

# COMP 5212 Machine Learning

Instructor: Junxian He

Course website: https://jxhe.github.io/teaching/comp5212s24

#### Teaching Team & Office Hours

Instructor: Junxian He

Office Hour: Wed 430pm - 530pm, Room 3512

TA1: Kashun Shum

Office Hour: Wed 2pm-3pm, outside room 3657

TA2: Chi Xu

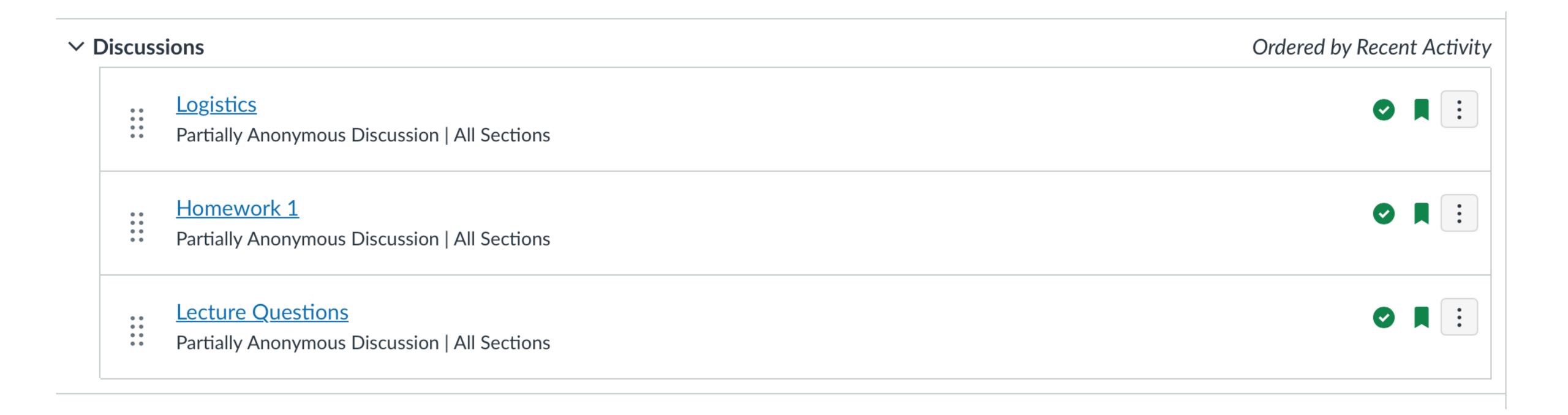
Office Hour: Fri 6pm-7pm, Room 4033

### This is an in-person Class

- Lectures are recorded, but the videos will be released only twice during the midterm and before the final
  - I may release the respective lecture videos in special cases, e.g., the lectures around Lunar New Year Eve

Lecture slides will be available after each lecture

#### Communication and Discussion



#### Pre-requisite

- Probability
  - Distribution, random variable, expectation, conditional distribution, variance

- Linear algebra
  - Matrix multiplication
- Python programming

This is not an easy course! It may be mathematically intense for you. Difficulty is often rewarding:)

### Grading

- Attendance (10%)
- 5 assignments (50%)
  - 4 Written + programing assignments
  - 1 programming-only assignment (Kaggle alike competition)
  - 3 free late days in total, for additional late days, 20% penalization applied for each day late
  - No assignment will be accepted more than 3 days late
- Mid-term Exam, open-note (20%)
- Final exam, open-note (20%)

#### Attendance

Occasional quiz questions

80% of attendance will give you full grade

Correctness of quiz answers does not influence attendance grading

#### Assignments

- 5 assignments (50%)
  - 4 Written + programing assignments (4 \* 9%)
  - 1 programming-only assignment (14%, Kaggle alike competition)
  - 3 free late days in total, for additional late days, 20% penalization applied for each day late
  - No assignment will be accepted more than 3 days late

#### **Honor Code**

#### Do's

- Form study groups to discuss (e.g. homework)
- Write down the homework solutions independently
- Write down the names of people with whom you've discussed the homework
- You are encouraged to use generative AI (e.g. ChatGPT) to assist you

#### Don'ts

- Copy, refer to, or look at solutions from previous years, online, or others
- Copy ChatGPT's answers
- Longer versions on the course website

We have zero tolerance — in the case of honor code violation for a single time, you will fail this course directly

#### Waiting List & Audit

We will let everyone on the waiting list in as long as the total enrollment does not exceed 90 (currently 60 enroll + 20 WL)

