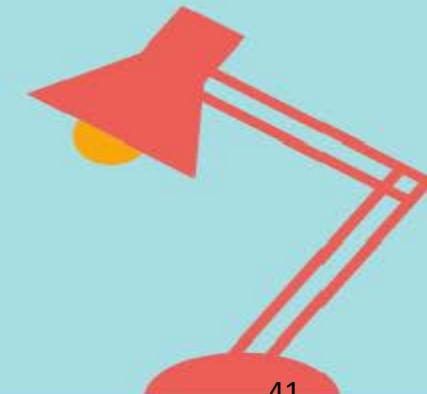
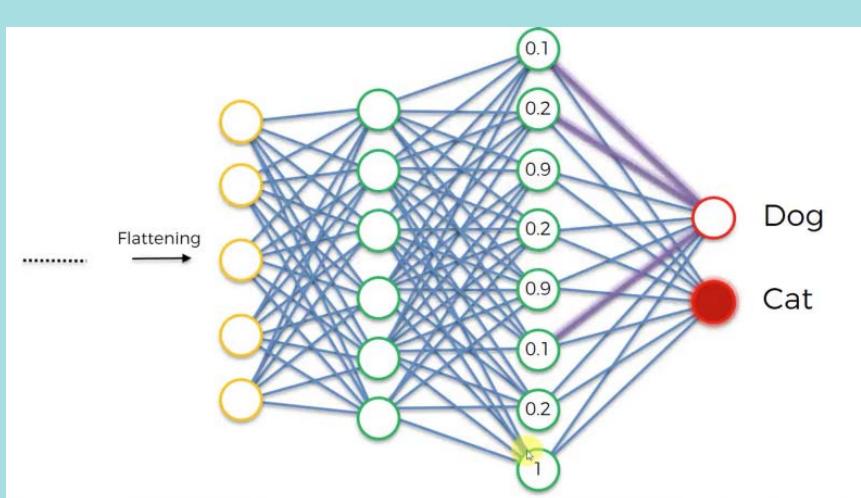
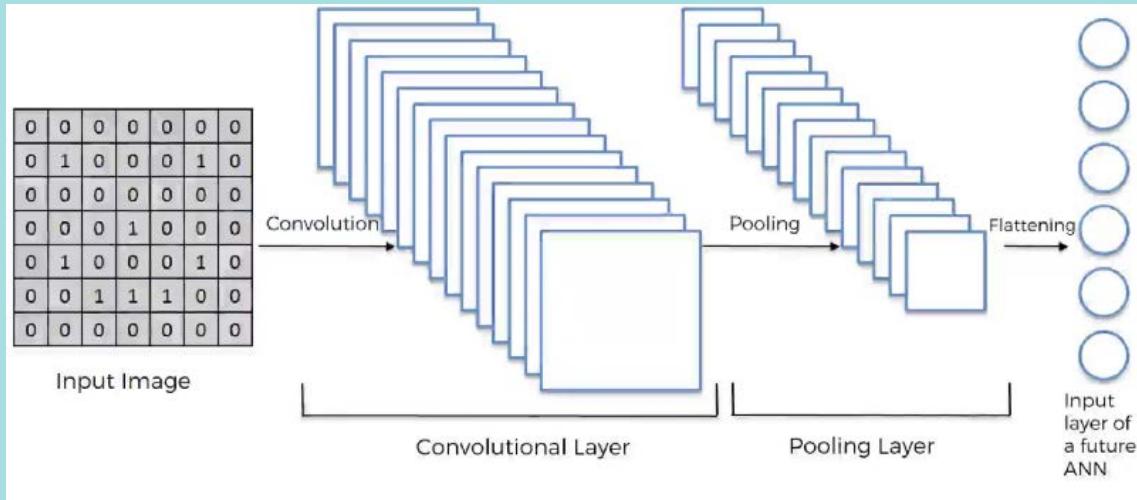
Concepts



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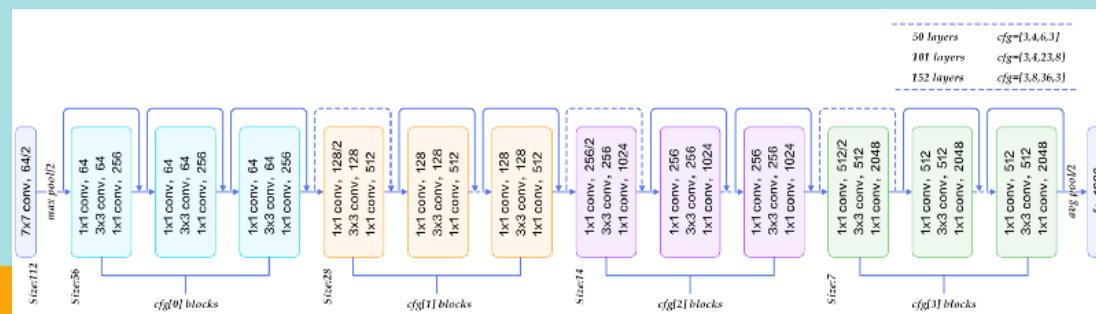
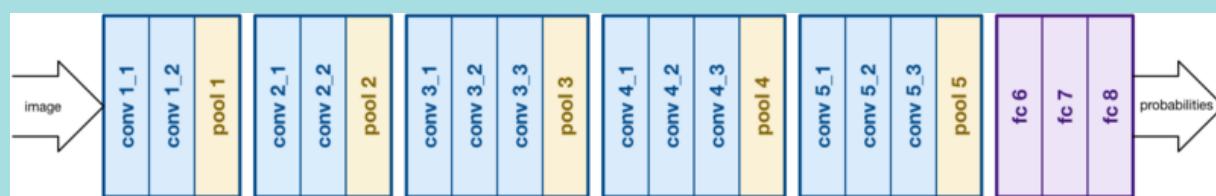
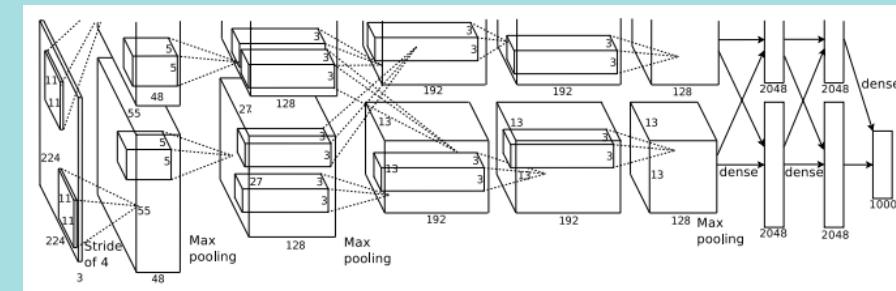
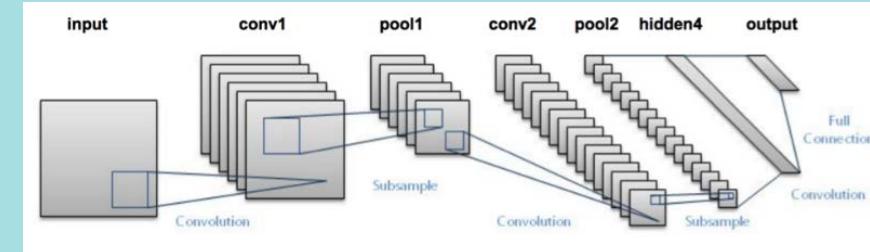
Convolutional Neural Network

- Convolution Layer
 - Feature Detector
- Pooling Layer
 - Max Pooling
- Fully Connected Layer



