

# ECBM E4040

## Neural Networks and Deep Learning

### Introduction to Deep Learning (DL)

**Zoran Kostić**

Columbia University

Electrical Engineering Department &

Data Sciences Institute



**COLUMBIA ENGINEERING**  
The Fu Foundation School of Engineering and Applied Science



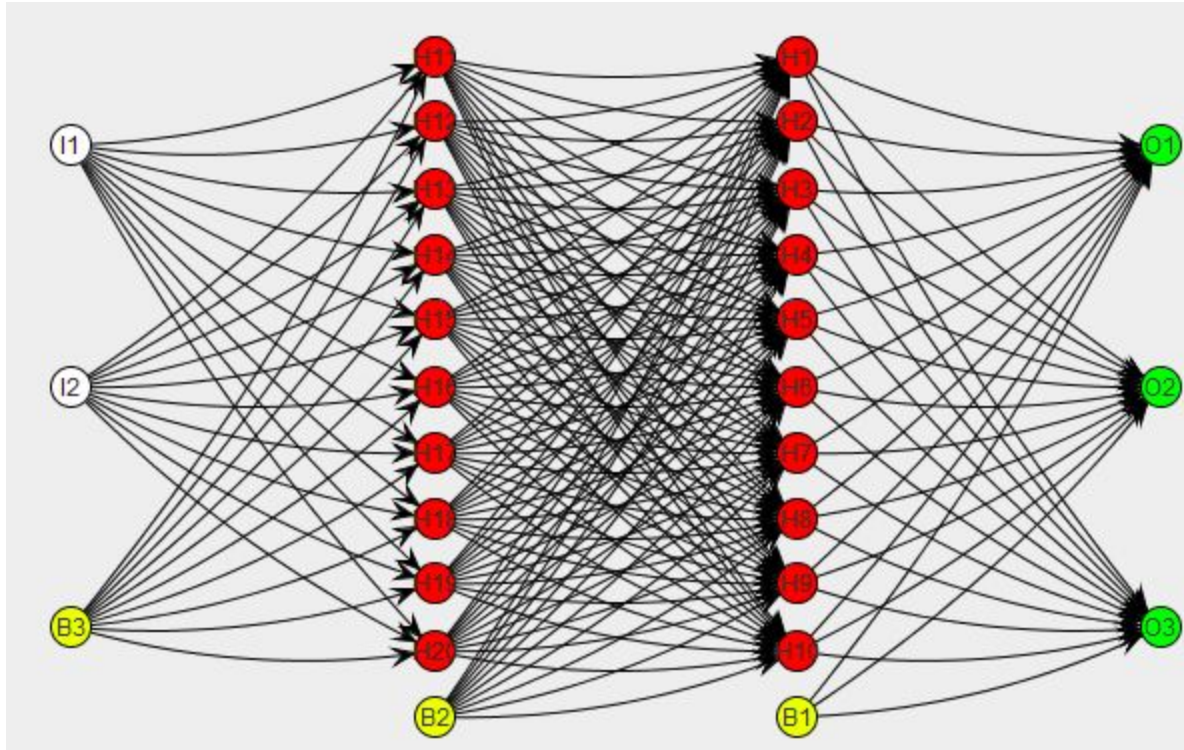
# References and Acknowledgments

- Lecture material by bionet group / Prof. Aurel Lazar (<http://www.bionet.ee.columbia.edu/>)

# Outline

- Artificial Neural Networks (ANNs) for Deep Learning
- What can deep learning do
- Biological Neural Networks
- What is Deep Learning / Artificial Neural Networks
- Historical Trends in DL

# Artificial Neural Networks (ANN)



source <https://www.codeproject.com/Articles/477689/JavaScript-Machine-Learning-and-Neural-Networks-wi>

# Artificial Neural Network Structure

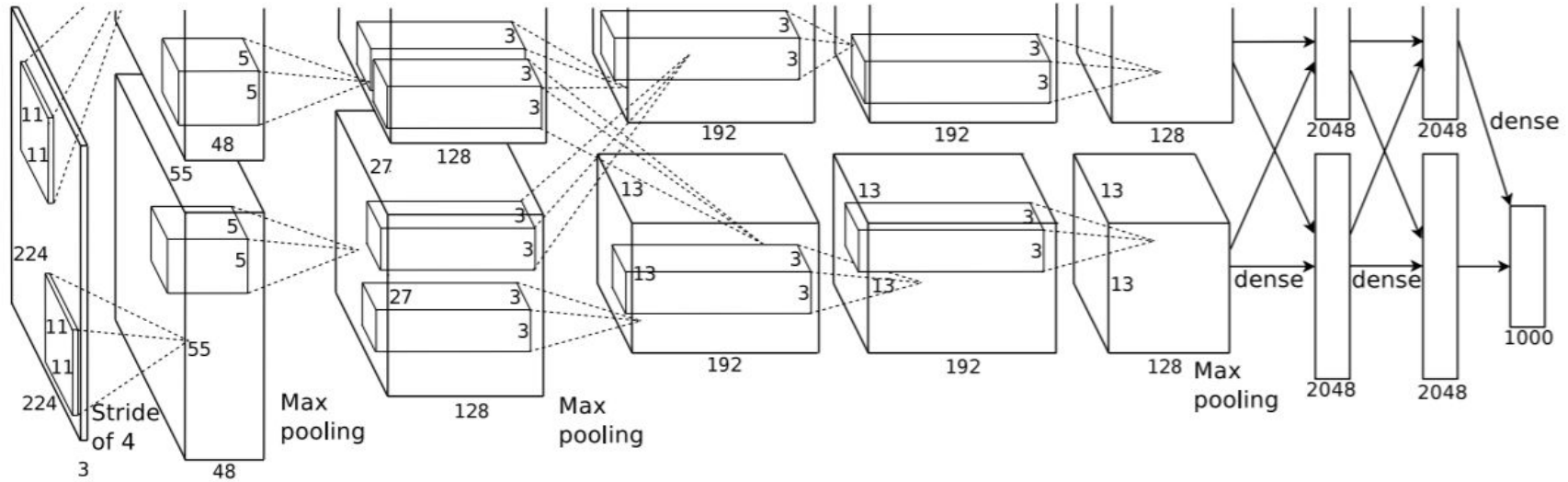


Figure 2: An illustration of the architecture of our CNN, explicitly showing the delineation of responsibilities between the two GPUs. One GPU runs the layer-parts at the top of the figure while the other runs the layer-parts at the bottom. The GPUs communicate only at certain layers. The network's input is 150,528-dimensional, and the number of neurons in the network's remaining layers is given by 253,440–186,624–64,896–64,896–43,264–4096–4096–1000.

# What do Neural Networks Do ?

In a nutshell:

- ANN can learn to **represent any function**

# Artificial Neural Networks for Deep Learning

Deep learning uses architectures of  
**many-layer**  
artificial neural networks

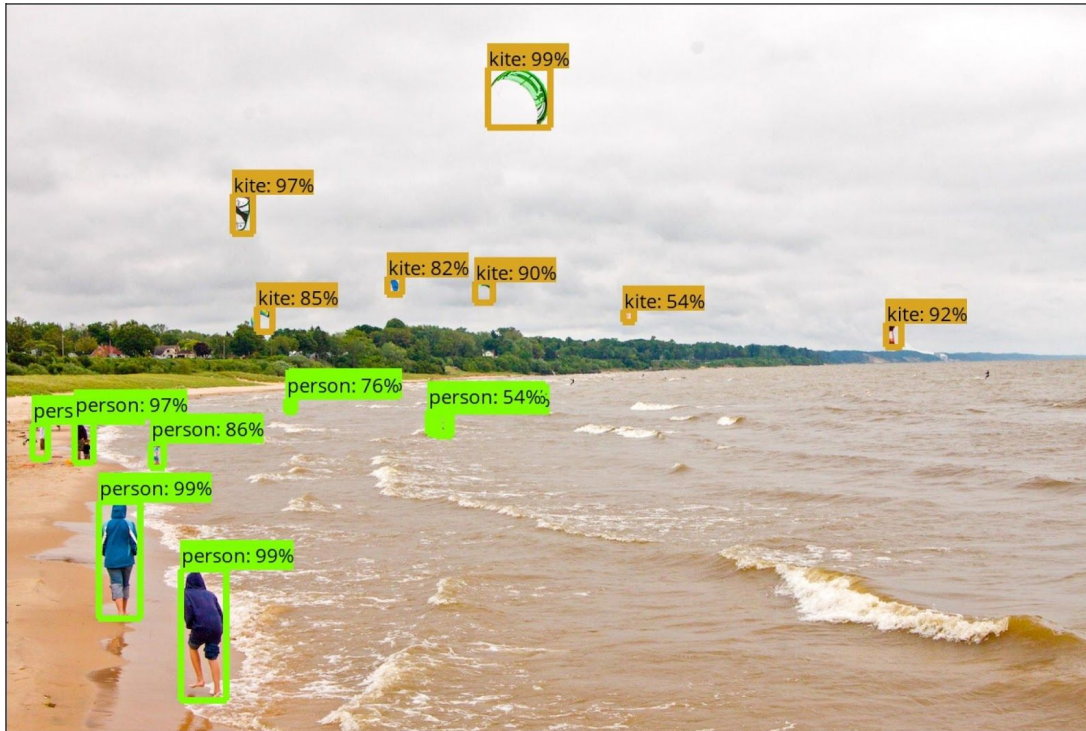
# Artificial Neural Networks

WHAT CAN DEEP LEARNING DO?



