

# Computational Intelligence

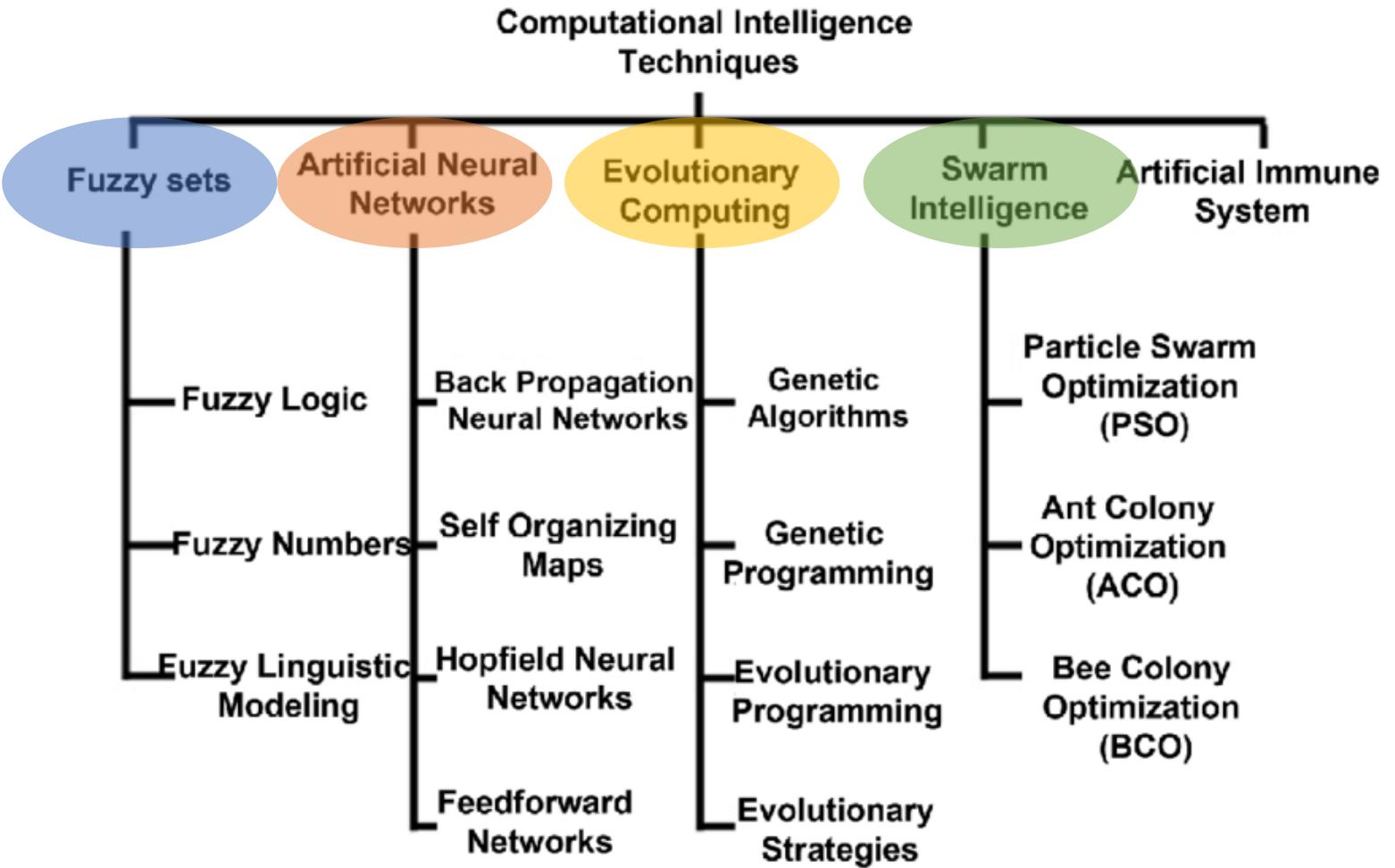
Samaneh Hosseini

Isfahan University of Technology

# Outline

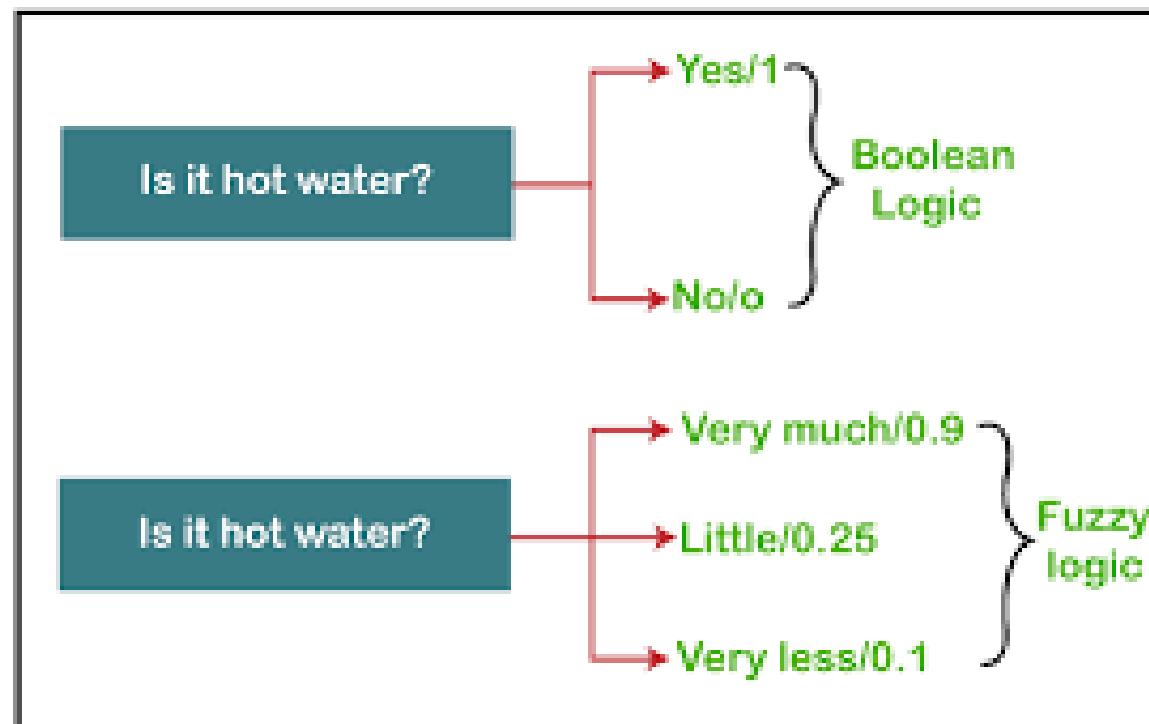
- What is Computational Intelligence?
- Computational Intelligence Applications
- Why Computational Intelligence and Why now?
- Course administrations

