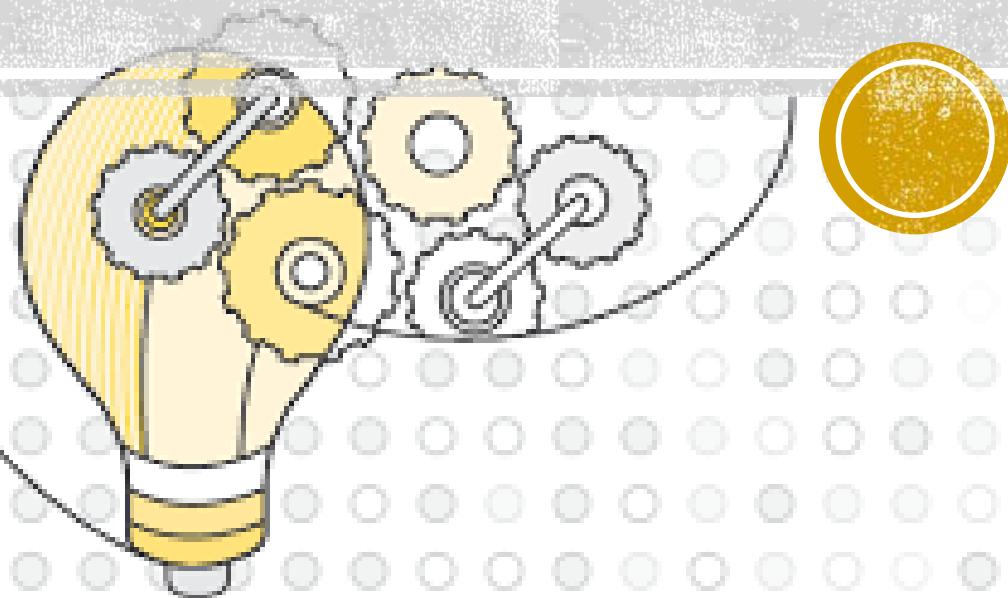
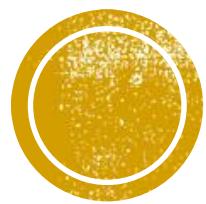
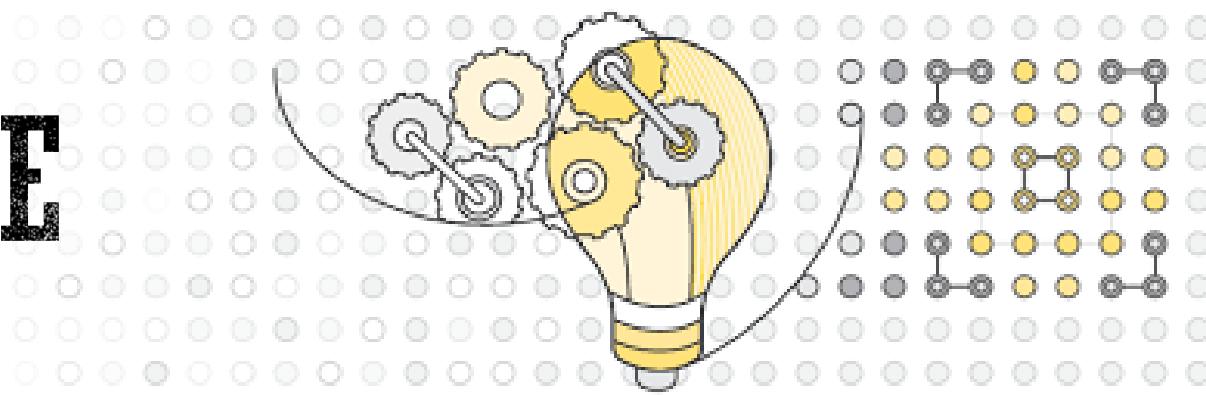


INTRODUCTION





PEOPLE



WHO ARE WE?



Prof. Shaowei Lin
Instructor
Weeks 1-7



Prof. Lu Wei
Instructor
Weeks 8-14

Teaching Assistants

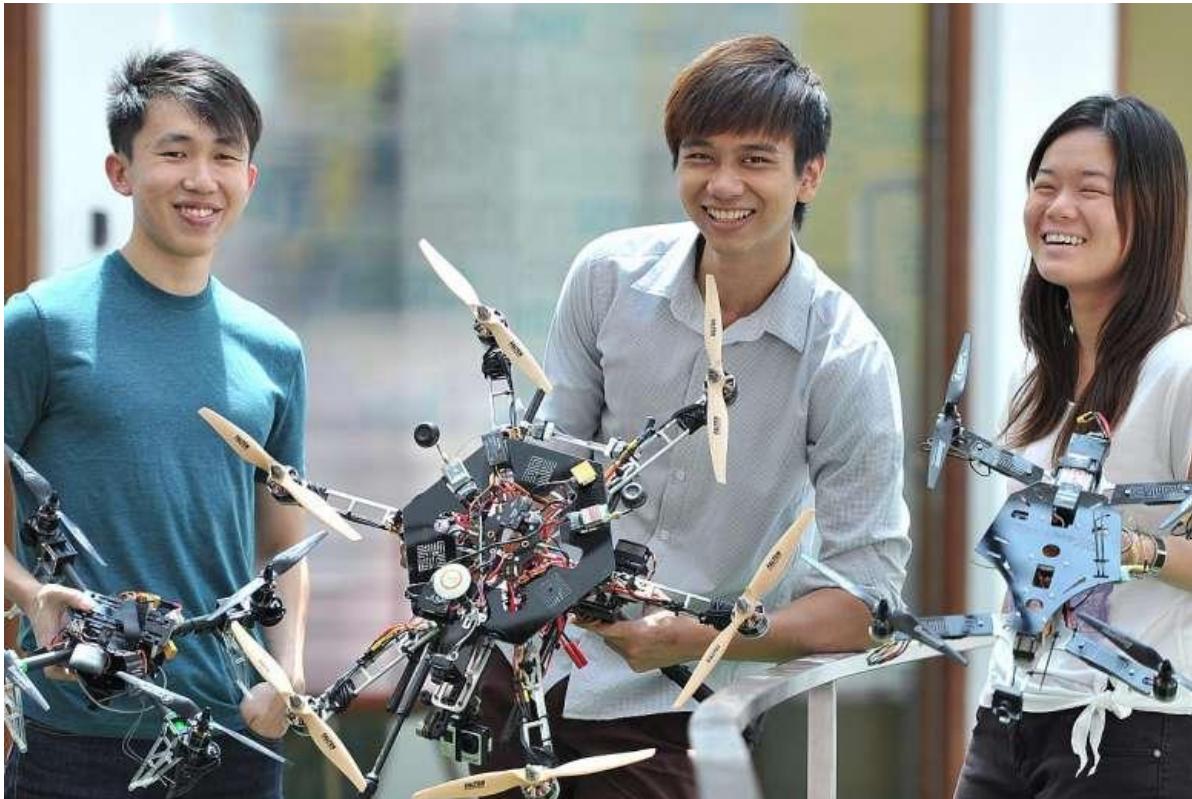
Zhang Yan
Mon PM, Thu AM

Gary Phua
Mon AM, Thu PM

Allan Jie
Mon AM, Thu PM

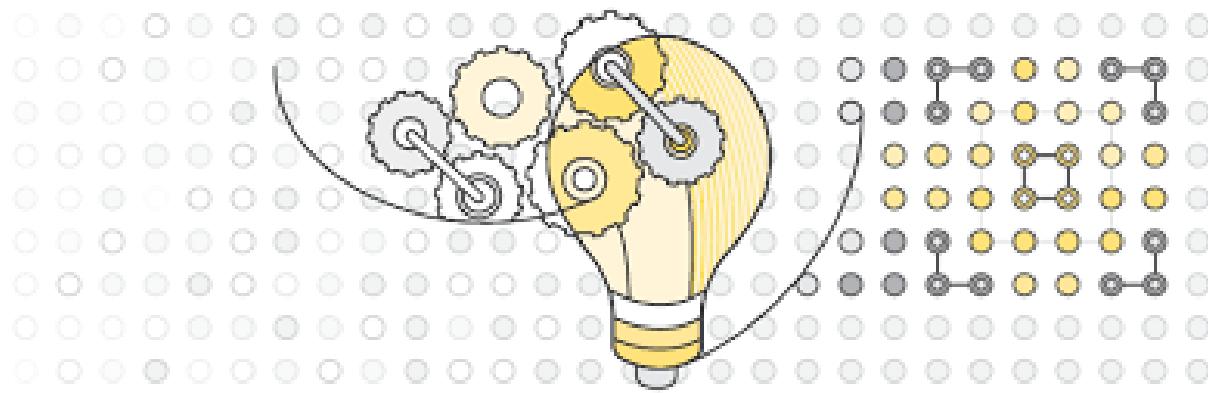


WHO ARE YOU?





