

*COMP 1433: Introduction to Data Analytics &  
COMP 1003: Statistical Tools and Applications*

# Lecture 11 – Machine Learning: An Introduction

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The Hong Kong Polytechnic University

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# Roadmap

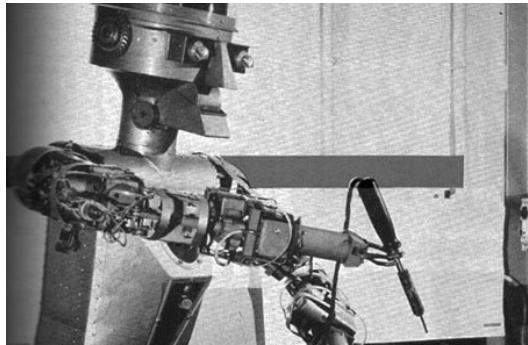
- Artificial Intelligence (*AI*) vs. Machine Learning (*ML*)
- Machine Learning Taxonomy
  - Supervised Learning, Unsupervised Learning, and Reinforcement Learning
- Important AI Areas based on ML methods:
  - Computer Vision (*See*)
  - Speech Processing (*Listen and Speak*)
  - Natural Language Processing (*Read and Write*)
- More ML applications

# Roadmap

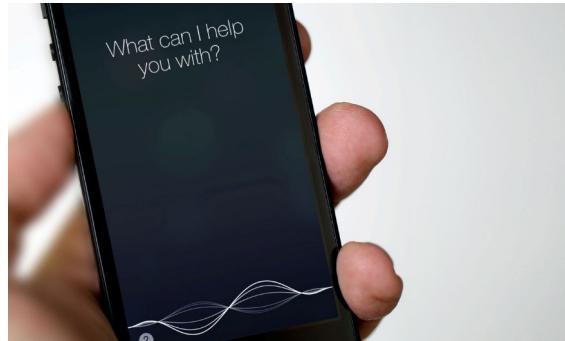
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# What is Artificial Intelligence (AI)?

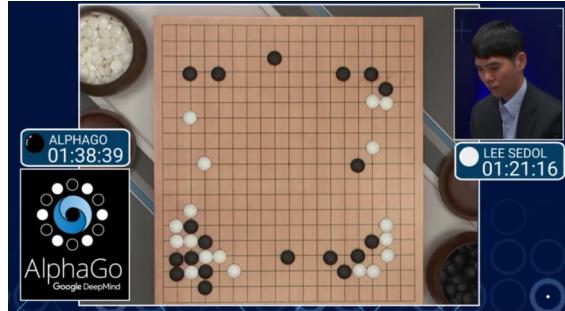
1950s-1990s



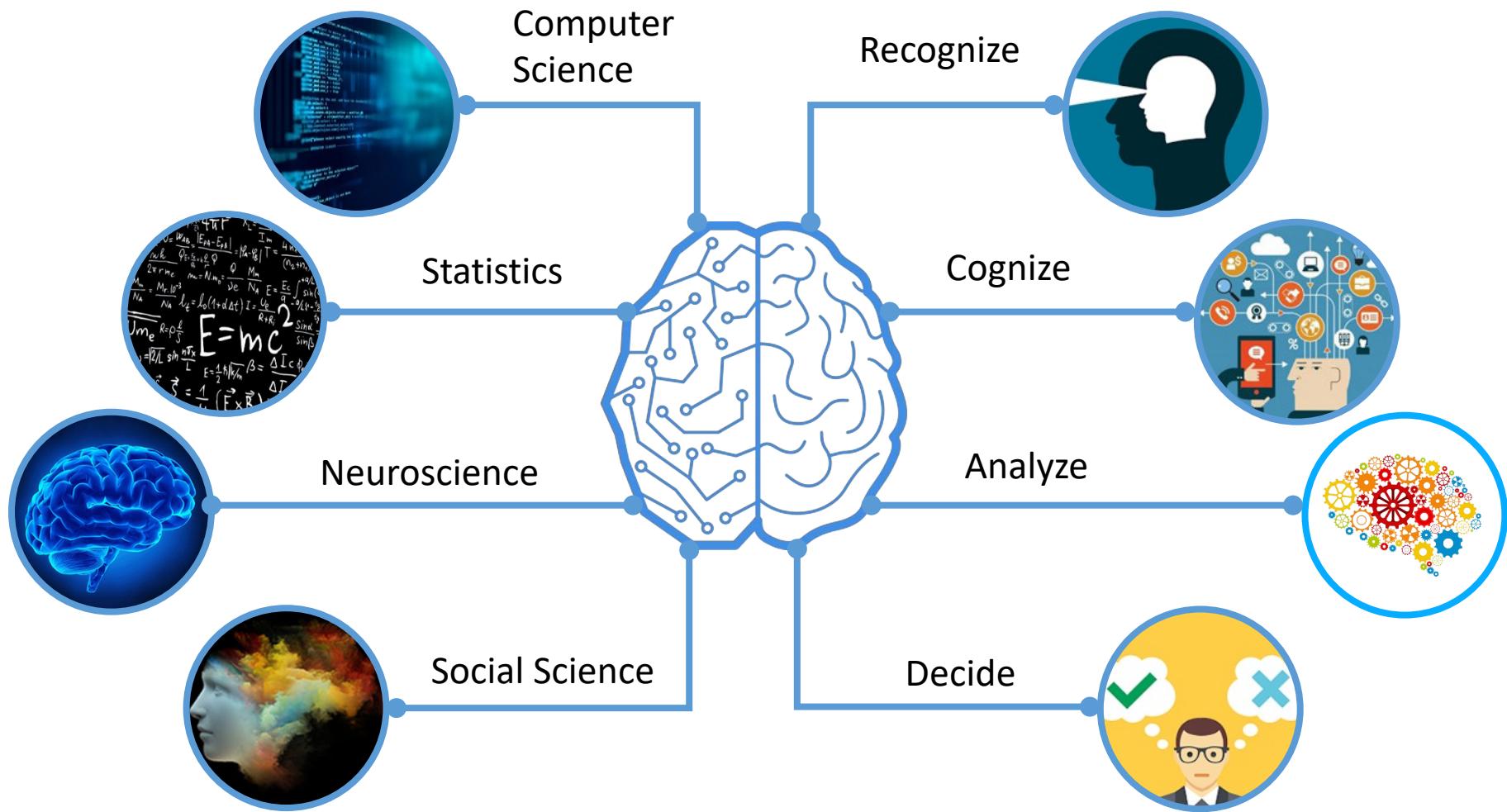
2000s-2010s



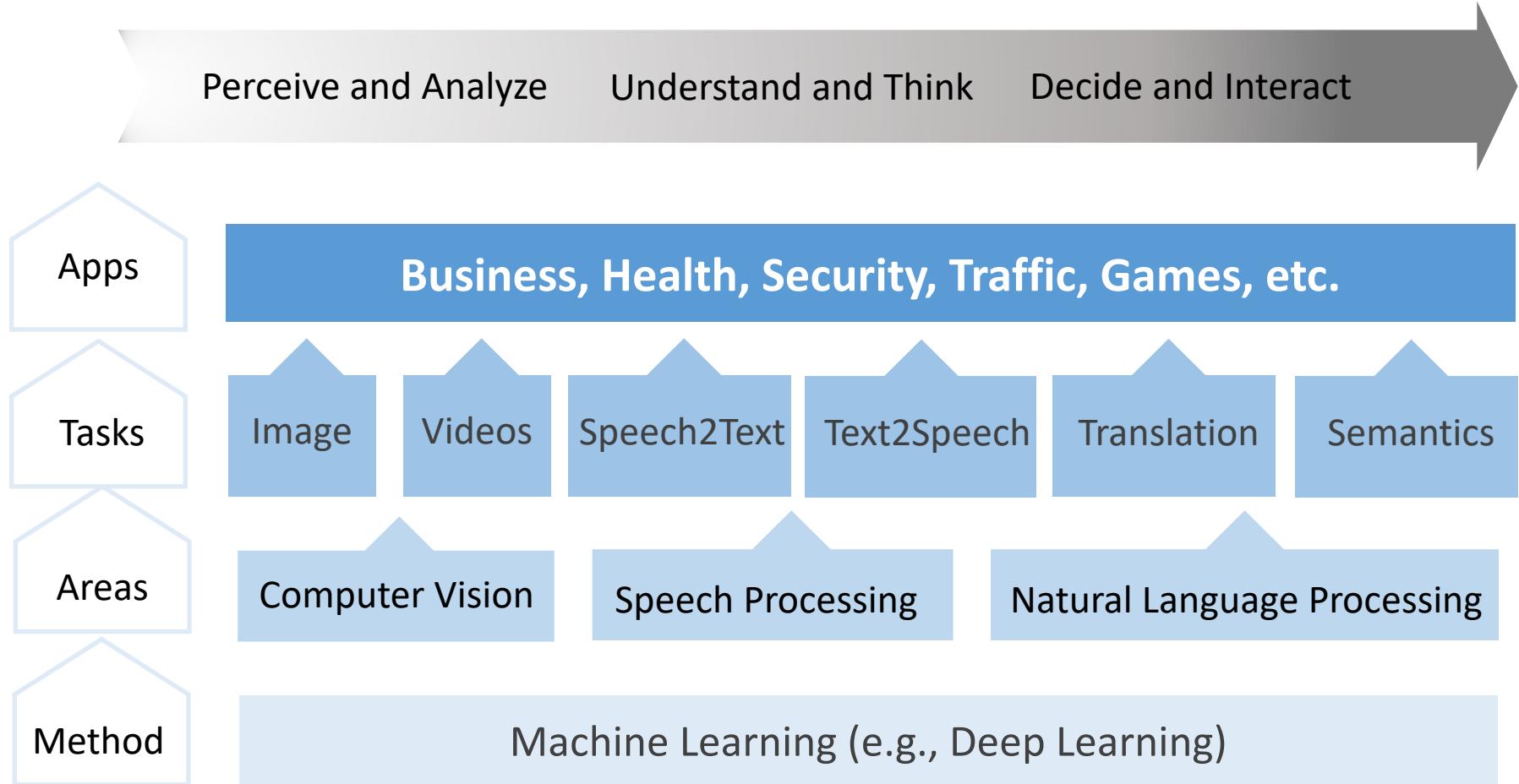
???