# What is Computational Intelligence?



# Fuzzy logic introduction

- The term fuzzy refers to things that are not clear or are vague

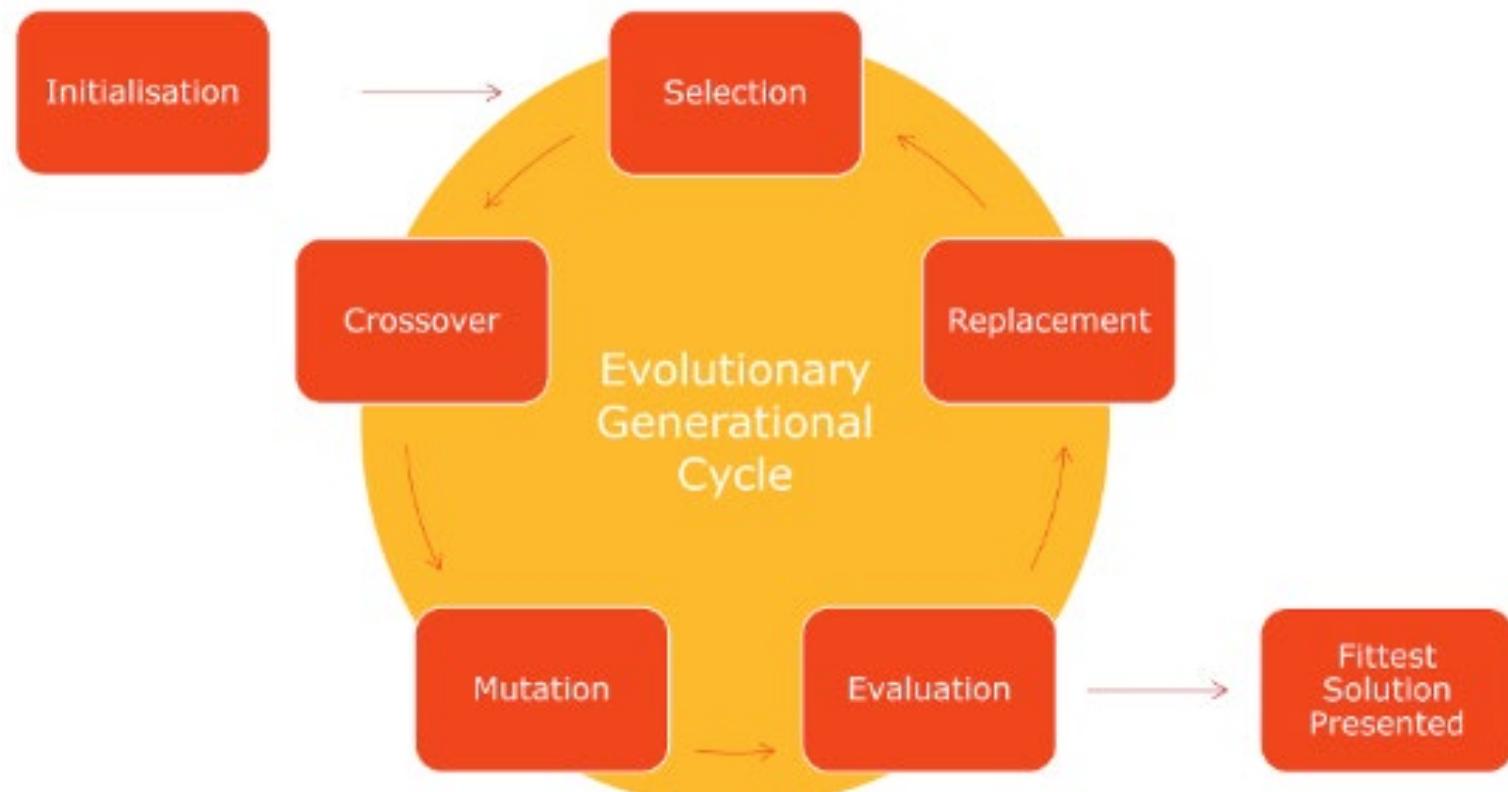


# Evolutionary Computing

- simulate physical and/or biological behavior in nature to solve optimization problems.
- Try to solve problems that:
  1. Very difficult to model mathematically.
  2. Computationally expensive to solve.
  3. Involves a large number of parameters.

# Genetic Algorithms

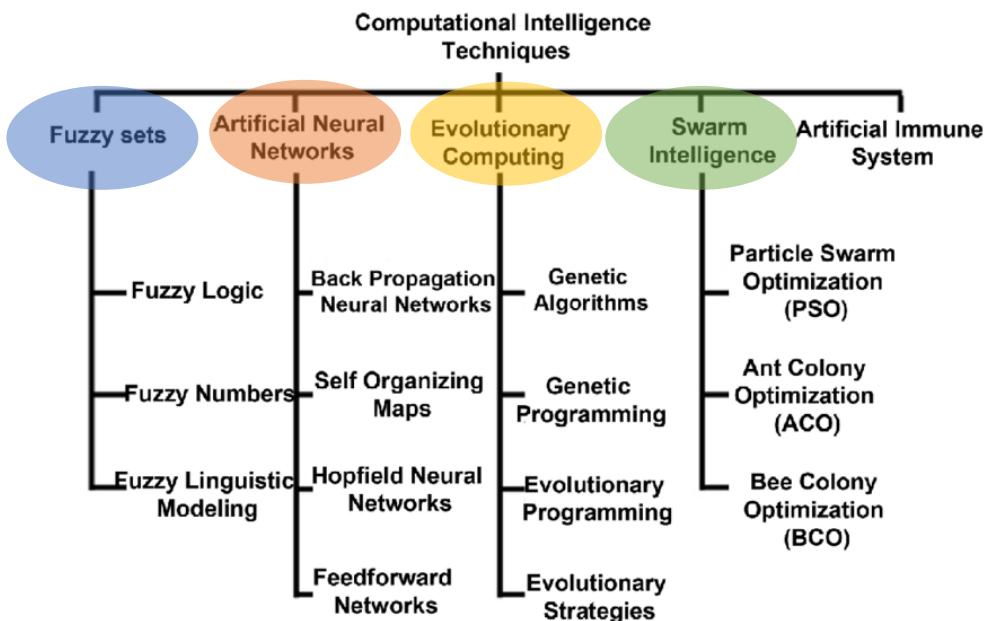
- a subset of evolutionary algorithms
- simulates Genetics and Evolution (biological behavior)



# Swarm intelligence



# What is Computational Intelligence?



و کامپیوتر - سماوه حسینی سمنانی

- L1-Introduction
- L2-Neuron math model
- L3-Perceptron
- L4-Building and Applying NN
- L5-Gradient descent
- L6-Vectorization
- L7-Overfitting
- L8-Regularization I
- L9-Regularization II
- L10-Optimization Algorithms I (mini-batches)
- L11-Optimization Algorithms II(exponentially weighted averages)
- L12-Hyperparameter Tuning
- L13-Batch Normalization
- L14-Softmax
- L15-Convolutional Neural Networks
- L16-Padding, Strided convolution
- L17-Simple Convolutional Network
- L18- Genetic Algorithms
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- L6-Particle Swarm Intelligence (PSO)
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- L8-Discrete PSO
- L9-PSO Hyper-parameters tuning
- L10-PSO Applications
- L11-PSO Applications
- L17-Ant Colony Optimization (ACO)
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- L19-Ant Colony Optimization (ACO)
- L20-ACO applications
- L21-ACO applications
- L22-ACO applications