CLASS



COURSE INFORMATION

- Office Hours
- Lessons
- Prerequisites
- Assessment
- Schedule
- Textbooks
- Syllabus
- Software
- Project
- Homework
- eDimension
- Observers



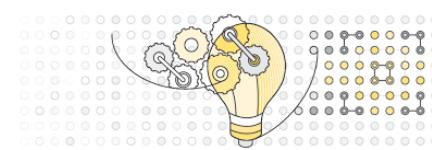
COURSE GOALS

1. Curious to discover more
2. Confident of doing it yourself
3. Contemplative of the theory
4. Cautious of the dangers



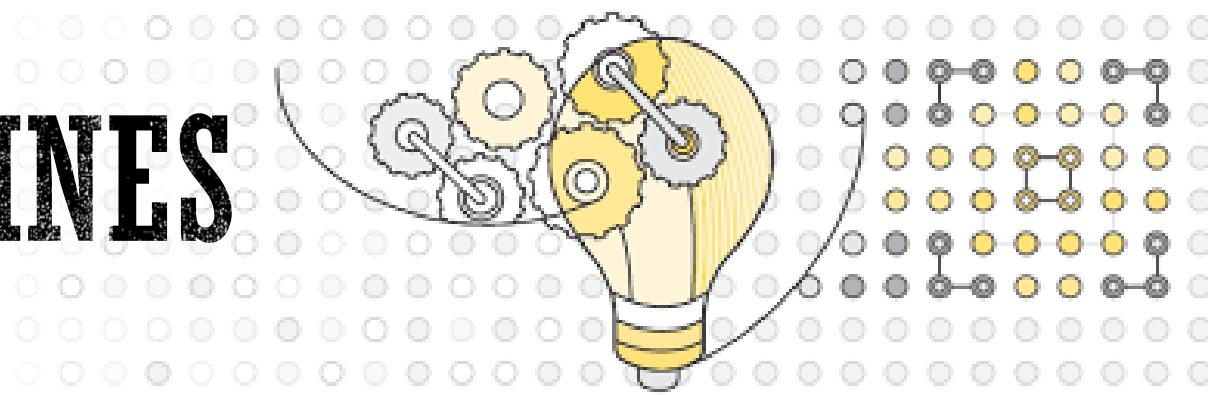
ACKNOWLEDGEMENTS

- MIT 6.036 Introduction to Machine Learning
- SUTD 50.007 Machine Learning (Alex Binder)
- Stanford CS229 Machine Learning





MACHINES



WHAT IS MACHINE LEARNING?



Hard-Coded



Trained

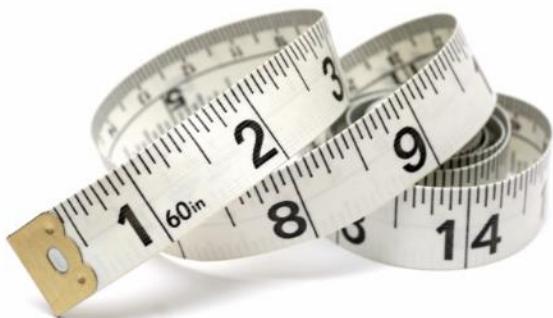
Giving computers the ability to learn
without being explicitly programmed
– Arthur Samuel (1959)



WHAT IS MACHINE LEARNING?



Task



Performance



Experience

Algorithms that improve their performance
at some task with experience

– Tom Mitchell (1998)



TYPES OF MACHINE LEARNING



Supervised Learning



TYPES OF MACHINE LEARNING



Unsupervised Learning



TYPES OF MACHINE LEARNING



Playing is more
fun than watching!

Reinforcement Learning



TYPES OF MACHINE LEARNING

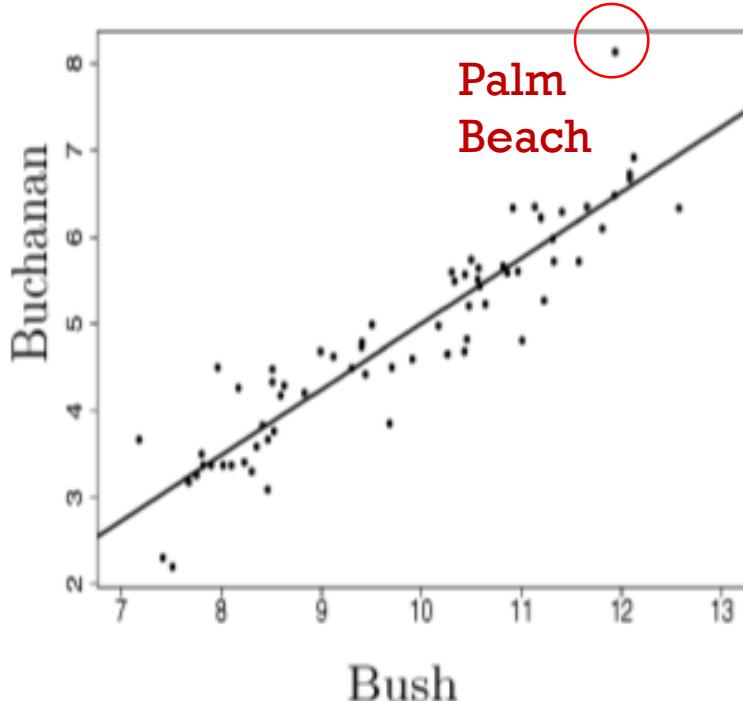


Transfer Learning



SUPERVISED LEARNING

Regression (Linear)



Learning a function

$$y = f(x)$$

$$x \in \mathbb{R}$$

$$y \in \mathbb{R}$$

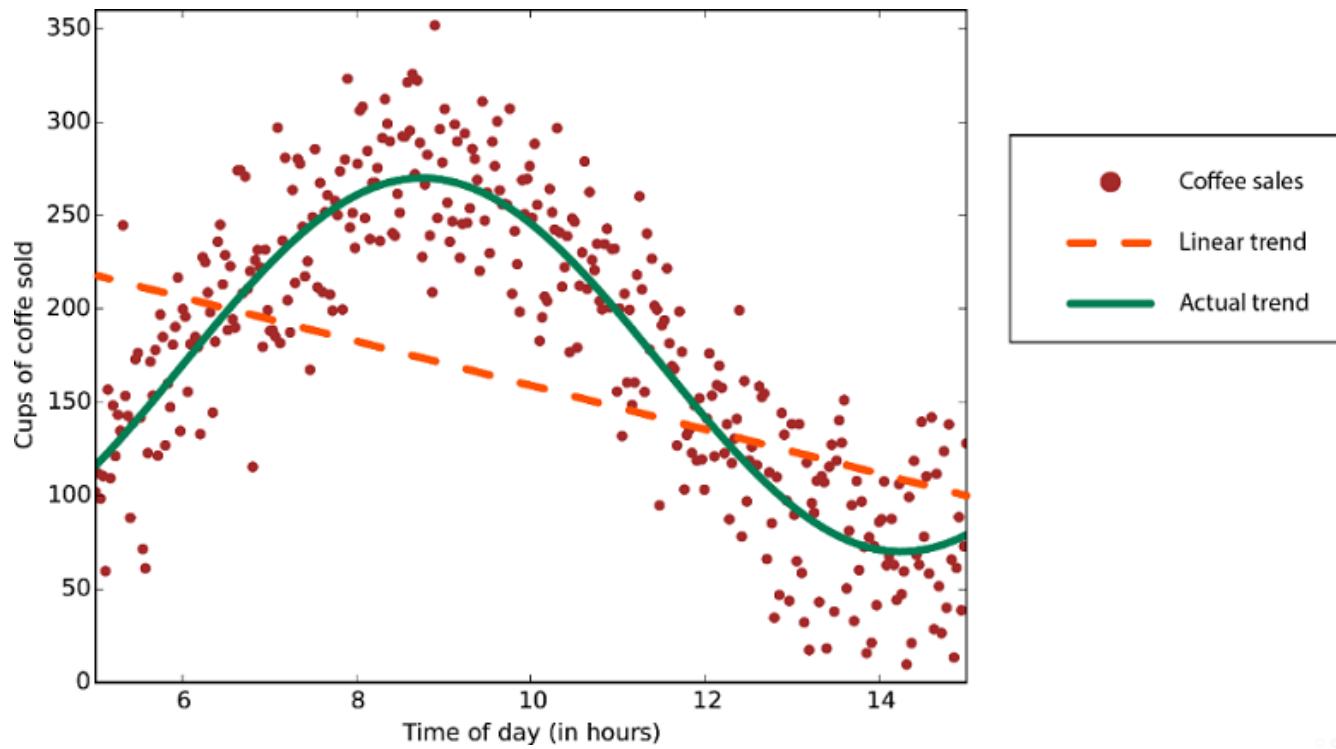
2000 USA Presidential Elections.