# What can Deep Learning Do - Examples

## Detect Objects in a Sample Image



from  
Google research blog

<https://research.googleblog.com/2017/06/supercharge-your-computer-vision-models.html>

Original image: Mike Miley

[https://www.flickr.com/photos/mike\\_miley/](https://www.flickr.com/photos/mike_miley/)

# What can Deep Learning Do - Browser Demos

Classify MNIST digits with a Convolutional Neural Network



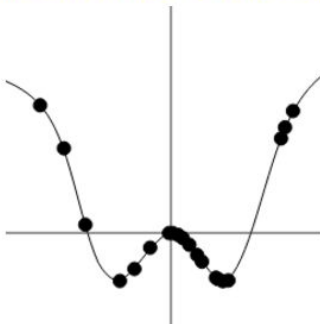
Classify CIFAR-10 with Convolutional Neural Network



Interactively classify toy 2-D data with a Neural Network



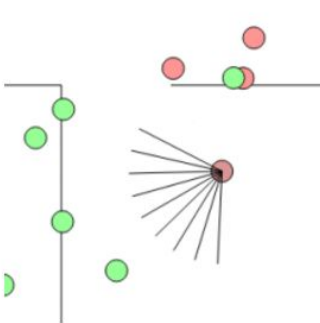
Interactively regress toy 1-D data



Train an MNIST digits Autoencoder



Reinforcement Learning with Deep Q Learning



<http://cs.stanford.edu/people/karpathy/convnetjs/>

# What can Deep Learning Do - Examples

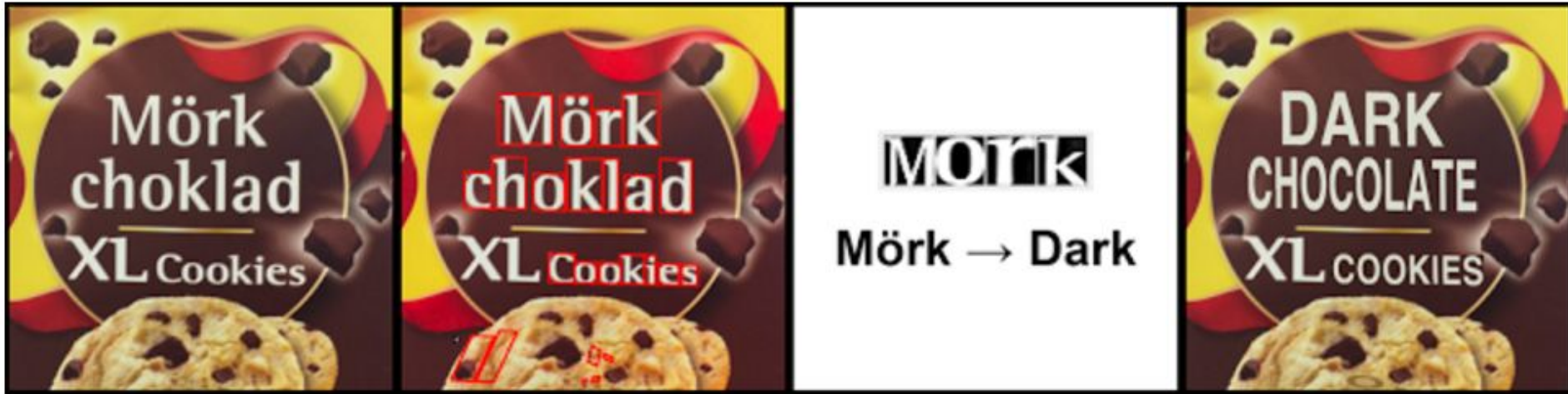
## Handwriting

- Take the breath away where they are
- He dismissed the idea
- prison welfare Officer complement
- She looked closely as she
- at Hinterscombe is being adapted for

<http://www.cs.toronto.edu/~graves/handwriting.html>

# What can Deep Learning Do - Examples

## Automatic Machine Translation



<https://research.googleblog.com/2015/07/how-google-translate-squeezes-deep.html>

# What can Deep Learning Do - Examples

## Project COSMOS: Cloud Connected Vehicles



Prof. Kostic lab

<https://cosmos-lab.org/>

Videos:

<https://youtu.be/SvXrQPRBkHE>

<https://www.dropbox.com/s/p69pg079a48ylv6/radarMap.190619.slowCompare.mp4?dl=0>

# What can Deep Learning Do

## ImageNet - [www.image-net.org](http://www.image-net.org)

Image database organized according to the WordNet hierarchy (the nouns), in which each node of the hierarchy is depicted by hundreds and thousands of images.

**22K categories and 14M images**

**Animals: Bird, Fish, Mammal, Invertebrate**

**Plants: Tree, Flower**

**Food**

**Materials**

**Artifact: Tools, Appliances, Structures**

**Person**

**Scenes: Indoor, Geological, Formations**

**Sport Activities**



# What can Deep Learning Do

## ImageNet - Large Scale Visualization Challenge

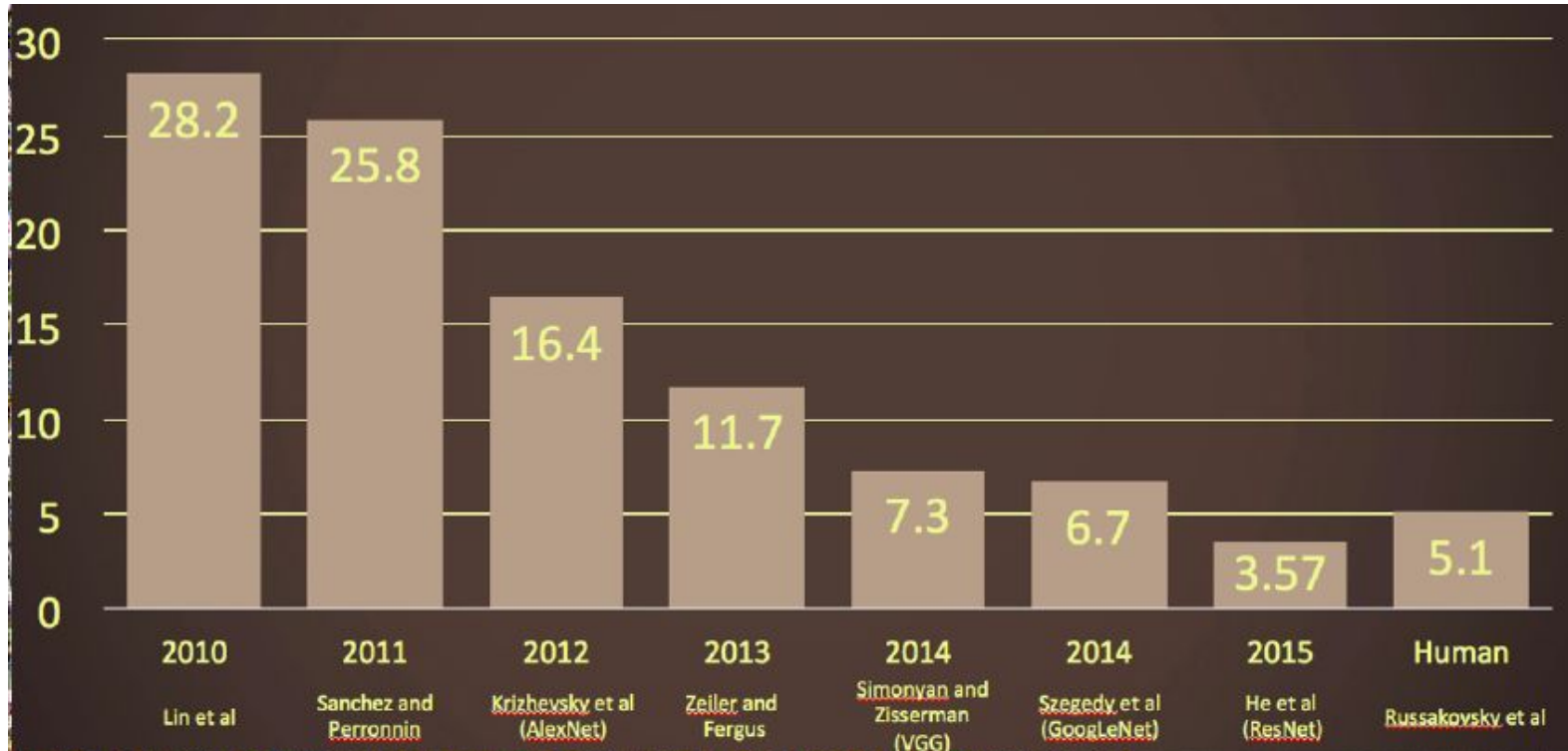
Evaluates algorithms for object localization/detection from images/videos and scene classification/parsing at scale.