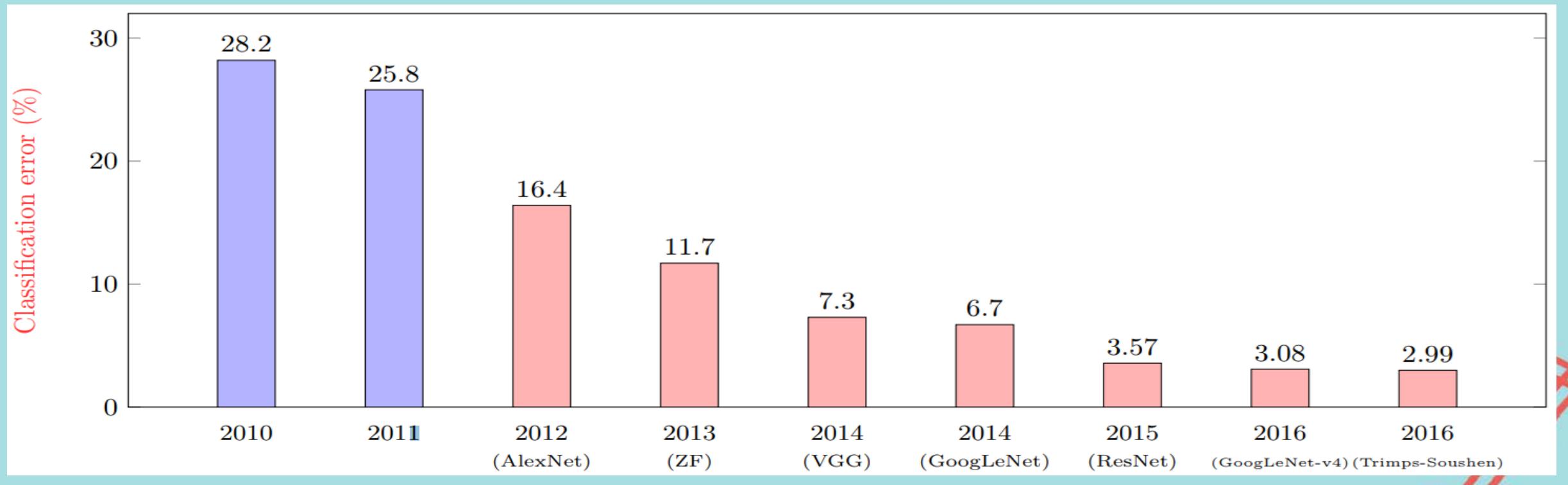
CNN Architectures

- LeNet-5 (1998)
- AlexNet (2012)
- VGG (2014)
- ResNet (2015)

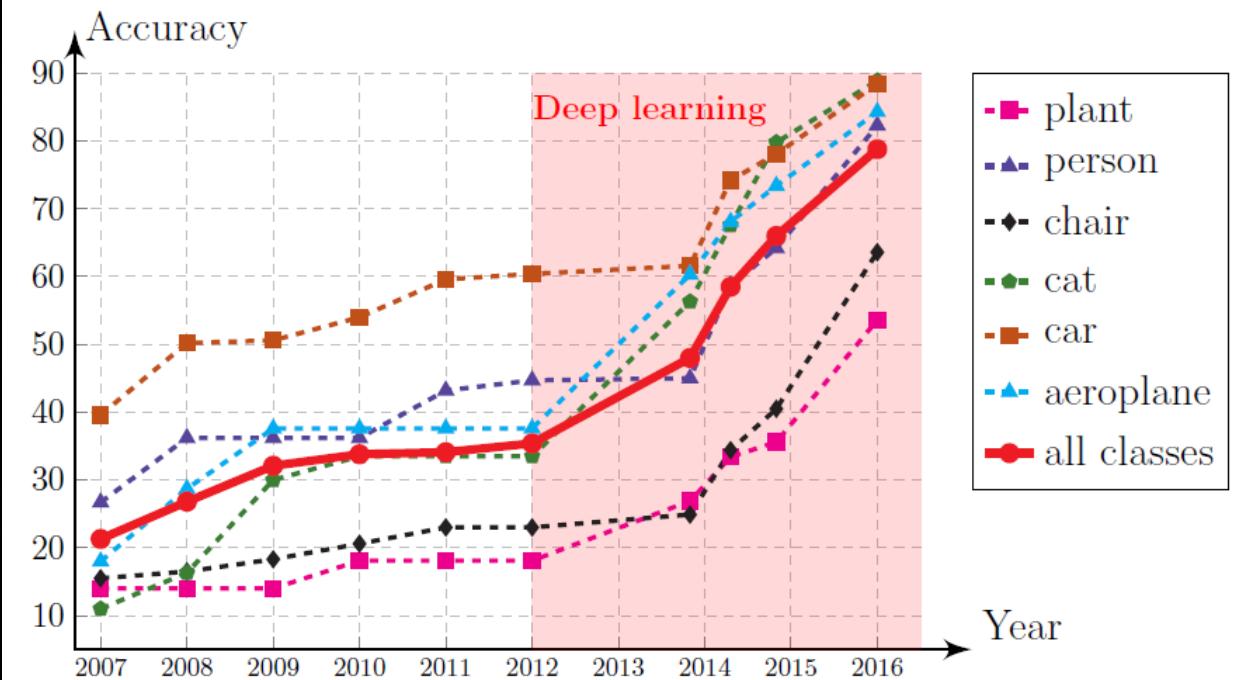
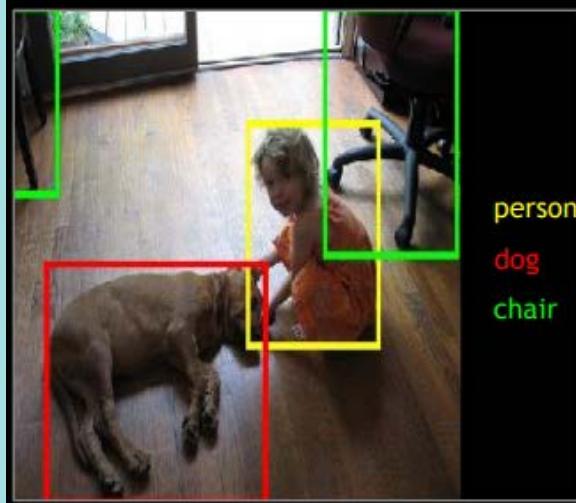
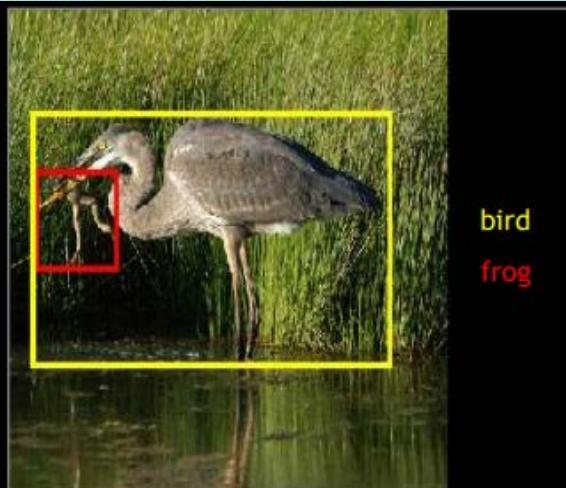
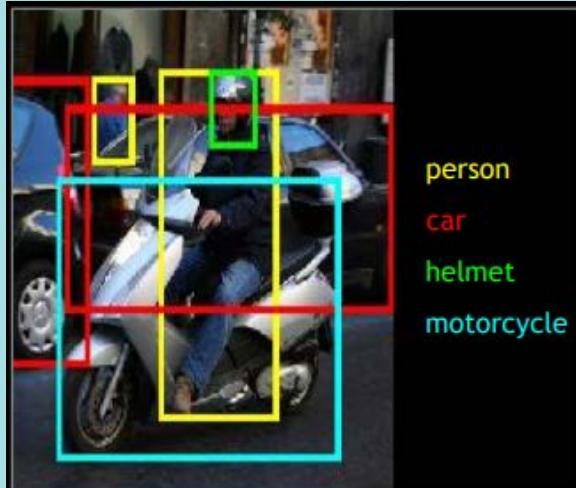


ImageNet Large Scale Visual Recognition Challenge (ILSVRC)

- 1000 object classes, 1.4 million labelled training images



Object Localization - Neural Networks

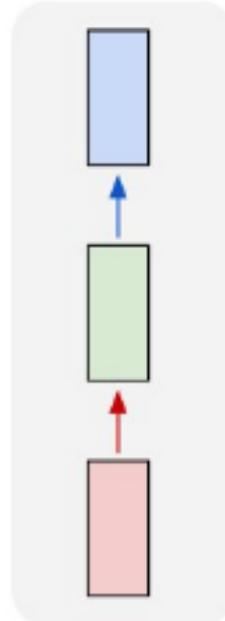


Progress on the Pascal VOC 2007 challenge.

