## Deep Machine Learning

Department of Computer Engineering

College of Engineering

University of Diyala

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# Deep Learning

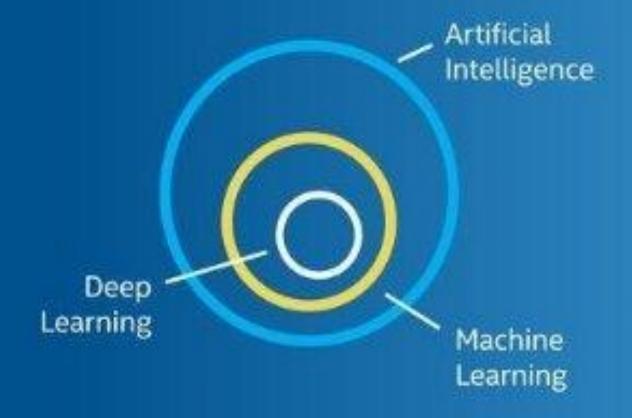


#### ARTIFICIAL INTELLIGENCE MACHINE Early artificial intelligence stirs excitement. **LEARNING** Machine learning begins DEEP to flourish. **LEARNING** Deep learning breakthroughs drive Al boom. (e")'. e" (e")'. e 1970's 2010's 1950's 1960's 1980's 1990's 2000's

Since an early flush of optimism in the 1950's, smaller subsets of artificial intelligence - first machine learning, then deep learning, a subset of machine learning - have created ever larger disruptions.



# ARTIFICIAL INTELLIGENCE 101



Deep learning is a branch of ML that uses neural network models to understand large amounts of data. It can accelerate processes like image and speech recognition, and natural language recognition.

#### **ARTIFICIAL INTELLIGENCE**

A program that can sense, reason, act, and adapt

#### **MACHINE LEARNING**

Algorithms whose performance improve as they are exposed to more data over time

#### DEEP LEARNING

Subset of machine learning in which multilayered neural networks learn from vast amounts of data

## Machine Learning

Input: X Output: Y

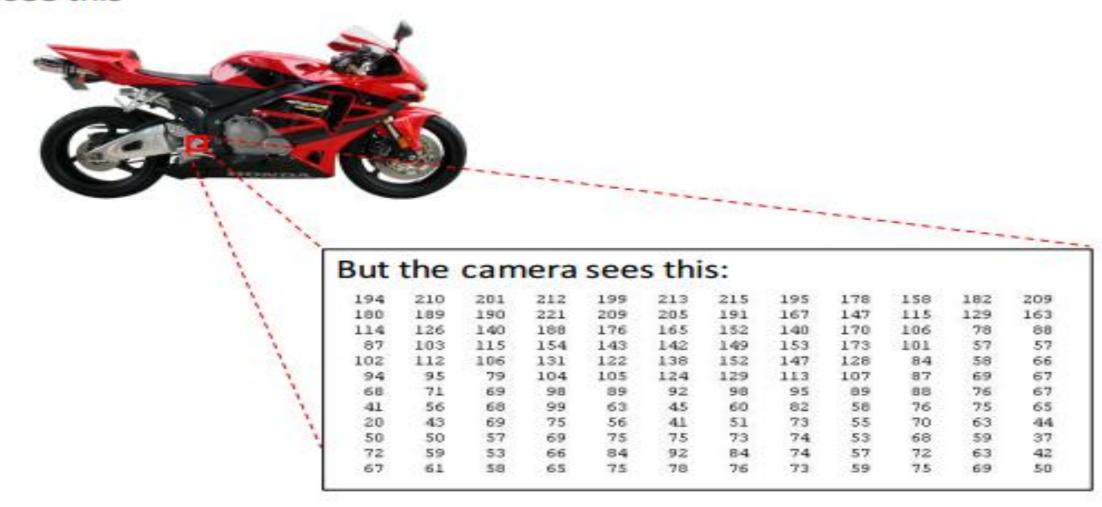




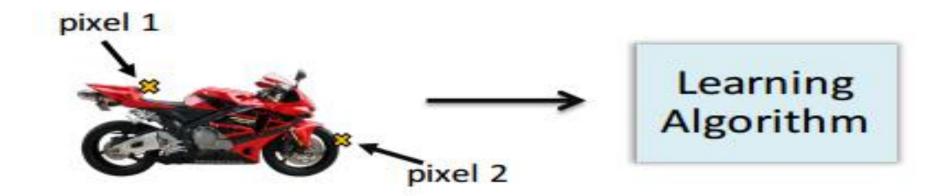
Label" motorcycle"

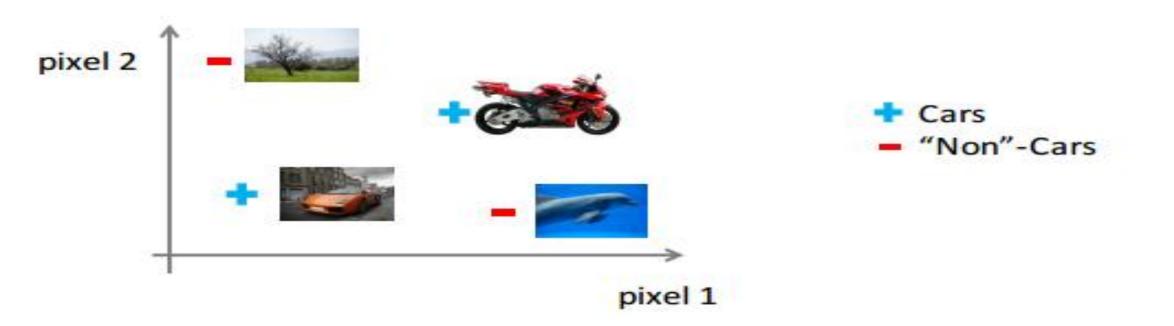
## Why is it hard?