- Object localization for 1000 categories.
- Object detection for 200 fully labeled categories.
- Object detection from video for 30 fully labeled categories.
- Scene classification for 365 scene categories (Joint with MIT Places team) on Places2 Database <http://places2.csail.mit.edu>.
- Scene parsing

<http://image-net.org/challenges/LSVRC/2016/index>

# What can Deep Learning Do

## ImageNet - Image Classification



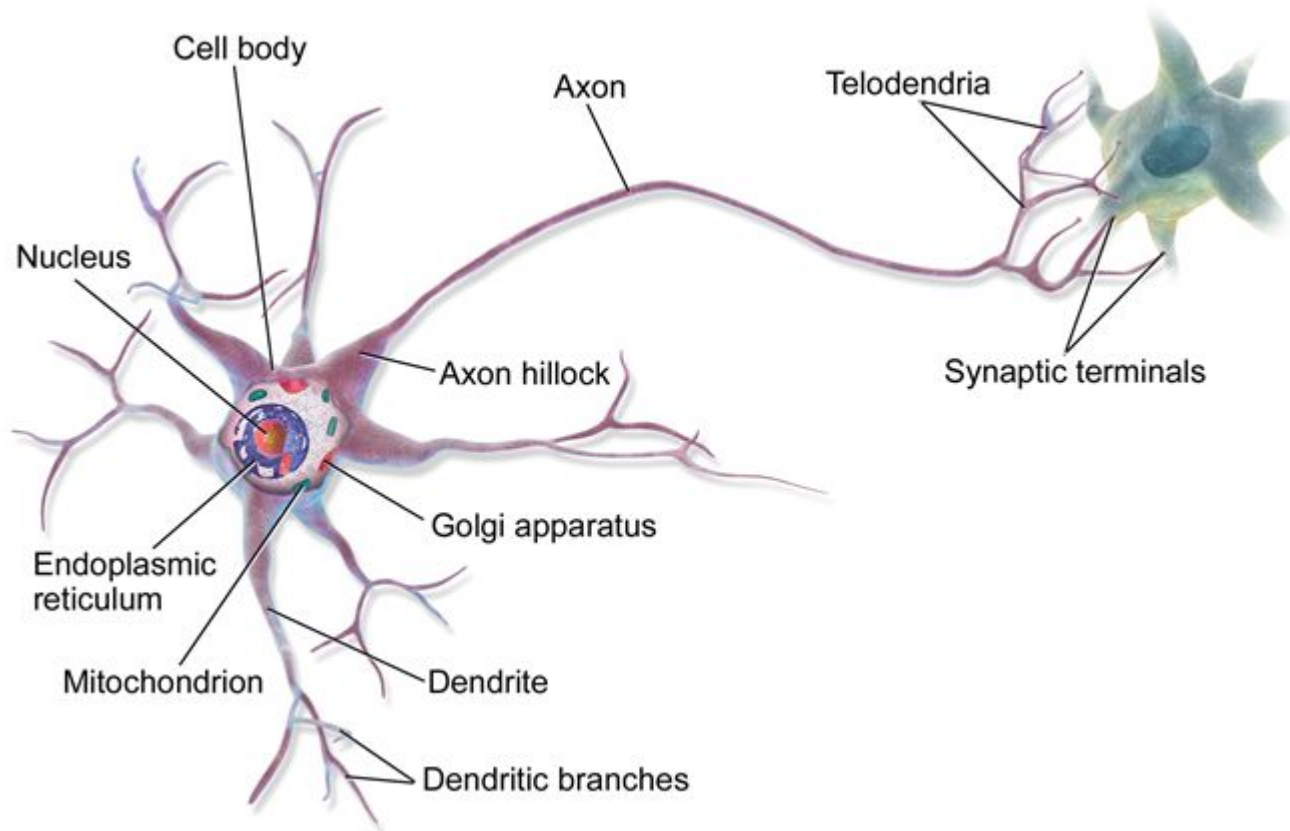
Russakovsky et al. arXiv, 2014



# Biological Neural Networks

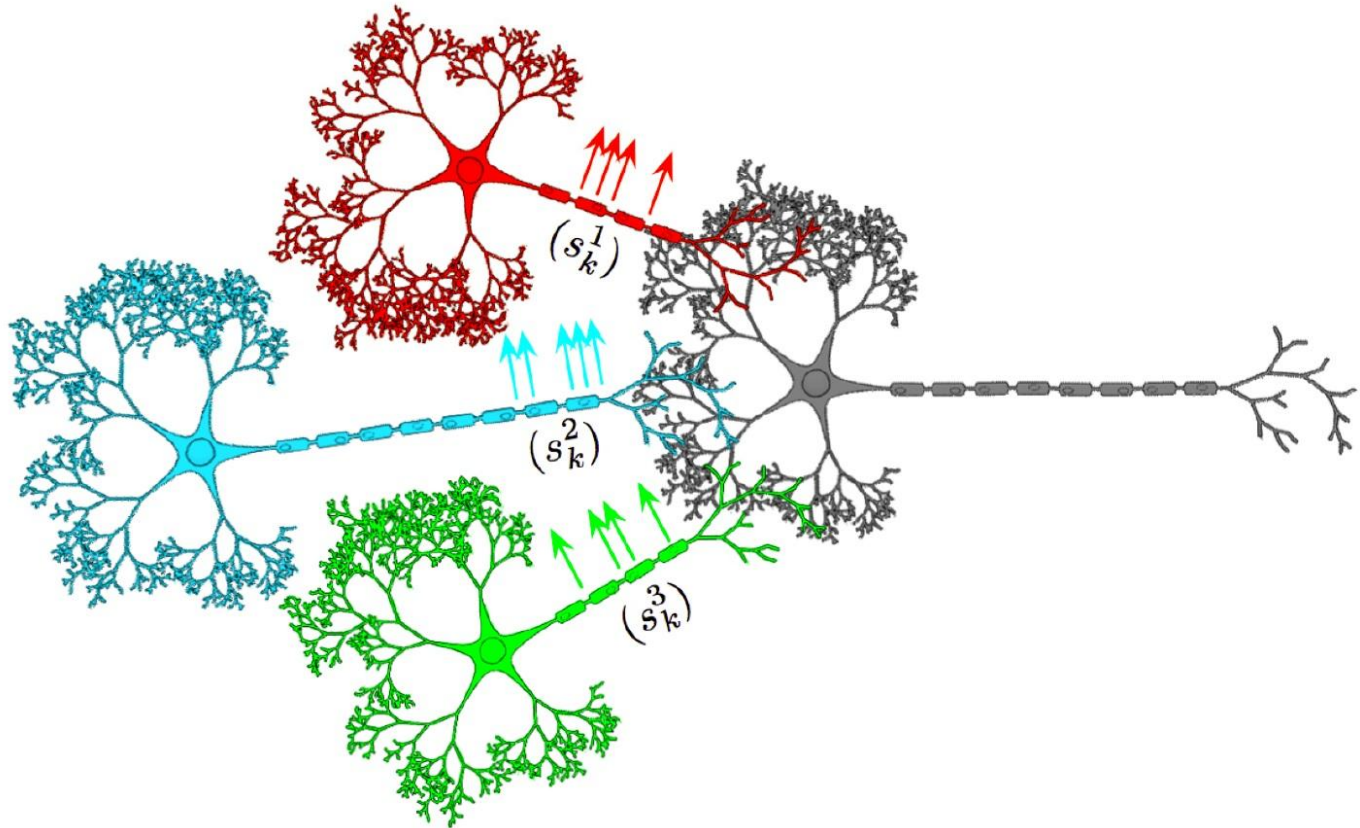
relationship to ANNs

# Biological Neural Networks - Neurons



A **neuron** is an electrically excitable cell that processes and transmits information through **electrical** and **chemical** signals.