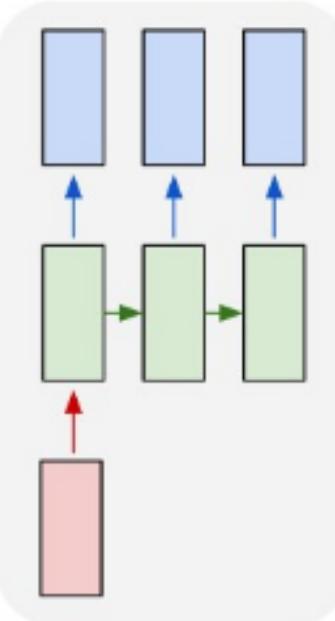
Recurrent Neural Networks

- Sequences - Learn and Generate

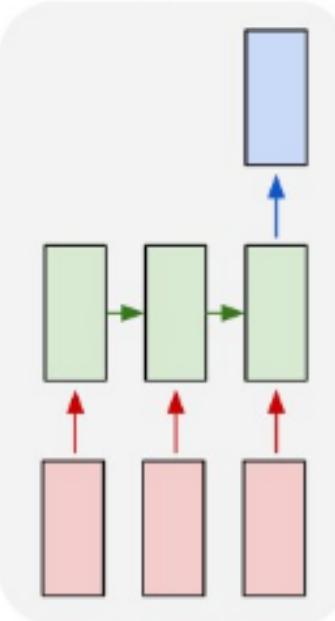
one to one



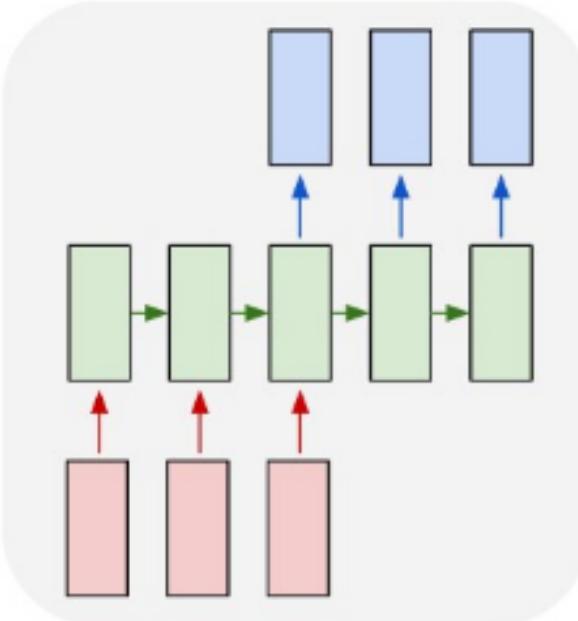
one to many



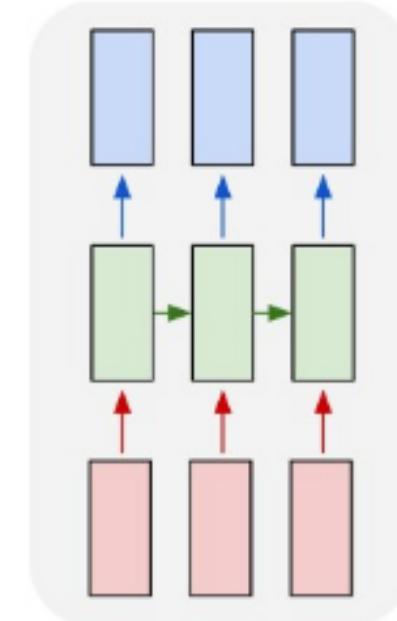
many to one



many to many

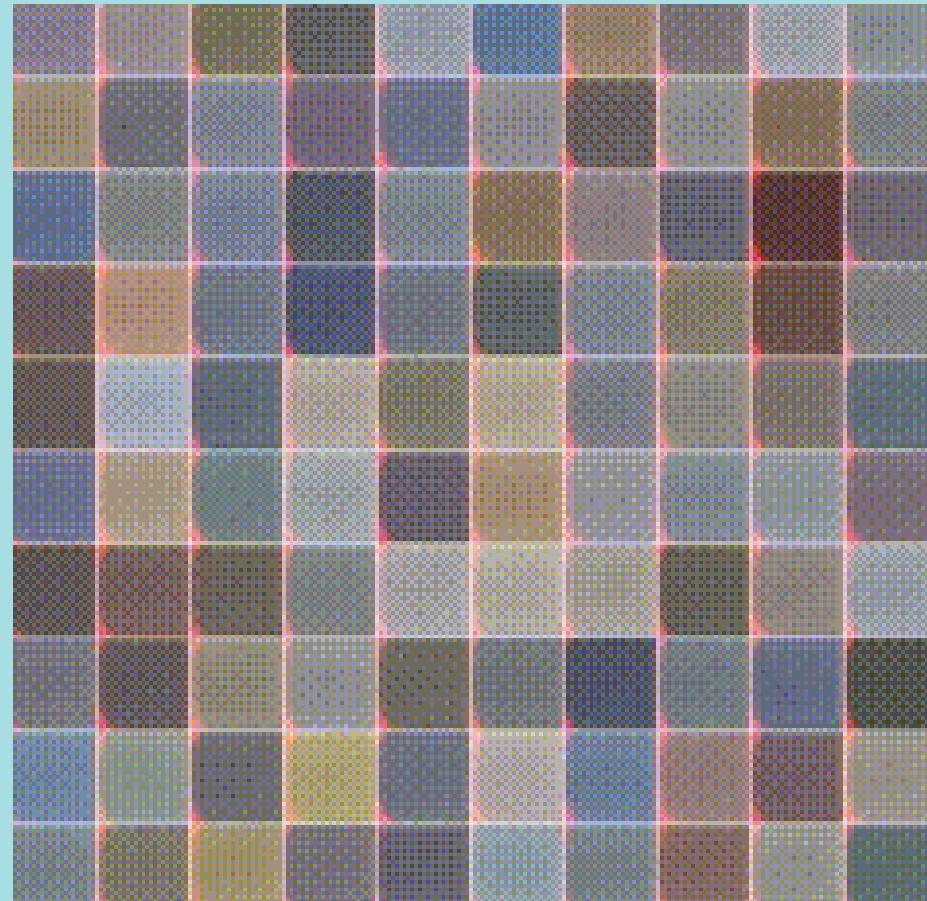
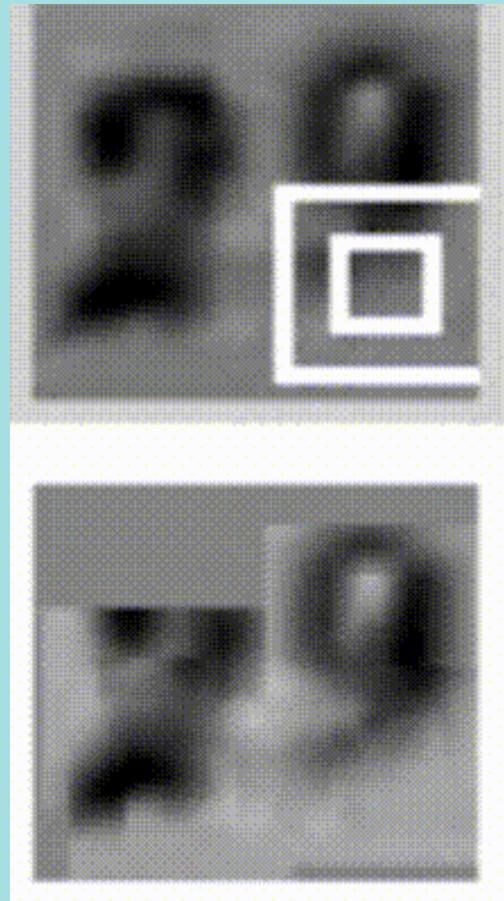