#### You see this



## Raw Image Representation





#### Things we want to do with data

**Images** 



Label image

Audio



Speech recognition

**Text** 



Web search

#### Features for machine learning

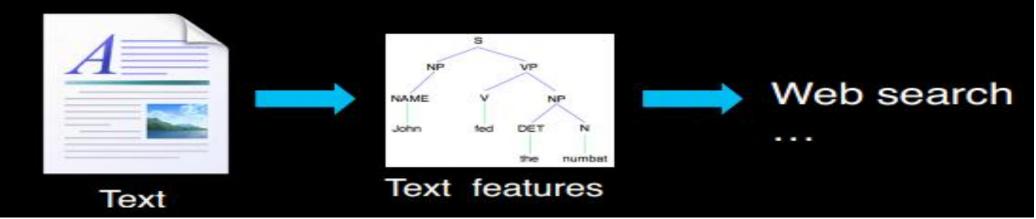




Audio



Text



#### The short answers

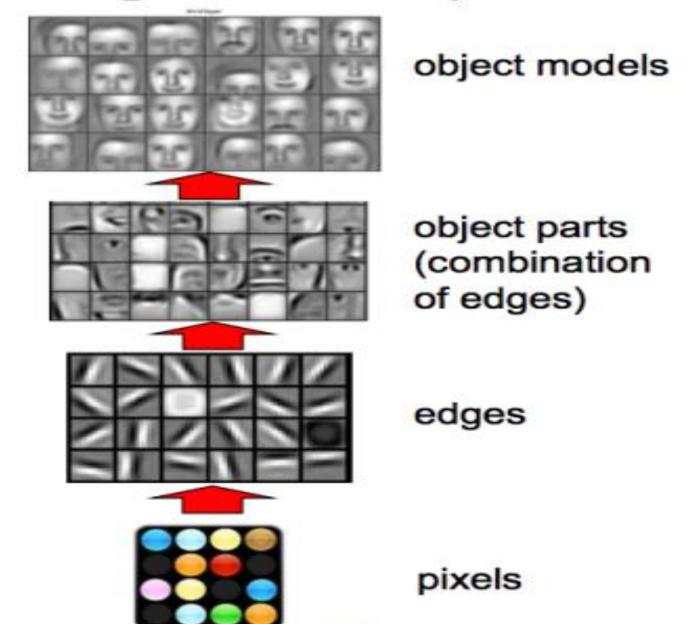
'Deep Learning' means using a neural network
 with several layers of nodes between input and output

the series of layers between input & output do feature identification and processing in a series of stages, just as our brains seem to.

## How do We Train Deep Architectures?

- Inspiration from mammal brain
- Multiple Layers of "neurons" (Rumelhart et al 1986)
- Train each layer to compose the representations of the previous layer to learn a higher level abstraction
  - Ex: Pixels → Edges → Contours → Object parts → Object categories
  - Local Features → Global Features
- Train the layers one-by-one (Hinton et al 2006)
  - Greedy strategy

