

# INTRODUCTION TO DEEP LEARNING

COMP4660/8420

Adapted from a lecture by Christopher Chow and Josephine Plested

Presented by Nathan Elazar

# WHAT CAN'T NEURAL NETWORKS DO?

# COMPLICATED DATA



Images



Text



Audio

- Inputs have very high dimension.
- Individual features are not informative.
- Traditional machine learning models will struggle to learn.

# OVERVIEW

What is deep learning?

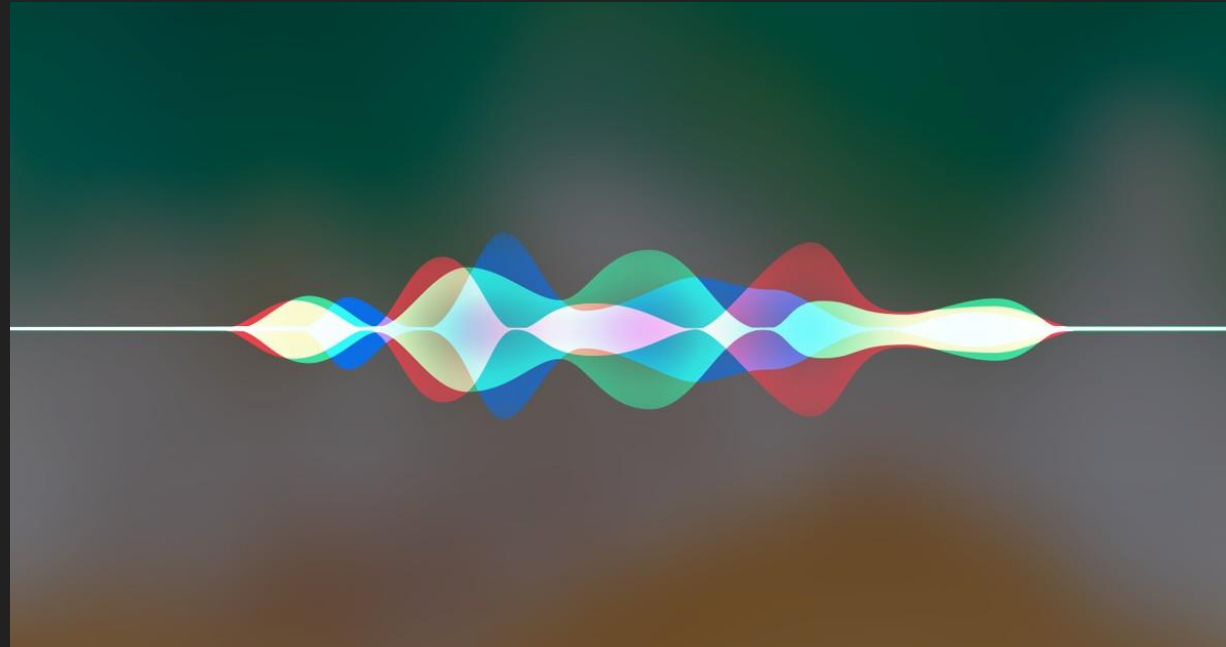
Why is it successful?

Where is it applied?

Who uses it and how?

# WHAT IS DEEP LEARNING?

# EXAMPLE: IMPROVEMENTS TO SIRI'S VOICE



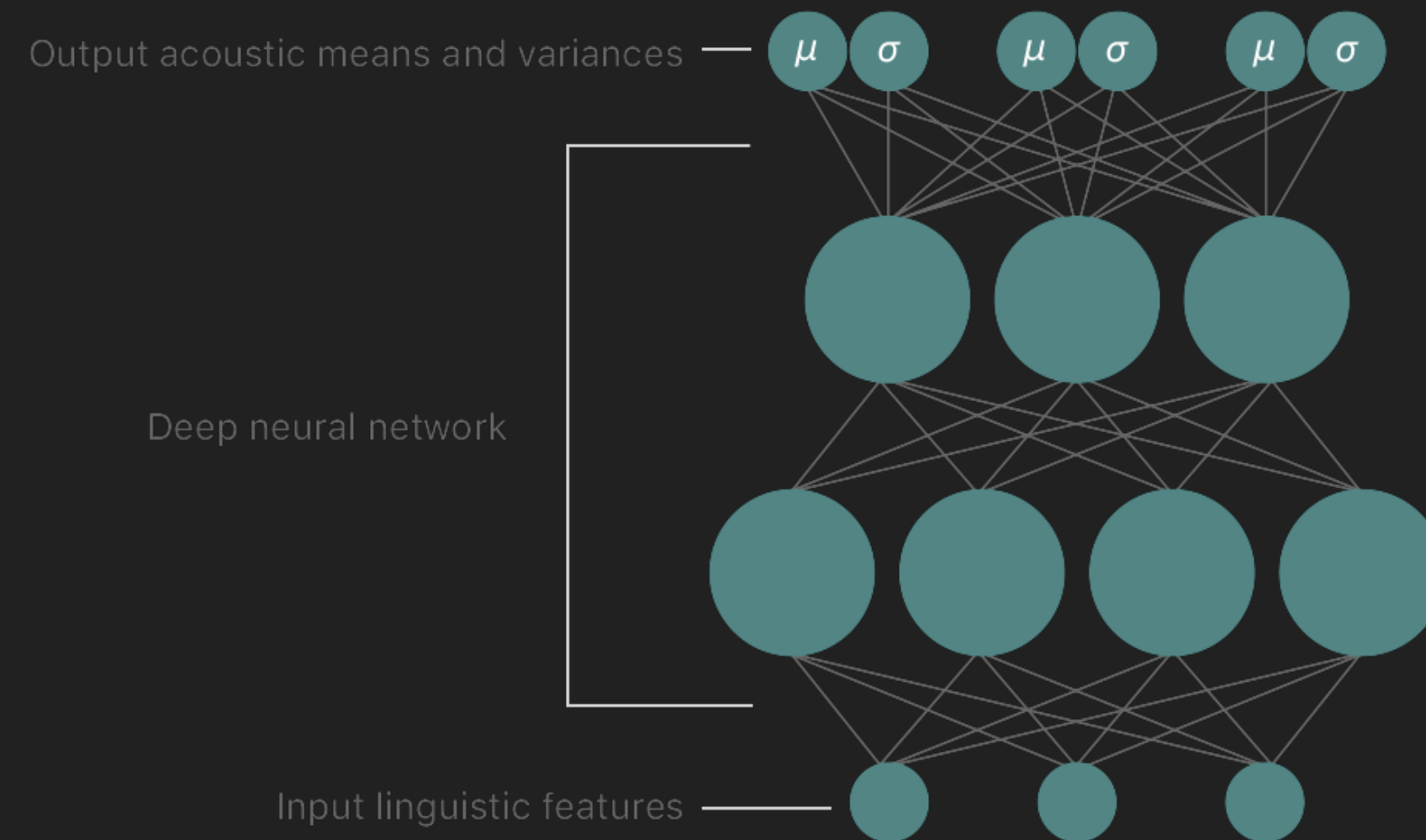
Starting in iOS 10 and continuing with new features in iOS 11, we base Siri voices on deep learning. The resulting voices are more natural, smoother, and allow Siri's personality to shine through

It has been deployed into hundreds of millions of desktop and mobile devices (e.g. iPhone, iPad, Mac, etc.) via iOS and macOS in multiple languages.

iOS 9 🔊 iOS 10 🔊 iOS 11 🔊



# EXAMPLE: IMPROVEMENTS TO SIRI'S VOICE

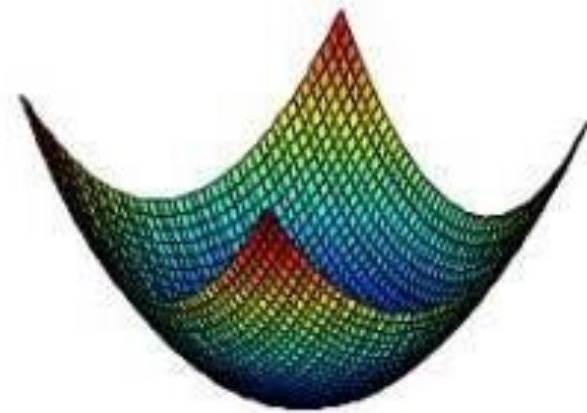


Read more in the [paper](#)

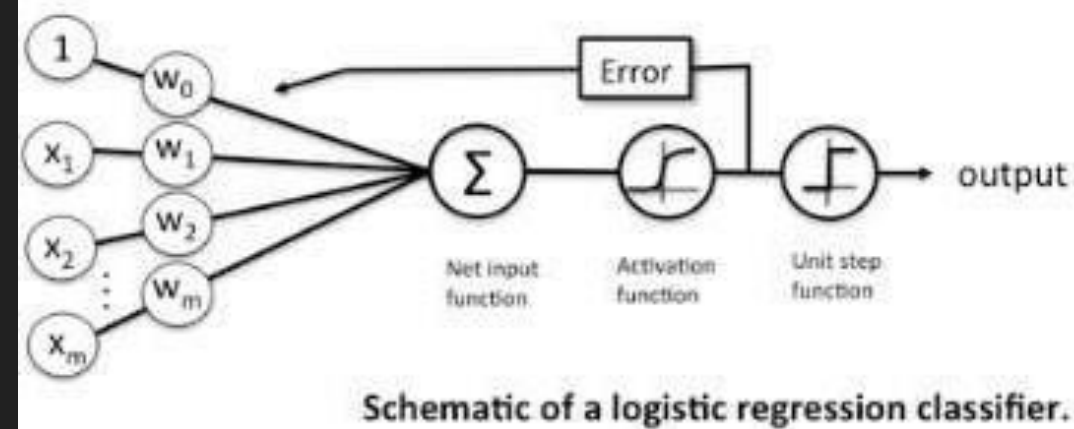
# FROM ANNS TO DNNS

The concepts learnt from ANNs are the basis for deep learning

You

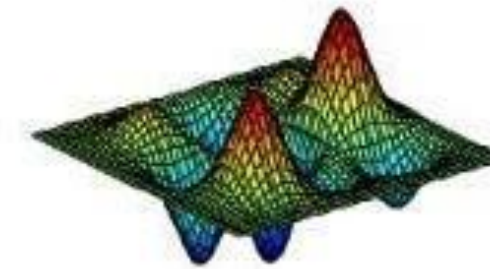


- Unique optimum: global/local.