many to many

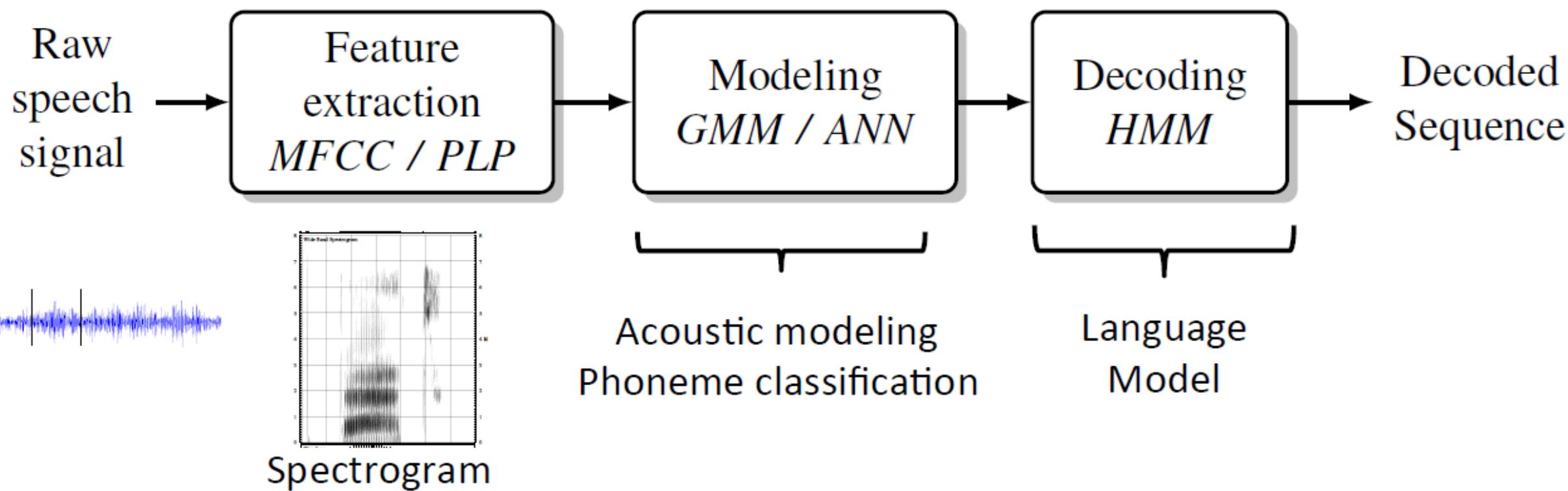


Sequential processing in absence of sequences

- Left: RNN learns to read house numbers. Right: RNN learns to paint house numbers.

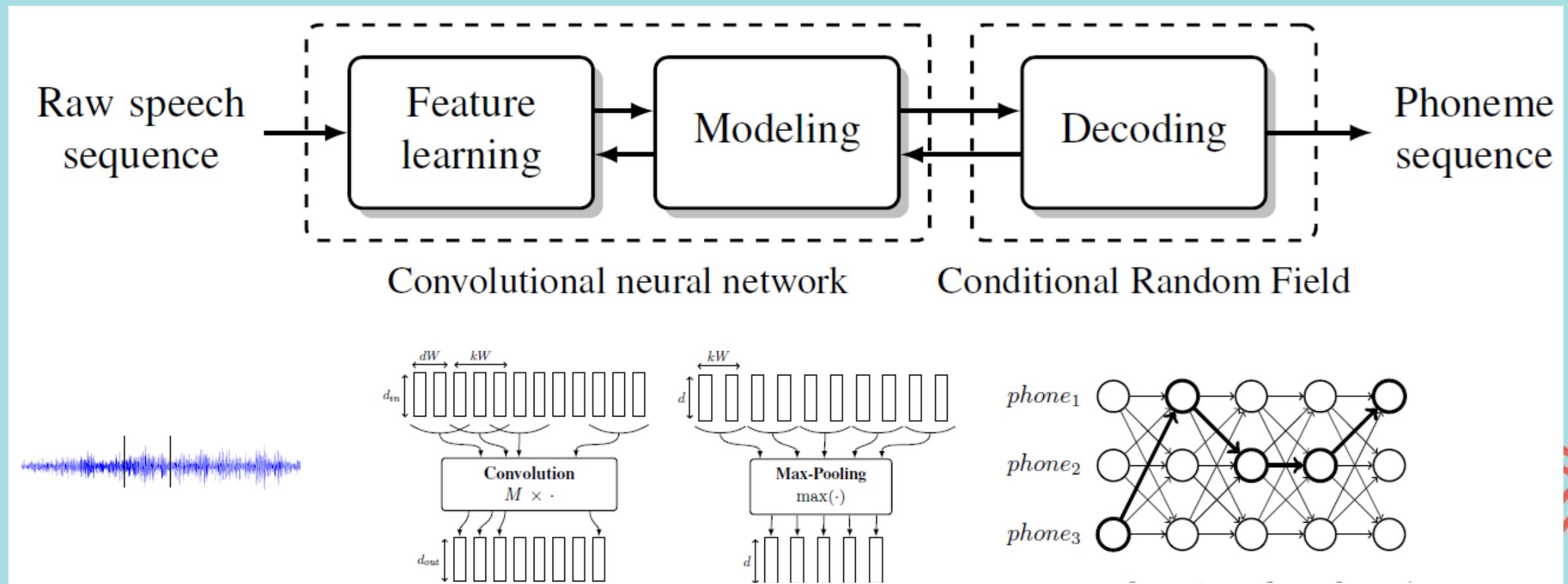


Speech recognition



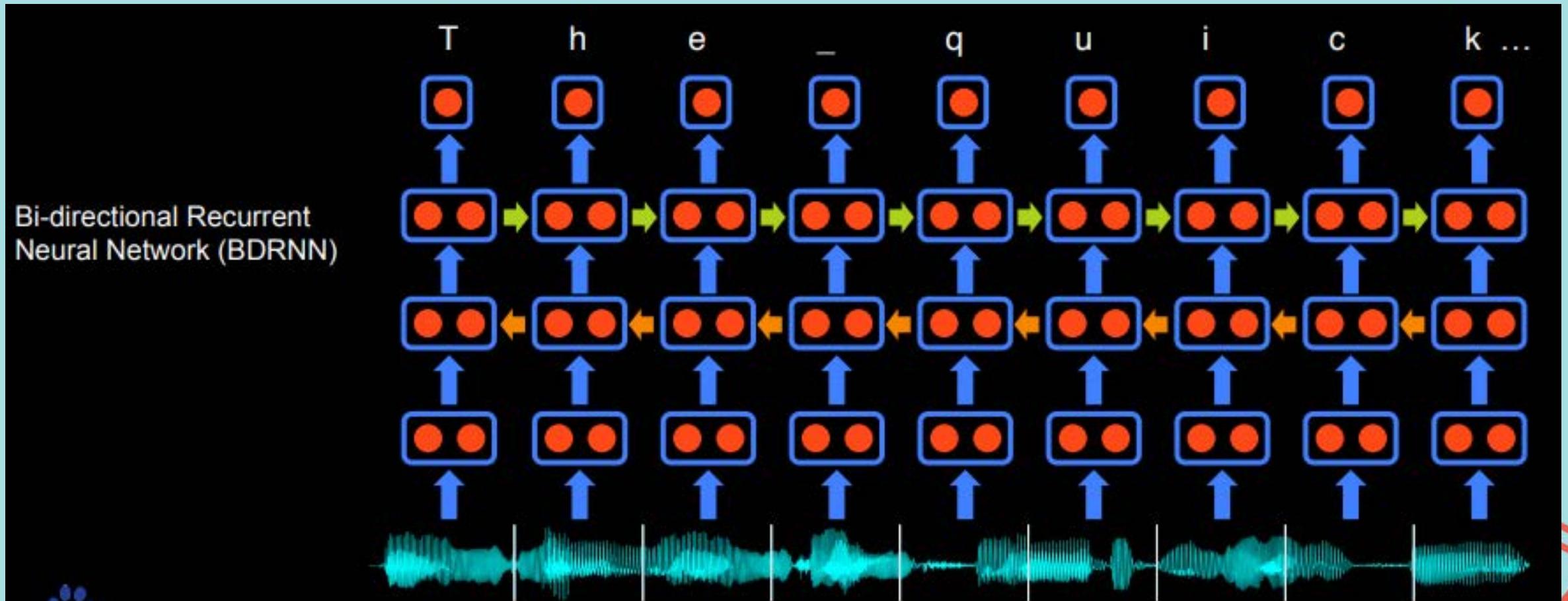
Speech recognition - Neural Networks

- End-to-End Recognition



Palaz, Dimitri, Mathew Magimai Doss, and Ronan Collobert. "Convolutional neural networks-based continuous speech recognition using raw speech signal." *Acoustics, Speech and Signal Processing (ICASSP), 2015 IEEE International Conference on*. IEEE, 2015.