#### Deep Learning: learn representations!



Source:Lee et.al. ICML2009

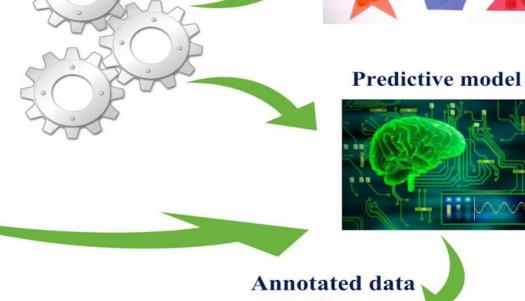
#### Machine learning workflow



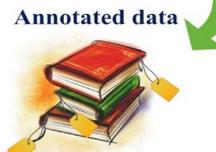












## Deep Learning trends

Now

0-2 years Tagged data 3-5 years Tagged & untagged data





#### Learning from tagged data (supervised)



Coffee mug



Coffee mug



Coffee mug



Coffee mug



Coffee mug

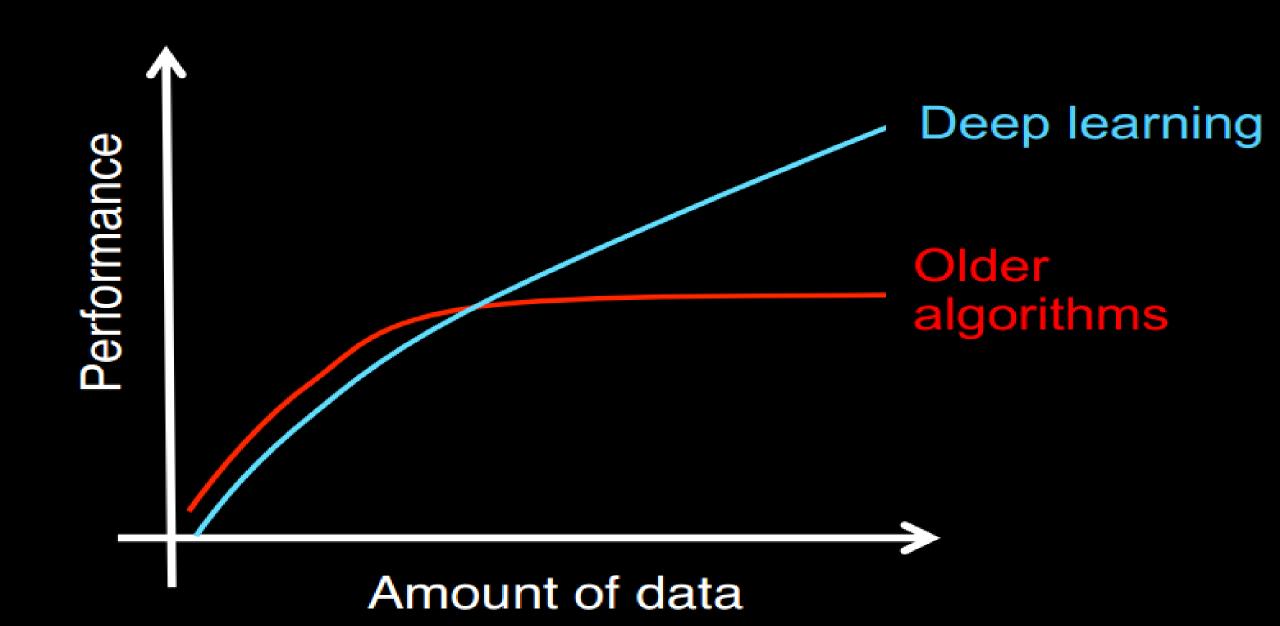


Coffee mug

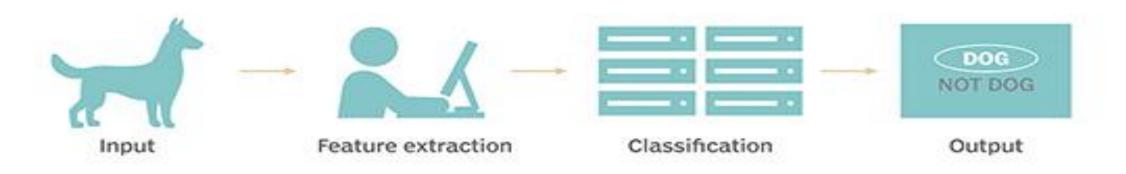
Testing: What is this?



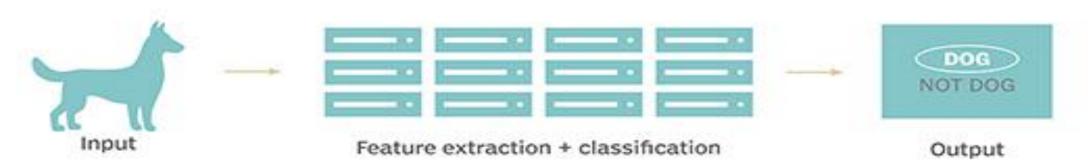
#### Learning from tagged data



#### TRADITIONAL MACHINE LEARNING

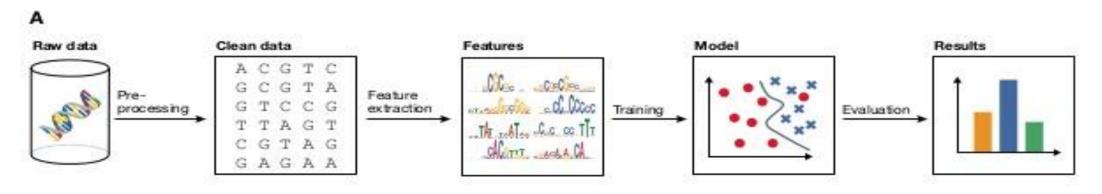


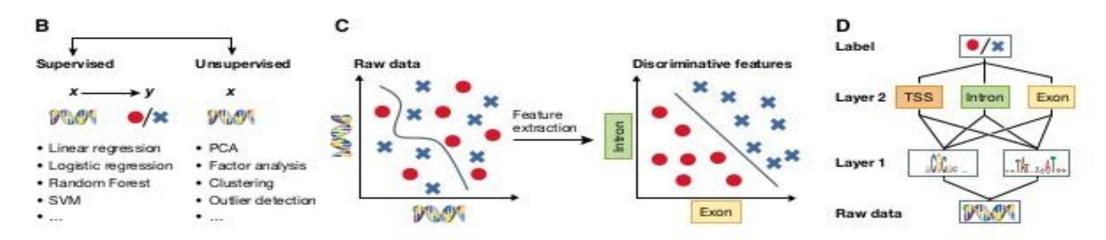
#### **DEEP LEARNING**



#### Deep learning applications

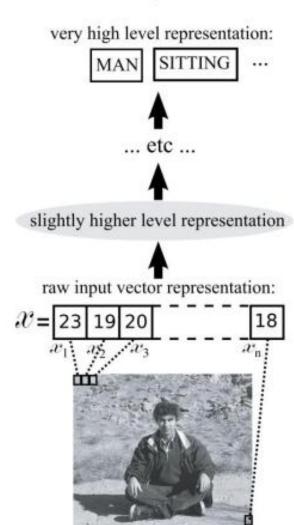
More traditional Machine Learning Applications to Deep Learning Application