# Biological Neural Networks - Inputs via Dendrites



# Biological Neural Networks - Neurons

**Dendrites receive electrochemical impulses from other neurons, and carry them inwards and towards the soma, while axons carry the impulses away from the soma.**

**Dendrites are short and heavily branched in appearance, while axons are much longer.**

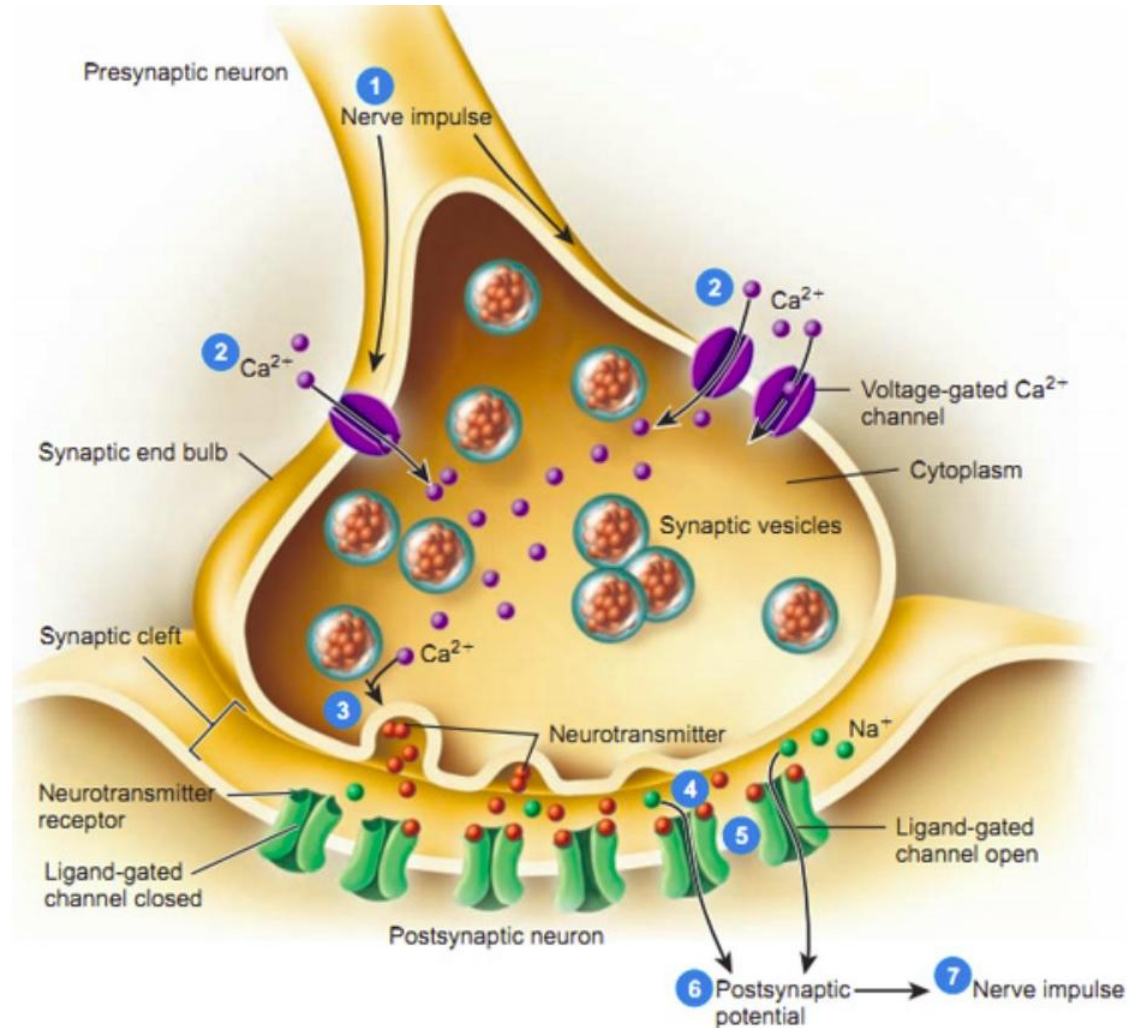
**Generally, dendrites receive neuron signals, and axons transmit them.**

**Most neurons have a lot of dendrites and only have one axon.**

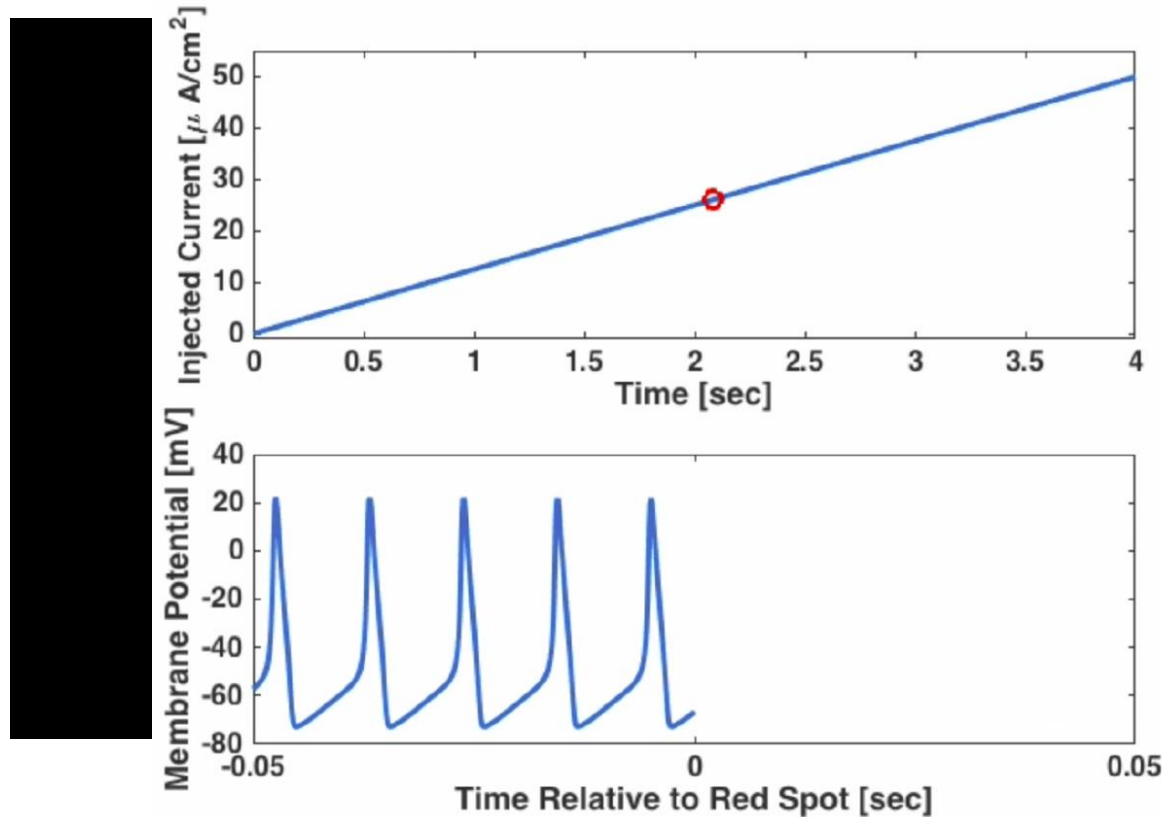
**Dendrites' radius tapers, while axons' remain constant.**

# Biological Neural Networks

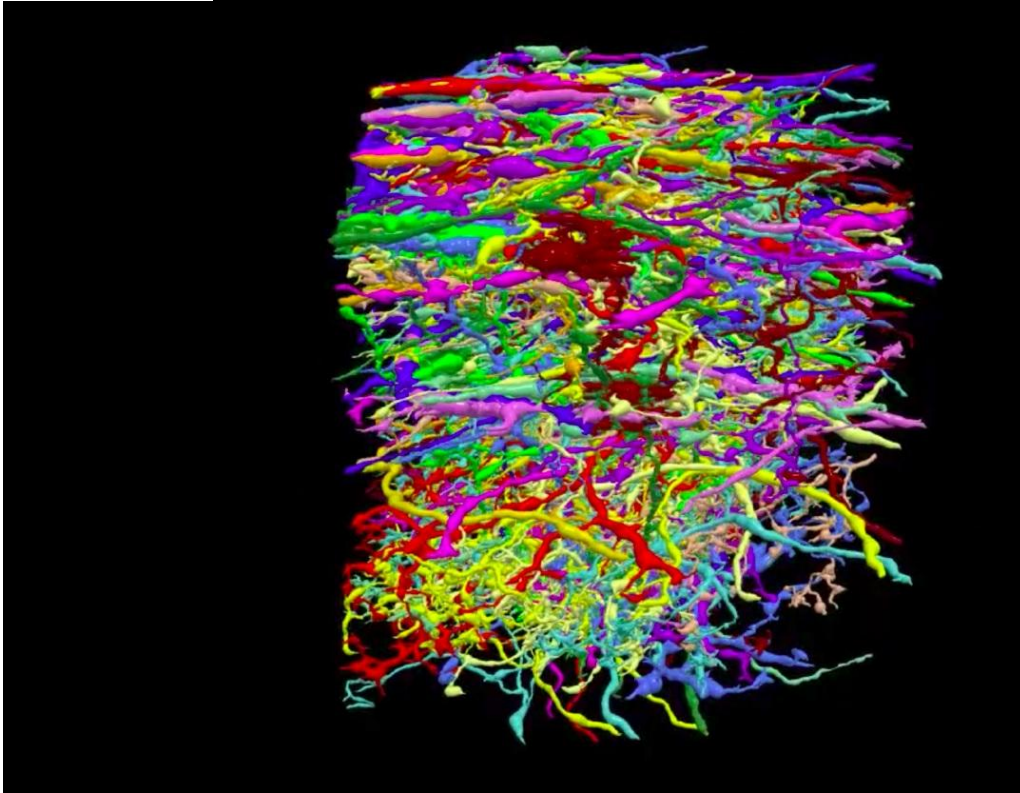
Signal transmission between neurons are through synapses.