# What is Neural Network?

## ARTIFICIAL INTELLIGENCE

Any technique that enables computers to mimic human behavior



## MACHINE LEARNING

Ability to learn without explicitly being programmed



## DEEP LEARNING

Extract patterns from data using neural networks



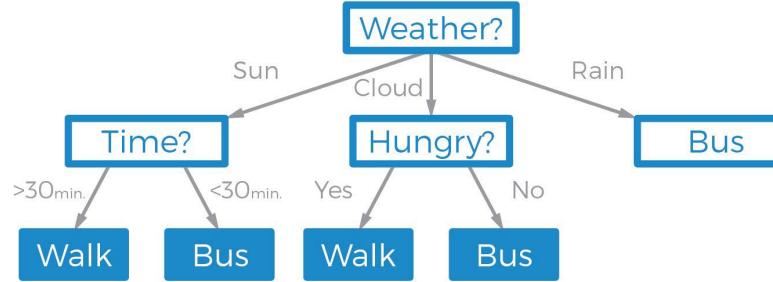
# Why do we need to neural networks?

- ML algorithms are mathematical algorithms that allow machines to learn by imitating the way humans learn,
- Machine learning is basically a way to get artificial intelligence.
- Instead of writing a program by hand, we collect lots of examples that specify the correct output for a given input.
- A machine learning algorithm then takes these examples and produces a program that does the job.

# Machine Learning



Input



Decision tree

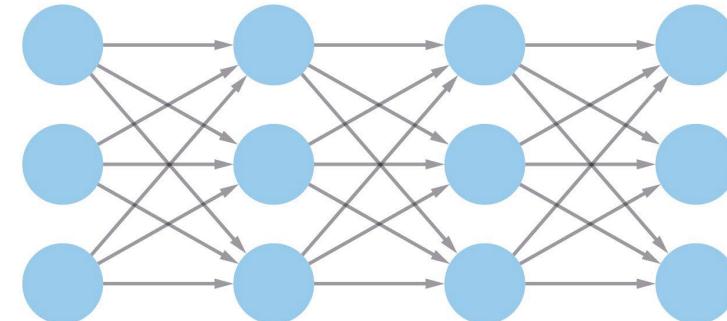


Output

# Deep Learning



Input



Feature extraction + Classification



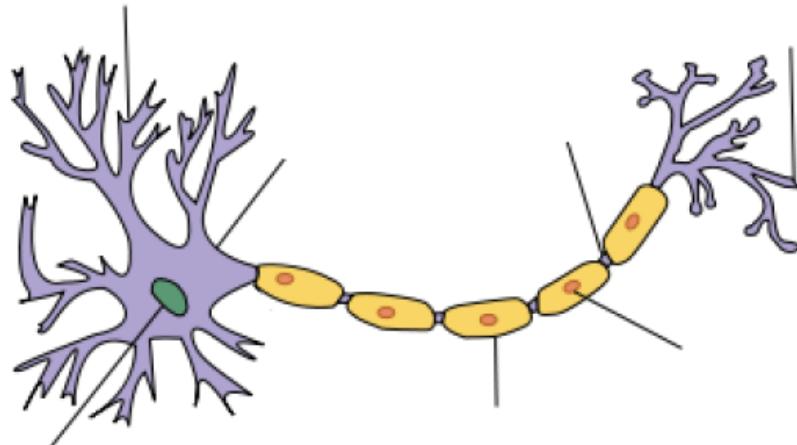
Output

# What is Neural Networks?

- Neural Networks attempts to learn representations of data with multiple levels of abstraction.
- Neural Networks usually refers to a set of algorithms and computational models that are composed of multiple processing layers.
- These methods have significantly improved the state-of-the-art in many domains including, speech recognition, classification, pattern recognition, drug discovery, and genomics.

# Human brains

- ▶ A brain is a set of densely connected neurons.
- ▶ Components of a neuron: dendrites, soma, axon, synapse
- ▶ Depending on the input signals, the neuron performs computations and decides to fire or not.



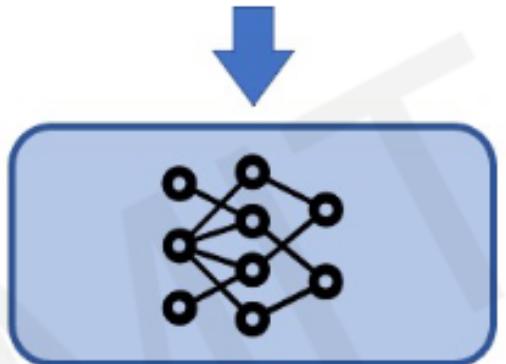
# Applications

- ▶ Medical diagnosis
- ▶ Spam filtering
- ▶ Facial recognition
- ▶ Speech understanding
- ▶ Handwriting recognition

# Success Stories

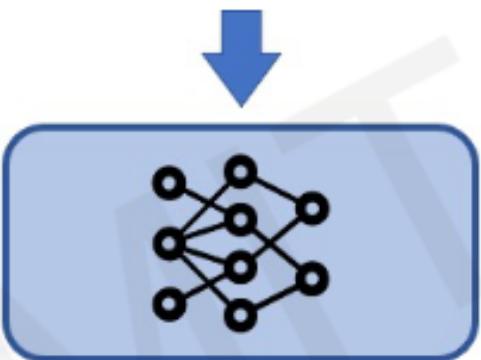
# Generating Images from natural language

“A photo of an astronaut  
riding a horse.”



# Generating Images from natural language

“Write code in  
TensorFlow to train a  
neural network.”



Certainly! Here is an example of how you can use TensorFlow to train a neural network for binary classification:

```
import tensorflow as tf

# Load the data
(x_train, y_train), (x_test, y_test) = load_data()

# Define the model
model = tf.keras.Sequential([
    tf.keras.layers.Input(shape=x_train.shape[1:]),
    tf.keras.layers.Dense(16, activation='relu'),
    tf.keras.layers.Dense(1, activation='sigmoid')
])

# Compile the model
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])

# Train the model
model.fit(x_train, y_train, epochs=10, batch_size=32, validation_data=(x_test, y_test))
```

This code assumes that you have a function `load\_data()` that returns the training and test datasets as tuples of NumPy arrays. The input shape of the model is determined by the shape of the training data, and the model has two dense layers with 16 and 1 units, respectively, with ReLU and sigmoid activations. The model is then compiled with an Adam optimizer and a binary cross-entropy loss function, and is trained using the `fit()` method.

# Video Generation

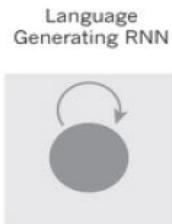
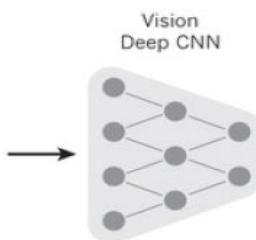


# Games

Deep Learning Machine Teaches Itself Chess in 72 Hours, Plays at International Master Level.