Votes for Buchanan and Bush in cities of Florida on a log scale.



SUPERVISED LEARNING

Regression (Non-linear)



SUPERVISED LEARNING

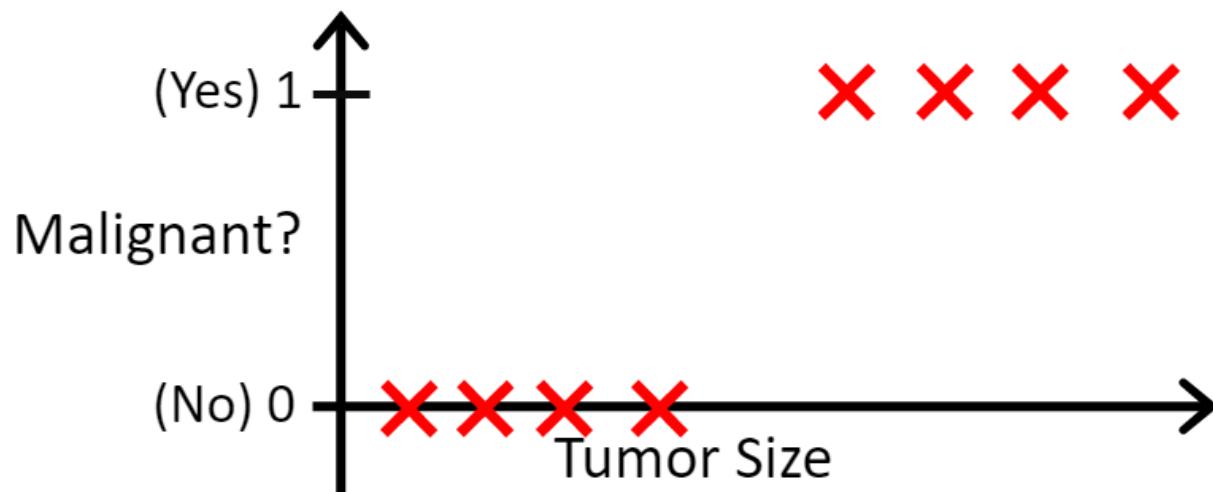
Classification

Learning a function

$$y = f(x)$$

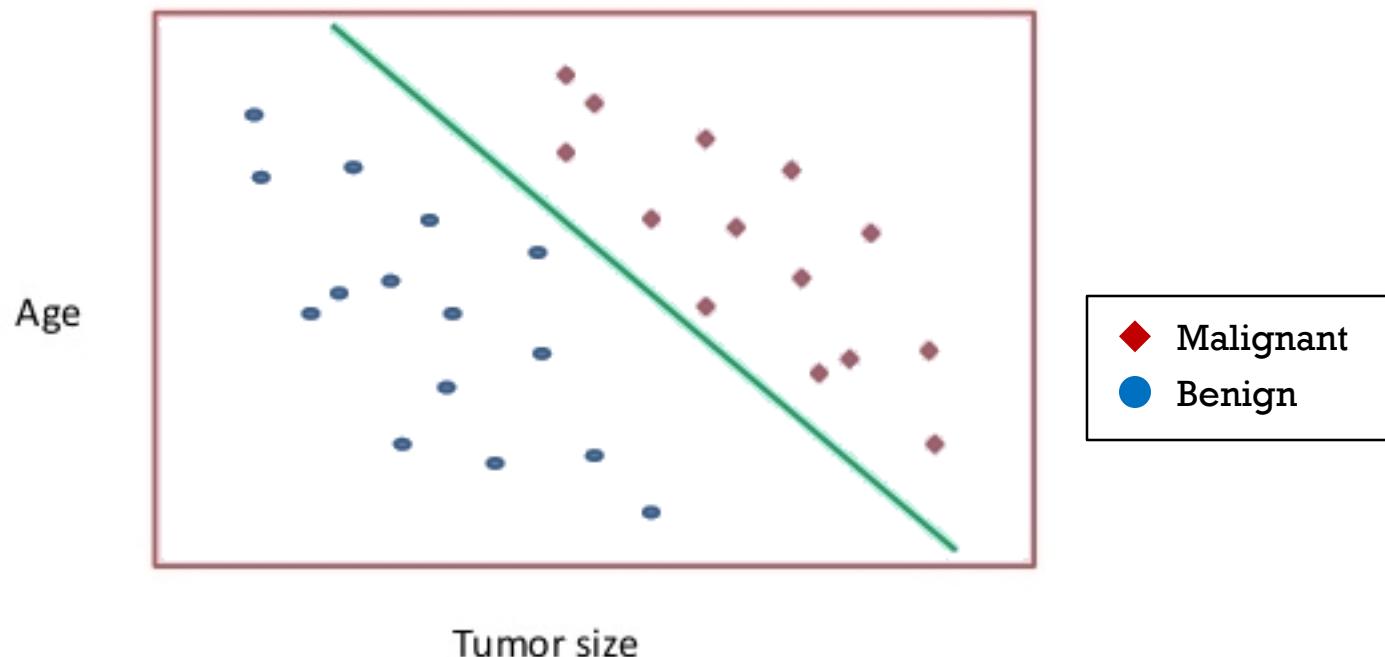
$$x \in \mathbb{R}$$

$$y \in \{1, 2, \dots, k\}$$



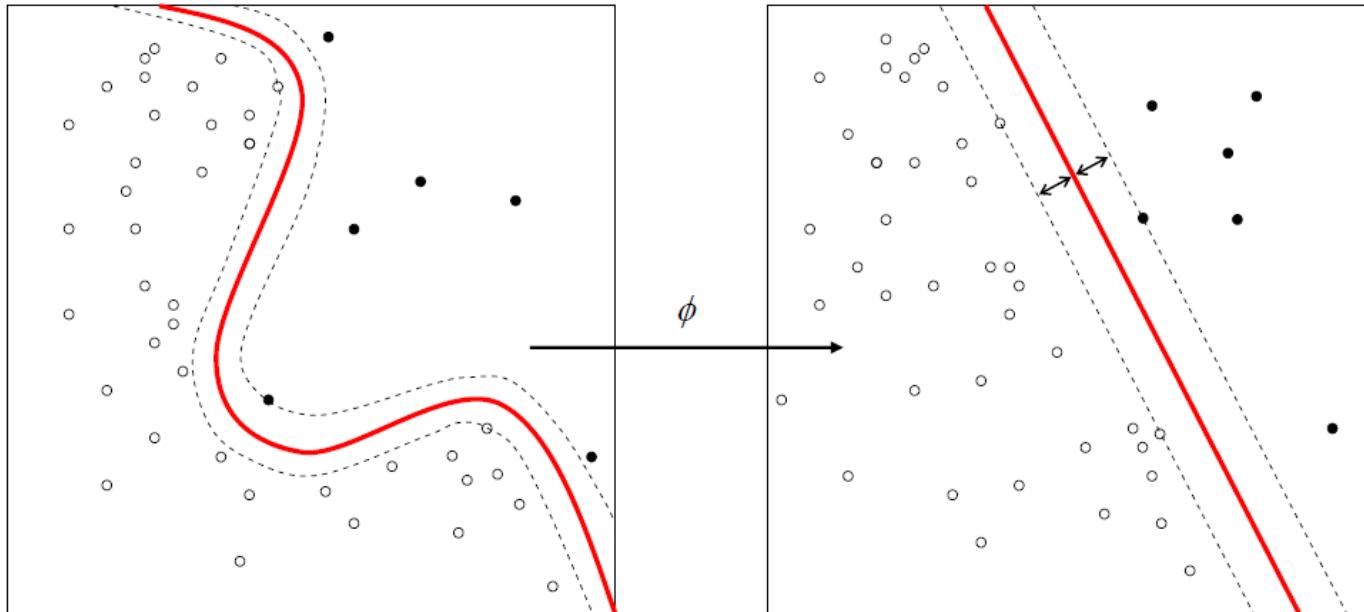
SUPERVISED LEARNING

Classification (Linear)



SUPERVISED LEARNING

Classification (Non-linear)