# Biological Neural Networks - Neurons Generate Action Potentials when Excited



# Biological Neural Networks - An Example of a Reconstructed Neural Network



# What is Deep Learning

and artificial neural networks

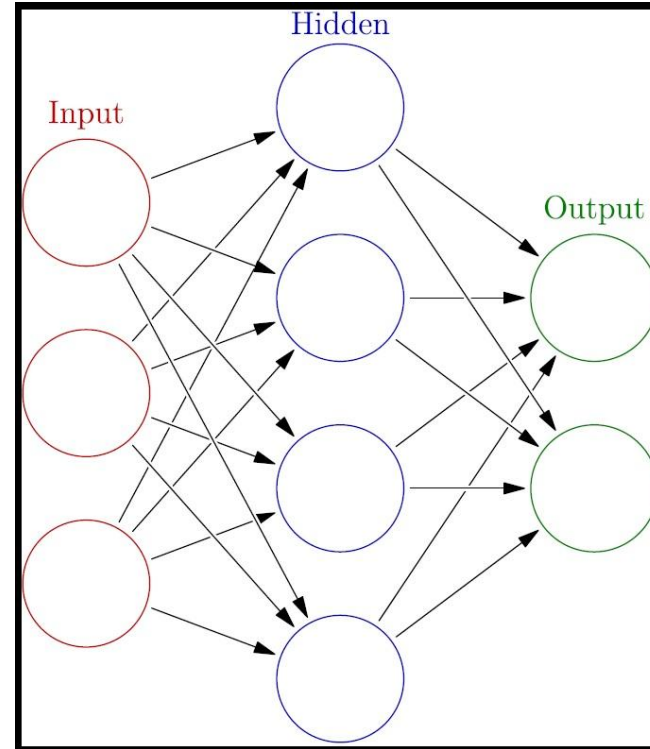


# What is Deep Learning? Starting with Standard Artificial Neural Networks (ANNs)

A standard ANN consists of

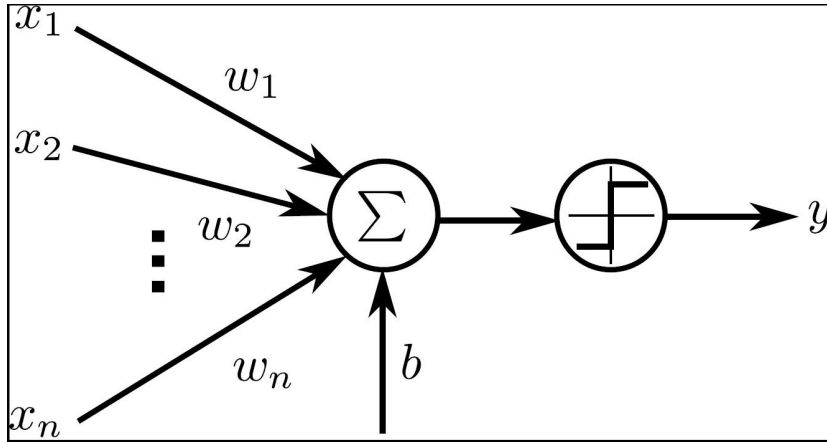
- many **simple, connected ``neurons''**
- each neuron has an **activation function**
- input neurons get activated through **sensors** perceiving the environment
- other neurons are activated through **weighted connections** from previously activated neurons

Figure: Glosser.ca Wikipedia



# What is Deep Learning?

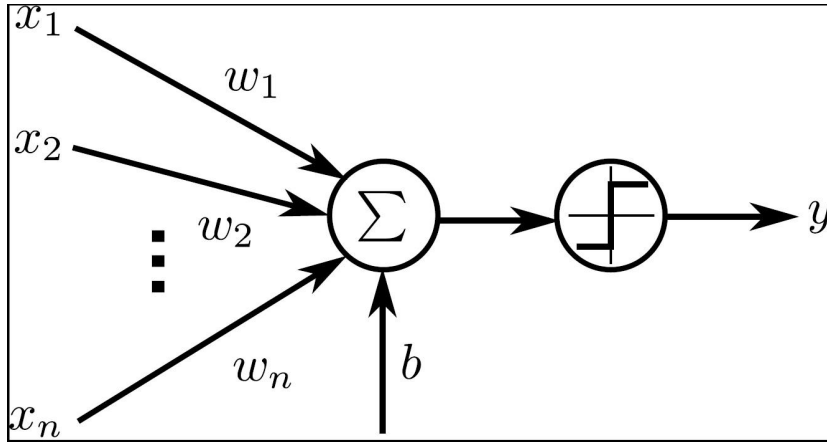
## Perceptron



$$y(\mathbf{x}) = \begin{cases} 1, & \text{if } \mathbf{w}^T \mathbf{x} + b > 0 \\ -1, & \text{otherwise} \end{cases}$$

# What is Deep Learning?

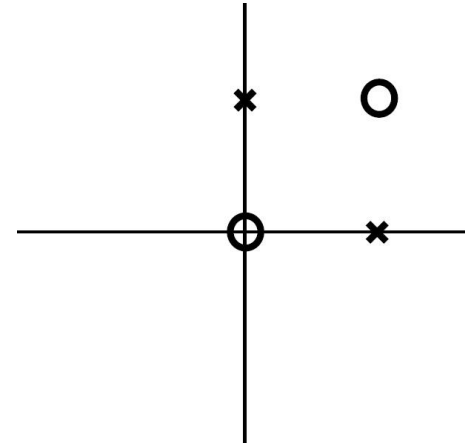
## Perceptron



$$y(\mathbf{x}) = \begin{cases} 1, & \text{if } \mathbf{w}^T \mathbf{x} + b > 0 \\ -1, & \text{otherwise} \end{cases}$$

Linear classifier

Does not work on the following case (two classes are star and circle):



# Deep Neural Networks

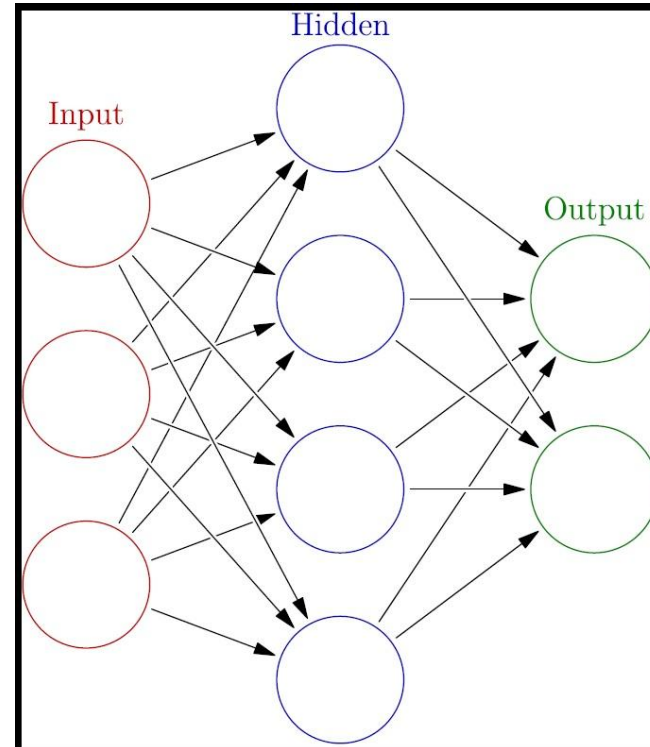
## Multilayer Perceptron (MLP)

### Universal approximation theorem:

A feed-forward network with a single hidden layer containing a finite number of neurons can approximate continuous functions on compact subsets of  $n$ .