An artificial intelligence machine plays chess by evaluating the board rather than using brute force to work out every possible move.

# Caption Generation



A group of people shopping at an outdoor market.

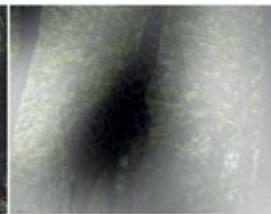
There are many vegetables at the fruit stand.



A woman is throwing a **frisbee** in a park.

A dog is standing on a hardwood floor.

A **stop** sign is on a road with a mountain in the background



A little **girl** sitting on a bed with a teddy bear.

A group of **people** sitting on a boat in the water.

A giraffe standing in a forest with **trees** in the background.

# Word embedding

**Word2vec** , Mikolov, 2013.

$$\text{king} - \text{man} + \text{woman} = \text{queen}$$

# Word embedding



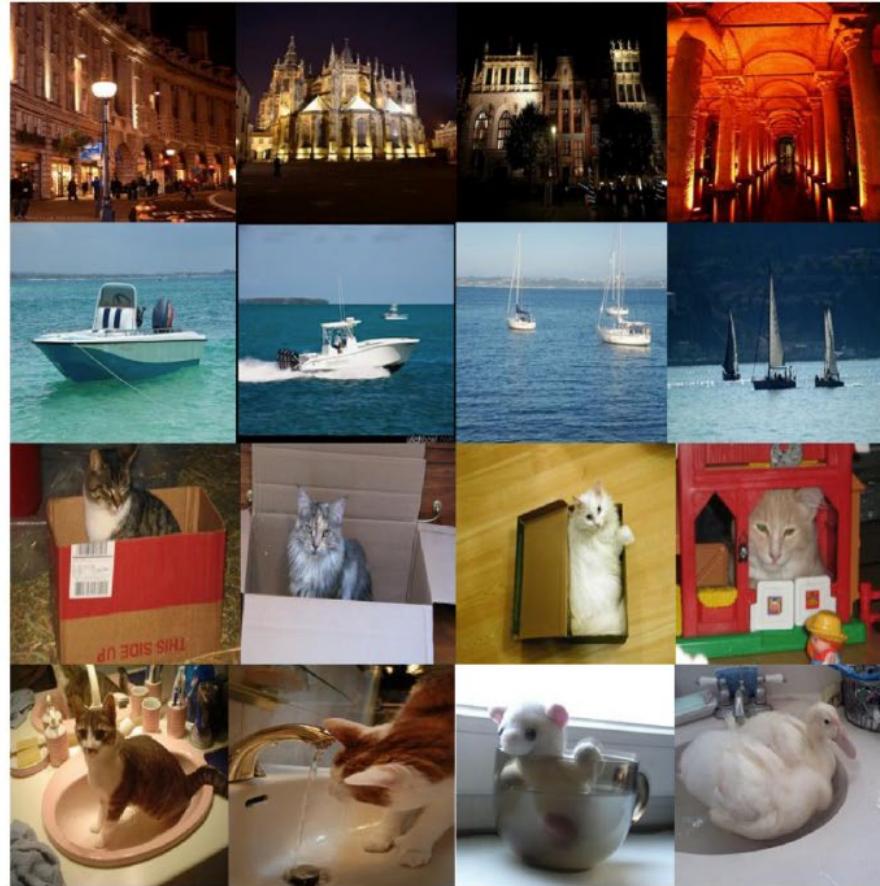
- day + night =

- flying + sailing =

- bowl + box =

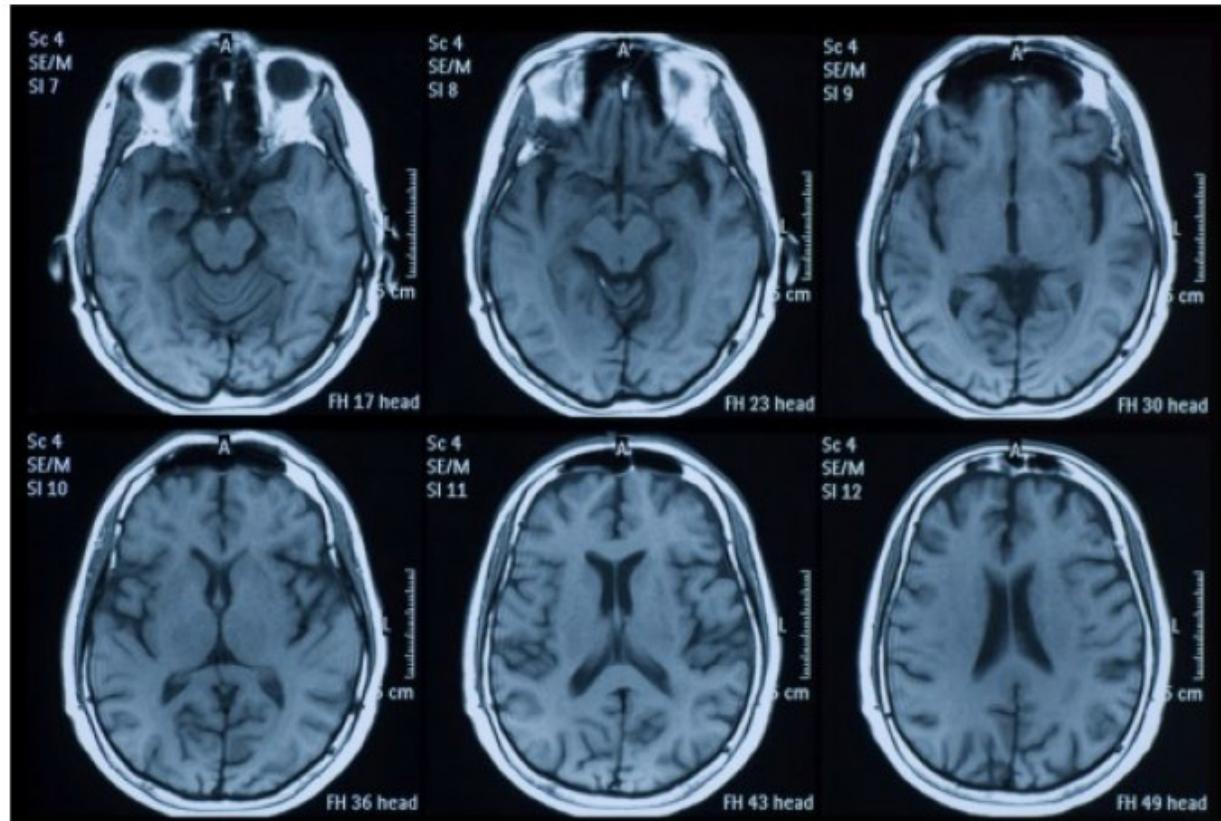
- box + bowl =

Nearest Images



(Kiros, Salakhutdinov, Zemel, TACL 2015)