# Multidisciplinary in AI

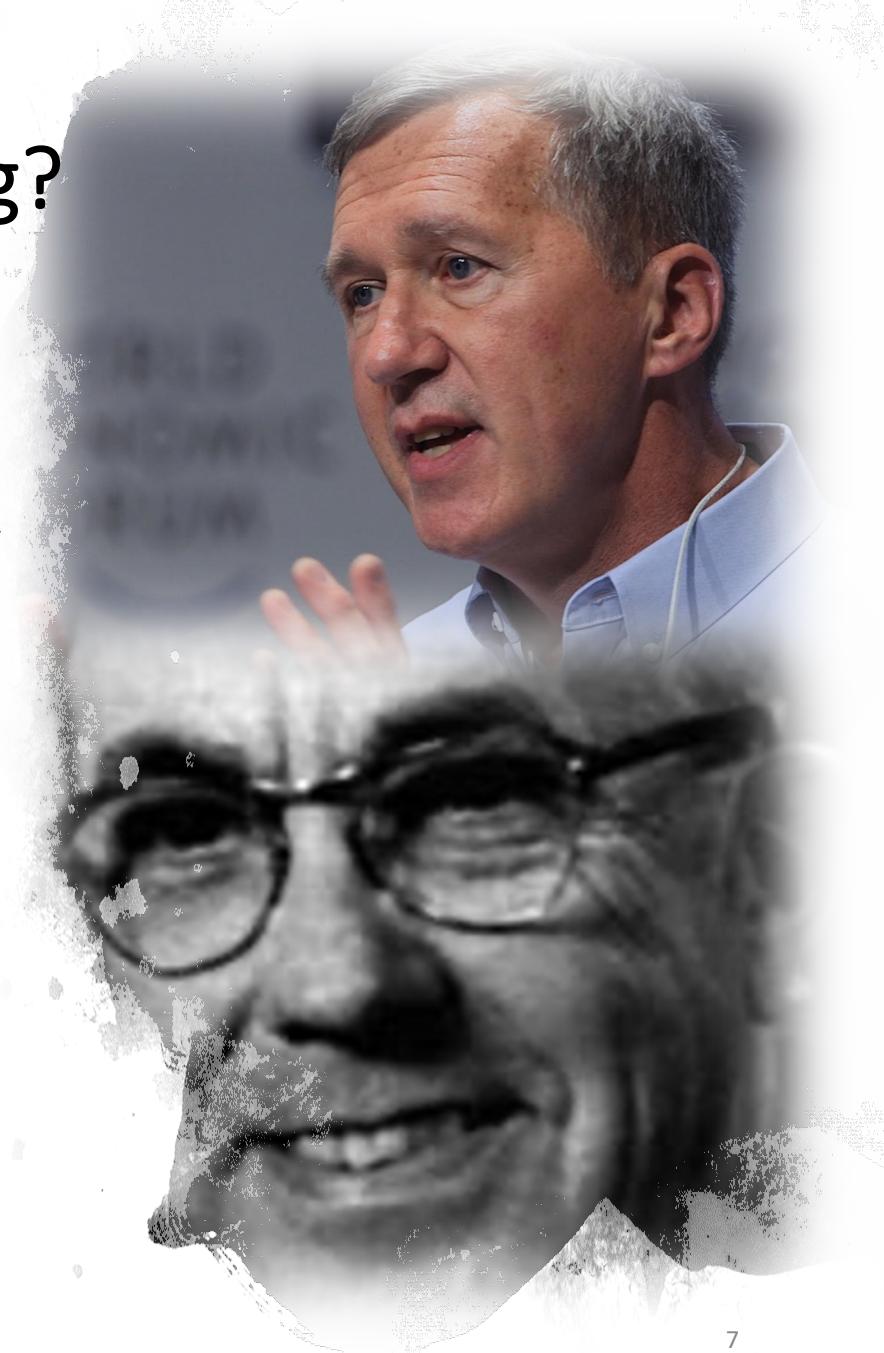


# Varying Tasks in AI

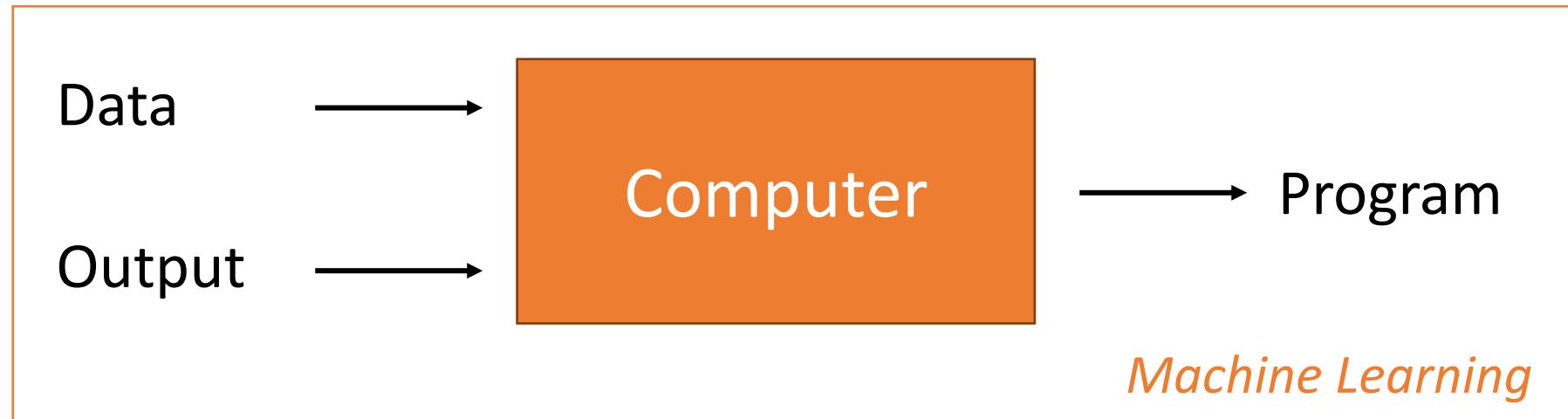
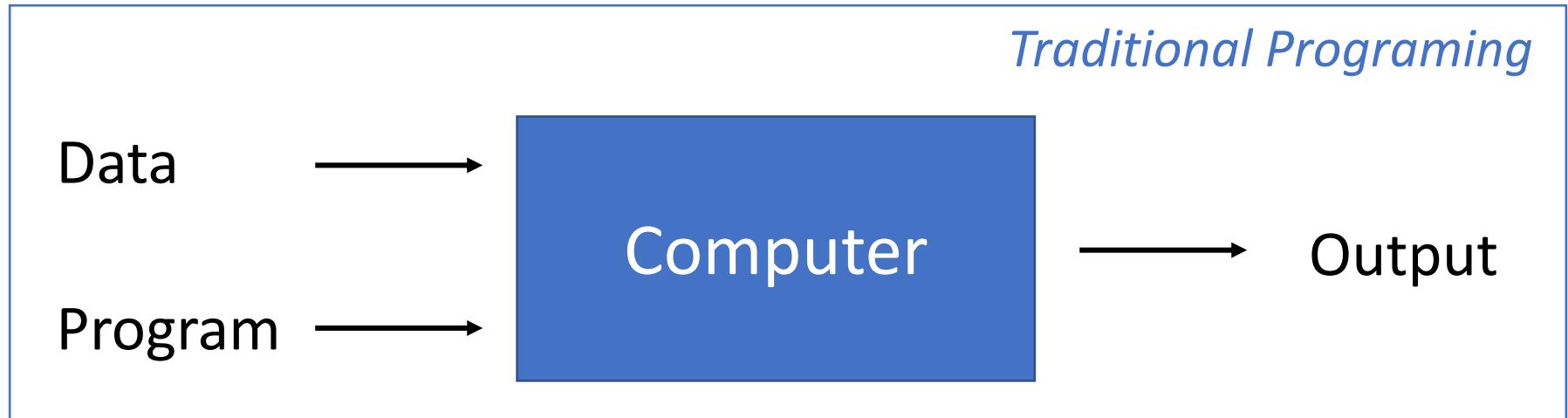


# What is Machine Learning?

- Machine Learning is the field of study that gives the computer *the ability to learn without being explicitly programmed*. --- By *Arthur Samuel* (1959):
- Definition by *Tom Mitchell* in 1998:
  - Machine Learning is the study of algorithms that
    - Improve their *performance*  $P$
    - At some *task*  $T$
    - With *experience*  $E$
  - Hence a well-defined learning task is given by three elements  $\langle P, T, E \rangle$ .

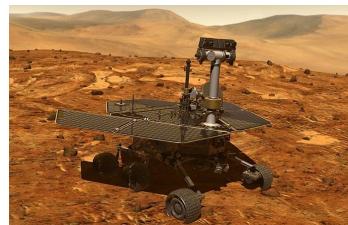


# Machine Learning Process



# When machine learning is helpful?

- We let machine to learn when:
  - Human expertise does not exist, e.g., navigate on Mars.
  - Human cannot explain their expertise, e.g., speech recognition.
  - Models must be customized, e.g., personalized recommendation.
  - Something needs to be learned from huge volume of data, e.g., social media fake news detection.



- Machine learning may sometimes not help:
  - There is no need to “learn” to calculate the scores.

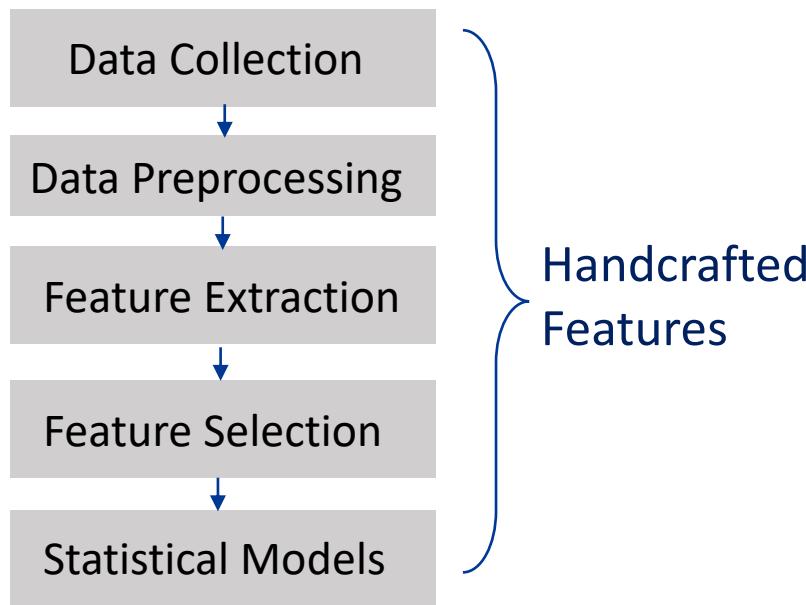
# Example task with machine learning



Hard to say  
what makes  
a hand-  
written 2.