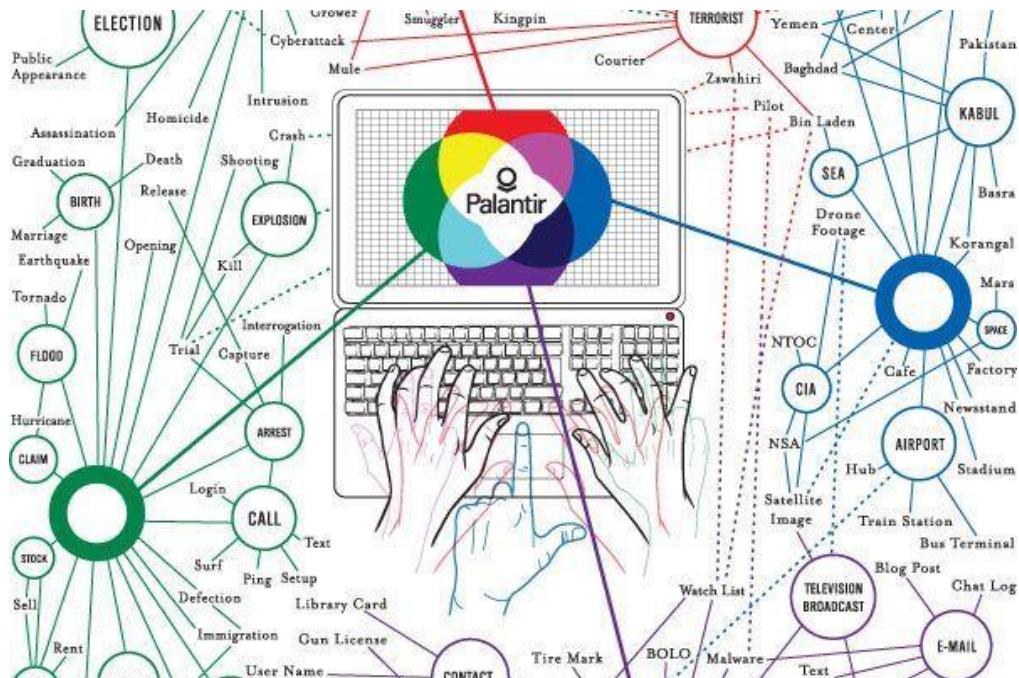
SPAM FILTERS



Bayesian
Networks



FRAUD DETECTION



P PayPal
eBay

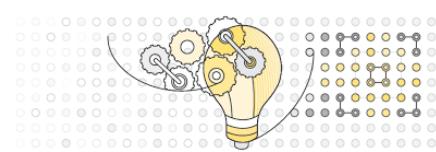
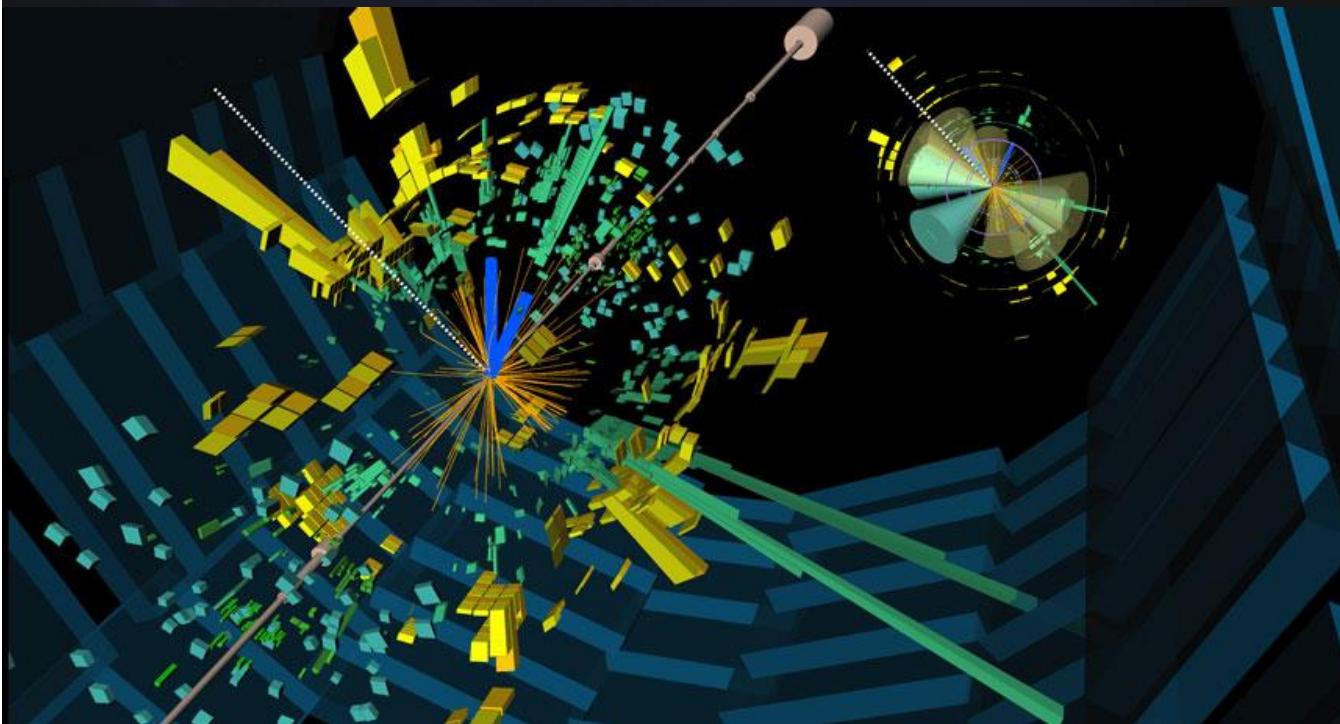


Higgs
challenge

the HiggsML challenge

May to September 2014

When **High Energy Physics** meets **Machine Learning**

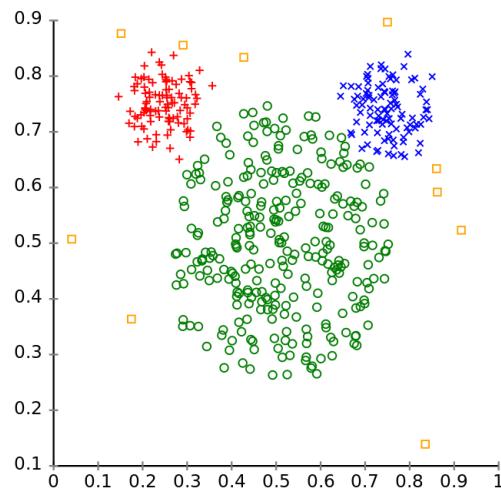


UNSUPERVISED LEARNING

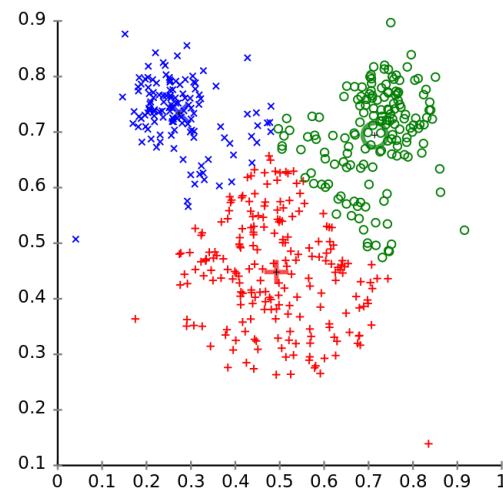
Clustering

Different cluster analysis results on "mouse" data set:

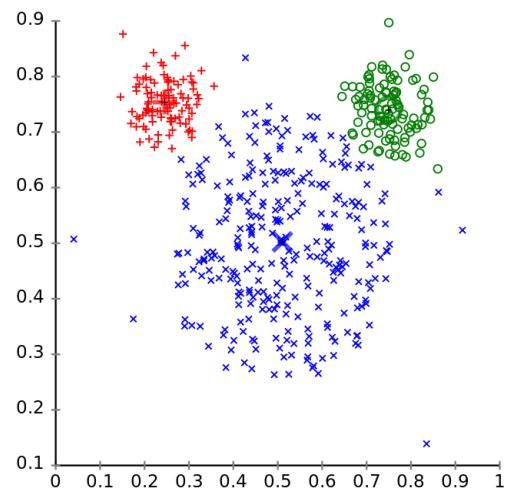
Original Data



k-Means Clustering

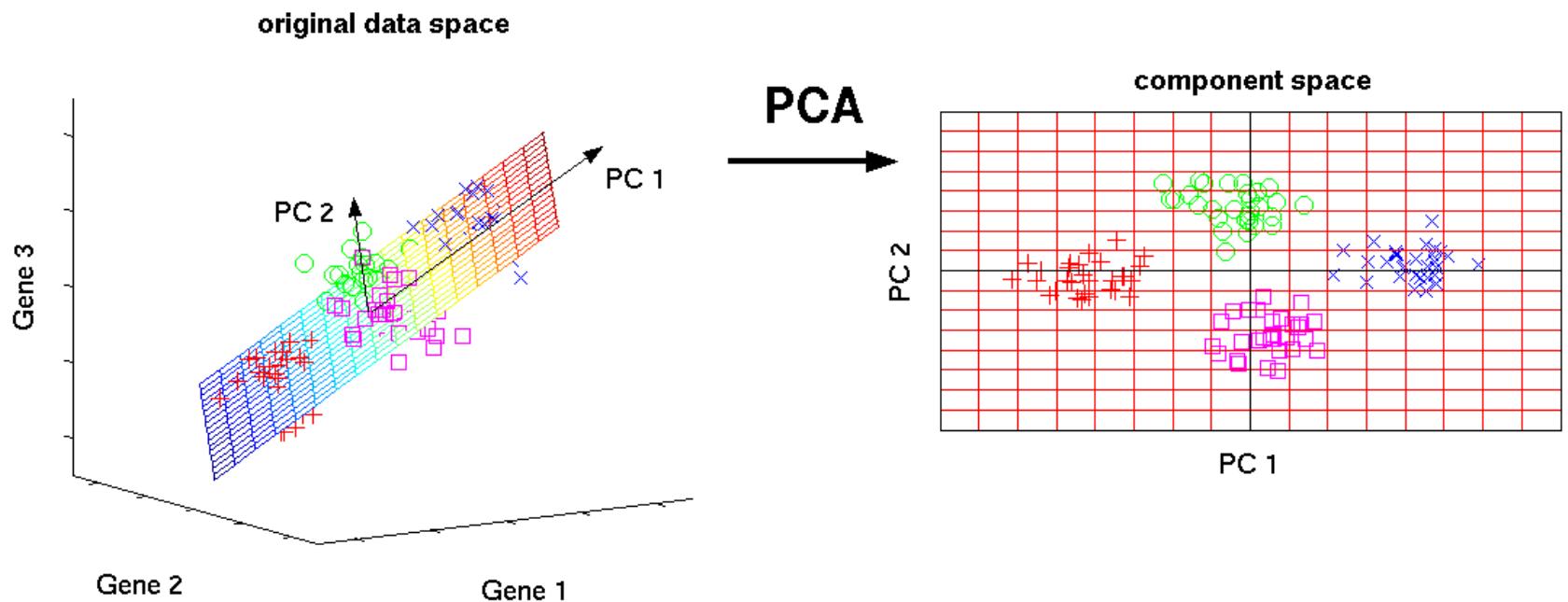


EM Clustering



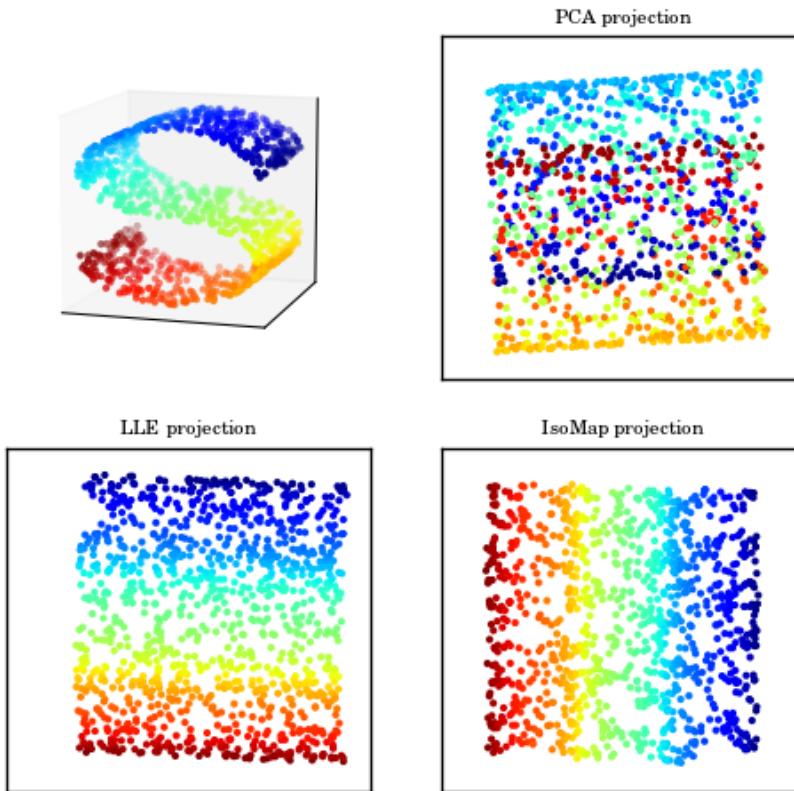
UNSUPERVISED LEARNING

Dimensionality Reduction: Subspace Learning

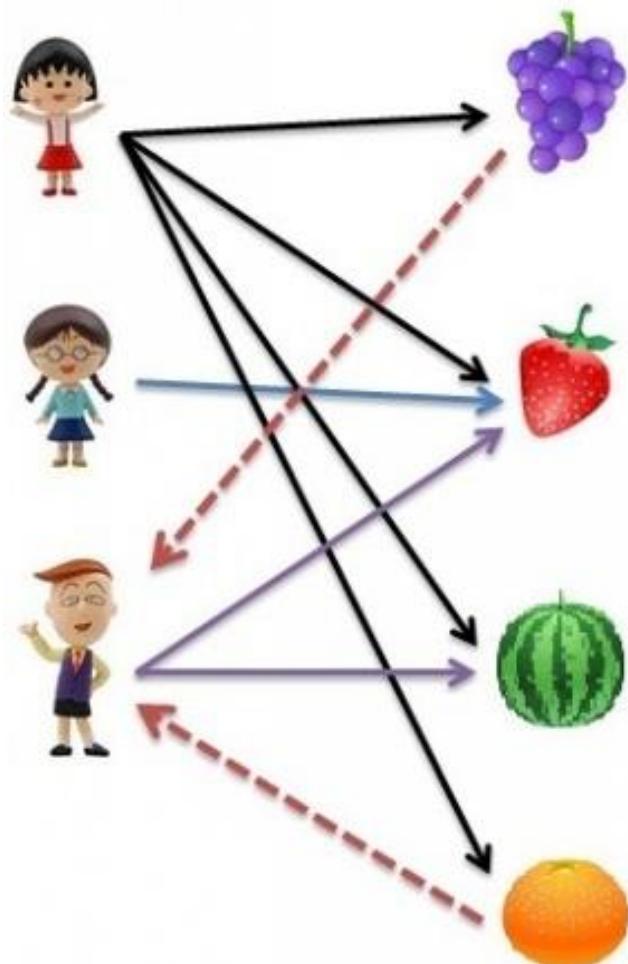


UNSUPERVISED LEARNING

Dimensionality Reduction: Manifold Learning

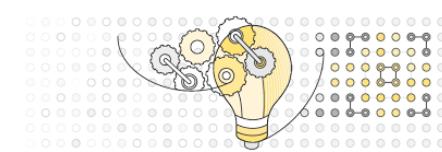