Speech recognition - Recurrent Neural Networks



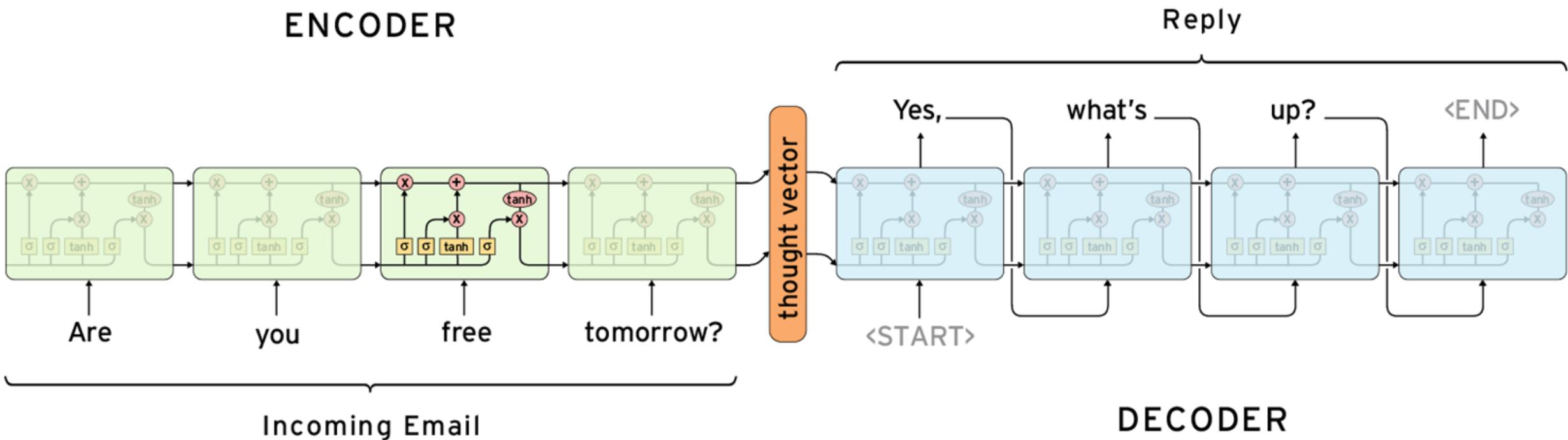
Progress of speech recognition

- After no improvement for 10+ years by the research community . . . MSR reduced error from ~27% to < 13% (and under < 7% for Rick Rashid's demo in 2012)!



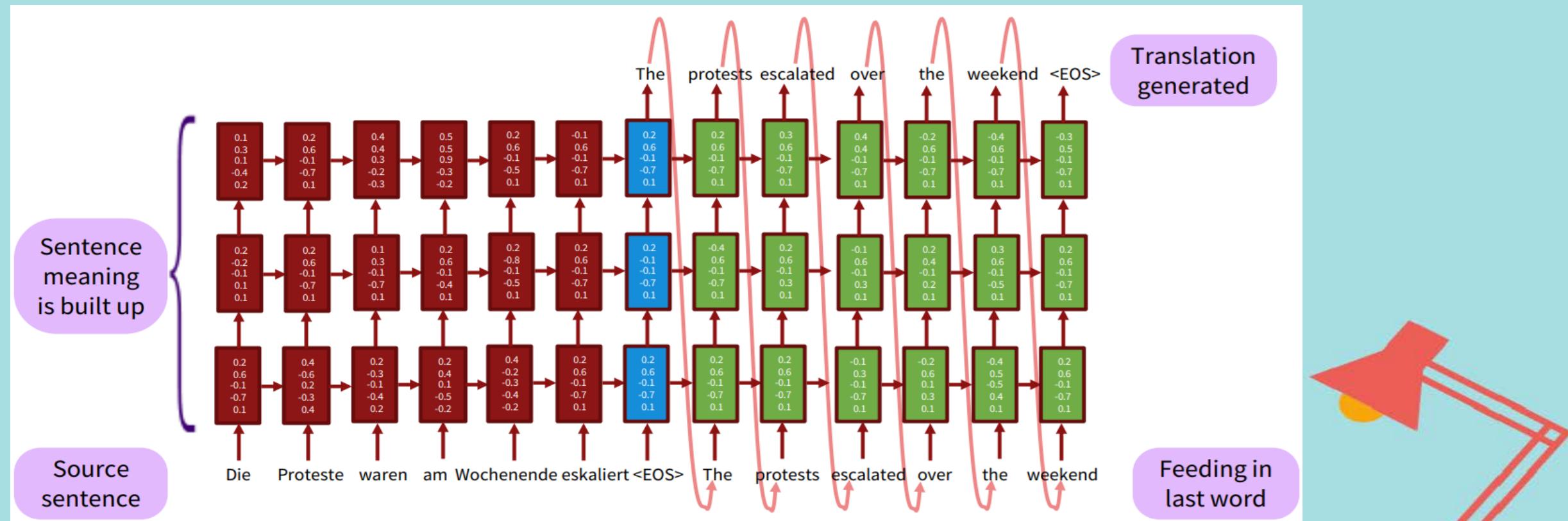
Dialog Systems - Neural Networks

- Sequence-to-Sequence Model



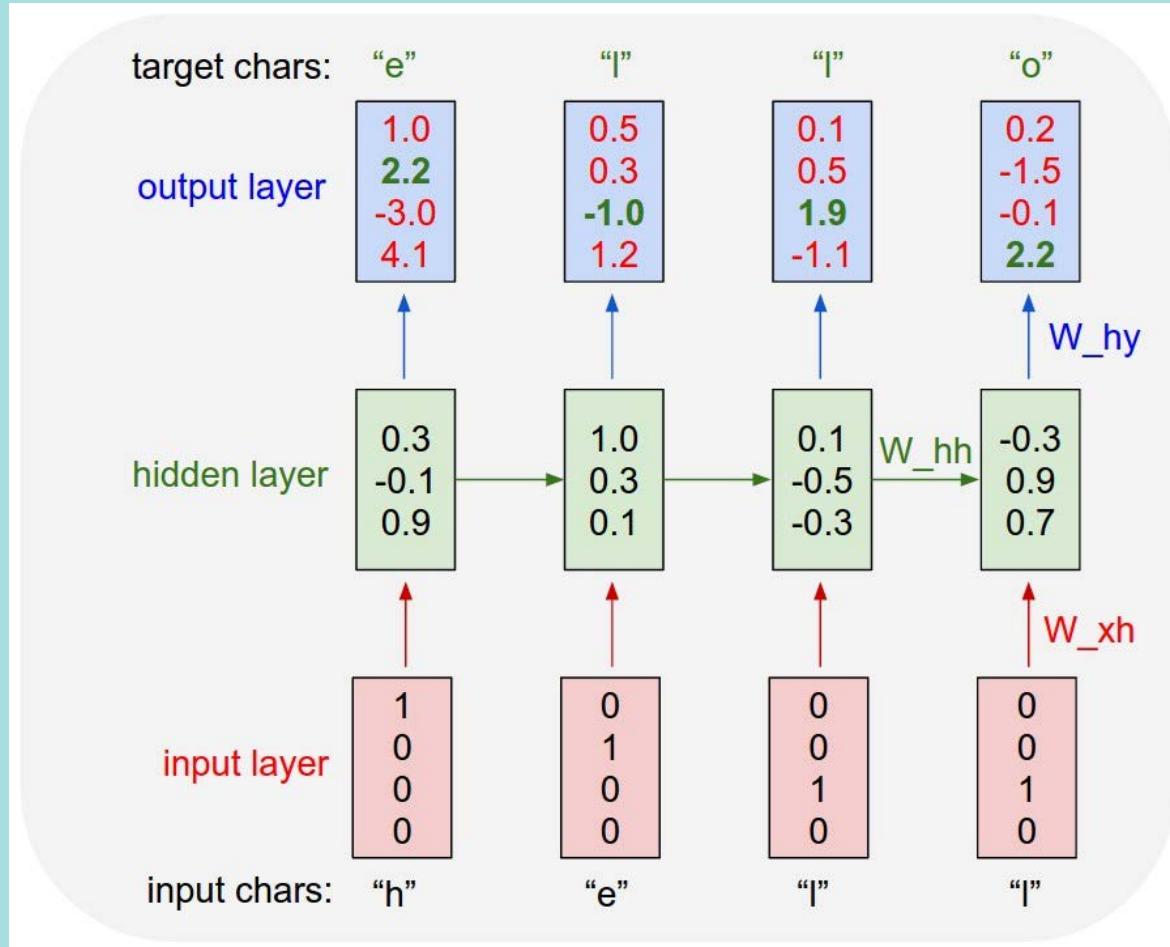
Natural Language Translation - Neural Networks

- Sequence-to-Sequence Model



Sutskever, Ilya, Oriol Vinyals, and Quoc V. Le. "Sequence to sequence learning with neural networks." *Advances in neural information processing systems*. 2014.