Figure created by *Geoffrey Hinton*

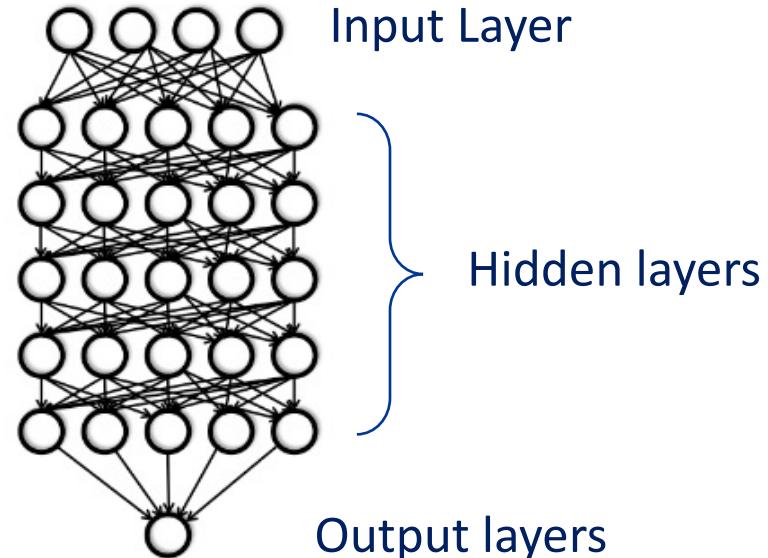
# How to learn?



Classical Machine Learning

Feature Extraction to Fit Data

- **Machine Learning**: Allows machine to learn new things from the data.
- **Deep Learning**: More complex model to fit large-scale data.



A *multi-layer perceptron model*

Deep Learning

Automatic Feature Learning

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# Taxonomy of Machine Learning Methods

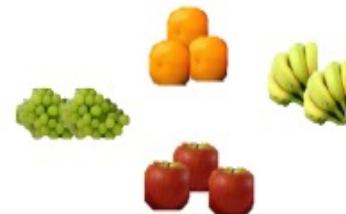
Supervision

Unsupervised Learning

Unlabeled Data



Clusters

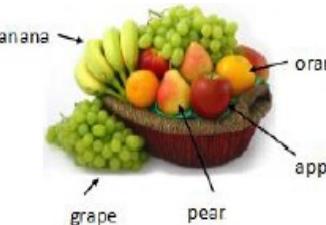


Topic Discovery



Supervised Learning

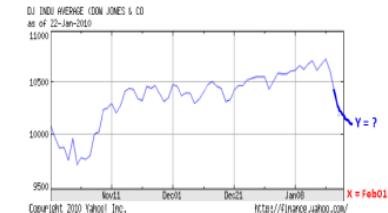
Labeled Data



Predict Labels

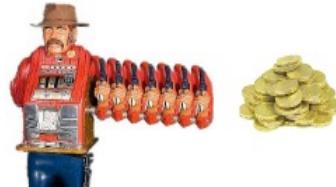


Price Prediction



Reinforcement Learning

Rewards



Good Actions

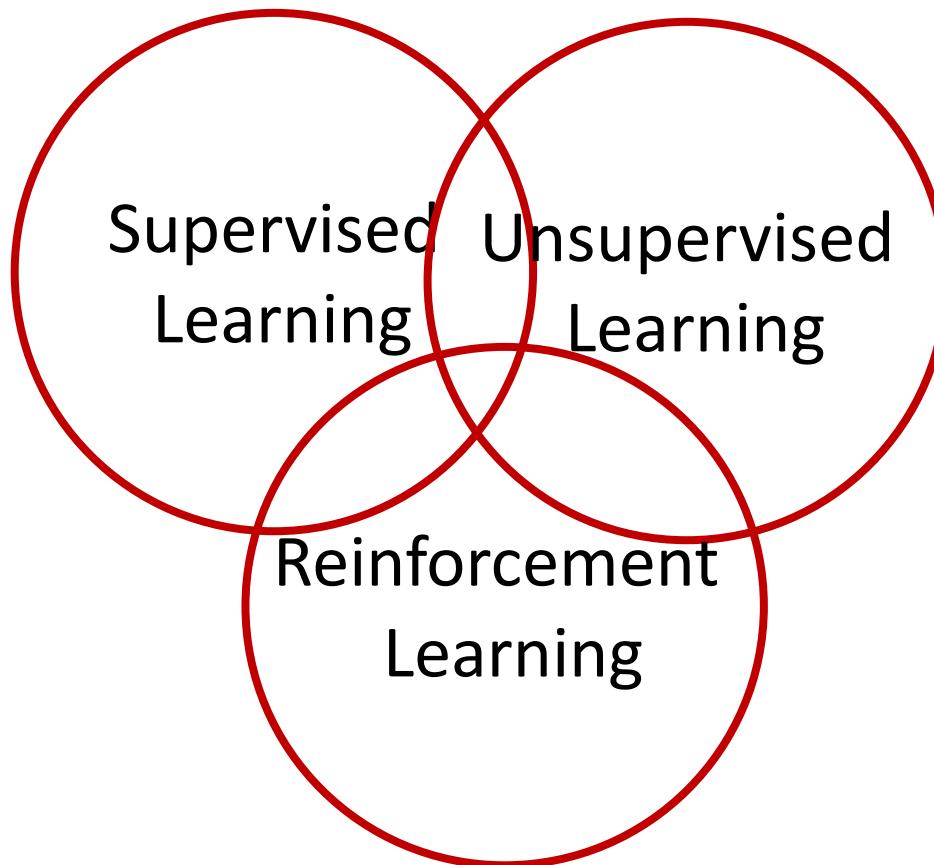


Dog Training



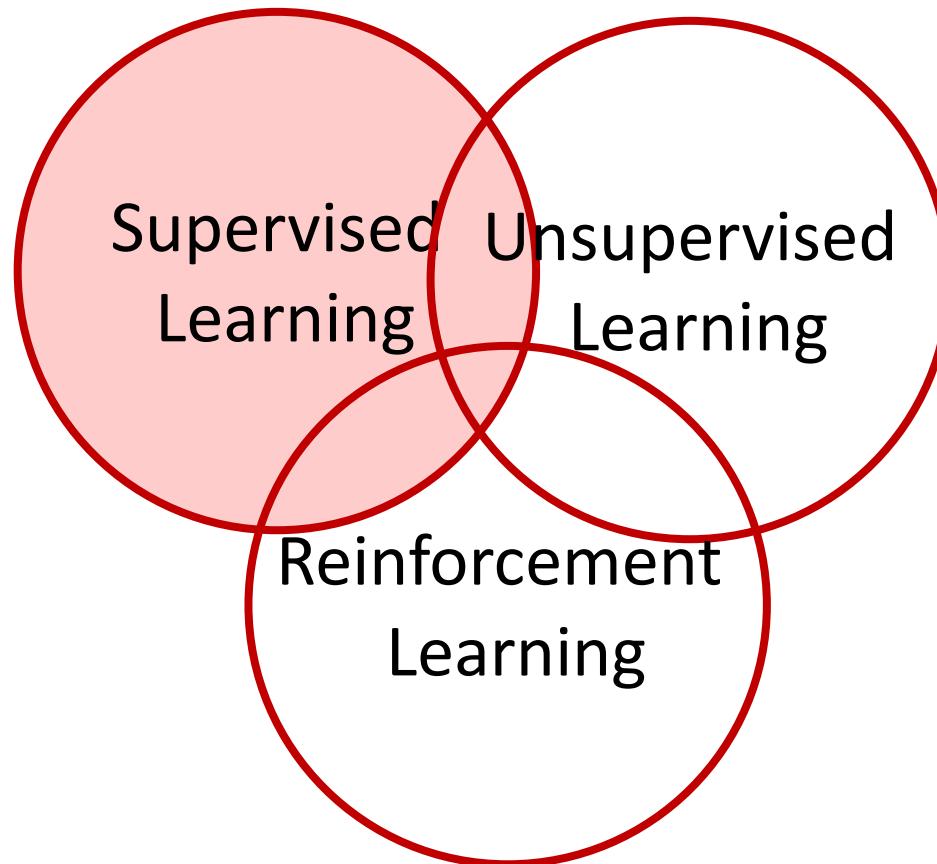
# Taxonomy of Machine Learning Methods

A simplified view based on tasks



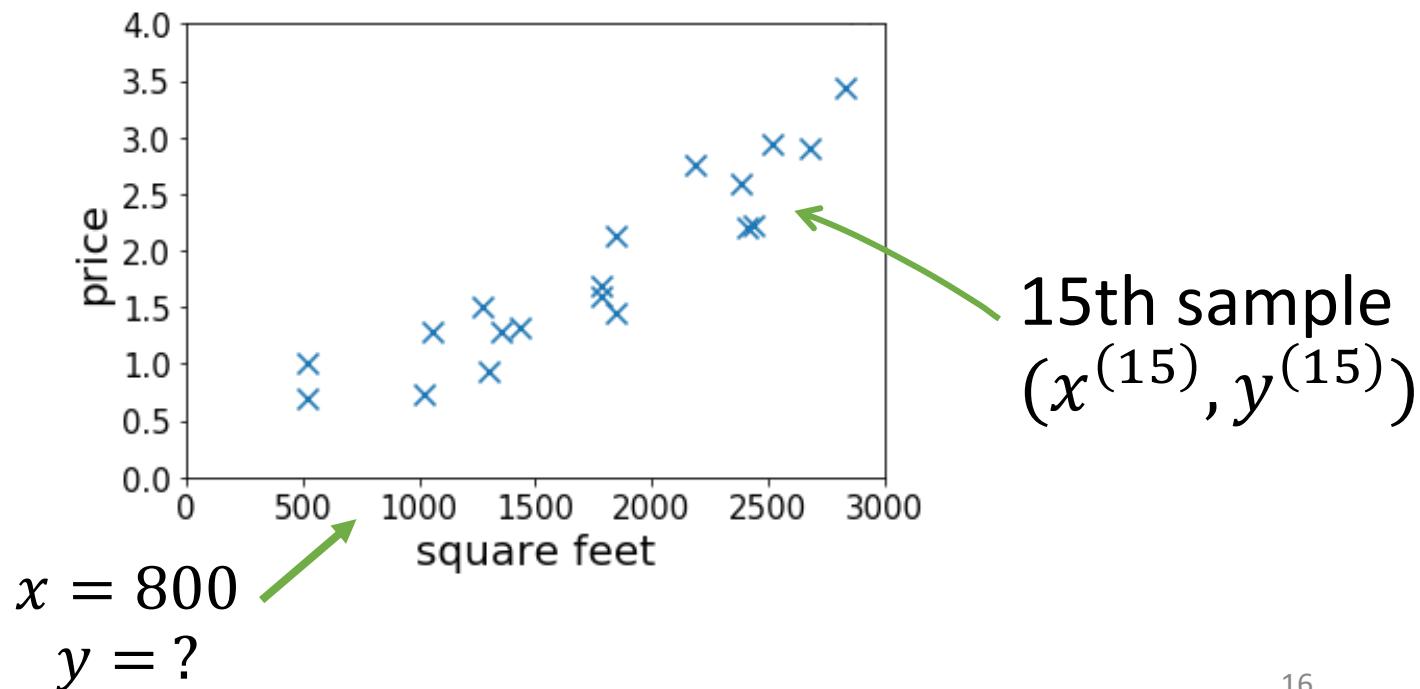
# Taxonomy of Machine Learning Methods

A simplified view based on tasks



# Supervised Learning (Regression)

- Given: a dataset that contains  $n$  samples
$$(x^{(1)}, y^{(1)}), \dots (x^{(n)}, y^{(n)})$$
- House Price Prediction Task:** if a guy has a house with  $x$  square feet, how to predict its price?