#### A Motivational Task: Percepts -> Concepts

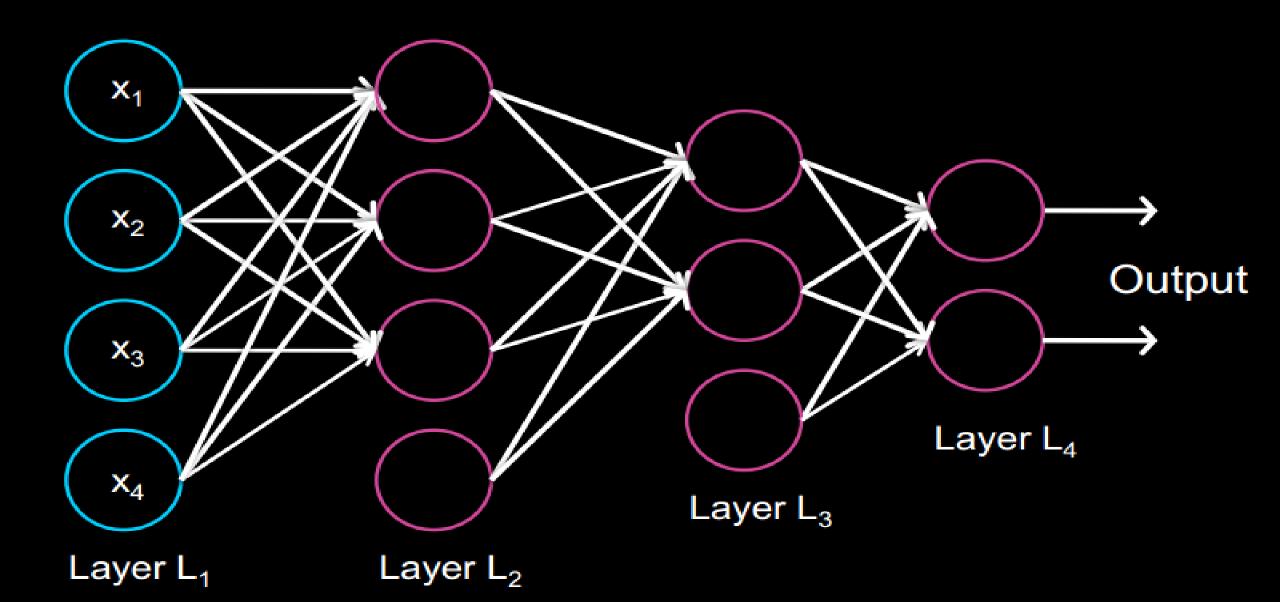
- Create algorithms
  - that can understand scenes and describe them in natural language
  - that can infer semantic concepts to allow machines to interact with humans using these concepts
- Requires creating a series of abstractions
  - Image (Pixel Intensities) → Objects in Image → Object Interactions → Scene Description
- Deep learning aims to automatically learn these abstractions with little supervision