Character-Level Language Models

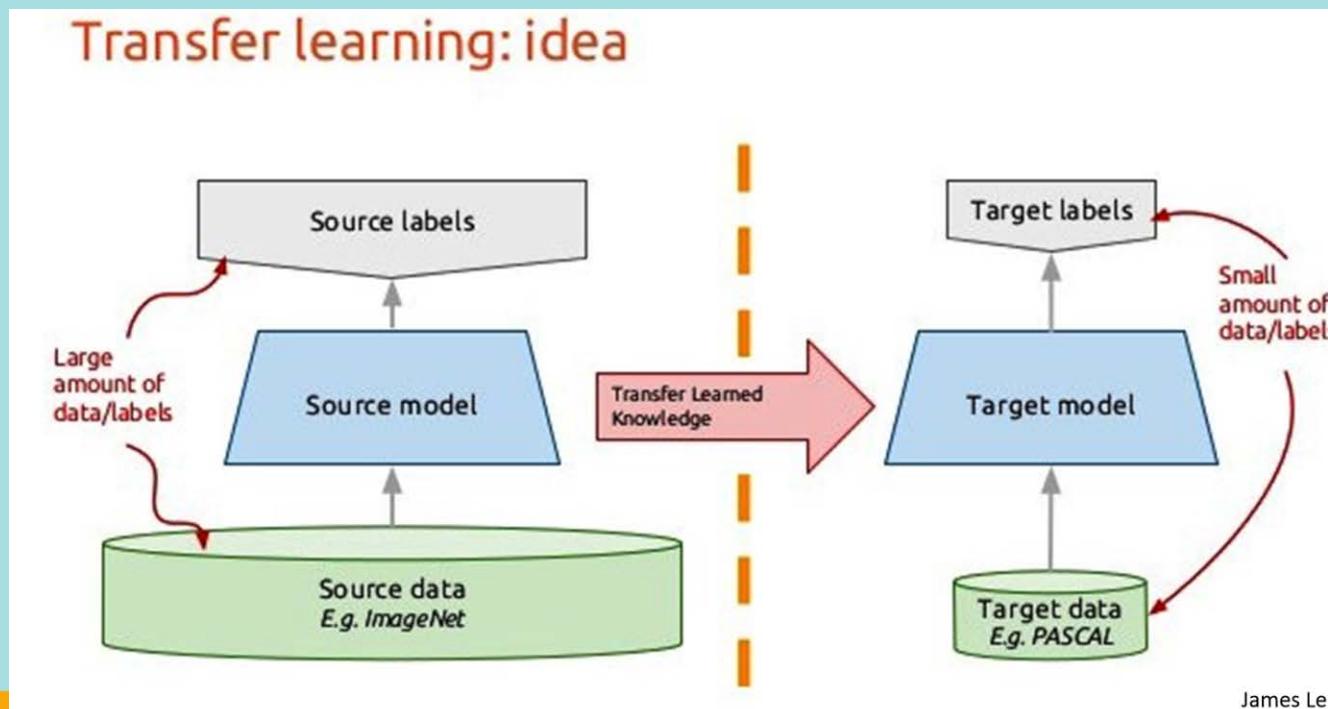


- An example RNN with 4-dimensional input and output layers, and a hidden layer of 3 units (neurons).
- This diagram shows the activations in the forward pass when the RNN is fed the characters "hell" as input.
- The output layer contains confidences the RNN assigns for the next character (vocabulary is "h,e,l,o");
- We want the green numbers to be high and red numbers to be low.



Transfer Learning for Deep Learning

- Transfer learning make use of the knowledge gained while solving one problem and applying it to a different but related problem.



Deep Neural Networks

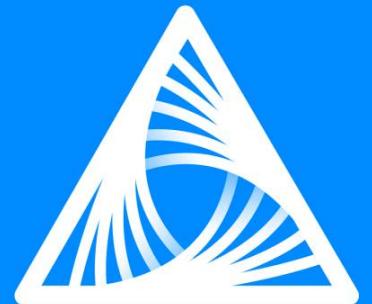
- Image Detection
- Face Recognition
- Gesture Recognition
- Video Search & Analytics
- Speech Recognition & Translation
- Recommendation Engines
- Indexing & Search



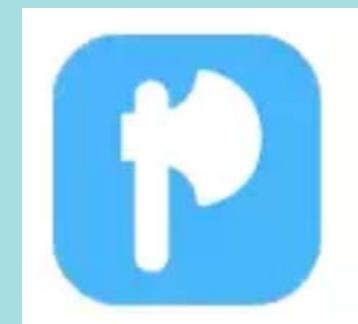
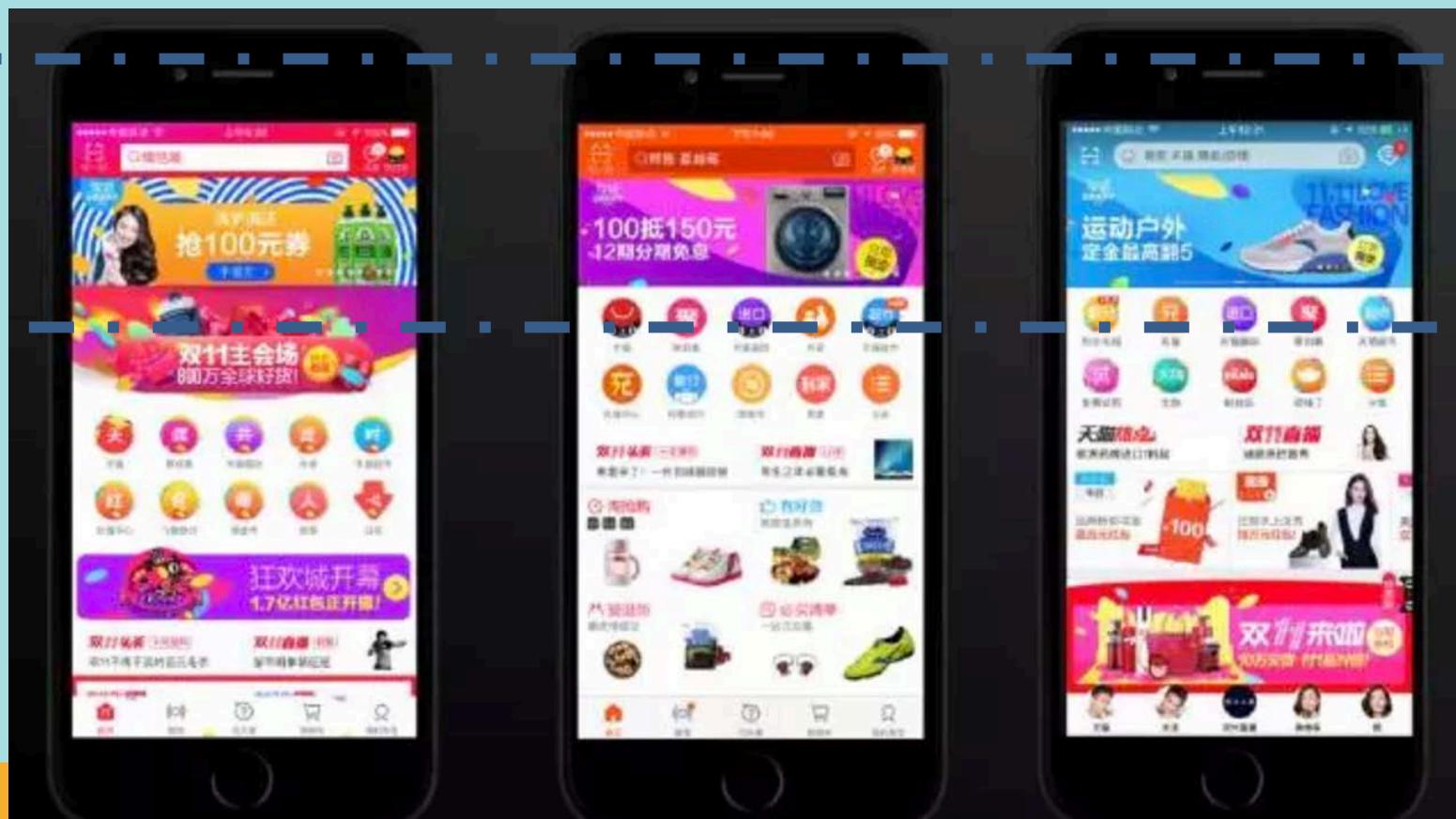
(5) APPLICATION: AI DESIGNER



魯班 (Alibaba Luban AI banner)