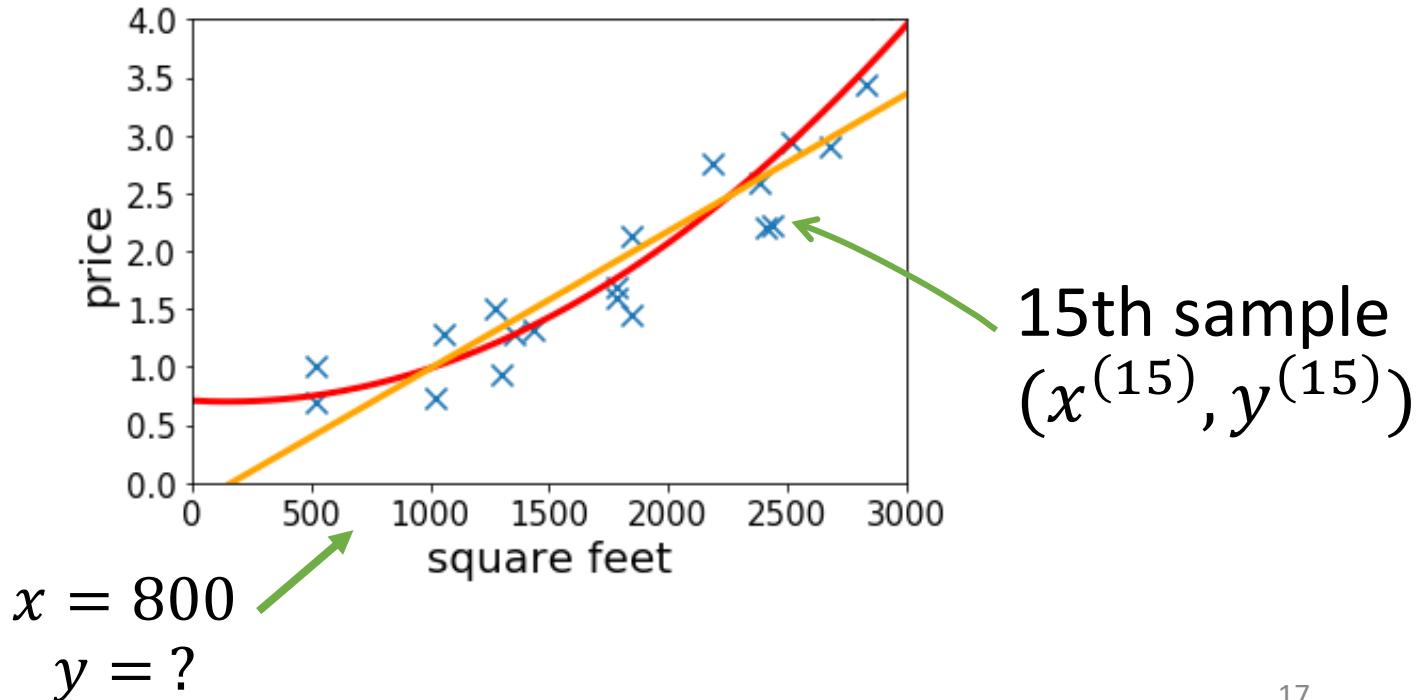


# Supervised Learning (Regression)

- Given: a dataset that contains  $n$  samples

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- House Price Prediction Task:** if a guy has a house with  $x$  square feet, how to predict its price?



# Supervised Learning (More Features)

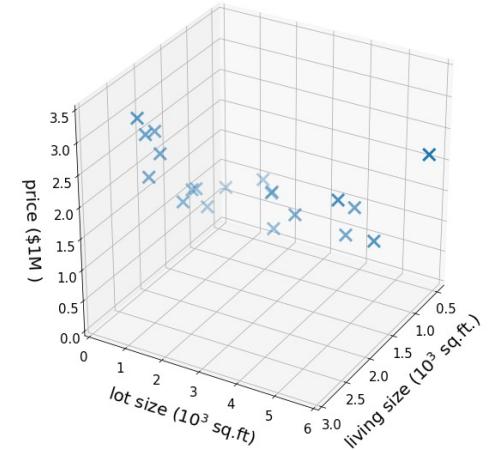
- Suppose we also know the lot size
- **Task:** find a function that maps
  - $(\text{size}, \text{lot size}) \rightarrow \text{price}$

*features/input*

$$x \in R^2$$

*label/output*

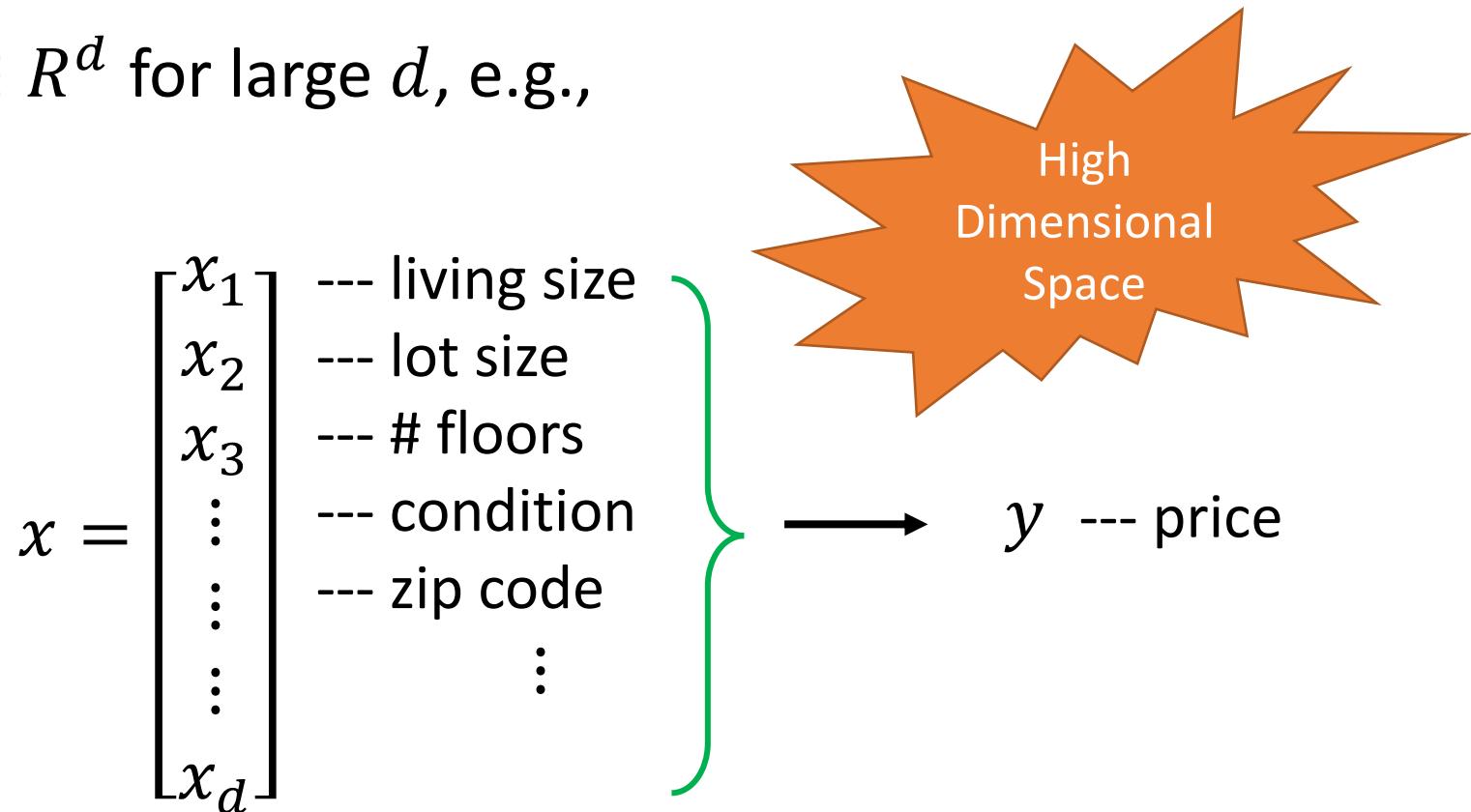
$$y \in R$$



- **Dataset:**  $(x^{(1)}, y^{(1)}), \dots, (x^{(n)}, y^{(n)})$  where the feature  $x^{(i)} = (x_1^{(i)}, x_2^{(i)})$
- “**Supervision**” refers to the labels  $y^{(1)}, \dots, y^{(n)}$

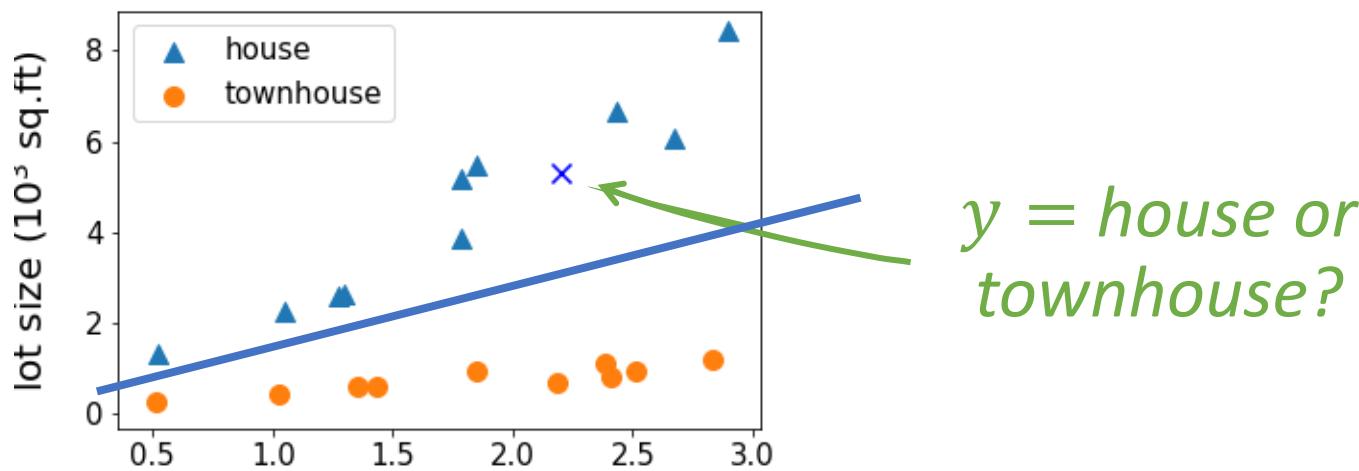
# Supervised Learning (A lot of features)