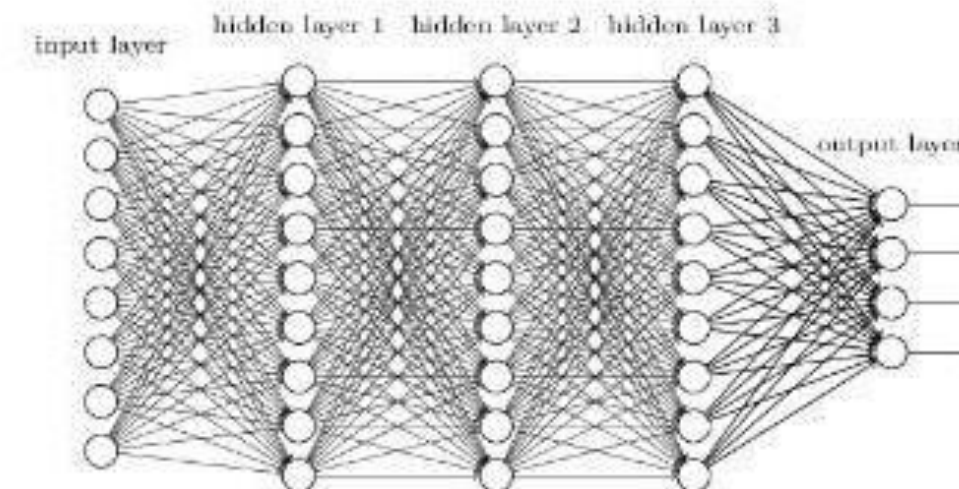
Schematic of a logistic regression classifier.

The guy she tells you  
not to worry about



- Multiple local optima
- In high dimensions possibly

Deep neural network





# DEEP LEARNING

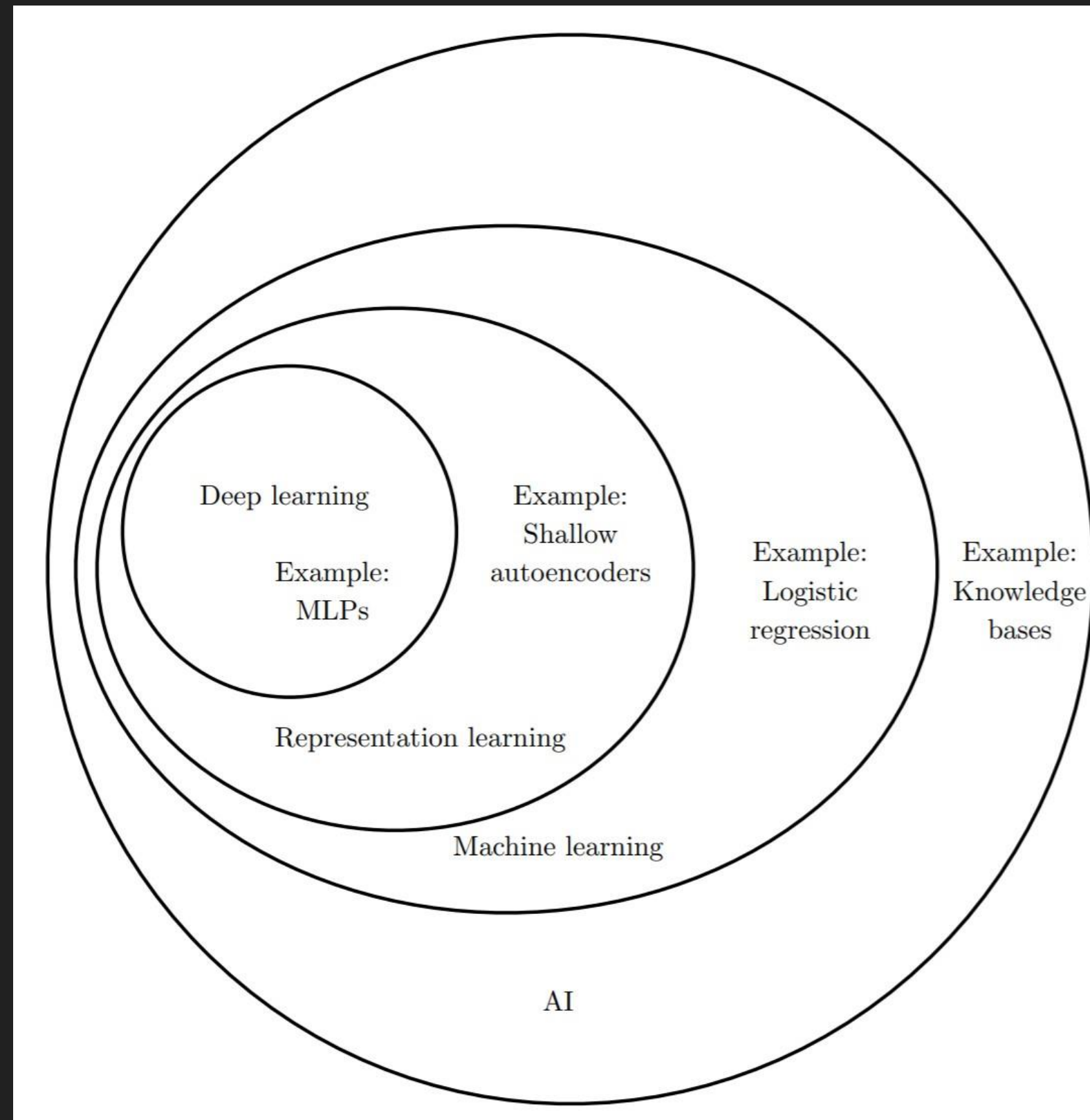
Neural network research suffers from a large amount of hype

OK, Deep Learning has outlived its usefulness as a buzz-phrase...Deep Learning was a rebranding of the modern incarnations of neural nets with **more than two layers**



- A **common view** is that deep learning is a class of machine learning techniques that are able to understand the world through a **hierarchy of concepts**. Data representation is learnt through **multiple levels of abstraction**
- **Schmidhuber (2015)**: proposed a **credit assignment path (CAP)** that distinguishes deep learning models as those that have at least **three layers of representational processing**. A CAP of 10 or more can be considered very deep
- The phrase can sometimes be misleading, as deep neural networks may not necessarily be an adequate distinction anymore

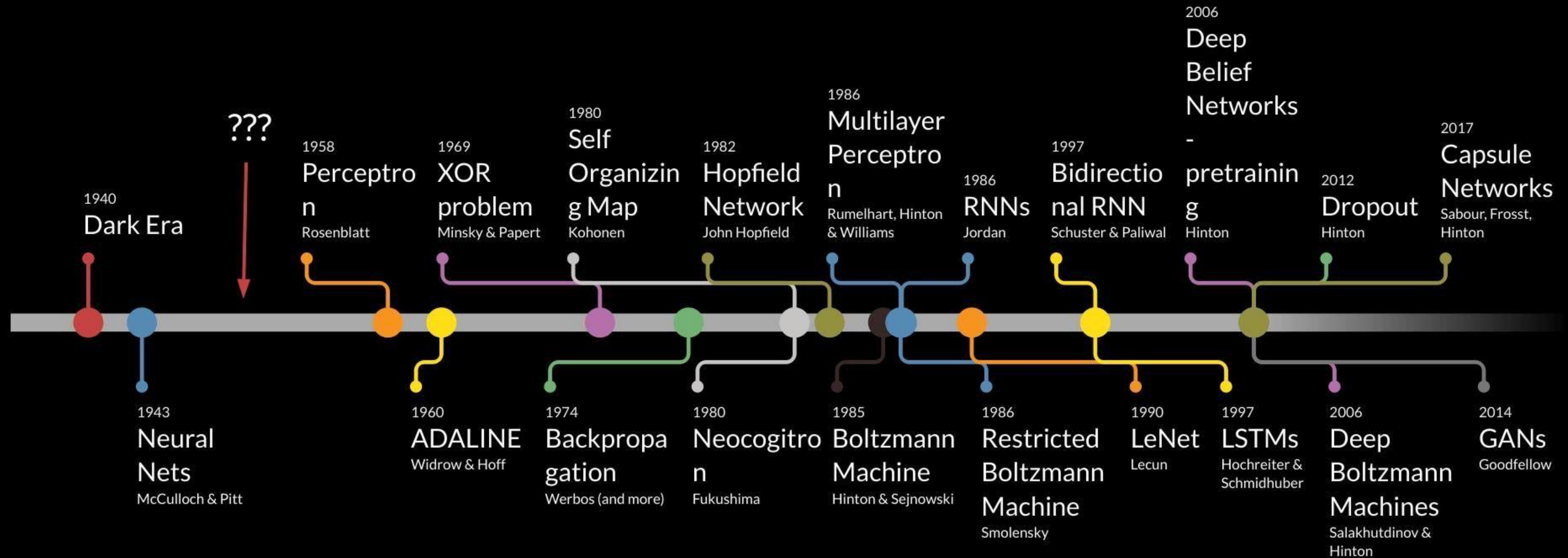
# AN APPROACH TO AI



# BIOLOGICAL INSPIRATION



# Deep Learning Timeline



# DEEP LEARNING MODELS

There are many different types of deep learning models including:

- Deep Neural Networks (DNN)
- Deep Belief Networks (DBNs)
- Convolutional Neural Networks (CNNs)
- Recurrent Neural Networks (RNNs)
- Long Short-Term Memory Networks (LSTMs)
- Generative Adversarial Networks (GANs)
- Autoencoders



# HOW IT WORKS

# A BLACK BOX



# FEATURE ENGINEERING AND SELECTION

Extracting and choosing suitable features to represent the data.  
Consider features to determine the difference between the digits 7 and 2.

Flat top  
Slightly curved line from top to bottom  
No bottom

7

2

Curved top  
Slightly curved line from top to bottom  
Flat bottom

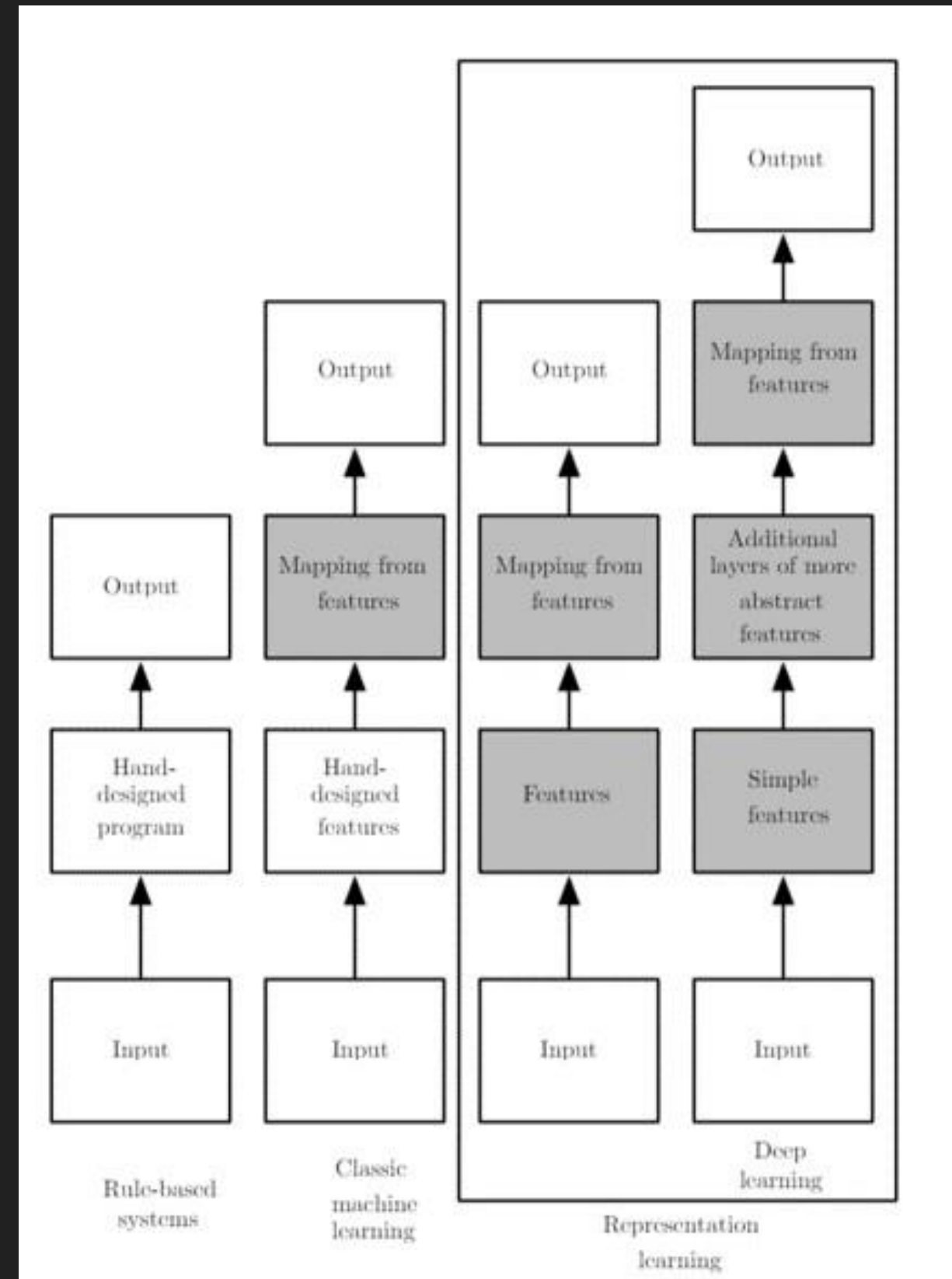
# A CHALLENGE

Feature engineering and selection has been a challenge for machine learning for a long time

What features should we use for classifying cats vs. dogs?