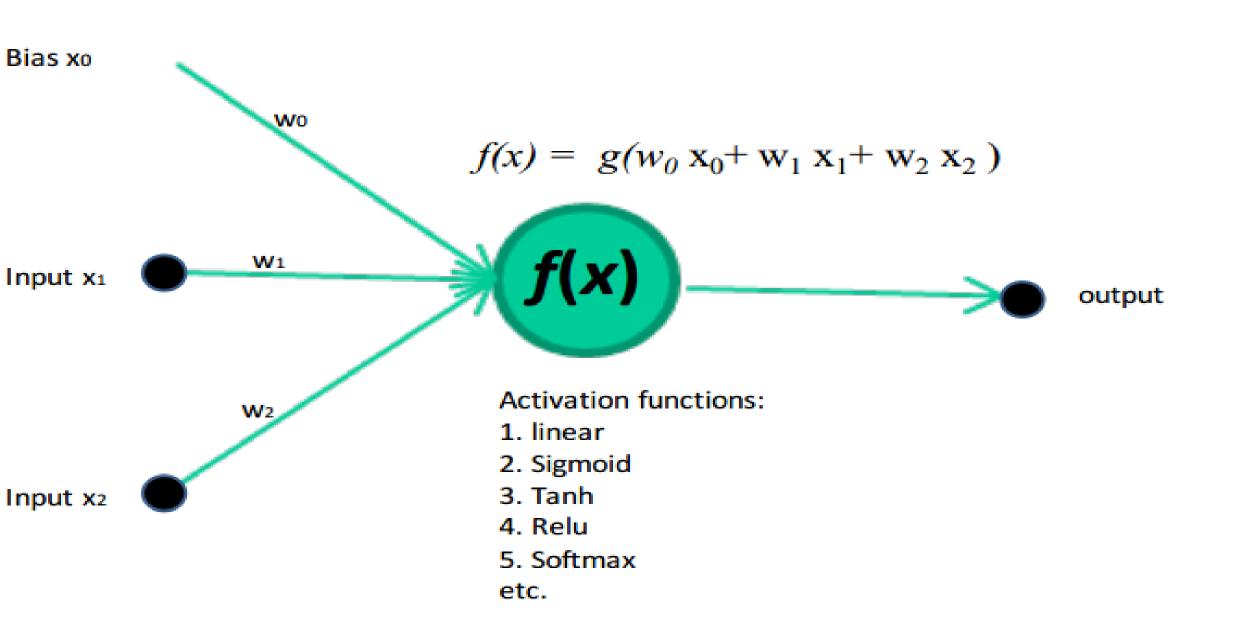
## Neurons in the brain



## Neural Network (Deep Learning)



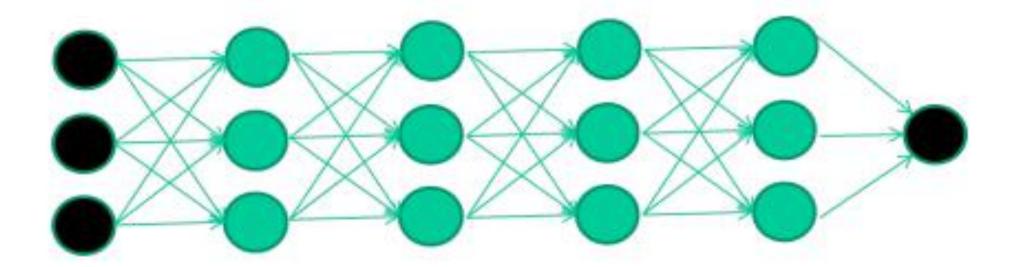
#### Single Unit, Input, weights, activation function, output



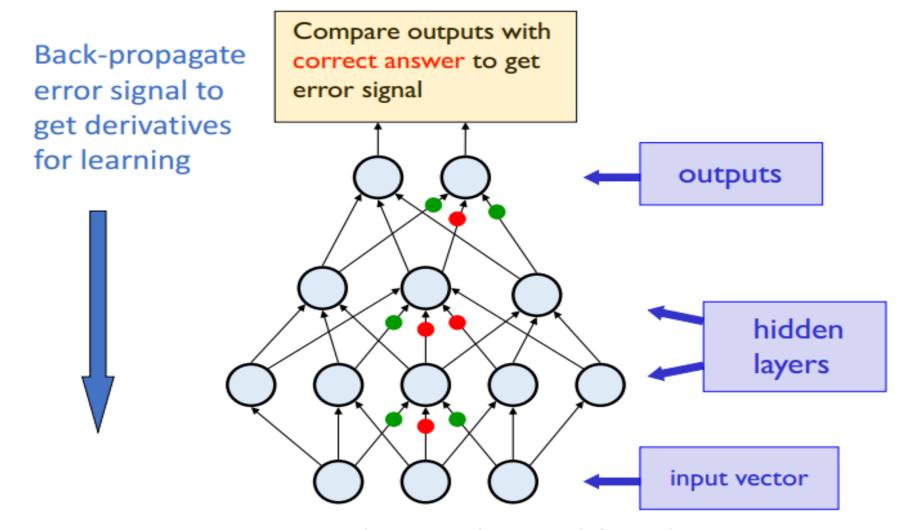
#### A dataset

Fields		class	
1.4 2.7	1.9	0	
3.8 3.4	3.2	0	
6.4 2.8	1.7	1	
4.1 0.1	0.2	0	
etc			

#### Train the deep neural network



#### Multilayer Perceptron with Back-propagation First deep learning model (Rumelhart, Hinton, Williams 1986)



Source: Hinton's 2009 tutorial on Deep Belief Networks

## DEEP LEARNING EVERYWHERE











INTERNET & CLOUD

Image Classification
Speech Recognition
Language Translation
Language Processing
Sentiment Analysis
Recommendation

MEDICINE & BIOLOGY

Cancer Cell Detection Diabetic Grading Drug Discovery MEDIA & ENTERTAINMENT

Video Captioning Video Search Real Time Translation SECURITY & DEFENSE

Face Detection Video Surveillance Satellite Imagery **AUTONOMOUS MACHINES** 

Pedestrian Detection Lane Tracking Recognize Traffic Sign

# Deep Visual-Semantic Alignments for Generating Image Descriptions (Karpathy, Fei-Fei; CVPR 2015)



"two young girls are playing with lego toy."



"boy is doing backflip on wakeboard."