- $x \in R^d$  for large  $d$ , e.g.,



# Regression vs. Classification

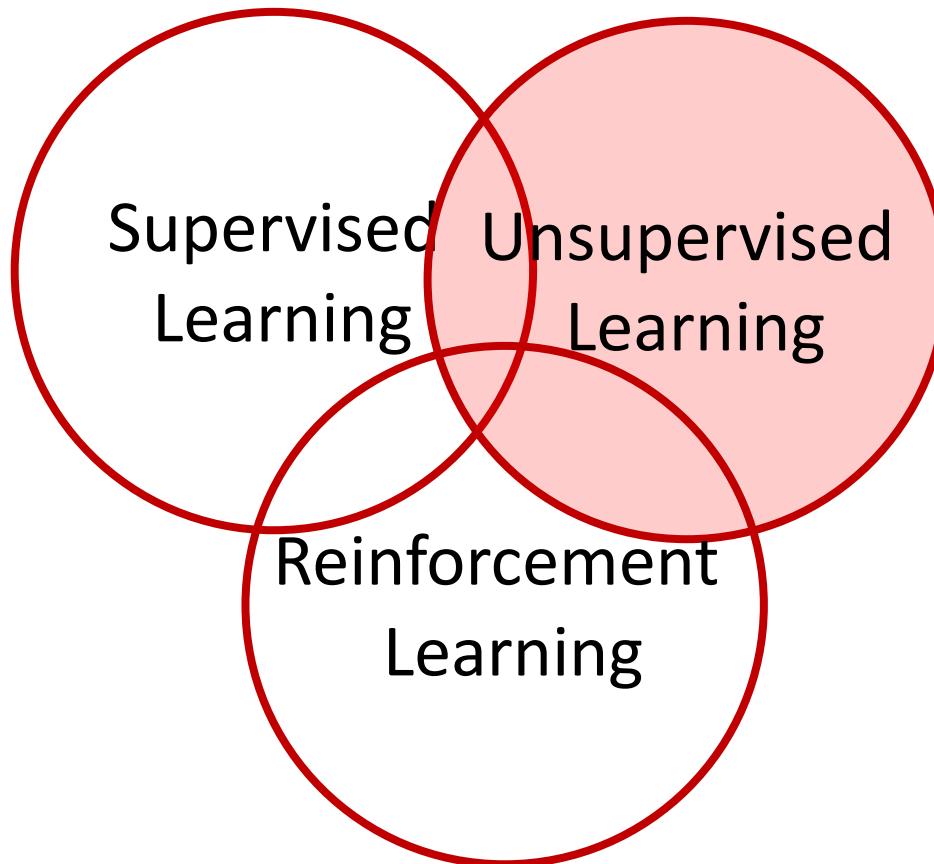
- **Regression**: if  $y \in R$  is a *continuous* variable
  - e.g., price prediction
- **Classification**: the label is a *discrete* variable
  - e.g., predicting the types of residence  
(size, lot size) → house or townhouse?



$y = \text{house or townhouse?}$

# Taxonomy of Machine Learning Methods

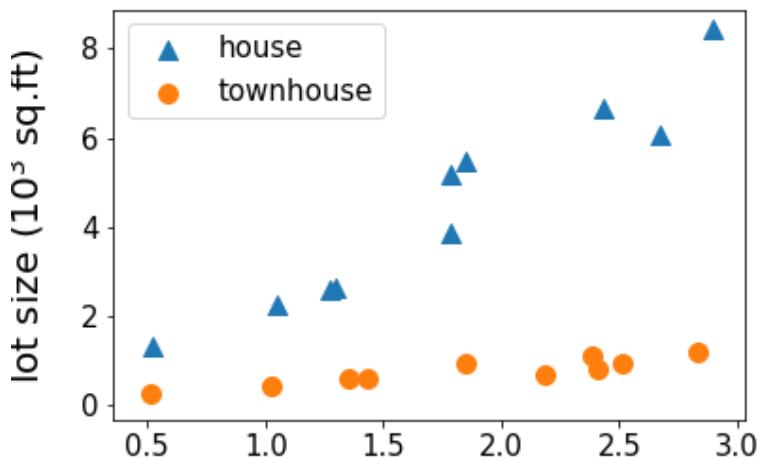
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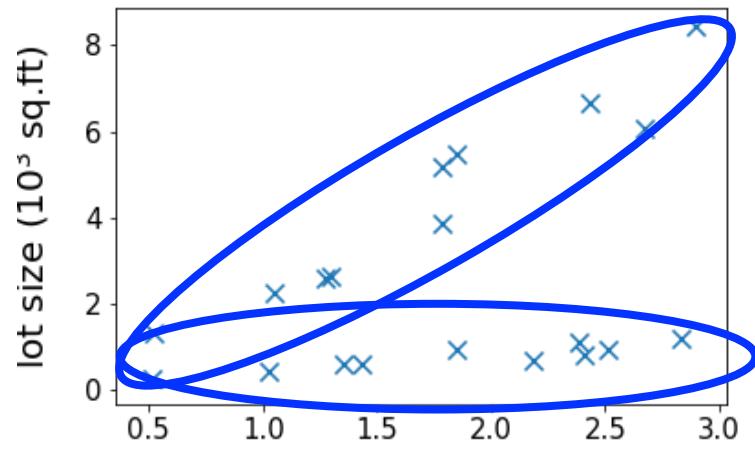
# Unsupervised Learning

- Dataset contains *no labels*:  $x^{(1)}, \dots x^{(n)}$
- **Goal:** to find interesting structures or patterns from the data

## Supervised Learning



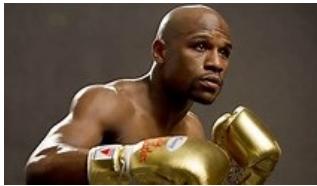
## Unsupervised Learning



23

# Cluster Wikipedia articles by K-means

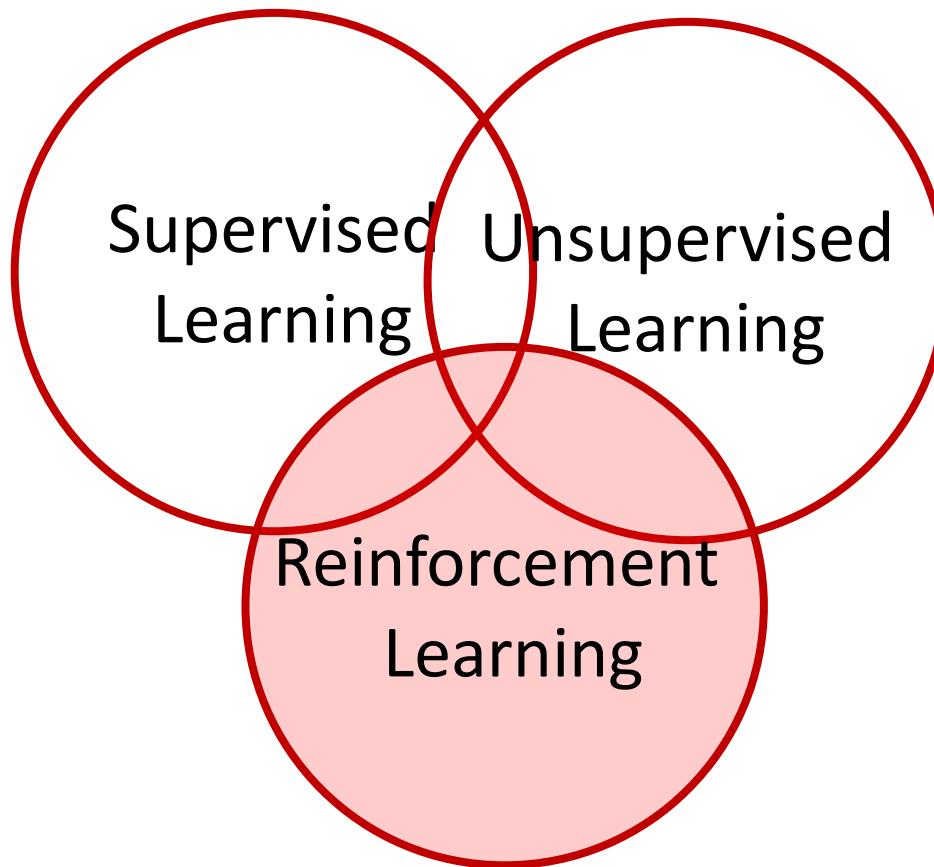
Cluster 1		Cluster 2		Cluster 3	
Word	Coef.	Word	Coef.	Word	Coef.
fight	0.038	holiday	0.012	united	0.004
win	0.022	celebrate	0.009	family	0.003
event	0.019	festival	0.007	party	0.003
champion	0.015	celebration	0.007	president	0.003
fighter	0.015	calendar	0.006	government	0.003



- “Floyd Mayweather, Jr”, “Kimbo Slice”, “Ronda Rousey”, “José Aldo”, “Joe Frazier”, “Wladimir Klitschko”, “Saul Álvarez”, “Gennady Golovkin”, “Nate Diaz”, ...
- “Halloween”, “Guy Fawkes Night” “Diwali”, “Hanukkah”, “Groundhog Day”, “Rosh Hashanah”, “Yom Kippur”, “Seventh-day Adventist Church”, “Remembrance Day”, ...
- “Mahatma Gandhi”, “Sigmund Freud”, “Carly Fiorina”, “Frederick Douglass”, “Marco Rubio”, “Christopher Columbus”, “Fidel Castro”, “Jim Webb”, ...