- represents a simple/parallel structure that can be easily executed for prediction
- learning is much more time involved

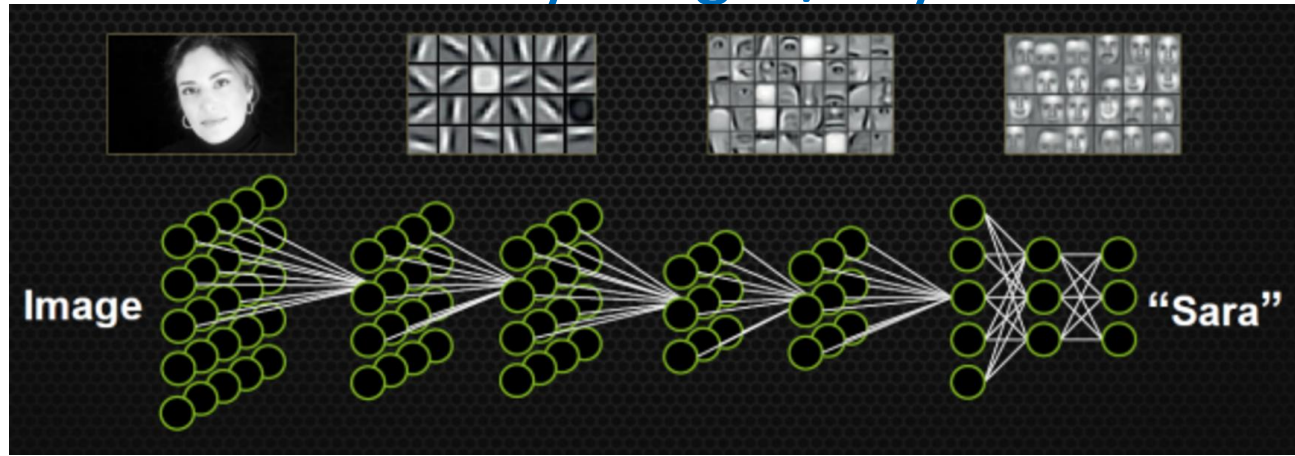
How to train a deep neural network becomes the central question.



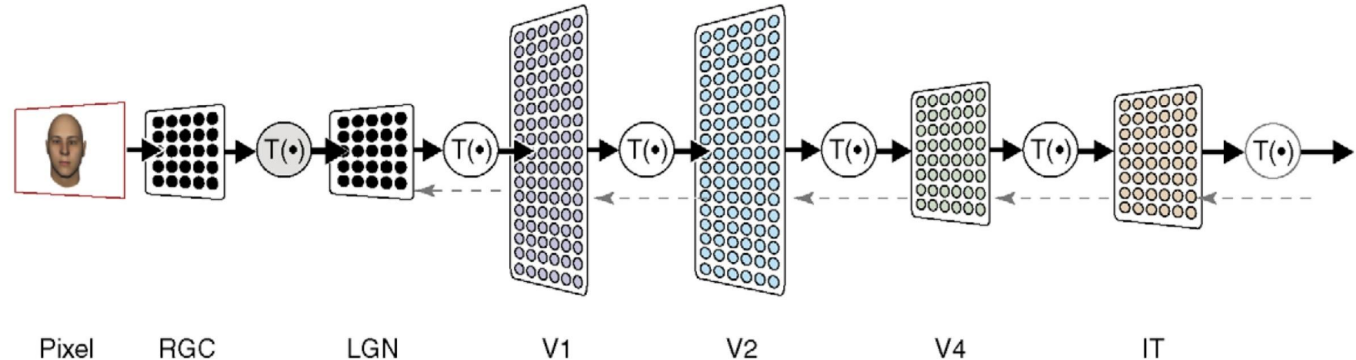
# Deep Neural Networks

## Neural Network with Many Stages/Layers

Deep Neural Network

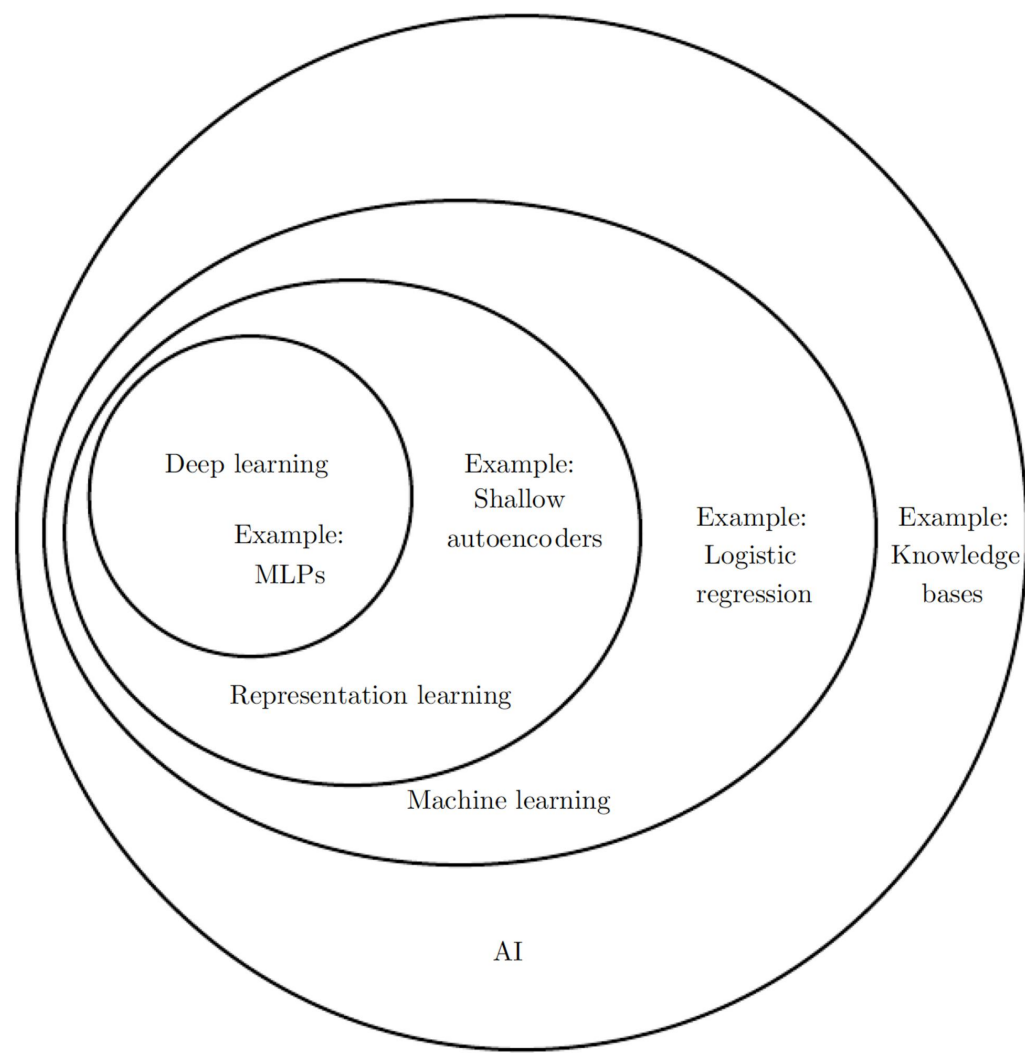


Ventral Visual Processing System



# Disciplines of Artificial Intelligence Venn Diagram

Deep Learning  
is only a subset of  
AI



# Features in Artificial Intelligence

## How to Identify Them?

Conventional machine-learning techniques are limited in their ability to process natural data in their raw form. Therefore, **features need to be extracted.**

### Feature identification:

- Requires careful engineering and considerable domain expertise to design a feature extractor which could **transform raw data into a suitable internal representation** or feature vector from which the learning subsystem (e.g. a classifier) can detect or classify patterns in the input

# Features in Artificial Intelligence

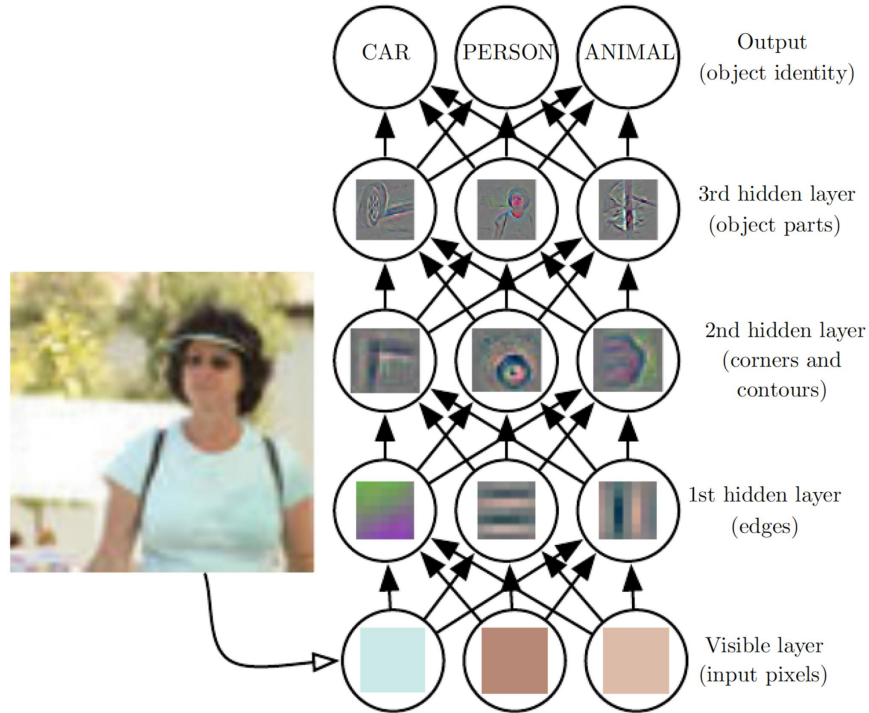
## How to Identify Them?