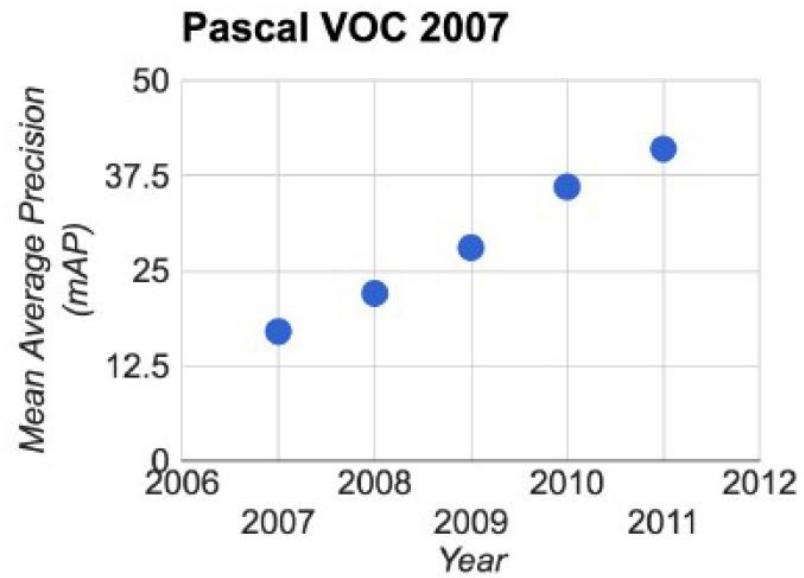
# Medical Image processing



# Success Stories

## PASCAL Visual Object Challenge (20 object categories)

[Everingham et al. 2006-2012]