Alibaba AI Design Lab



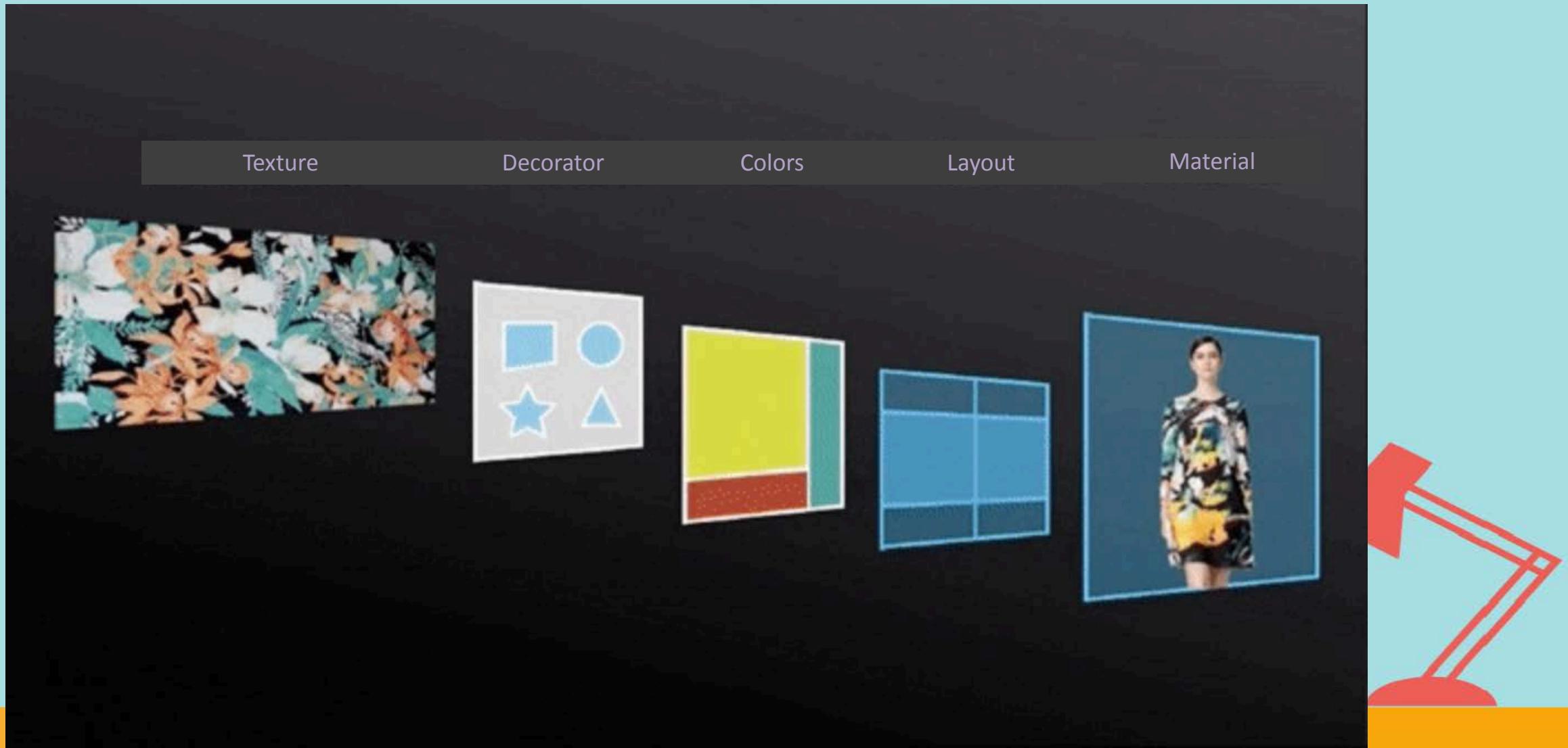
AI-Made banner

170 Million banners



www.dkai.com

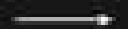
AI Design for banner



构建元素中心

元素中心

背景、主体、修饰



The dimensions of design



Scoring design

1. 设计数据化：更多维，机器可学习

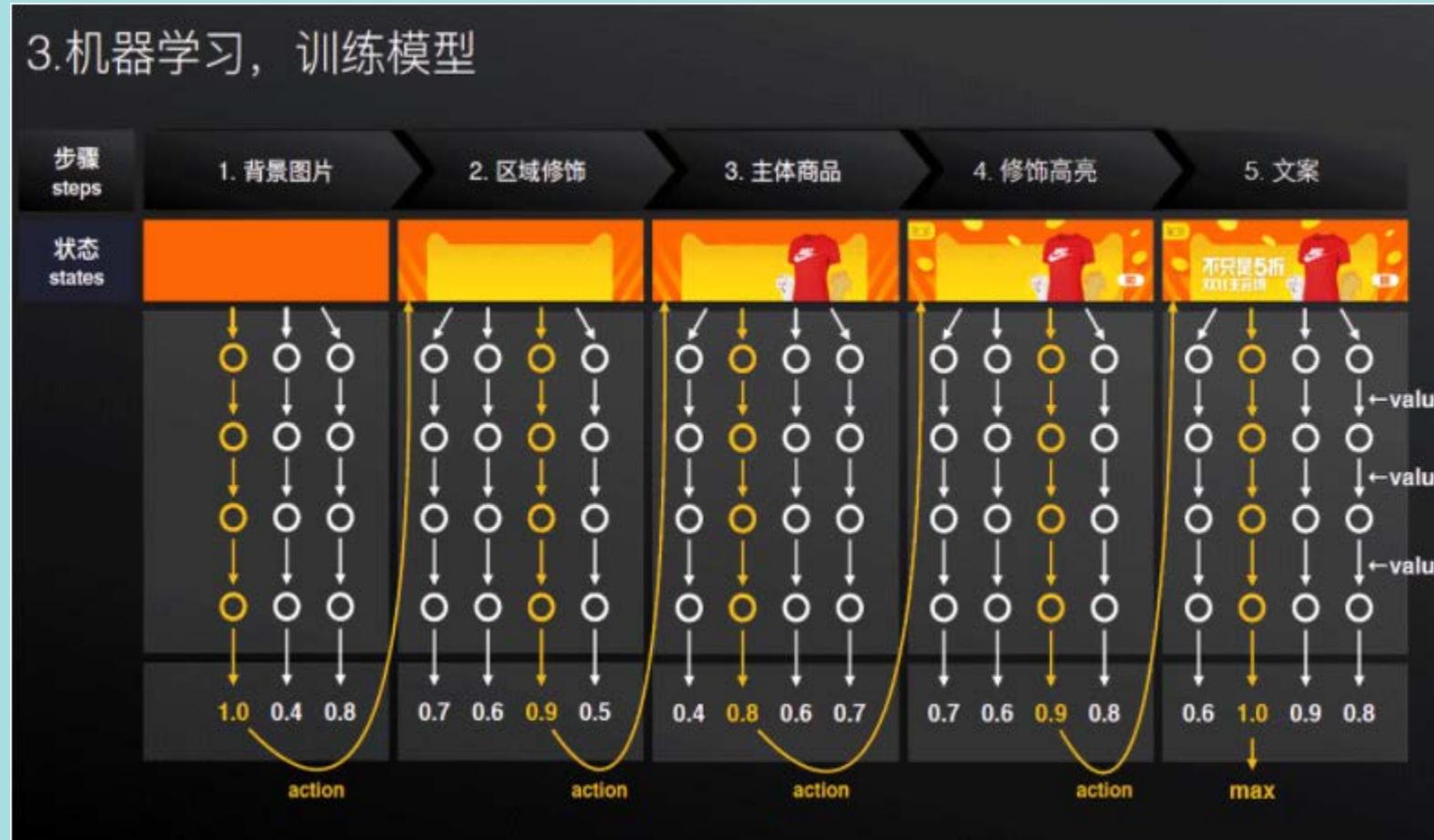


Define the quality of autonomous design

2. 数据规模与质量，标注让机器可理解



The model of deep learning



Training the AI system

4.生成树，生成设计结果并评估



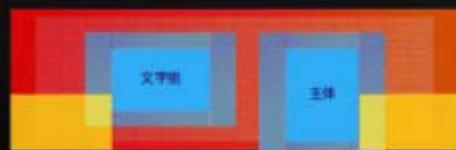
5分



4分



3分



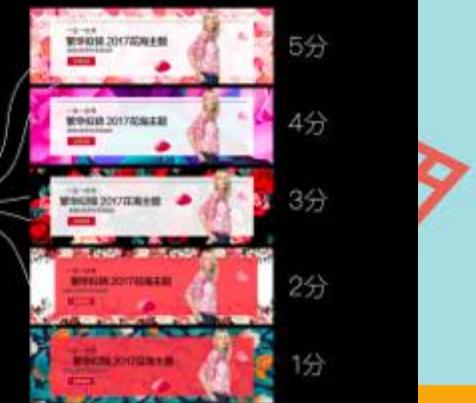
2分



1分



评估网络：人工+机器，对生成结果评估



(6) DISCUSSIONS



The UAE just appointed a minister for Artificial Intelligence



Robocops

Saudi Arabia to invest \$500 billion in fully automated city spanning three countries