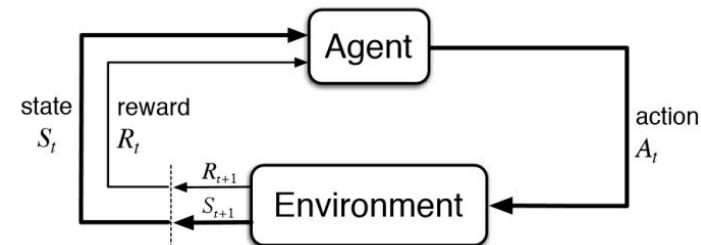
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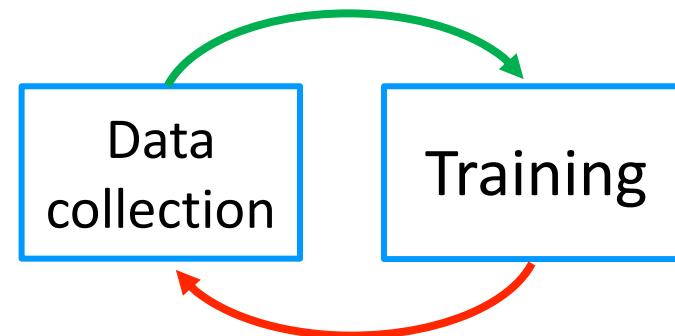


# Reinforcement Learning

- Agent and environment interact at discrete time steps:  $t = 1, 2, 3, \dots, T$ . At step  $t$ , the agent
  - Observes state (environment)
  - Produces actions
  - Gets resulting reward
  - And produces the next state



Try the policy and collect feedbacks



Improve the policy based on the feedbacks (rewards)

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# See: Computer Vision (CV)

Image Recognition



Image Classification

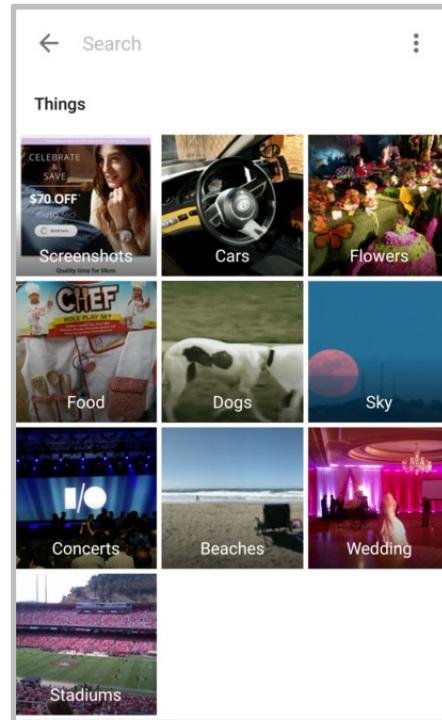


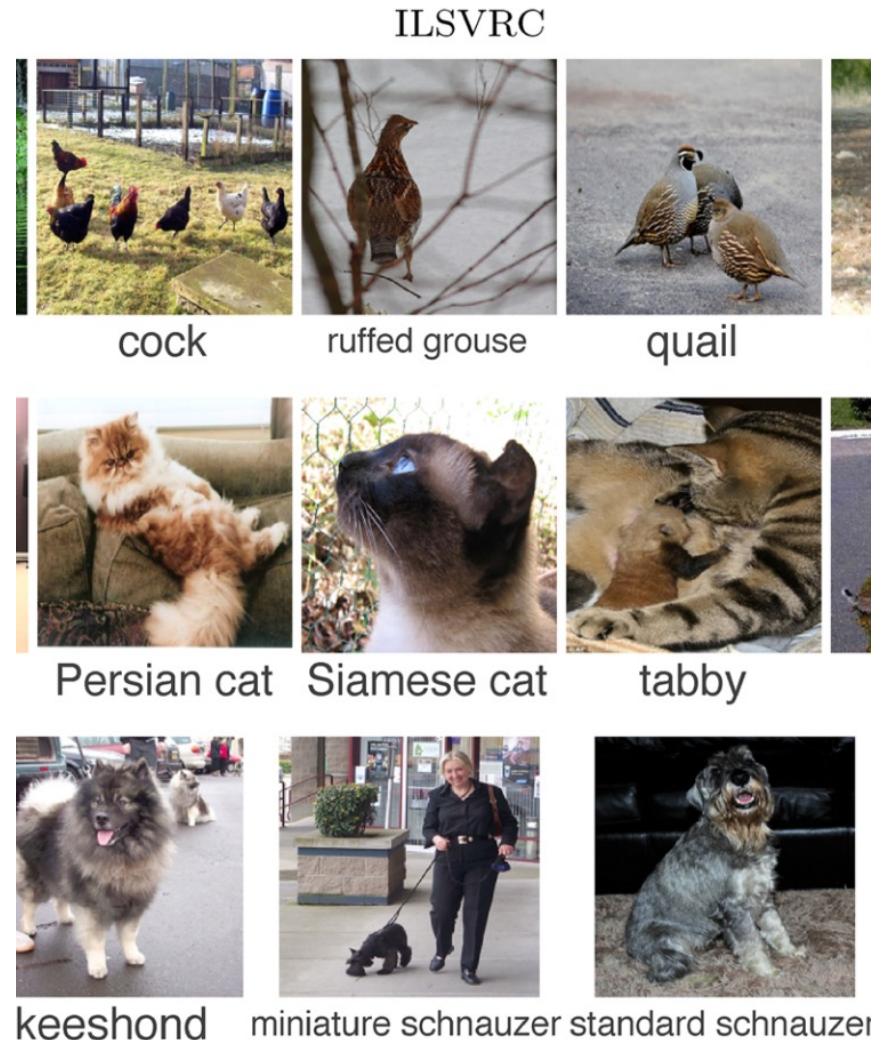
Image Understanding



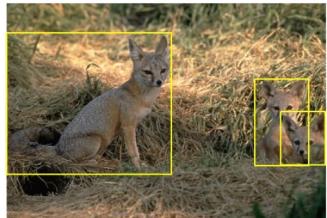
"two young girls are playing with lego toy."

# Supervised Learning (Computer Vision)

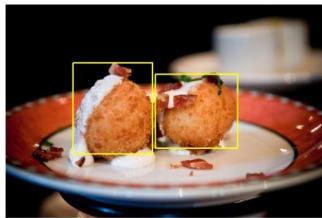
- Image Classification
  - $x = \text{raw pixels of the image}$
  - $y = \text{the main object}$



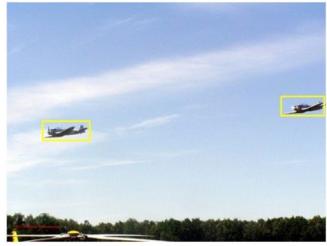
# Supervised Learning (Computer Vision)



kit fox



croquette



airplane



frog

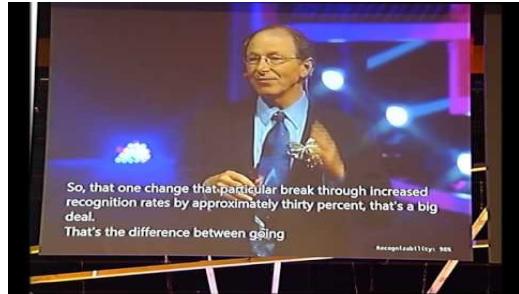
- Object localization and detection
  - $x$  = raw pixels of the image
  - $y$  = the bounding boxes

# Roadmap

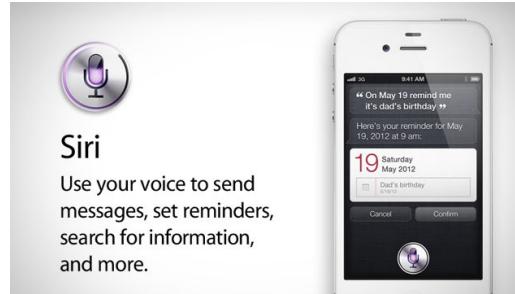
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# Listen and Speak: Speech Processing

## Audio Speech Recognition (Speech to Text)



## Voice Assistants



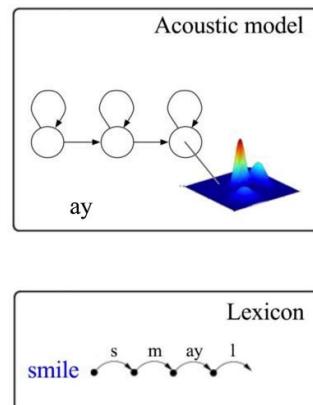
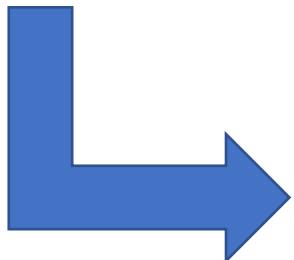
## Text to Speech



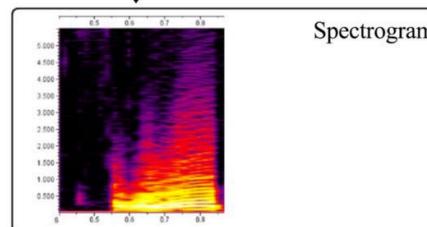
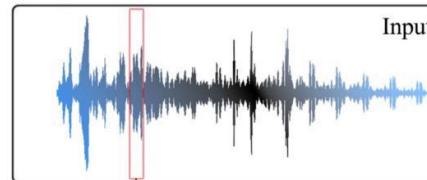
# Supervised Learning (Speech Processing)



## Audio Speech Recognition



$x$ : audio signals



39 features      Features extraction

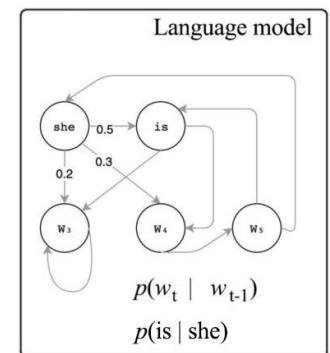
Decoding search

Word sequence

By Jonathan Hui

$$\mathbf{W}^* = \arg \max_{\mathbf{W}} P(\mathbf{W} | \mathbf{X})$$
$$\mathbf{W}^* = \arg \max_{\mathbf{W}} p(\mathbf{X} | \mathbf{W}) P(\mathbf{W})$$

word sequence      acoustic model      language model



$y$ : text signals (word sequence)

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# Read and Write: Natural Language Processing (NLP)



- Where natural language comes from?
- What is natural language processing?

# Read and Write: Natural Language Processing (NLP)

**Natural Language Understanding:**  
How to read

**Natural Language Generation:**  
How to write

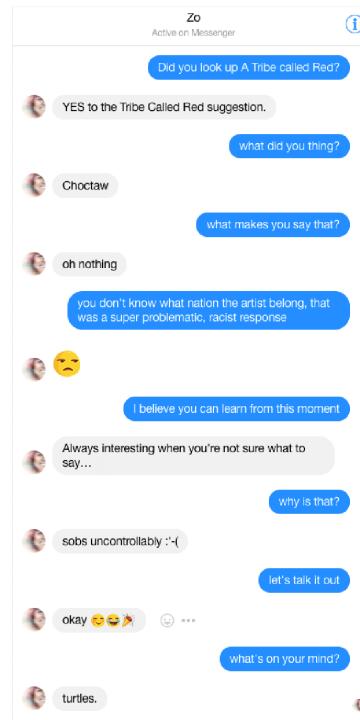
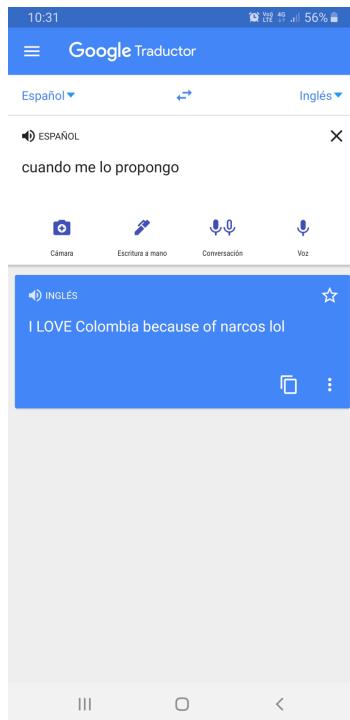