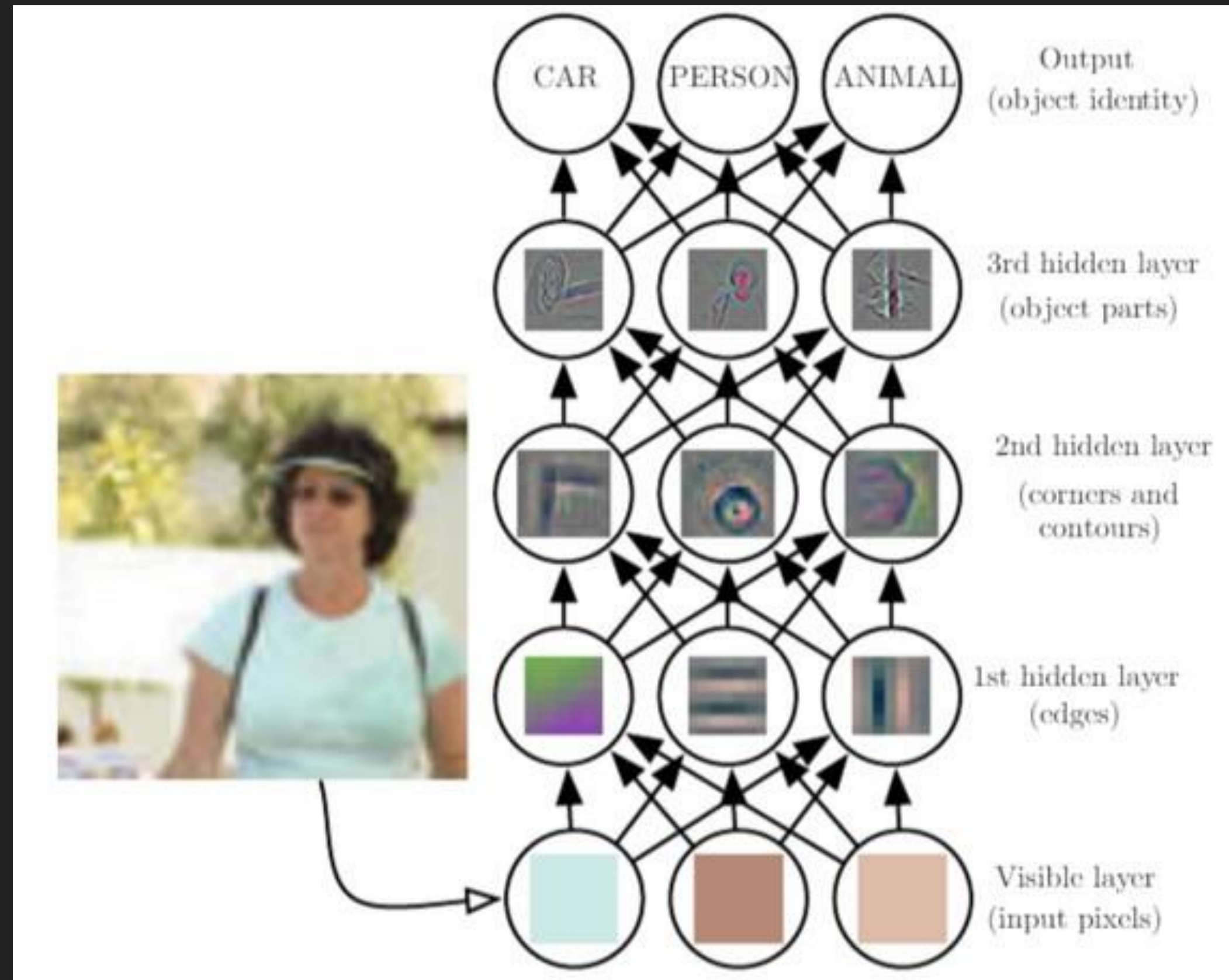


# REPRESENTATION LEARNING



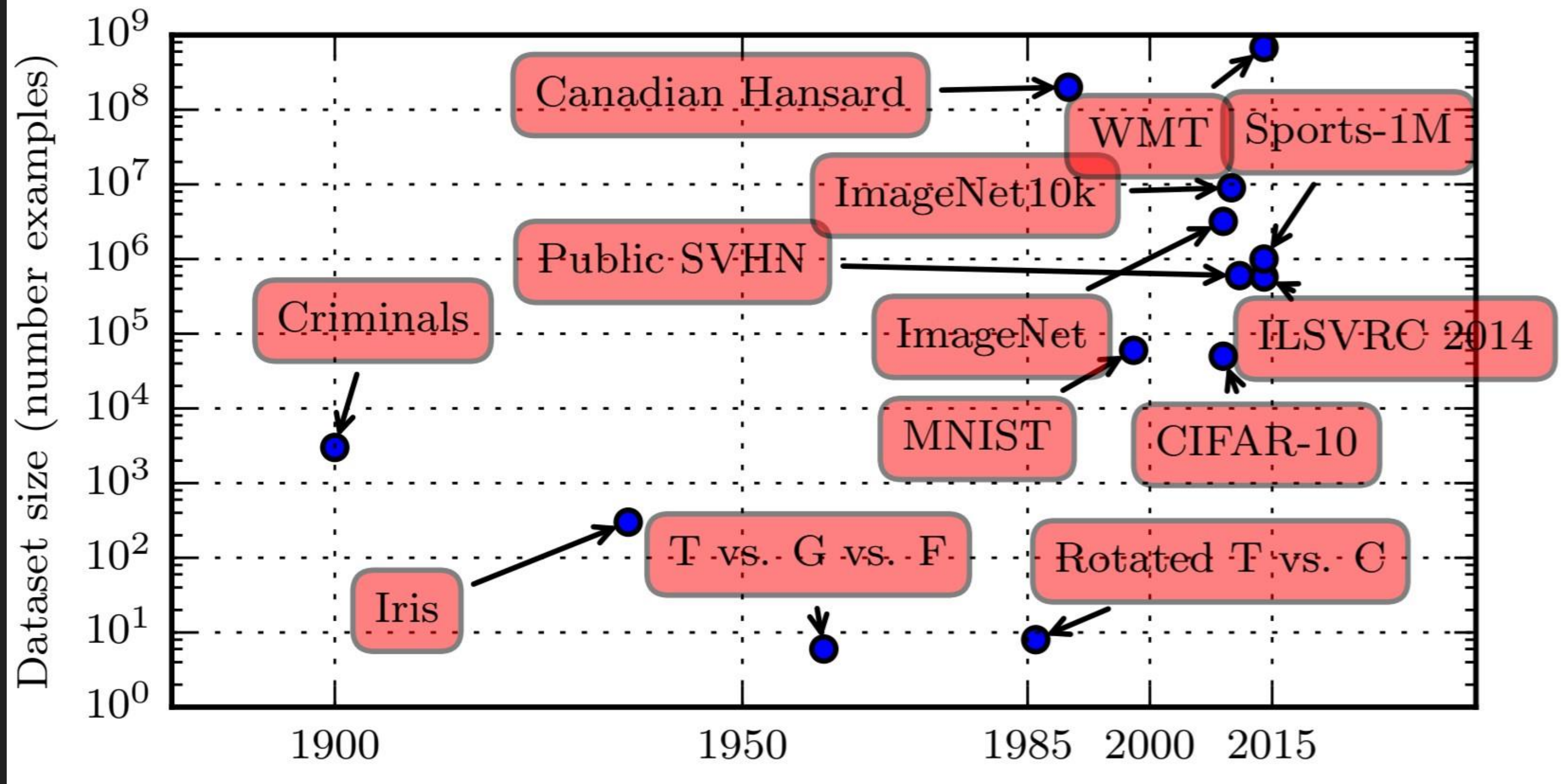


# HIERARCHICAL REPRESENTATION LEARNING



WHY NOW?

# BIG DATA



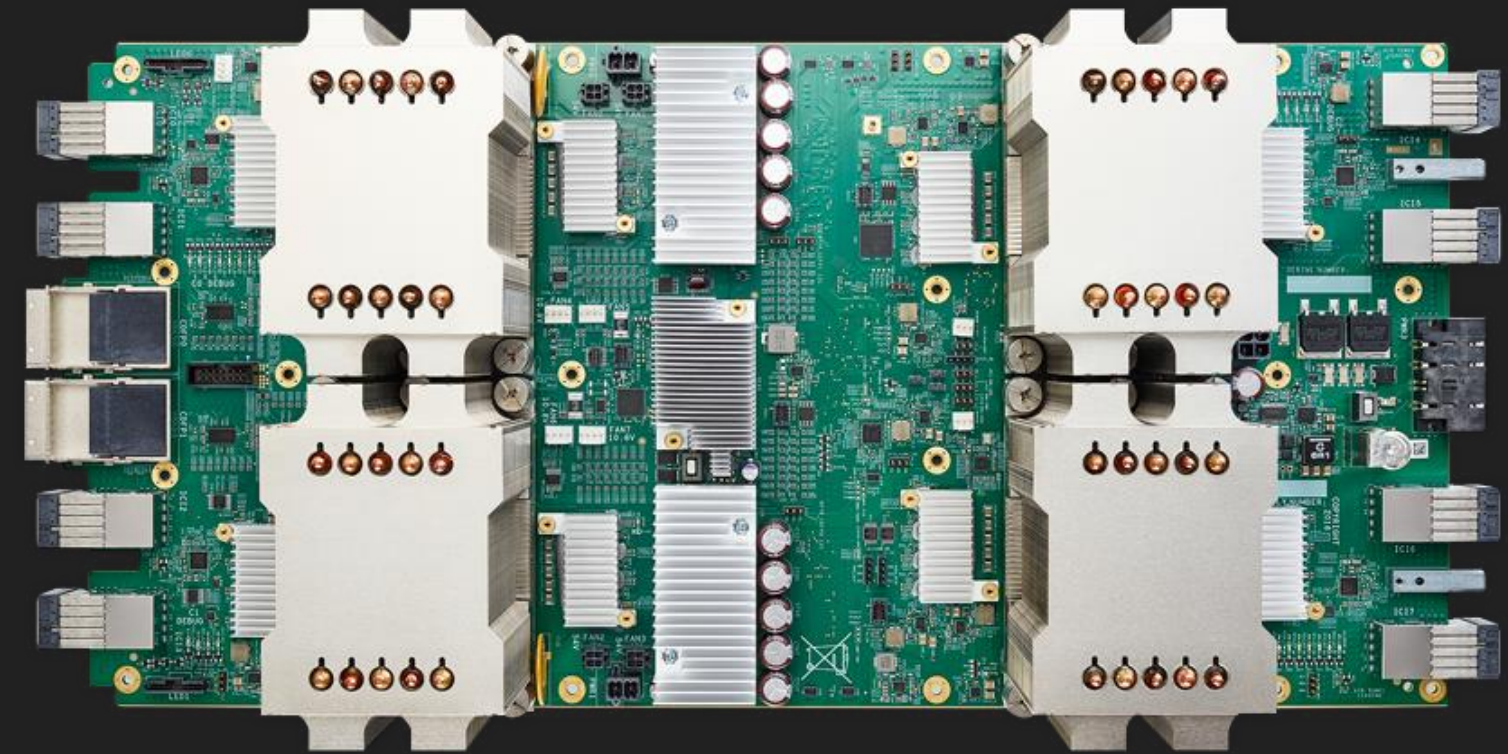


# SPECIALISED PROCESSORS

NVIDIA Graphics Processing Unit (GPU)