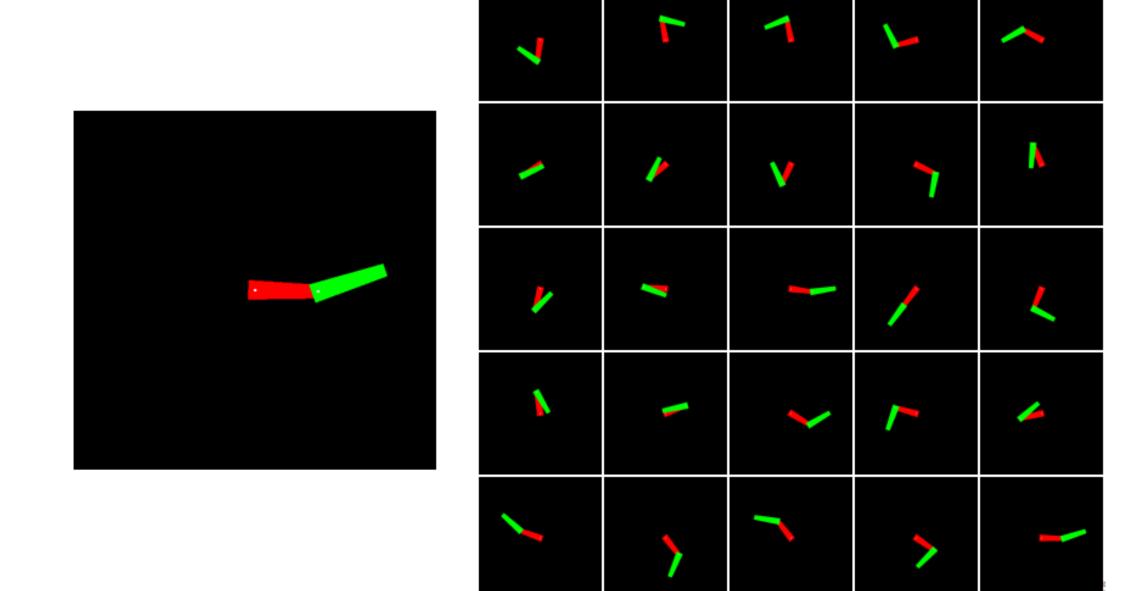


"construction worker in orange safety vest is working on road."

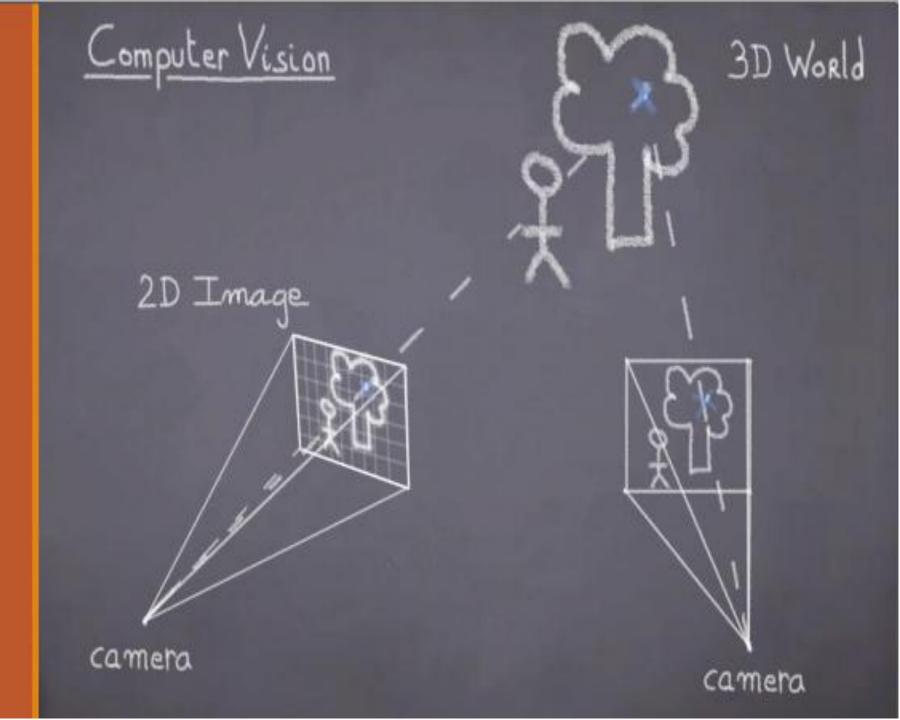


"man in black shirt is playing guitar."