**AI's Mission:** to pass Turing test

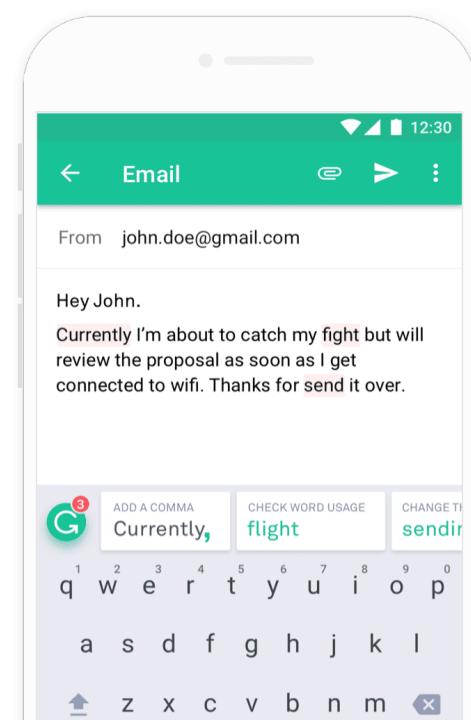


# Read and Write: Natural Language Processing (NLP)

Machine Translation   Dialogue System



Grammarly



# Supervised Learning (NLP)

≡ Google Translate



Text Documents  $x \longrightarrow y$

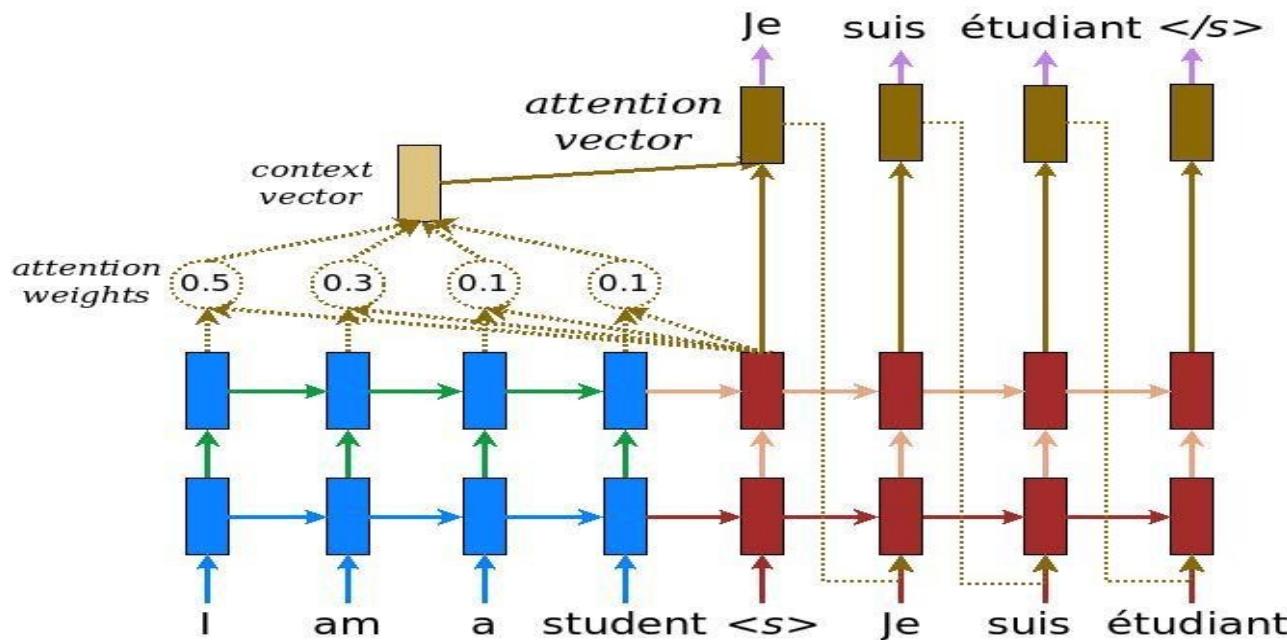
ENGLISH - DETECTED ENGLISH SPANISH FRENCH CHINESE (TRADITIONAL) ENGLISH SPANISH

The Hong Kong Polytechnic University 香港理工大學

Xiānggǎng lǐgōng dàxué

36/5000

From Tensorflow Tutorial



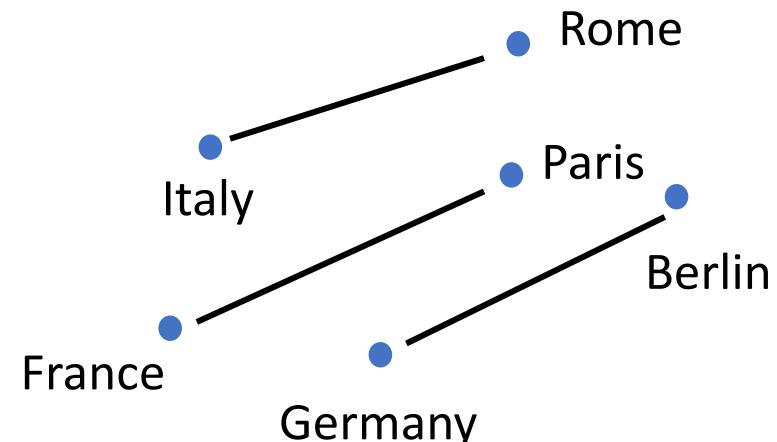
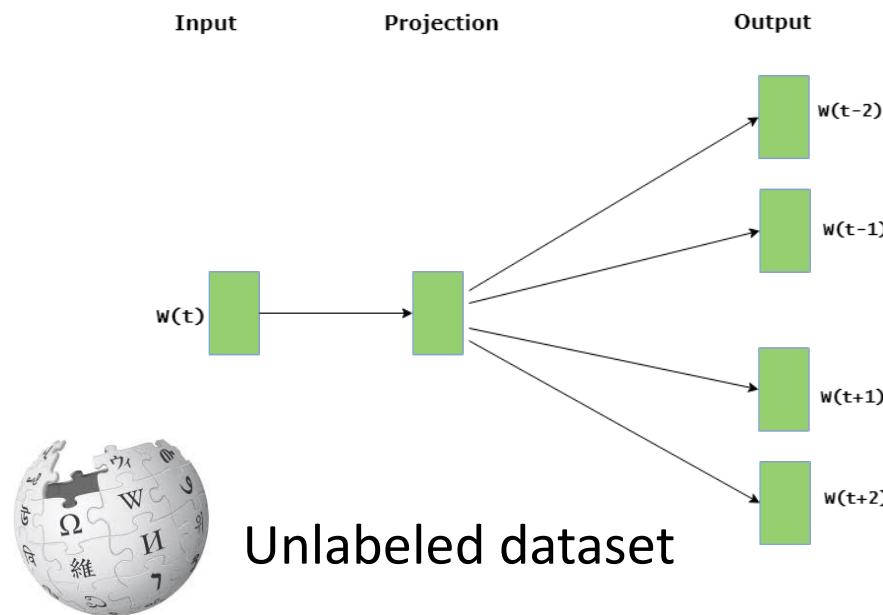
From Tensorflow  
Tutorial

# Unsupervised Learning (NLP)

Represent words by vectors (word embeddings)

word       $\xrightarrow{\text{encode}}$       vectors in high dimensional space

context     $\xrightarrow{\text{encode}}$       semantic meaning



Word2vec [Mikolov et al'13]  
GloVe [Pennington et al'14]

# Unsupervised Learning (NLP)

## Topic-word distribution

### Topics

gene 0.04  
dna 0.02  
genetic 0.01  
...

life 0.02  
evolve 0.01  
organism 0.01  
...

brain 0.04  
neuron 0.02  
nerve 0.01  
...

data 0.02  
number 0.02  
computer 0.01  
...

## Topic modeling from large-scale documents

### Documents

### Topic proportions and assignments

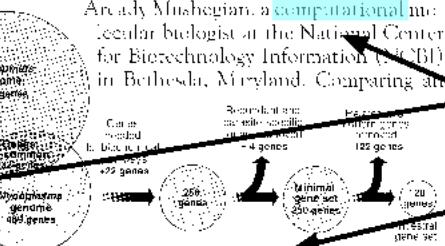
#### Seeking Life's Bare (Genetic) Necessities

COLD SPRING HARBOR, NEW YORK—How many genes does an organism need to survive? Last week at the genome meeting here,\* two genomic researchers with radically different approaches presented complementary views of the basic genes needed for life. One research team, using computer analyses to compare known genomes, concluded that today's organisms can be sustained with just 250 genes, and that the earliest life forms required a mere 128 genes. The other researcher mapped genes in a simple parasite and estimated that for this organism, 800 genes are plenty to do the job—but that anything short of 100 wouldn't be enough.

Although the numbers don't match precisely, those predictions

"are not all that far apart," especially in comparison to the 75,000 genes in the human genome, notes Siv Andersson, a postdoctoral fellow at Sweden's Uppsala University. "In fact, we also arrived at the 800 number." But coming up with a consensus answer may be mere chaff compared to the numbers some particularities more and more genomes are completely mapped and sequenced. "It may be a way of organizing any newly sequenced genome," explains

Arshad Minahegan, a computational molecular biologist at the National Center for Biotechnology Information (NCBI) in Bethesda, Maryland. Comparing all



\* Genome Mapping and Sequencing. Cold Spring Harbor, New York. May 8 to 12.