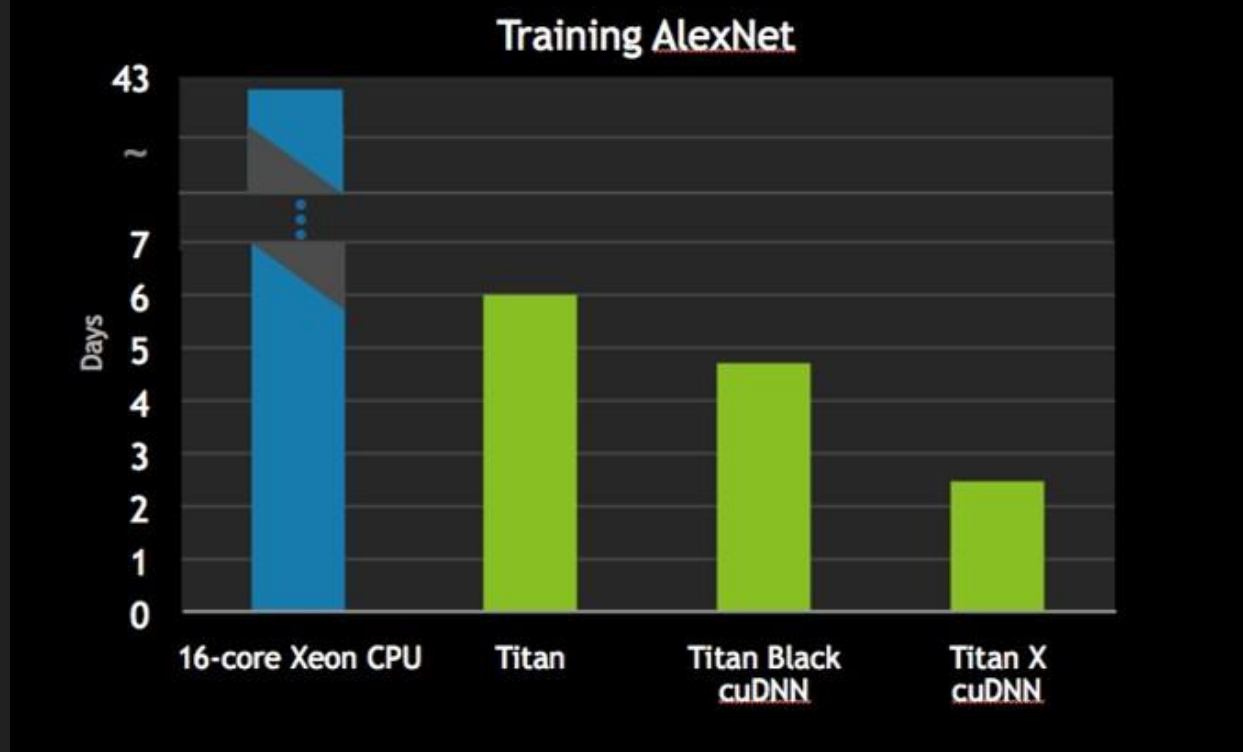
Google Tensor Processing Unit (TPU)



# GPU SPEED IMPROVEMENTS

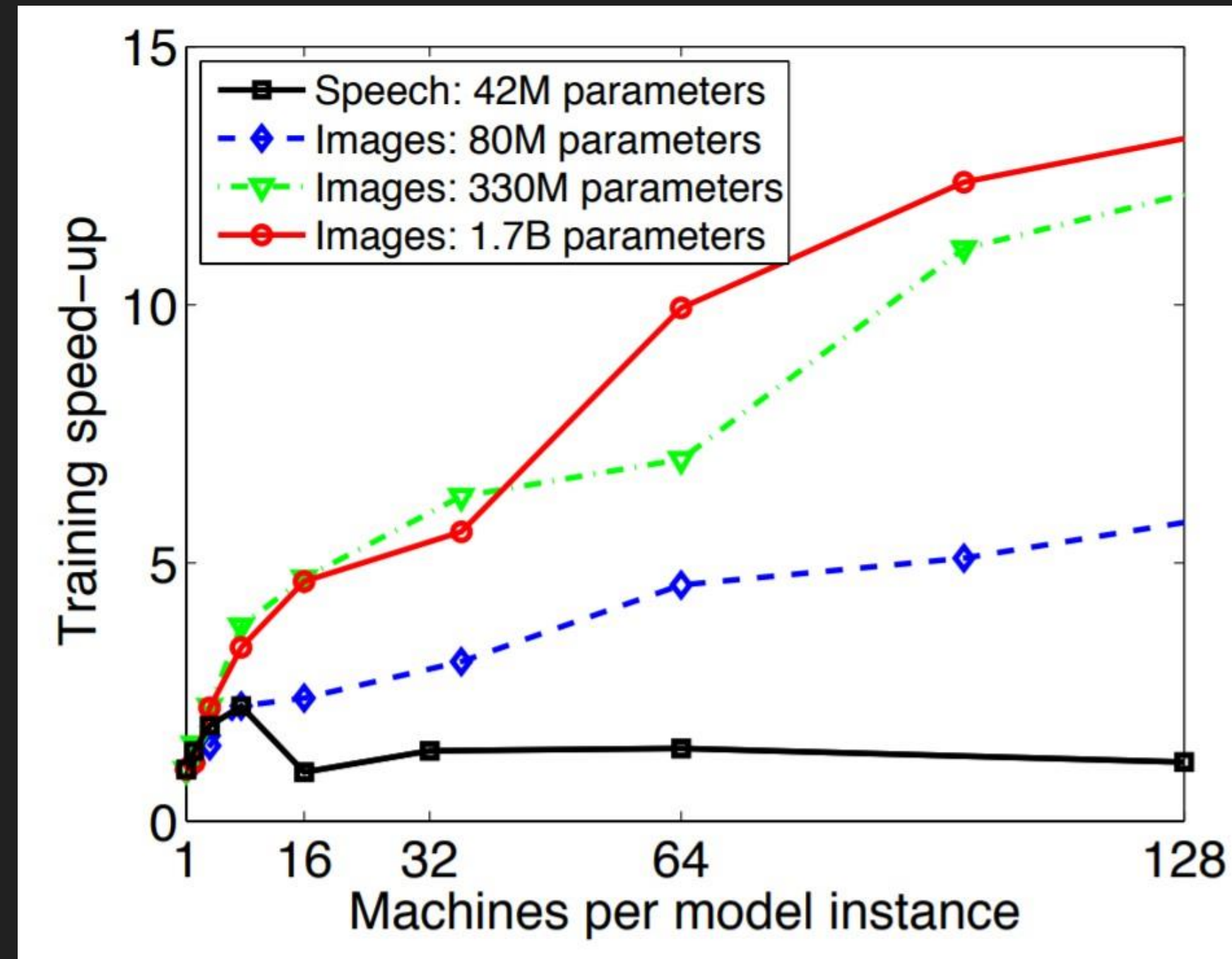
Package	Architecture	576x1024	1024x4096	2304x16000	4096x11008
Goto BLAS	Single CPU	563s	3638s	172803s	223741s
Goto BLAS	Dual-core CPU	497s	2987s	93586s	125381s
GPU		38.6s	184s	1376s	1726s
GPU Speedup		<b>12.9x</b>	<b>16.2x</b>	<b>68.0x</b>	<b>72.6x</b>