# Success Stories

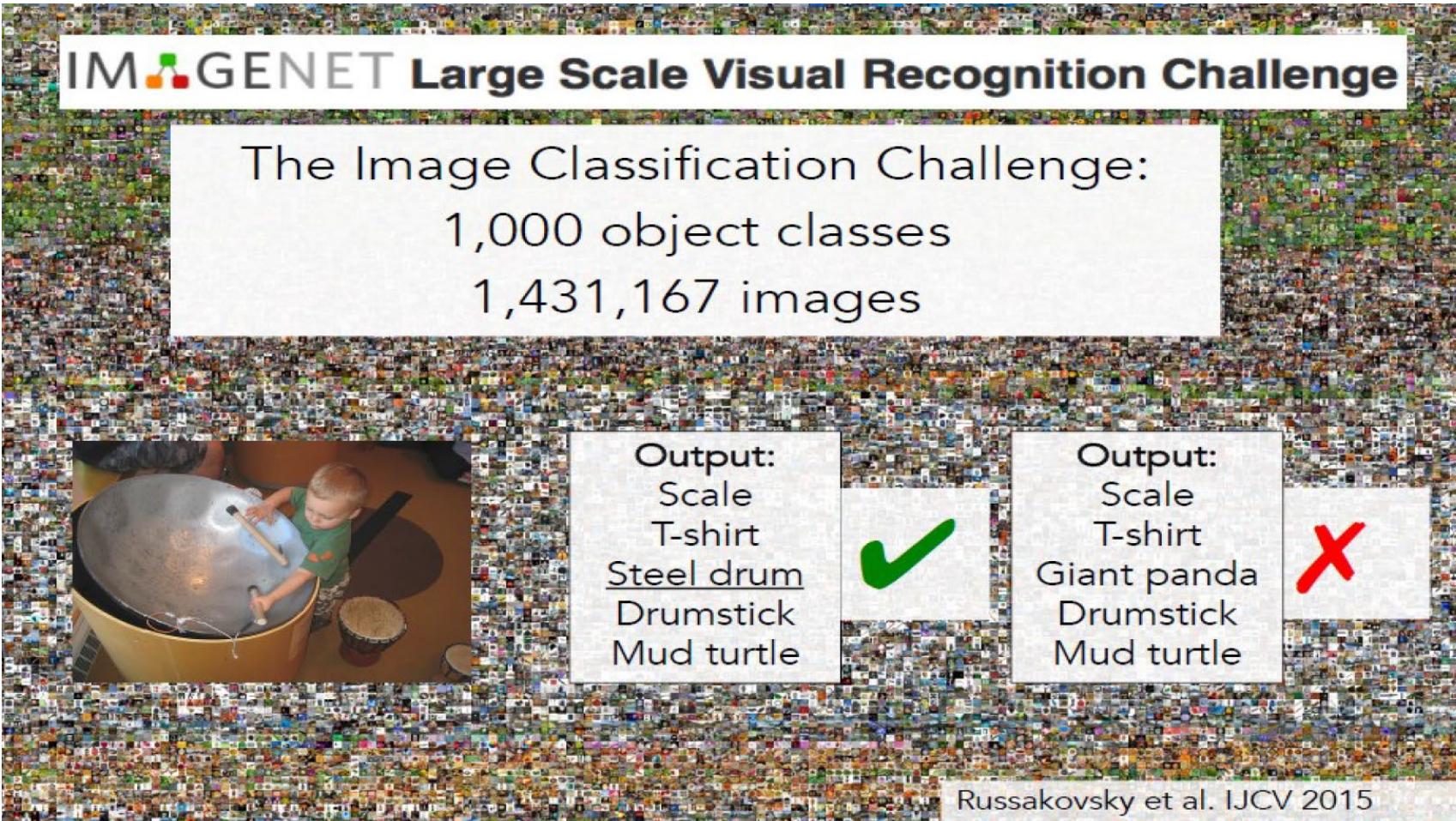


**IMAGENET** [www.image-net.org](http://www.image-net.org)

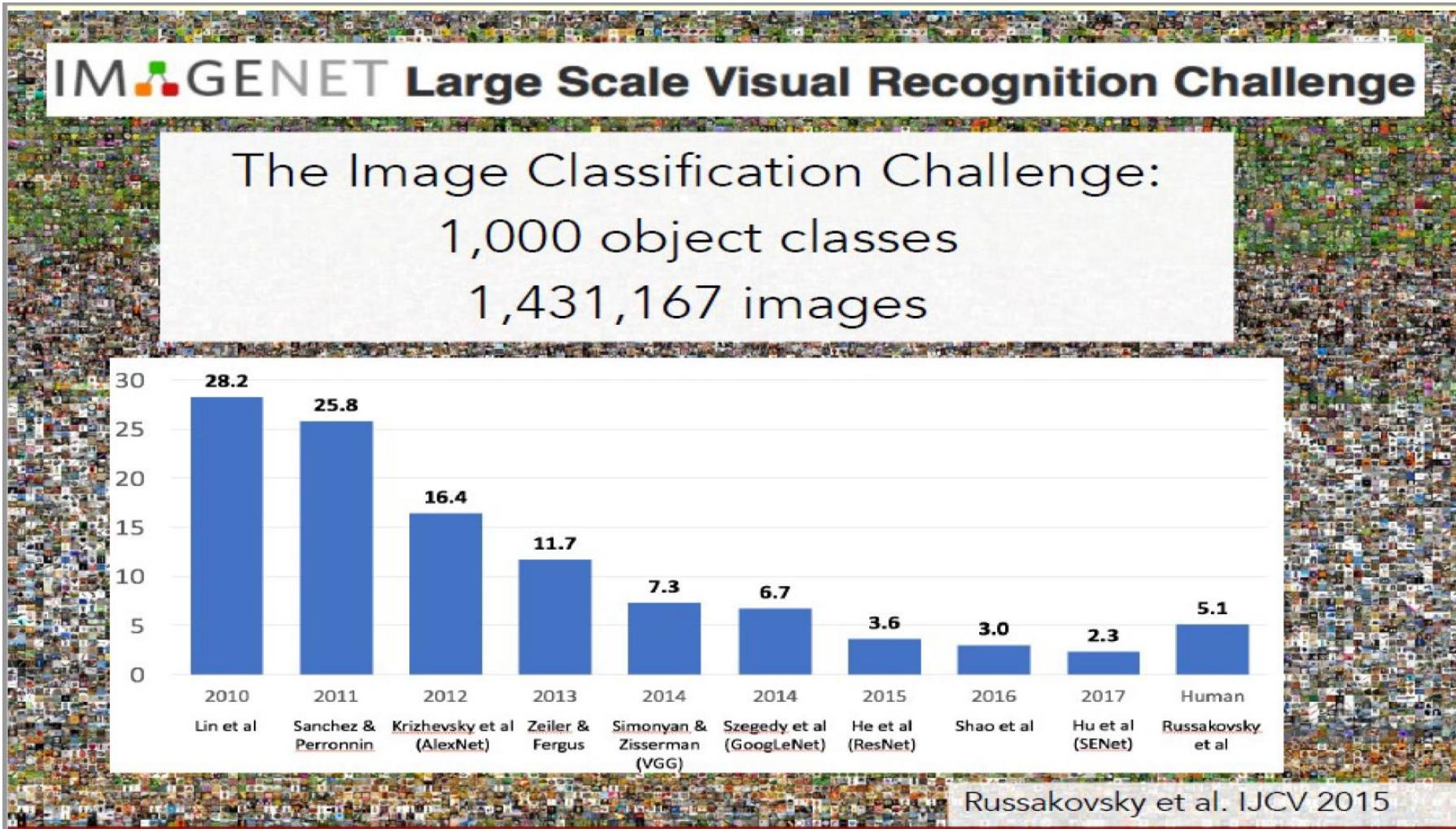
**22K categories and 14M images**

- Animals
  - Bird
  - Fish
  - Mammal
  - Invertebrate
- Plants
  - Tree
  - Flower
  - Food
  - Materials
- Structures
  - Artifact
  - Tools
  - Appliances
  - Structures
- Person
- Scenes
  - Indoor
  - Geological Formations
- Sport Activities

# Success Stories



# Success Stories

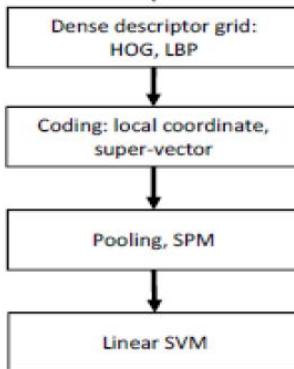


# Success Stories

## IMAGENET Large Scale Visual Recognition Challenge

Year 2010

NEC-UIUC

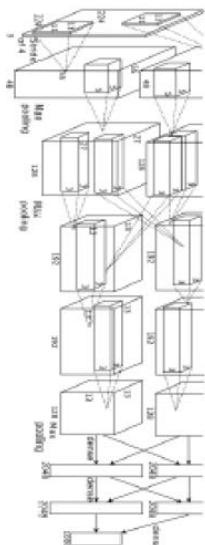


[Lin CVPR 2011]

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Year 2012

SuperVision



[Krizhevsky NIPS 2012]

Figure copyright Alex Krizhevsky, Ilya Sutskever, and Geoffrey Hinton, 2012. Reproduced with permission.

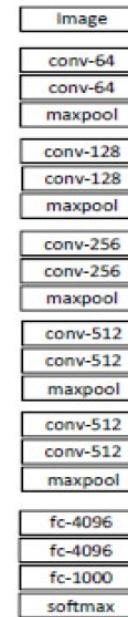
Year 2014

GoogLeNet



[Szegedy arxiv 2014]

VGG



[Simonyan arxiv 2014]

Year 2015

MSRA



[He ICCV 2015]

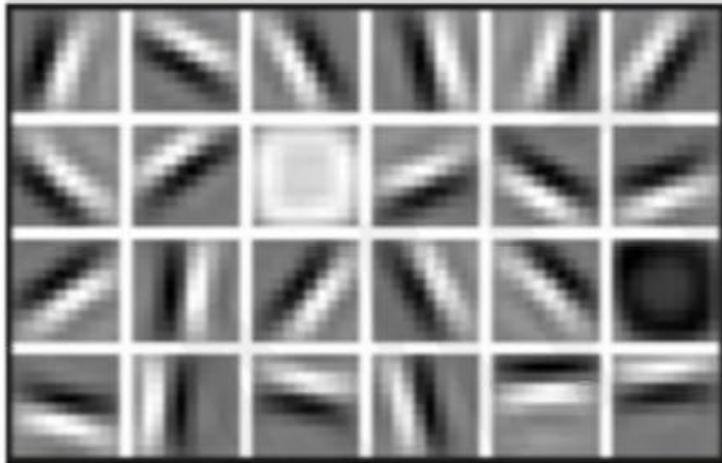
# Why Deep Learning and Why now?

# Why Neural Networks?

Hand engineered features are time consuming, brittle, and not scalable in practice

Can we learn the **underlying features** directly from data?

**Low Level Features**



Lines & Edges

**Mid Level Features**



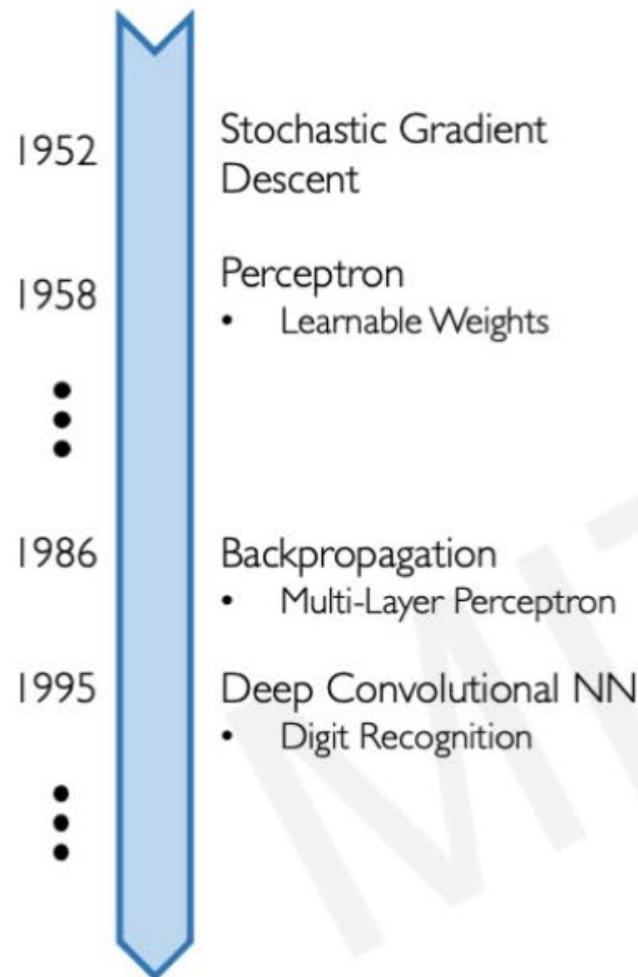
Eyes & Nose & Ears

**High Level Features**



Facial Structure

# Why Now?



Neural Networks date back decades, so why the resurgence?