RECOMMENDER SYSTEMS

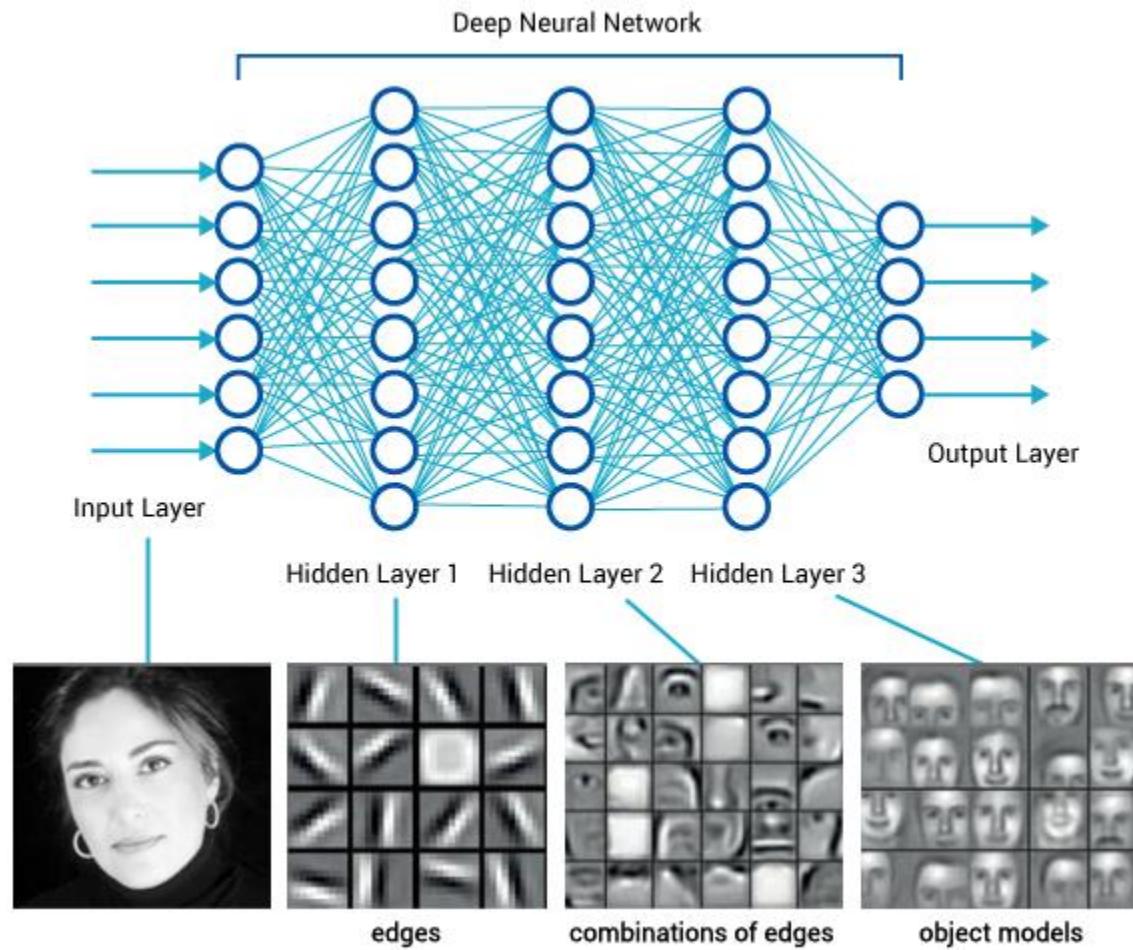


Low-Rank Matrix Factorization

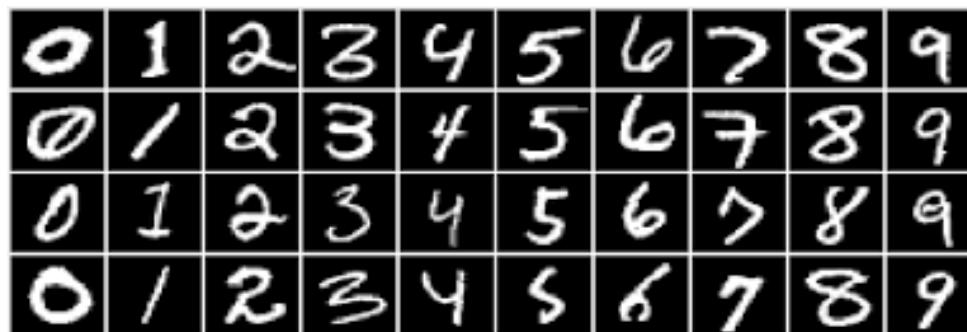
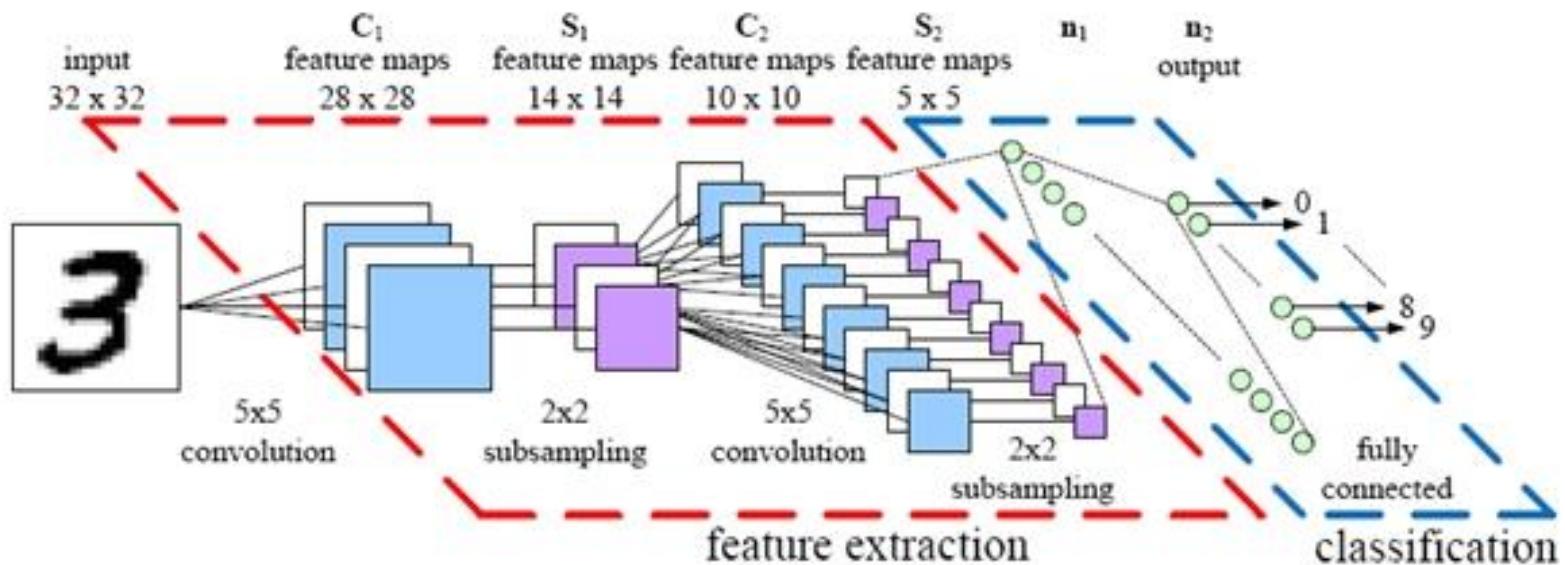
$$\text{Ratings} \approx \begin{matrix} \text{Users} \\ \times \\ \text{Movies} \end{matrix} \begin{matrix} f(i) \\ f(j) \end{matrix}$$



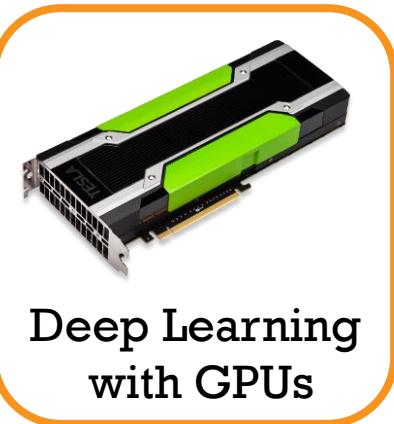
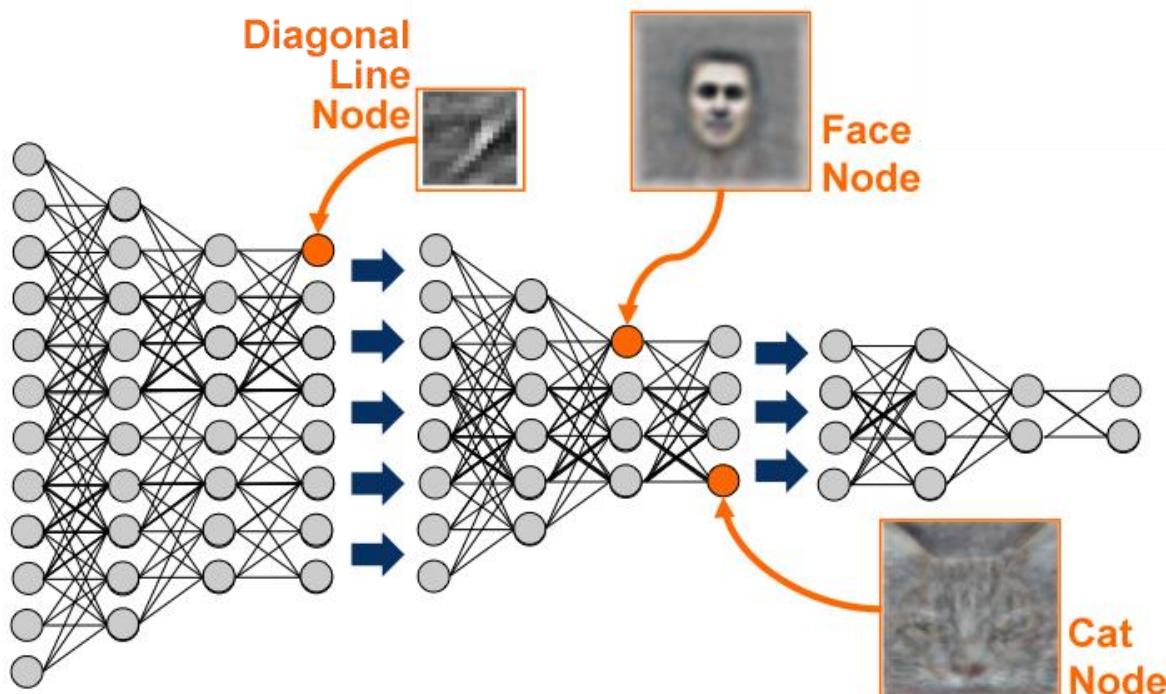
DEEP LEARNING