## TITAN X FOR DEEP LEARNING





# DISTRIBUTION SPEED IMPROVEMENTS



# MORE PLAUSIBLE ACTIVATIONS

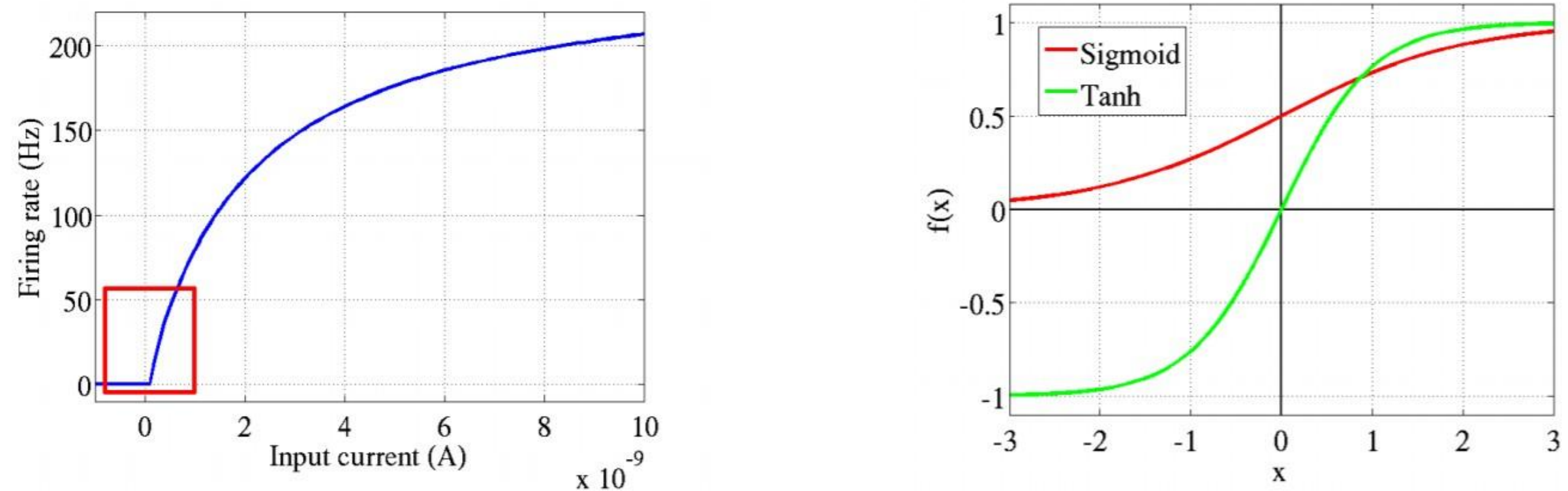
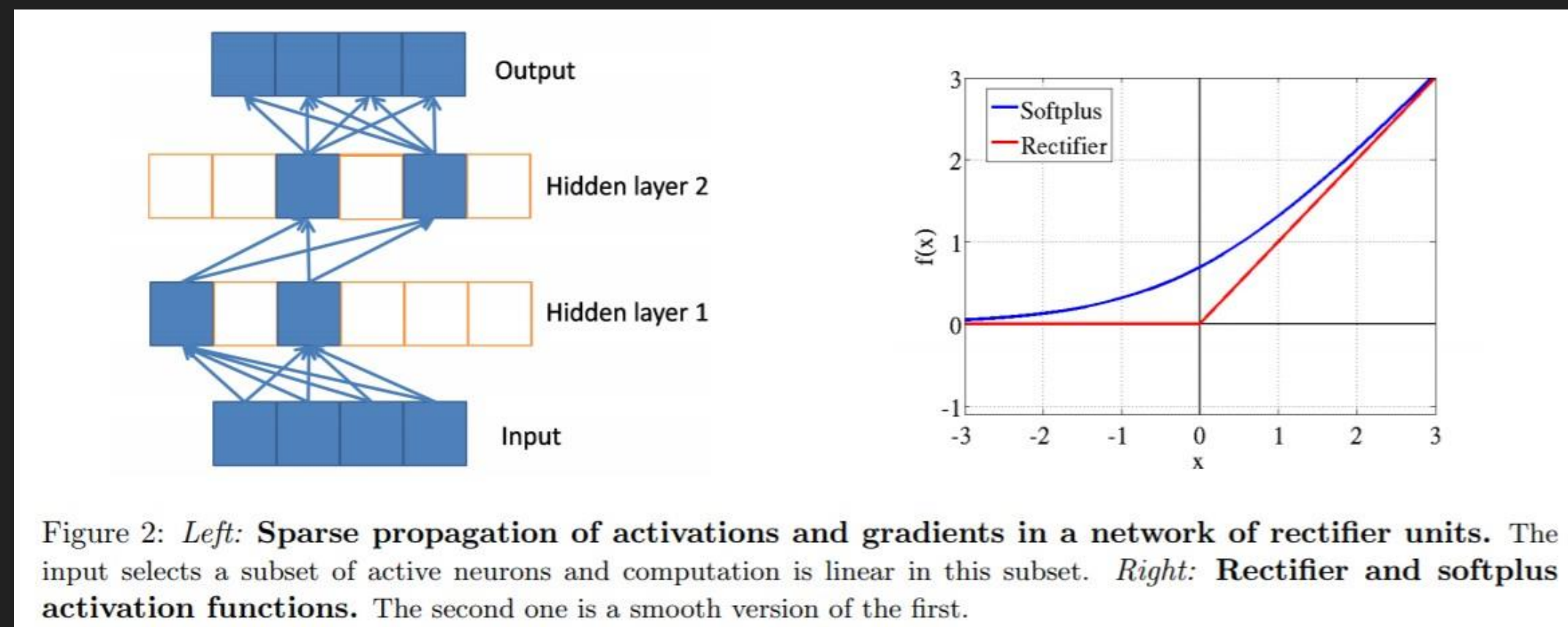
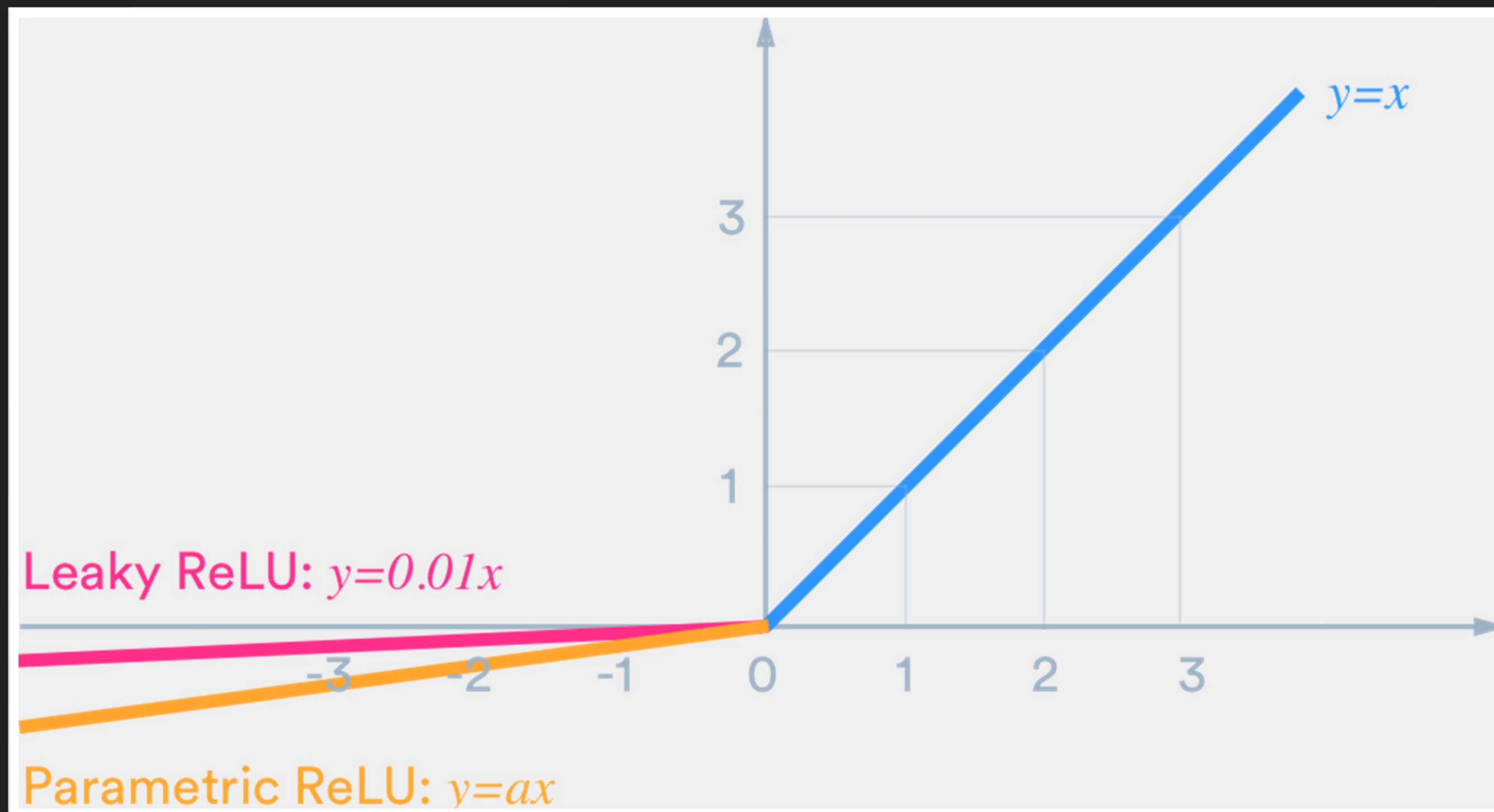


Figure 1: *Left: Common neural activation function motivated by biological data. Right: Commonly used activation functions* in neural networks literature: logistic sigmoid and hyperbolic tangent (*tanh*).