## I. Big Data

- Larger Datasets
- Easier Collection & Storage

IM<sup>2</sup>GENET



WIKIPEDIA  
The Free Encyclopedia



## 2. Hardware

- Graphics Processing Units (GPUs)
- Massively Parallelizable



## 3. Software

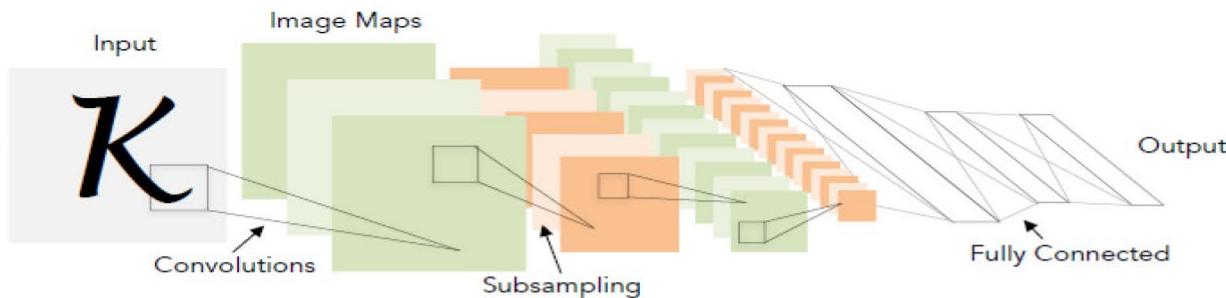
- Improved Techniques
- New Models
- Toolboxes



# Why now?

1998

LeCun et al.

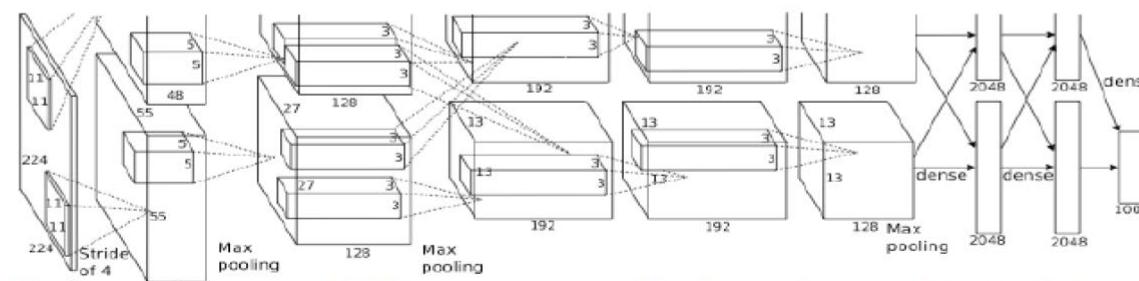


# of transistors  
pentium II  
 $10^6$

# of pixels used in training  
10<sup>7</sup> NIST

2012

Krizhevsky et al.