Example: Learning the Configuration Space of a Robotic Arm



Deep Learning in Computer Vision



# Deep Learning in Robotics



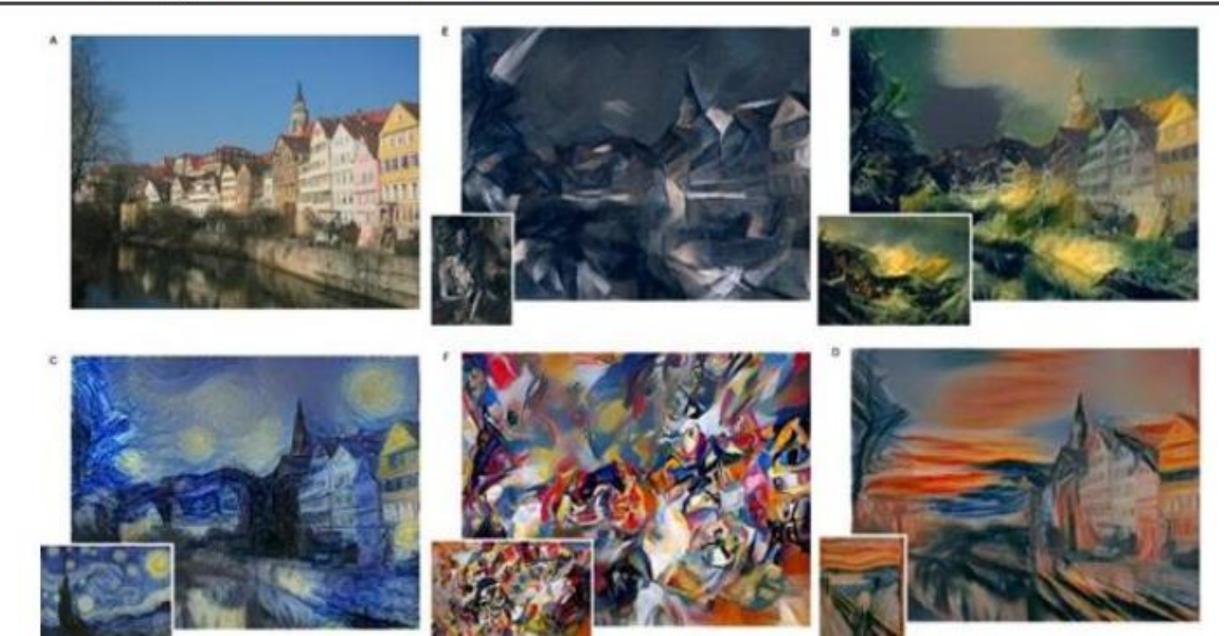
UVA DEEP LEARNING COURSE EFSTRATIOS GAVVES INTRODUCTION TO DEEP LEARNING AND NEURAL NETWORKS - 13



# Deep Learning in NLP and Speech



## Imitating famous painters



## Handwriting

Hi Motherboard readers!

This entire post was hand written by a neway network.

It probably writes better than you.)

If course, a neural network doesn't adually have hands

And the original text was typed by me, a human.

So what's going on here?

nounal network is a program that can learn to follow a set of rules

But it can't do it alone it needs to be trained.

This neural network was trained on a corpus of wating samples.

ut of the locations of a pen-tip as people write.

is how the natural searchs and creates different styles, from prior examples.

And it can use the buowledge to generate handwitten uses from inputed ket can evente its own style, or minis another's.

No two notes are the same.

It's the work of Alex Graves at the University of Toronto

And you can try it bo!

#### CIFAR 10 and Convolutional Neural Network



CIFAR 10 dataset:

50,000 training images

10,000 testing images

10 categories (classes)

Accuracies from different methods:

Human: ~94%

Whitening K-mean: 80%

.....