HANDWRITING RECOGNITION



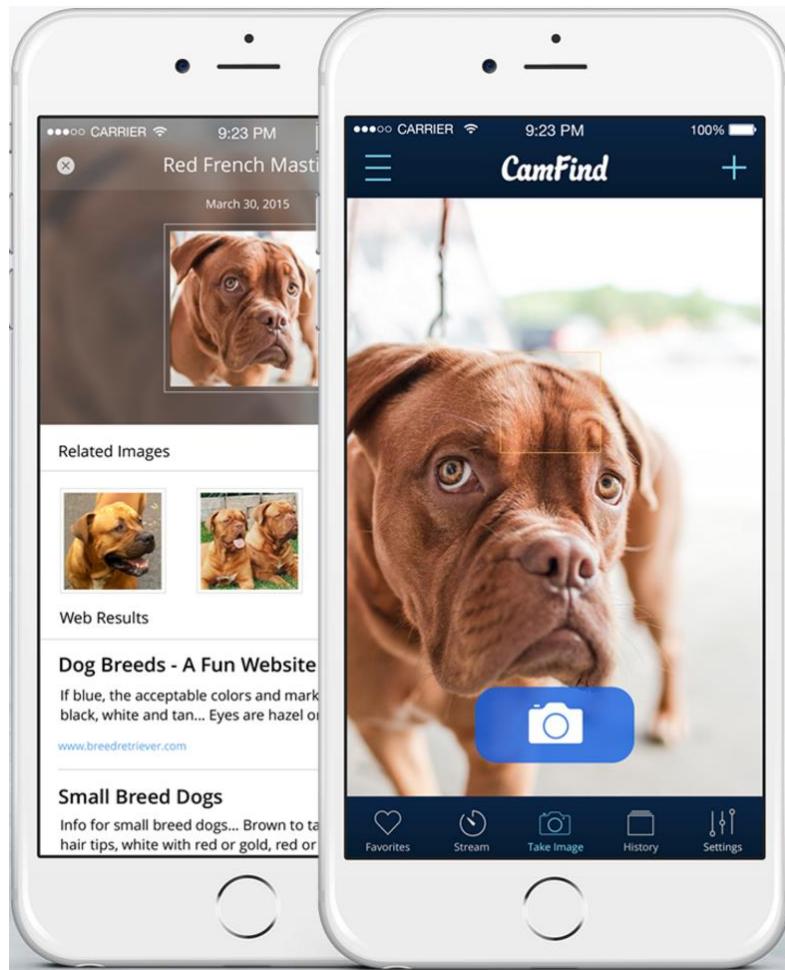
GOOGLE CAT VIDEOS



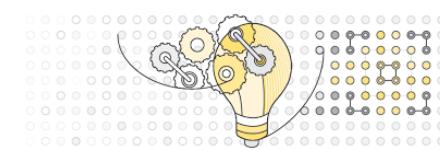
Deep Learning
with GPUs



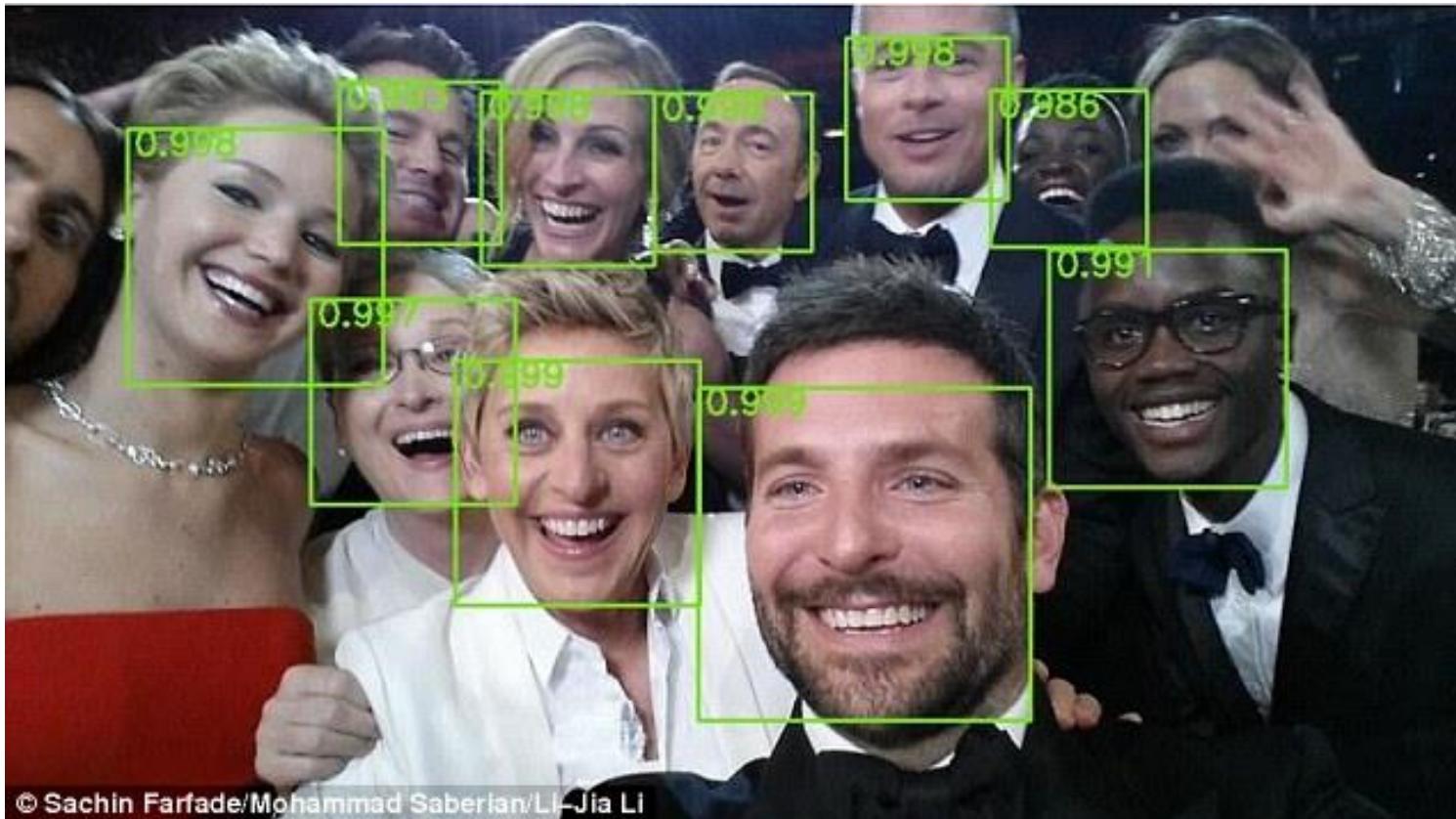
IMAGE RECOGNITION



CamFind
Visual Search Engine
(available on iOS, Android)