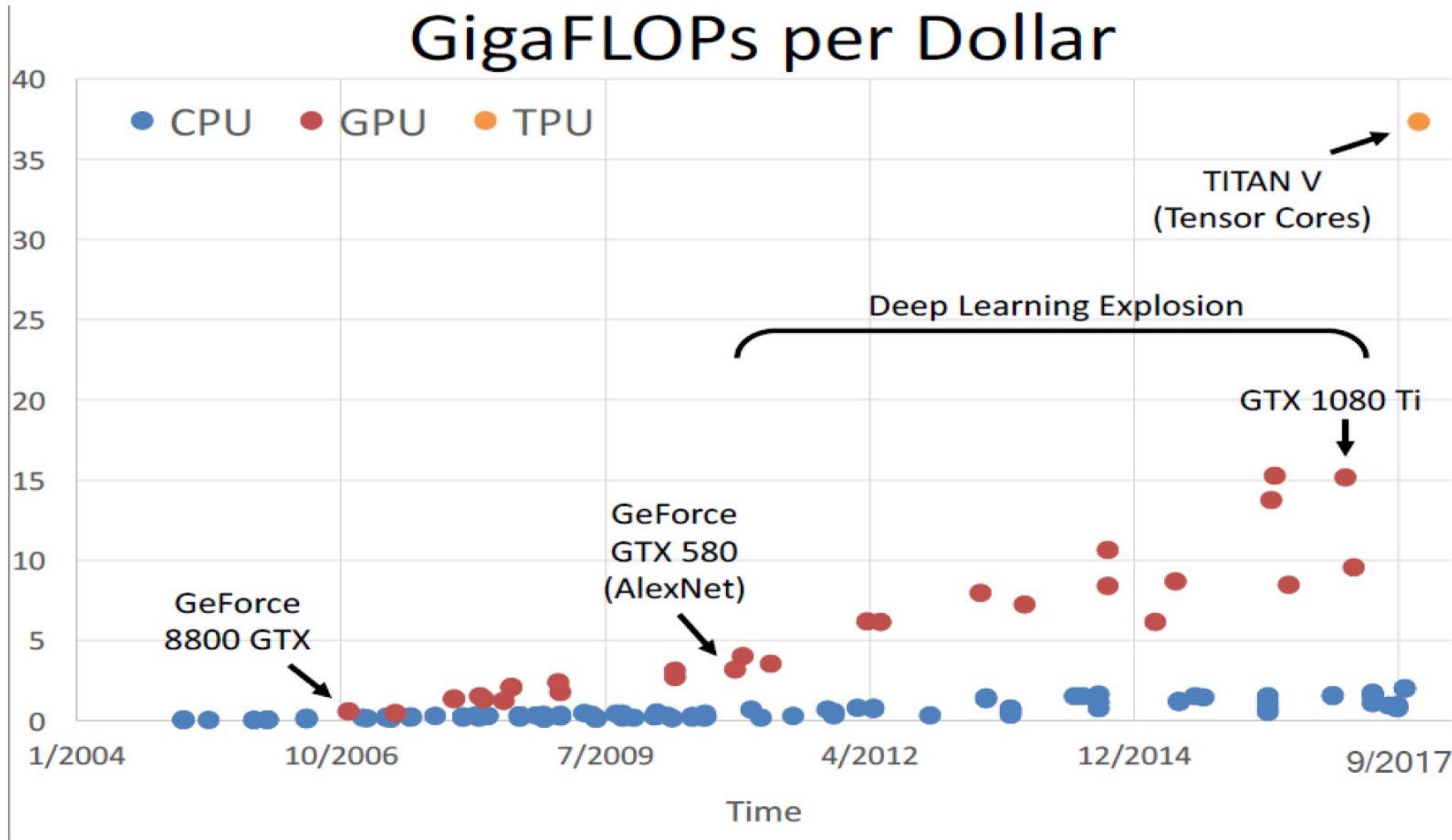


# of transistors  
GPUs  
 $10^9$

# of pixels used in training  
10<sup>14</sup> IMAGENET

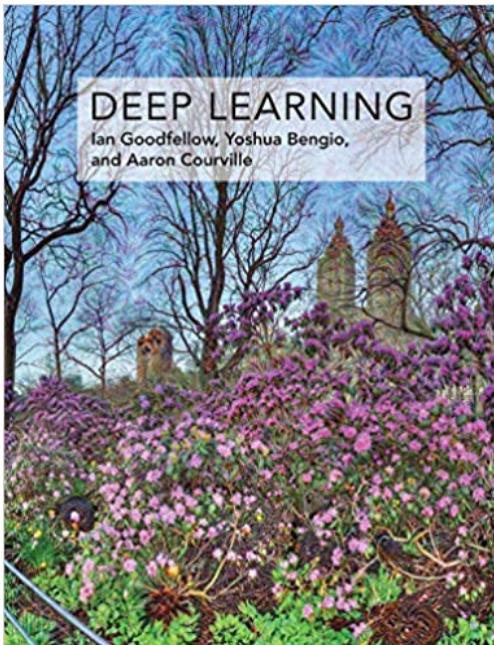
Figure copyright Alex Krizhevsky, Ilya Sutskever, and Geoffrey Hinton, 2012. Reproduced with permission.

# Why now?



# Text Books

- Deep learning, I. Goodfellow, j. Bengio MIT Press, 2016
- Neural Networks and deep learning, M. Nielsen Determination Press, 2015



# Syllabus

L1-Introduction

L2-Neuron math model

L3-Perceptron

L4-Building and Applying NN

L5-Gradient descent

L6-Vectorization

L7-Overfitting

L8-Regularization I

L9-Regularization II

L10-Optimization Algorithms I (mini-batches)

L11-Optimization Algorithms II(exponentially weighted averages)

L12-Hyperparameter Tuning

L13-Batch Normalization

L14-Softmax

L15-Convolutional Neural Networks

L16-Padding, Strided convolution

L17-Simple Convolutional Network

# Related Course

- Deep learning, Andrew Ng, Stanford University
- Intro to Deep Learning, MIT
- Neural Network for Machin Learning



# Course administrations

# Schedule

تاریخ	موضوع
یکشنبه, 14 بهمن 1403	L1-Introduction
سه شنبه, 16 بهمن 1403	L2-Neuron math model
یکشنبه, 21 بهمن 1403	L3-Perceptron
سه شنبه, 23 بهمن 1403	L4-Building and Applying NN
یکشنبه, 28 بهمن 1403	L5-Gradient descent
سه شنبه, 30 بهمن 1403	L6-Vectorization
یکشنبه, 5 اسفند 1403	L7-Overfitting
سه شنبه, 7 اسفند 1403	L8-Regularization I
یکشنبه, 12 اسفند 1403	L9-Regularization II
سه شنبه, 14 اسفند 1403	L10-Optimization Algorithms I (mini-batches)
یکشنبه, 19 اسفند 1403	L11-Optimization Algorithms II(exponentially weighted averages)
سه شنبه, 21 اسفند 1403	L12-Hyperparameter Tuning
یکشنبه, 26 اسفند 1403	L13-Batch Normalization
سه شنبه, 28 اسفند 1403	L14-Softmax
یکشنبه, 17 فروردین 1404	L15-Convolutional Neural Networks
سه شنبه, 19 فروردین 1404	L16-Padding, Strided convolution
یکشنبه, 24 فروردین 1404	L17-Simple Convolutional Network
سه شنبه, 26 فروردین 1404	L18- Genetic Algorithms
یکشنبه, 31 فروردین 1404	L19- Genetic Algorithms
سه شنبه, 2 اردیبهشت 1404	L1-Introduction
یکشنبه, 7 اردیبهشت 1404	L6-Particle Swarm Intelligence (PSO)
سه شنبه, 9 اردیبهشت 1404	L7-Particle Swarm Intelligence (PSO)
یکشنبه, 14 اردیبهشت 1404	L8-Discrete PSO
سه شنبه, 16 اردیبهشت 1404	L9-PSO Hyper-parameters tuning
یکشنبه, 21 اردیبهشت 1404	L10-PSO Applications
سه شنبه, 23 اردیبهشت 1404	L11-PSO Applications
یکشنبه, 28 اردیبهشت 1404	L17-Ant Colony Optimization (ACO)
سه شنبه, 30 اردیبهشت 1404	L18-Ant Colony Optimization (ACO)
یکشنبه, 4 خرداد 1404	L19-Ant Colony Optimization (ACO)
سه شنبه, 6 خرداد 1404	L20-ACO applications
یکشنبه, 11 خرداد 1404	L21-ACO applications
سه شنبه, 13 خرداد 1404	L22-ACO applications

# Course Staff



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# Assignments



# Calss Rules

- تمرین ها به صورت انفرادی انجام میشود.
- زمان تحويل هر تمرین ۲ هفته بعد از پست آن تمرین است
- زمان تحويل تمرین ها مطابق با فایل زمانبندی انجام میشود و قابل تغییر نیست
- تمرین ها با ۸ روز تاخیر در کل تمرین ها قابل تحويل هستند
- در صورتی که ۸ روز شما به اتمام رسید، تمرینهای بعدی را میتوانید هر کدام را تا ۳ روز به ازای هر روز کسر ۱۰ درصد تحويل دهید.
- بعد از ۸ روز جواب تمرین توسط تی ای ها در سامانه بازگزاری میشود.
- بعد از پست پاسخ تمرین ها تحويل تمرین مجاز نیست.

# Calss Rules

- نمره امتیازی برای هر تمرین عملی به صورتی انتخاب شود که رقابت بین گروه‌ها ایجاد شود.
- نمره امتیازی برای کل تکالیف بین ۰.۵ تا ۱ نمره خواهد بود
- تکالیف عملی توسط سیستم‌های چک تشابه بررسی می‌شوند و در صورت مشابهت بالا به هر دو نفر نمره ۰ تعلق می‌گیرد.
- تکالیف تئوری در صورت مشاهده تشابه بالا برای هر دو نفر نمره ۰ در نظر گرفته می‌شود.

# Marking Scheme

- Assignments and Project: 3-5 marks
- Midterm: 5-7 marks
- Final: 5-7 marks
- Clicker Questions (probely): 1 mark
- Attendance in class: 1-2 mark