This week, Saudi Arabia granted citizenship to a robot and unveiled plans for a fully autonomous city



Danil Boparai | 28 October 2017 | 3 comments



Kaggle

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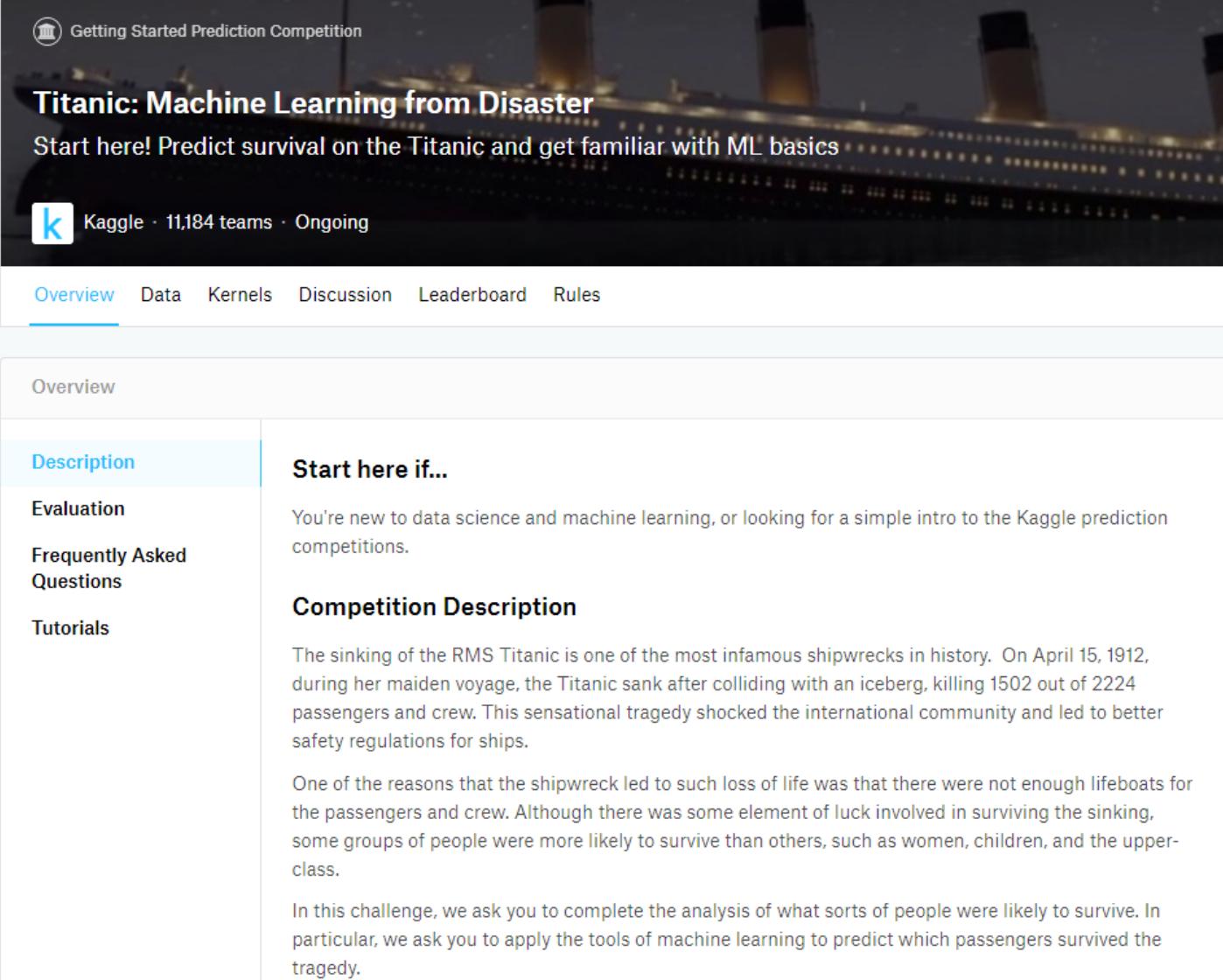
Email

Password

[Sign Up](#)



Titanic: Machine Learning from Disaster



The screenshot shows the Kaggle competition page for 'Titanic: Machine Learning from Disaster'. At the top, it says 'Getting Started Prediction Competition' and features a dark background image of the Titanic ship at night. Below the title, the competition name is displayed in bold. A sub-header reads 'Start here! Predict survival on the Titanic and get familiar with ML basics'. The Kaggle logo and a count of '11,184 teams · Ongoing' are shown. A navigation bar includes 'Overview' (which is underlined in blue), 'Data', 'Kernels', 'Discussion', 'Leaderboard', and 'Rules'. The main content area has a sidebar on the left with links for 'Description', 'Evaluation', 'Frequently Asked Questions', and 'Tutorials'. The main content on the right starts with a section titled 'Start here if...' which explains the competition's purpose for beginners. It then provides a 'Competition Description' detailing the sinking of the RMS Titanic and its historical significance. Following this is a section about the lack of lifeboats and the survival rates of different passenger groups. The final part of the visible content discusses the analysis required to predict survival.

Getting Started Prediction Competition

Titanic: Machine Learning from Disaster

Start here! Predict survival on the Titanic and get familiar with ML basics

Kaggle · 11,184 teams · Ongoing

Overview Data Kernels Discussion Leaderboard Rules

Overview

Description Start here if...

Evaluation You're new to data science and machine learning, or looking for a simple intro to the Kaggle prediction competitions.

Frequently Asked Questions

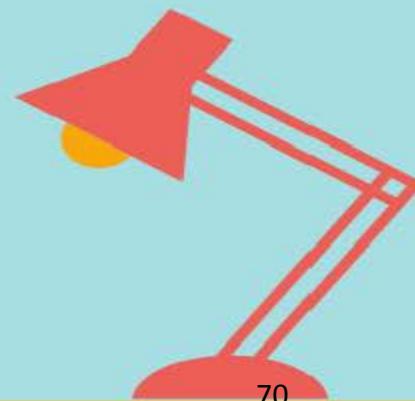
Tutorials

Competition Description

The sinking of the RMS Titanic is one of the most infamous shipwrecks in history. On April 15, 1912, during her maiden voyage, the Titanic sank after colliding with an iceberg, killing 1502 out of 2224 passengers and crew. This sensational tragedy shocked the international community and led to better safety regulations for ships.

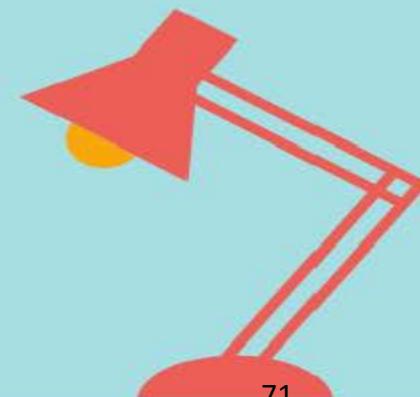
One of the reasons that the shipwreck led to such loss of life was that there were not enough lifeboats for the passengers and crew. Although there was some element of luck involved in surviving the sinking, some groups of people were more likely to survive than others, such as women, children, and the upper-class.

In this challenge, we ask you to complete the analysis of what sorts of people were likely to survive. In particular, we ask you to apply the tools of machine learning to predict which passengers survived the tragedy.



Assignment 1B

- Step 1: Open colab and create a new notebook
 - <https://colab.research.google.com>
- Step 2: Edit the notebook with text cells
 - <https://guides.github.com/features/mastering-markdown/>
- Step 3:
 - List AI Terms in this topics
 - Search AI challenge from the Internet and introduce one of them
- Step 4:
 - Rename the notebook: Untitled.ipynb->day1b.ipynb
 - Copy the notebook to GitHub
 - Check the notebook on GitHub





Thanks! Q&A