Deep CNN: 95.5%

## Recommender System

NETFLIX

Watch Instantly -

Just for Kids -

Your Queue

Personalize

DVDs -

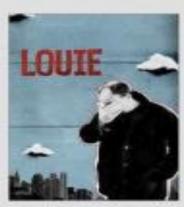
#### Friends' Favorites

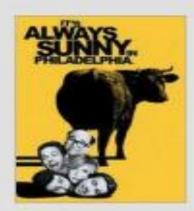












#### Watched by your friends



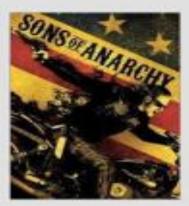


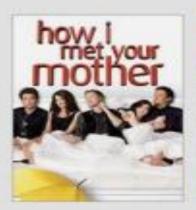


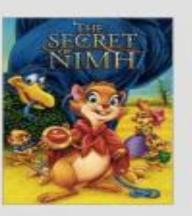


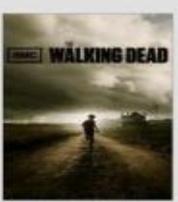
John Midgley





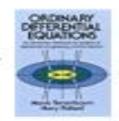






## Recommender System

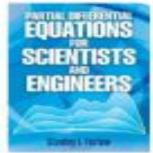






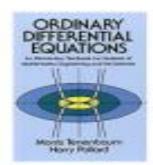
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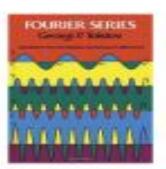
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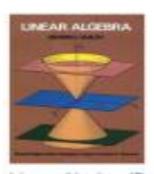
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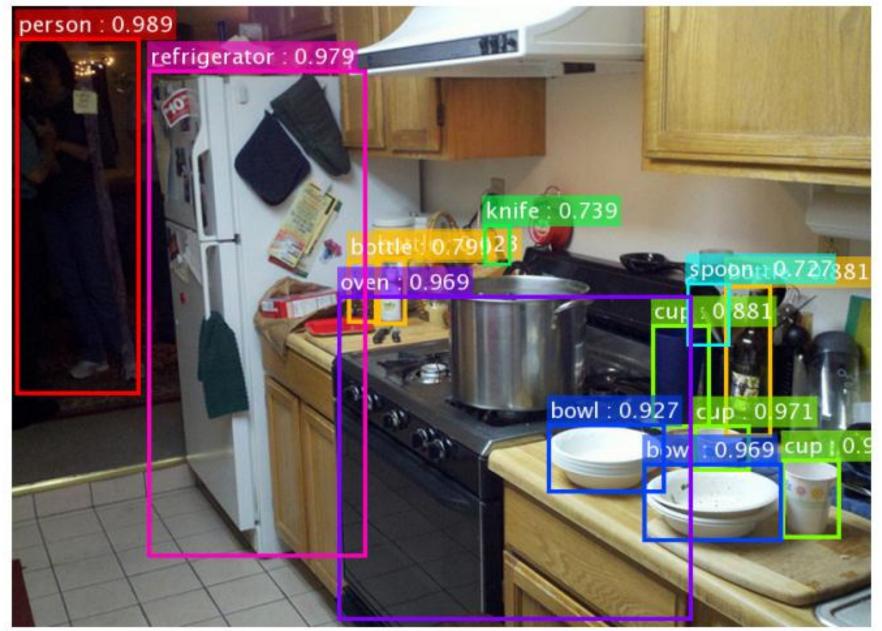


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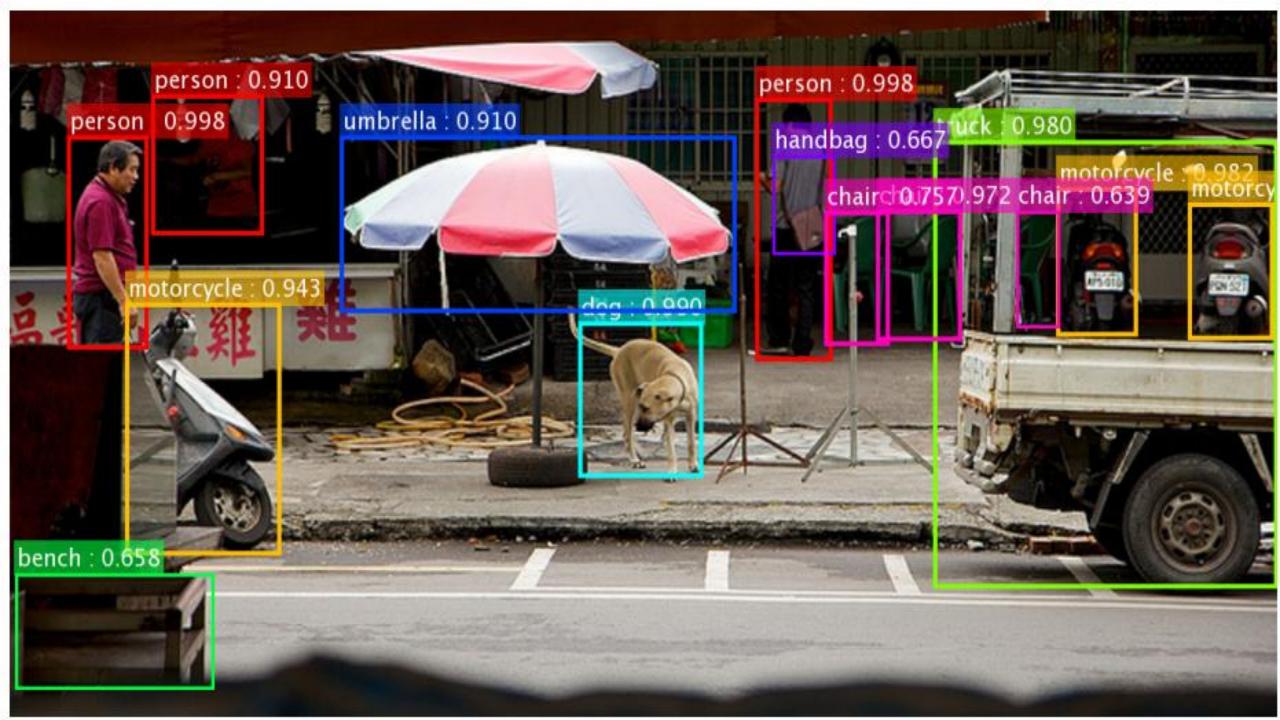


Linear Algebra (Dover Books on Mathematics) Georgi E. Shilov A A A A A A A Paperback \$14.35 Prime



\*the original image is from the COCO dataset

Kaiming He, Xiangyu Zhang, Shaoqing Ren, & Jian Sun. "Deep Residual Learning for Image Recognition". arXiv 2015. Shaoqing Ren, Kaiming He, Ross Girshick, & Jian Sun. "Faster R-CNN: Towards Real-Time Object Detection with Region Proposal Networks". NIPS 2015.



## Google Brain



## Building huge neural networks

10 million connections



1 billion connections



10 billion connections

## Desiderata for Learning Al

- Ability to learn complex, highly-varying functions
- Ability to learn multiple levels of abstraction with little human input
- Ability to learn from a very large set of examples
  - Training time linear in the number of examples
- Ability to learn from mostly unlabeled data
  - Unsupervised and semi-supervised
- Multi-task learning
  - Sharing of representations across tasks
- Fast predictions

## **Thanks for Listening**

**Any Questions?**