FACE RECOGNITION



SPEECH TRANSLATION



From Hidden Markov Models to Recurrent Neural Networks



HEALTHCARE

Watson correctly diagnoses woman after doctors were stumped

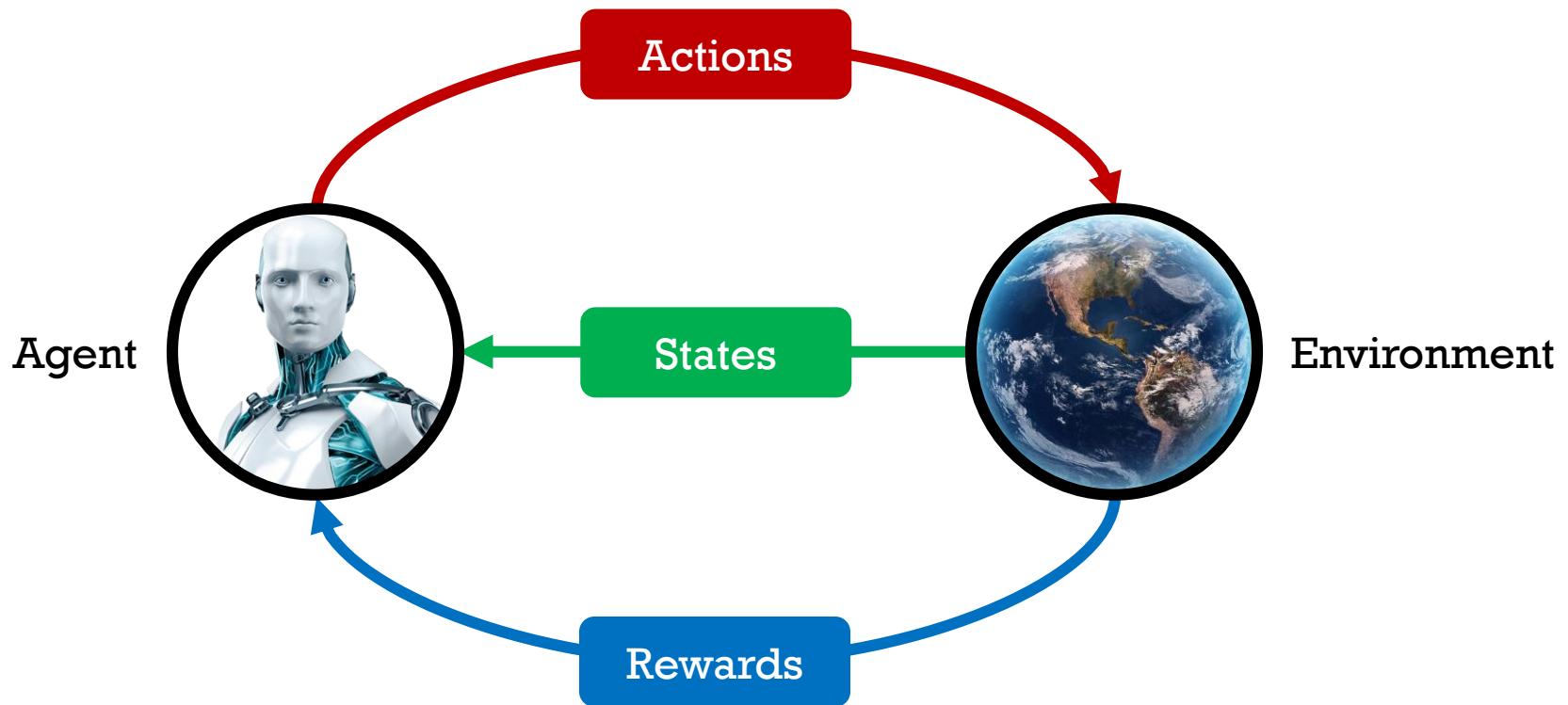
by Eric David | Aug 5, 2016 | 0 comments



After treatment for a woman suffering from leukemia proved ineffective, a team of Japanese doctors turned to IBM's Watson for help, which was able to successfully determine that she actually suffered from a different, rare form of leukemia than the doctors had originally believed.



REINFORCEMENT LEARNING



ATARI GAMES



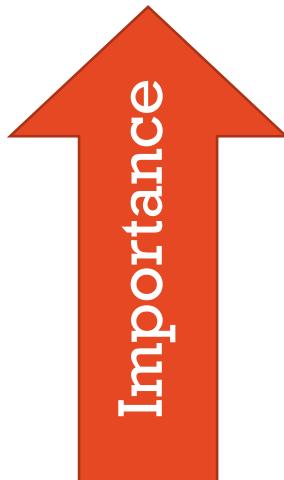
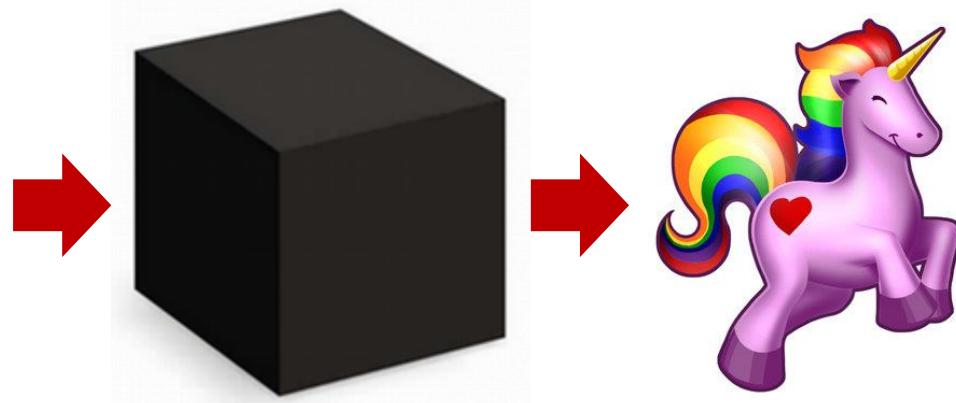
Google DEEPMIND



SELF-DRIVING CARS



NOT A BLACK BOX!



More Structure
More Data
Better Machines
Better Algorithms



DETECTING TANKS

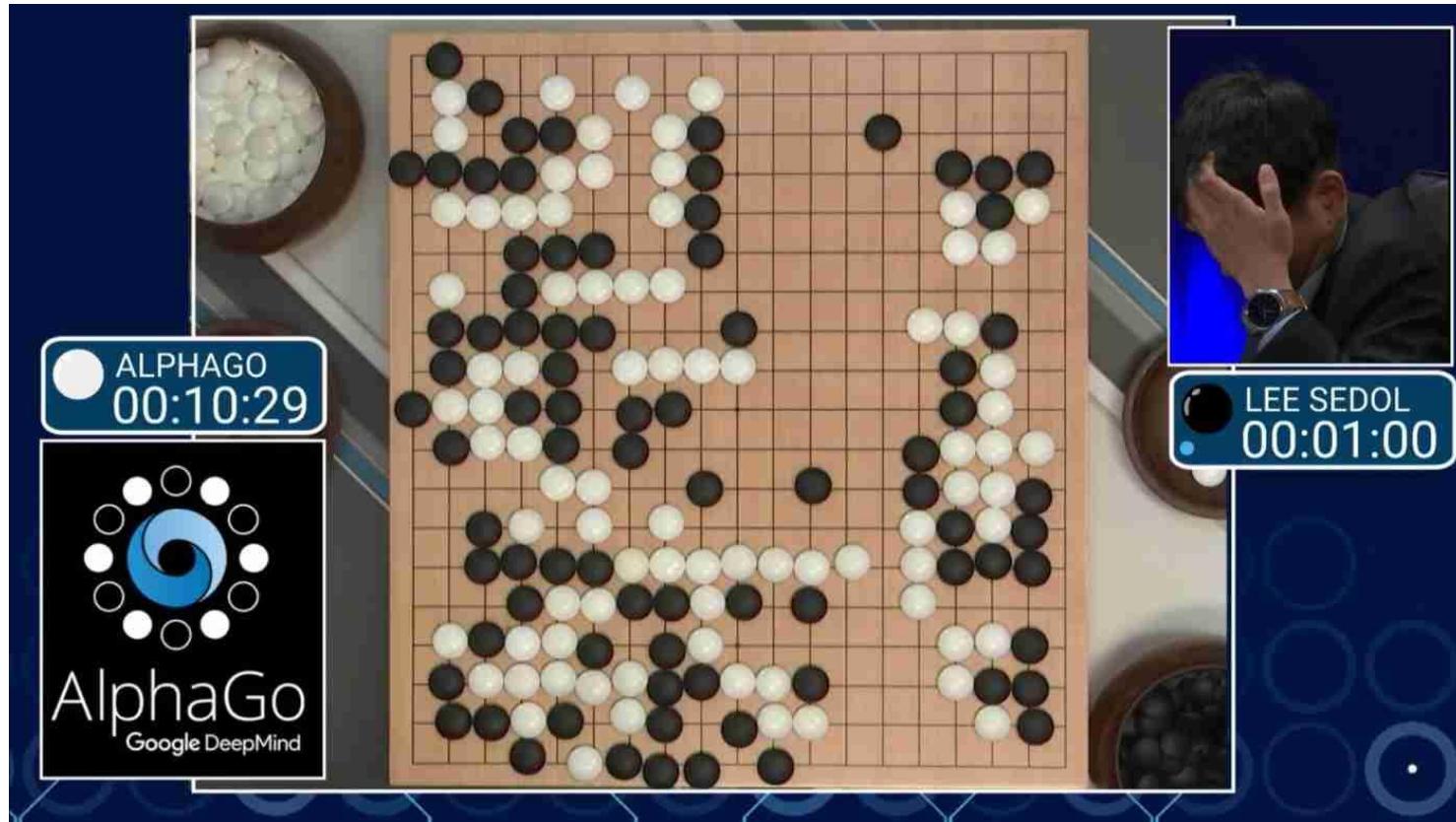


http://lesswrong.com/lw/7qz/machine_learning_and_unintended_consequences/

<https://www.jefftk.com/p/detecting-tanks>



ALPHAGO



INTENDED LEARNING OUTCOMES

Introduction

- Define machine learning in terms of algorithms, tasks, performance and experience.
- List four main types of machine learning, e.g. supervised, unsupervised, reinforcement, and transfer learning.
- Describe some potential dangers in machine learning, e.g. applying an algorithm without understanding its assumptions, forgetting that the training data could be biased.