# ReLU



# Leaky ReLU



# APPLICATION AREAS

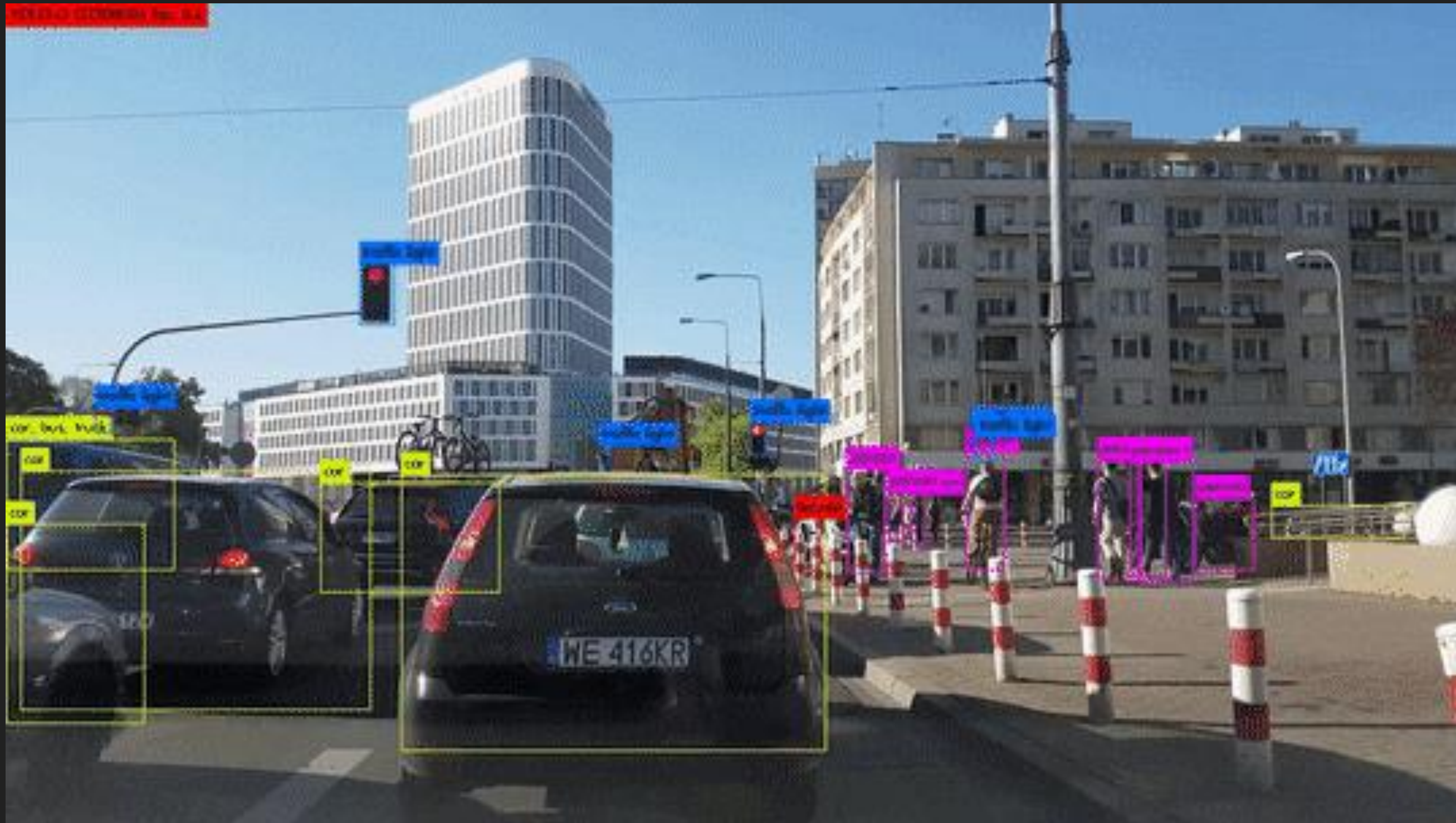


# OBJECT RECOGNITION





# OBJECT DETECTION

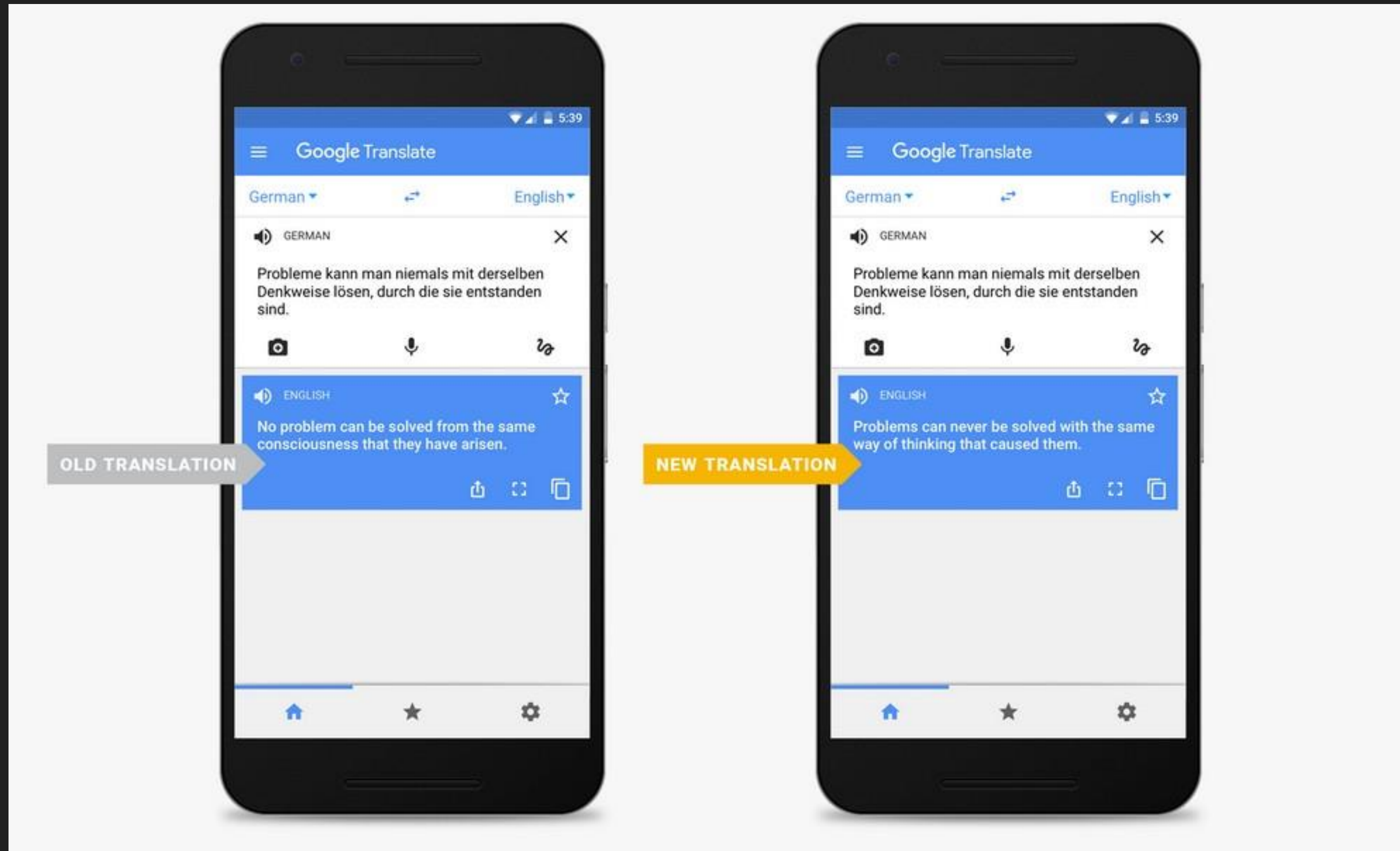


# OBJECT SEGMENTATION





# LANGUAGE TRANSLATION





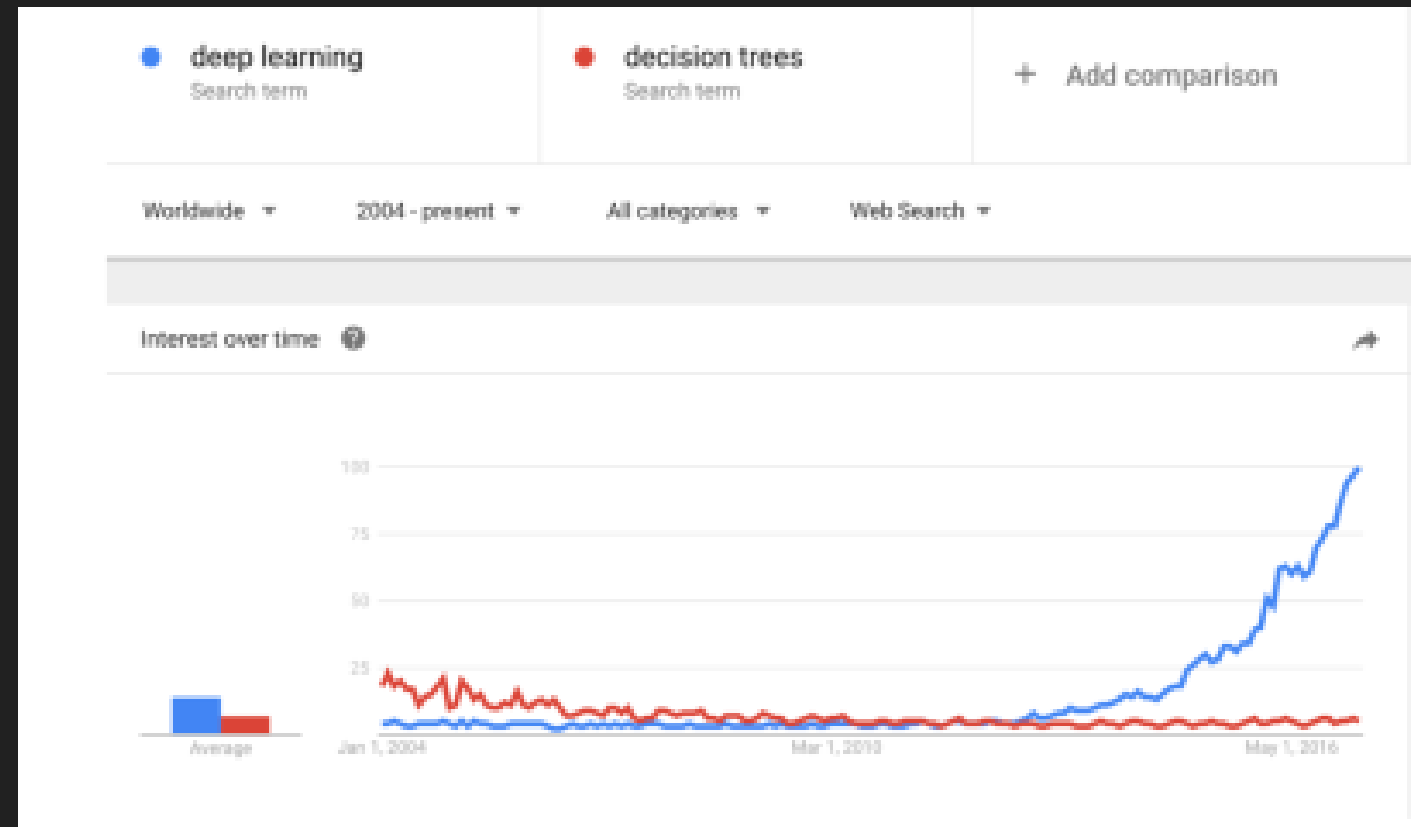
# OTHER APPLICATIONS

- Speech recognition
  - Deep Speech and Deep Speech 2
- Recommendation systems
  - Amazon and Netflix
- Games
  - AlphaGo
- Bio-informatics
  - Prediction of cardiovascular risk factors from retinal photographs