Doc-topic distribution

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# Example: Self-Driving Cars

- Nevada made it legal for self-driving cars to drive on roads in June 2011.
- As of 2013, four states (Nevada, Florida, California, and Michigan) have legalized self-driving cars

**Think:** What will happen if a self-driving car kills a person?



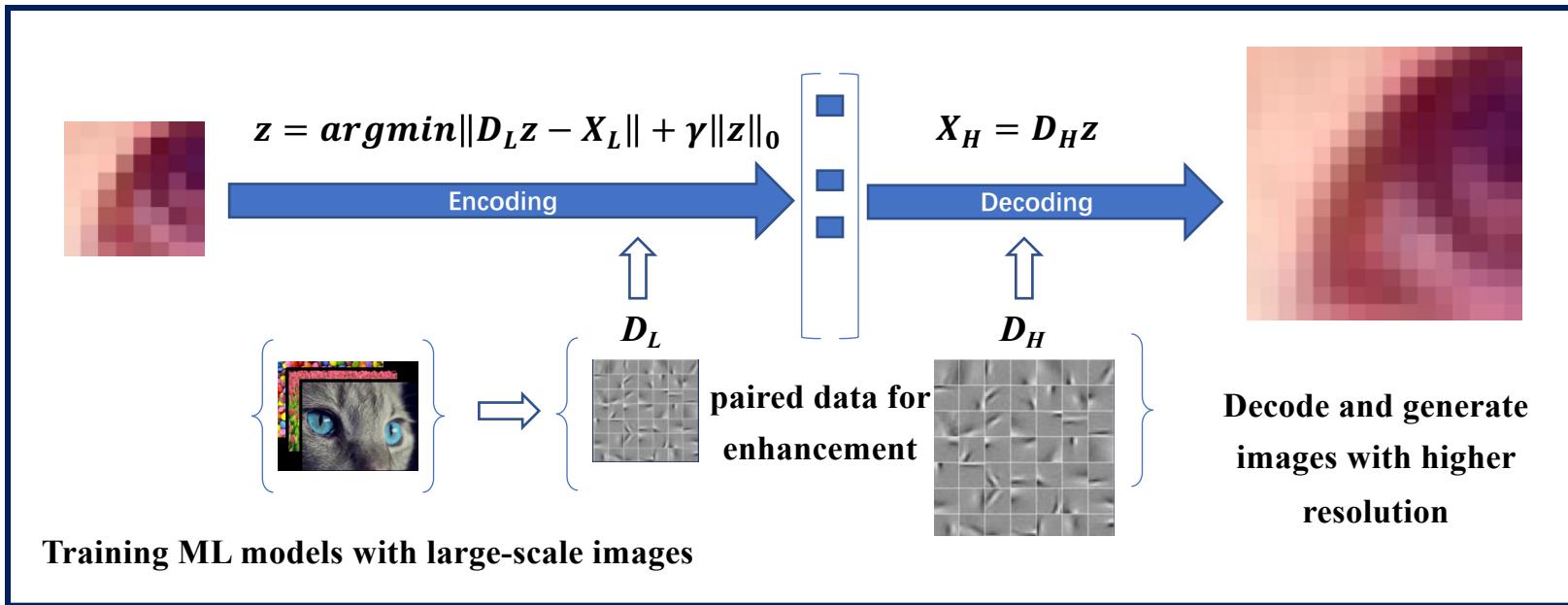
# Example: Drone (unmanned aerial vehicle)

Learn policy from user demonstrations with RL



*DJI was incubated from HKUST*

# Example: Super Resolution



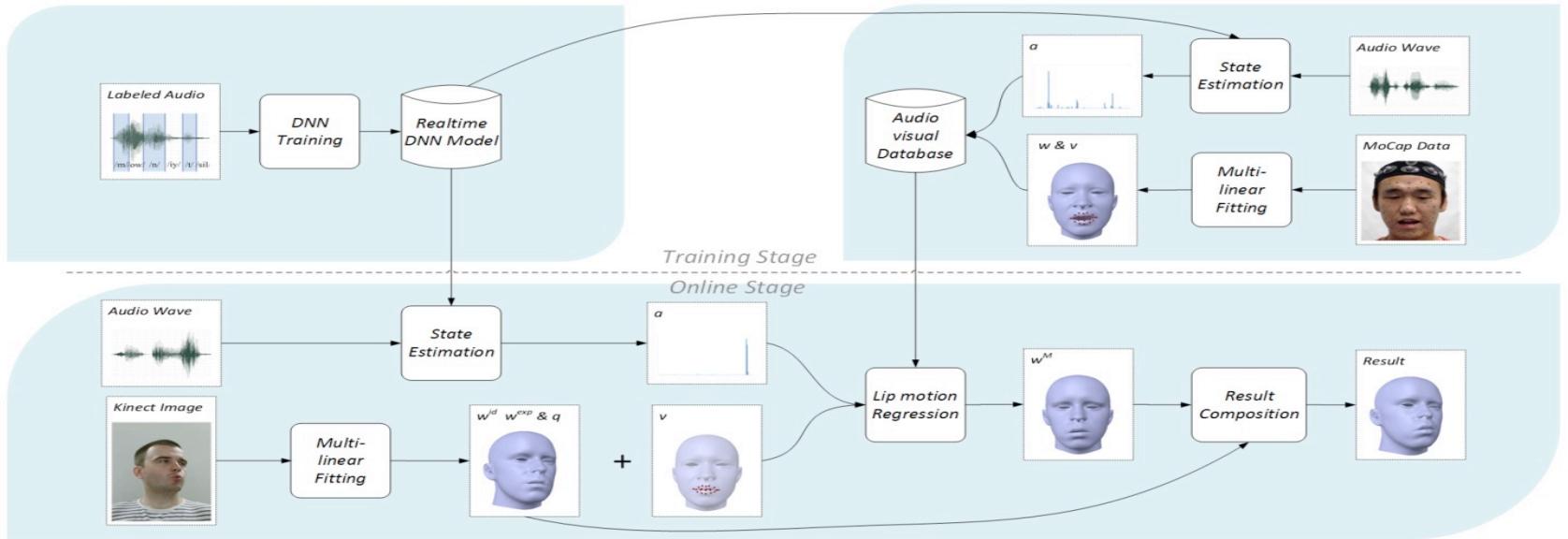
Max Payne

*The Elder Scrolls III:  
Morrowind*



Doom 3

# Example: Motion Capture

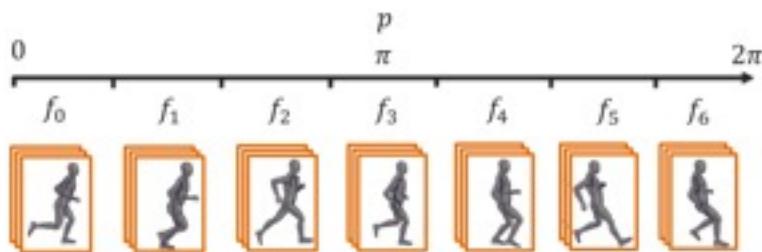
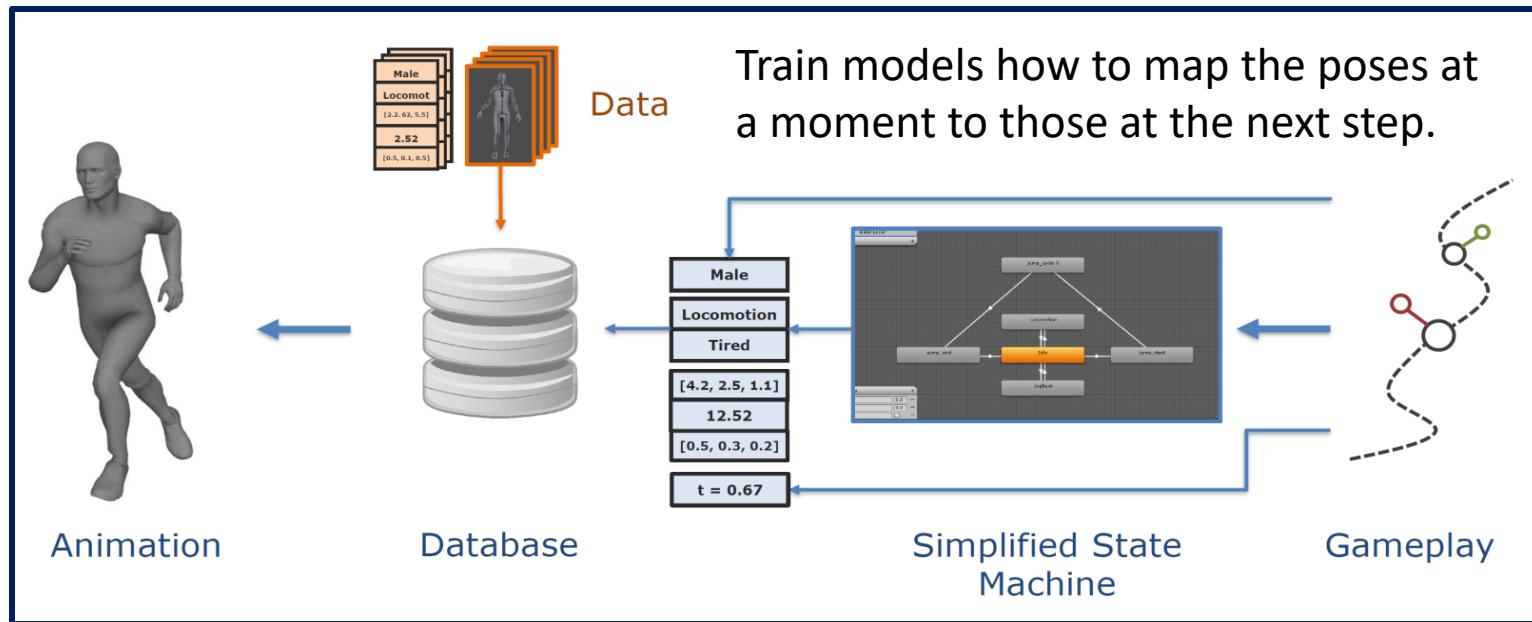


Capture 3D Facial Expression

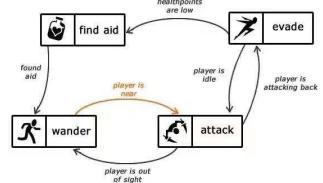
Person Pose Estimation

Style Transfer

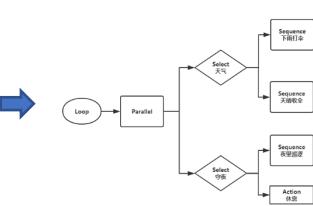
# Example: Automatic Character Animation



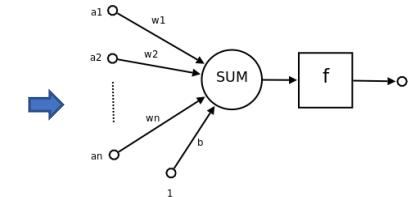
# Example: NPC/Bot Design



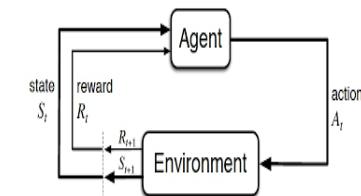
Finite State Machine



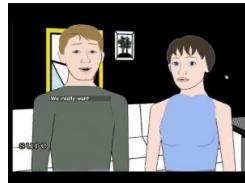
Behavior Tree



Supervised Learning



Reinforcement Learning



- **Early Attempt with NLP (Façade, 2005)**

Users can interact with NPCs with natural languages.



- **Virtual Player (The Crew)**

Allow AI players to race like human players.



- **Socialized NPC (The Elder Scrolls CiF-CK MOD)**

Allow NPC to have conduct social behavior and build social networks. They will act based on their social status.

# A slide to take away

- Machine Learning focuses on methods development to allow machines to gather some human capacities (AI's mission).
- ML taxonomies: supervised learning, unsupervised learning, and reinforcement learning.
- AI areas based on ML methods:
  - See (*Computer Vision*)
  - Listen and Speak (*Speech Processing*)
  - Read and Write (Natural Language Processing)
- AI technology benefit diverse scenarios.