Obtaining a representation may be as difficult as solving the original problem.

**Representation learning** is a set of methods that allows a machine to be fed with raw data and to automatically discover the **representation** needed for detection or classification.



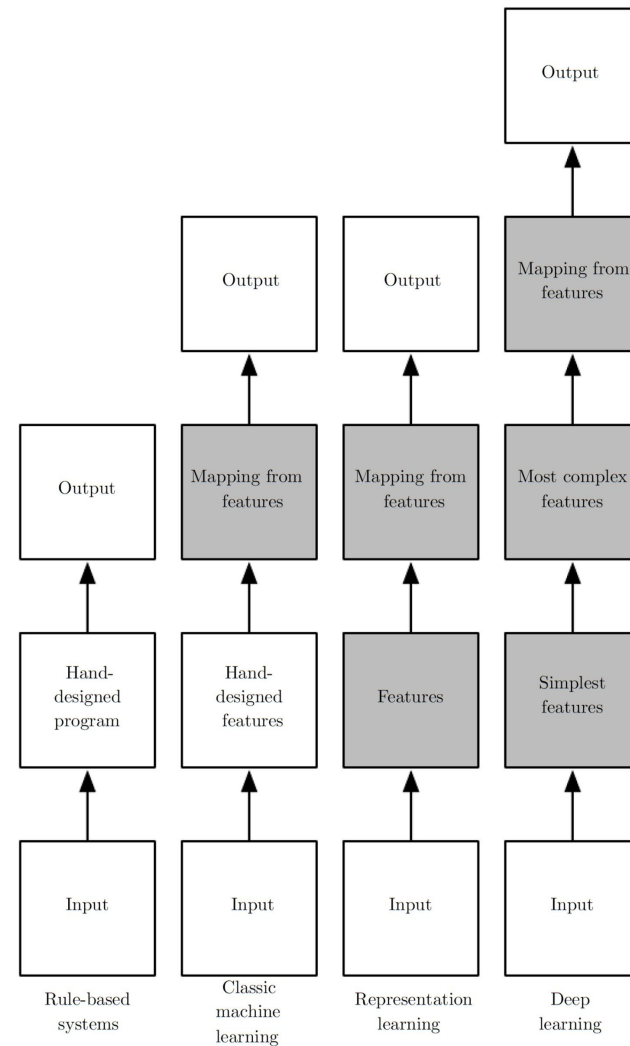
# Features and Representation Learning Implemented by Deep Learning



Deep-learning methods are **representation-learning methods** with multiple levels of representation, obtained by composing **simple, but non-linear modules** that each transform the representation at one level into a representation at a higher, slightly more abstract level.

# Deep Learning

## Comparison of Machine Learning Approaches



# Deep Learning History

McCulloch-Pitts Neuron (1943), Perceptron (1960s)

Back-propagation algorithm for training deep networks (1980s)

Deeper networks than before can be trained and have record-breaking results (2006+), emphasizing the importance of depth.

# Artificial Neural Networks (ANNs) and The Brain

**ANNs are influenced by the structure of the mammalian visual system.**

**Rectified linear activation function.**

**One deep learning algorithm solves many different tasks.**

**But ANNs/deep learning is not the same as the brain:**

- **It is not yet clear how**
  - **memory works in the brain**
  - **how values/functions are represented**
  - **what algorithms biological neural systems employ for learning.**

# Deep Learning Adoption

Google

Microsoft

Facebook

IBM

NVIDIA

Netflix

Baidu

Car manufacturers

(Tesla, BMW, Volvo,  
Ford self-driving cars)

and many more

# Deep Learning

## Impact - Breaking Records

- MNIST (Hand-Written Digits)
- Traffic Sign Contest
- ImageNet (Visual Object Detection and Image Classification)

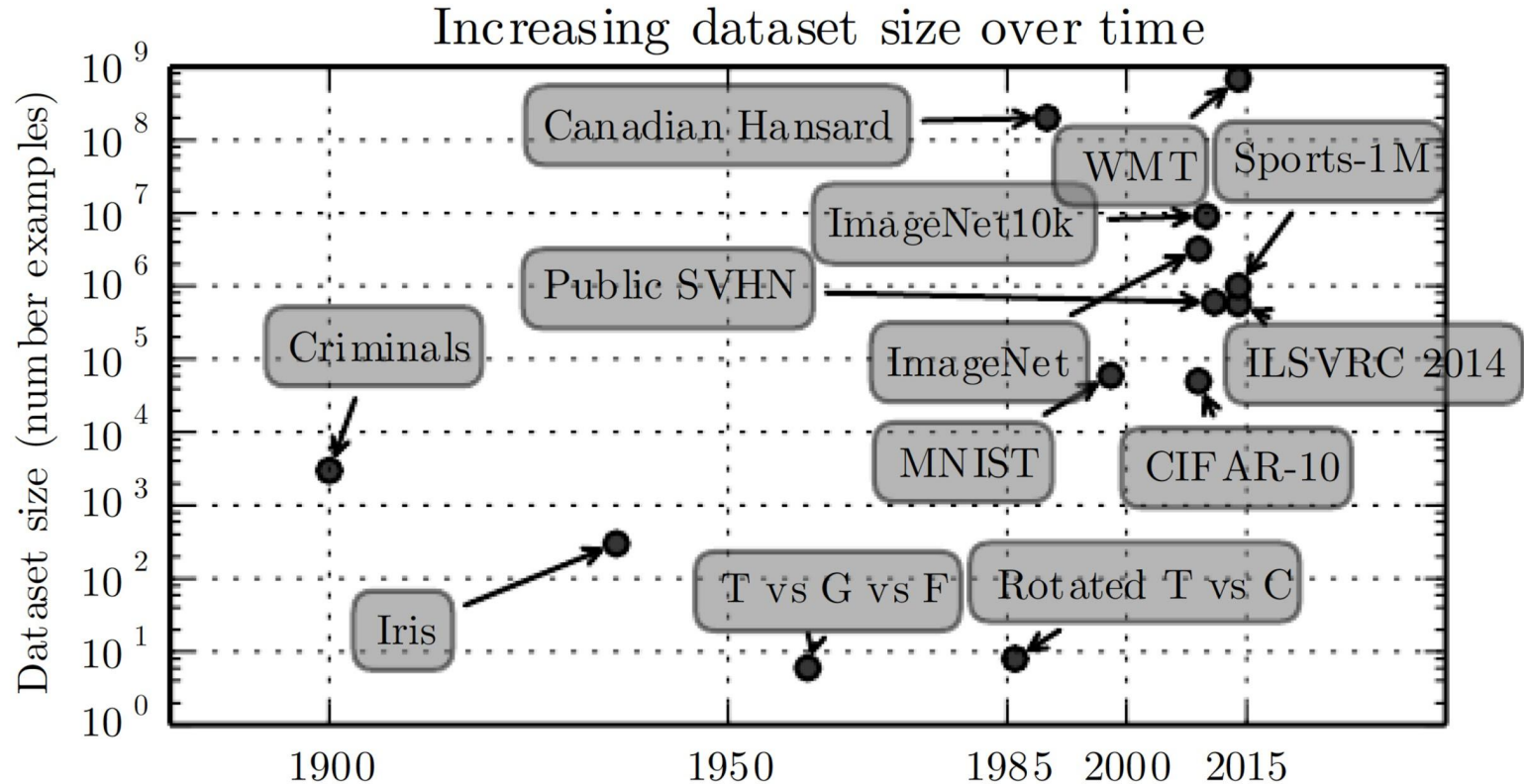
# Deep Learning

## Recent Upsurgence

Why did deep learning become popular again in recent years?