... plus more

# DEEP LEARNING IN PRACTICE

# AN ACTIVE AREA OF RESEARCH



## Neural Information Processing Systems Conference

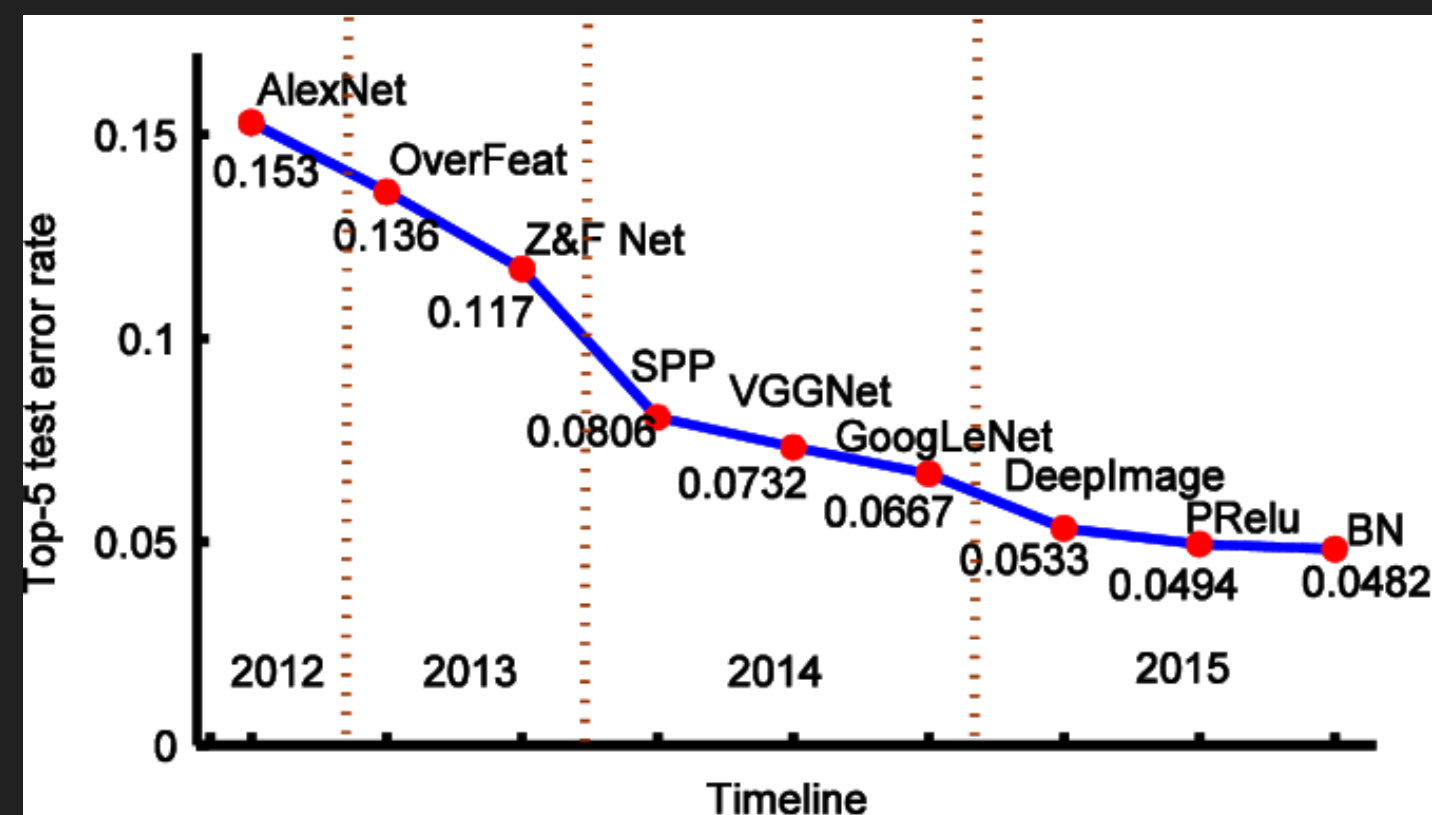
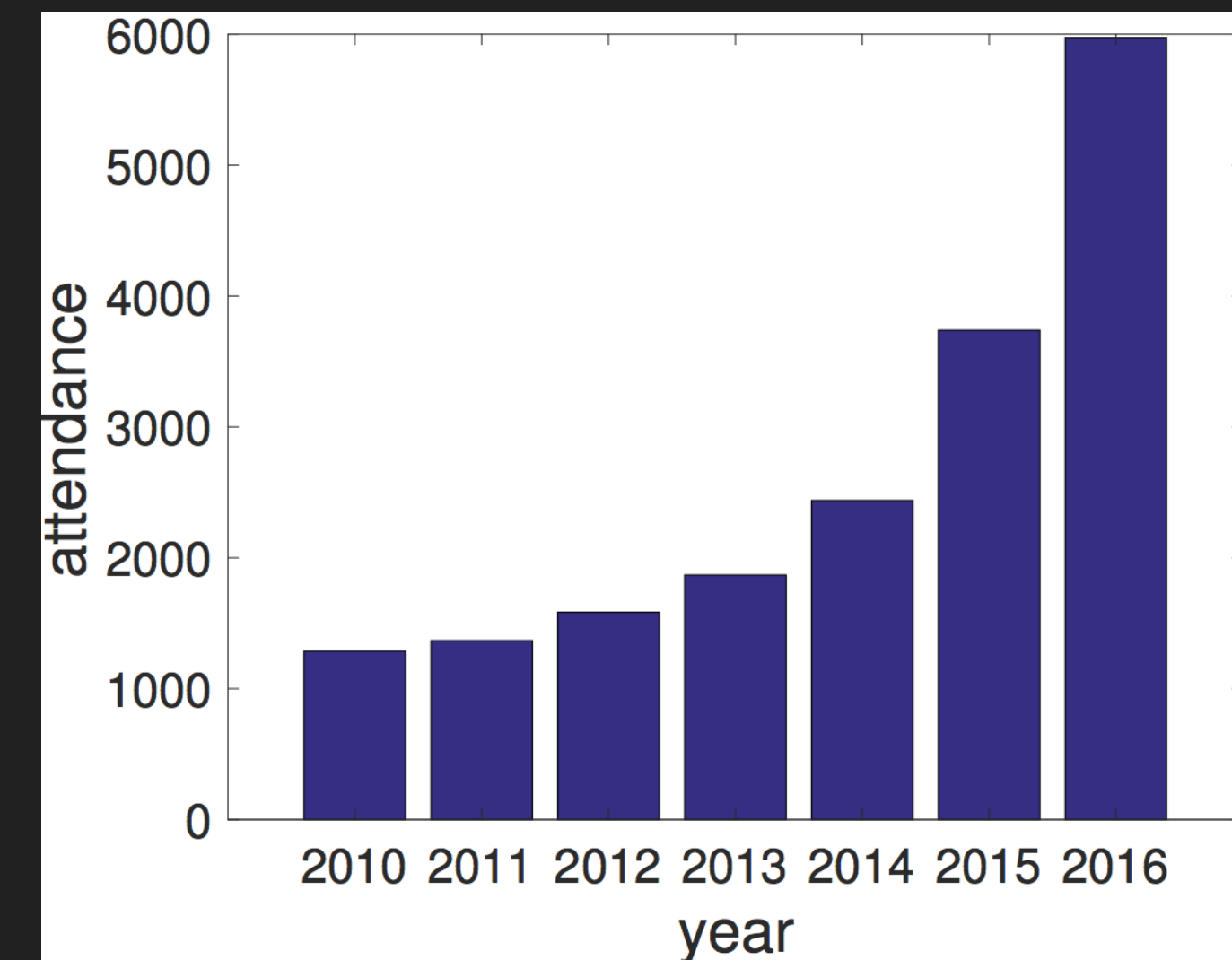


Fig. 18. ImageNet classification results on test dataset

## FRAMEWORKS AND LIBRARIES

theano

PYTORCH

mxnet

 TensorFlow

 Keras

 Caffe2

...plus many more!

[Software comparison](#)



# RESEARCH TRENDS

- Rush to apply deep learning algorithms to a variety of application domains
- Shift toward unsupervised/self-supervised and reinforcement learning
- Critical evaluation of deep learning approaches including fairness and interpretability
- Understanding representation learning

# PRACTICALITY

- A set of approaches among many; not necessarily a saving grace
- Typically large data sets
- Large parameter space can lead to overfitting
- Often too complex

# SUMMARY

- Hierarchical representation learning
- Inspired by the brain's highly-connected network structure
- Boosted by big data and GPUs
- Groundbreaking successes in many areas
- High popularity and hype

# UP NEXT

## Convolutional Neural Networks