# Deep Learning

## Recent Upsurgence <- Data Abundance





# Deep Learning

## Recent Upsurgence <- Computing Revolution

Graphics processing unit (**GPU**)

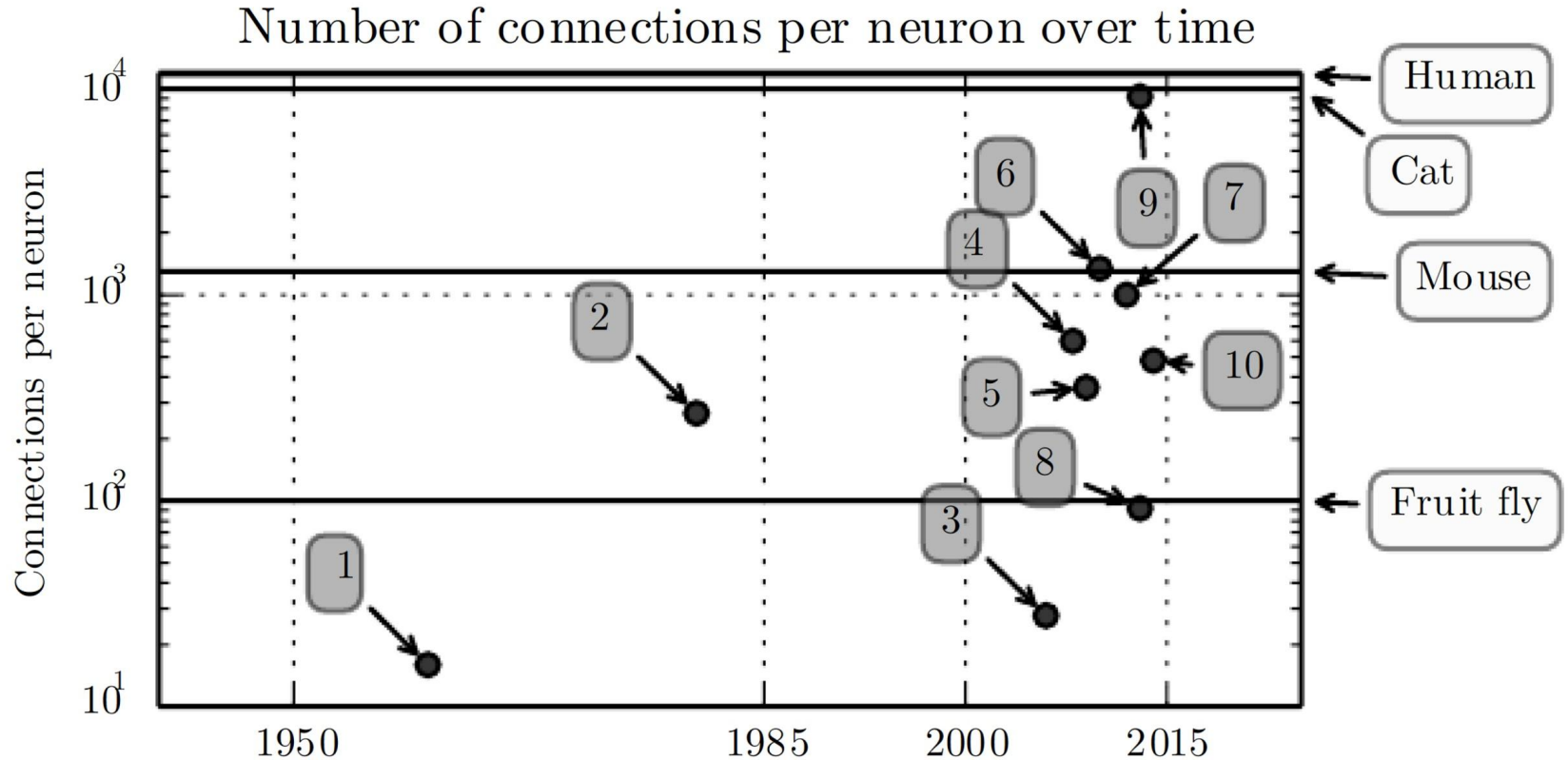
The **Role of GPUs** on Popularizing Deep Learning Algorithms

2010: a new MNIST record of 0.35% error rate. Made possible mainly through a GPU implementation of the **Back Propagation algorithm** that was up to **50 times faster** than standard CPU versions.

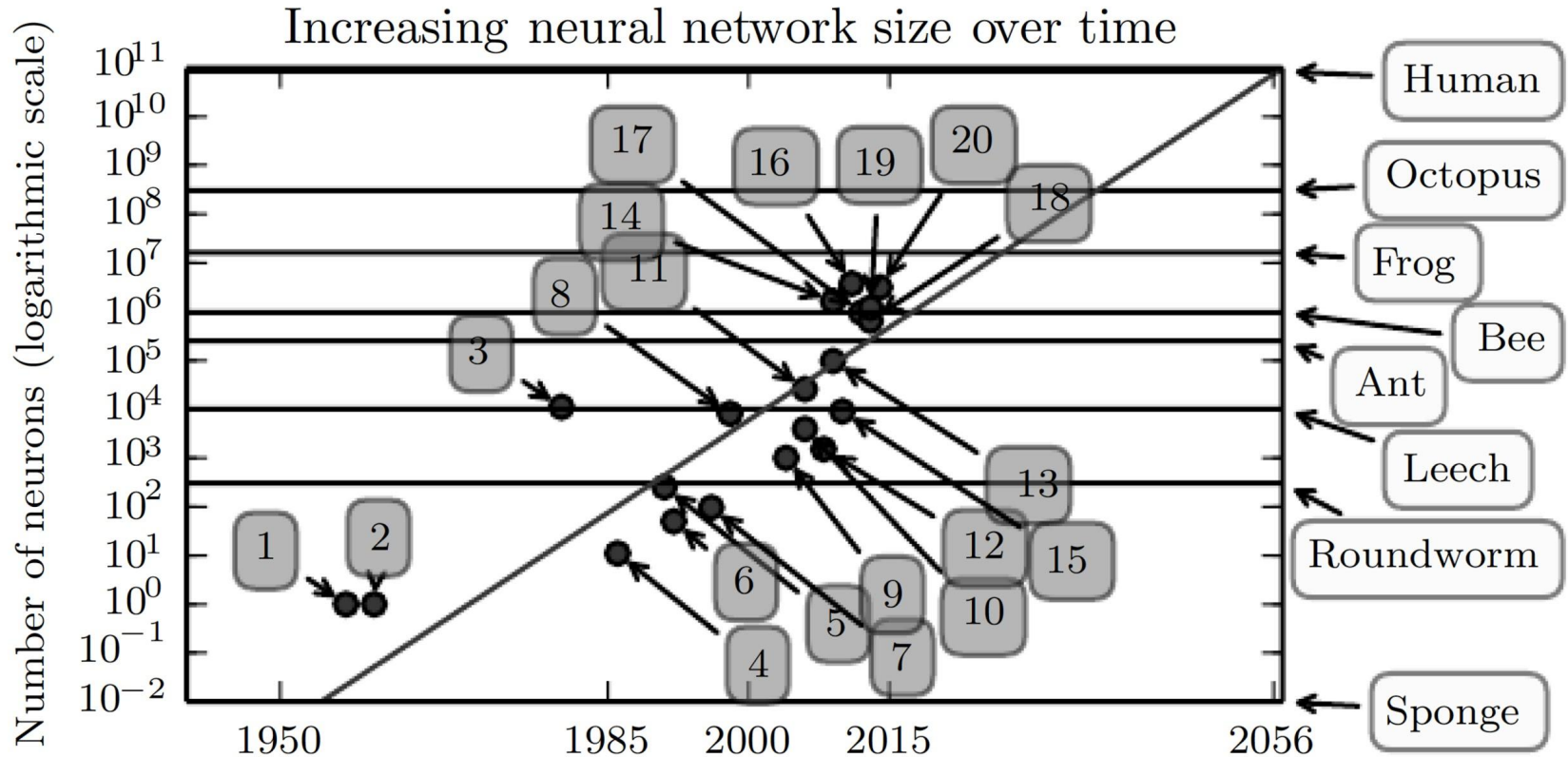
GPU based deep learning algorithms have won several pattern recognition competitions in the following years.

Recent ImageNet records from Microsoft and Google are all GPU based.

# Deep Learning: Larger and Faster Computers Support Increased ANN Model Size



# Deep Learning: Larger and Faster Computers Support Increased ANN Model Size



# Deep Learning Enablers

- Data abundance
- GPUs

# Deep Learning Applications in Science

Processing of massive amounts of data:

- Reconstructing neuron morphology from electron-microscopy raw data
- Analyzing particle accelerator data
- Predicting the activity of potential drug molecules
- Genetic research

# Backup Slides

**Various**