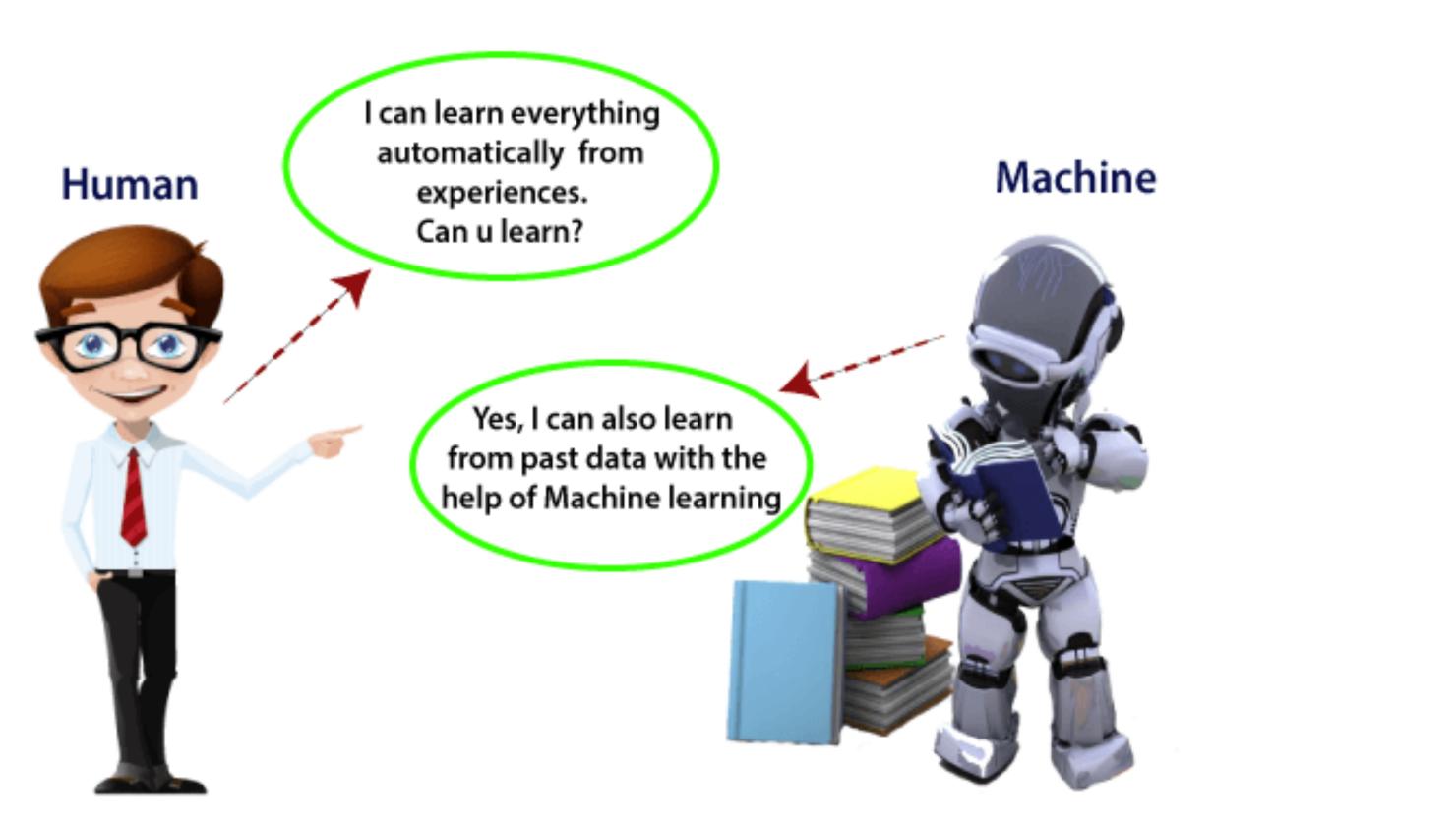
Audit is allowed, and will be graded in the same criterion

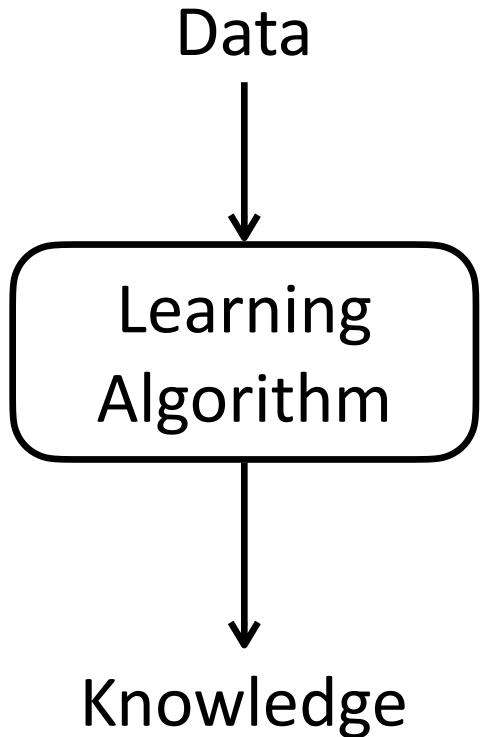
#### More Info on Course Website

- Canvas is the main platform for announcement, discussion, homework submission
- Recorded videos on canvas
- Syllabus, slides and relevant reading materials
- Detailed course logistics

# Overview of Machine Learning

## What is Machine Learning





# Machine Learning is Trending 🚀 🔴

- Everywhere wide application
  - Finance, data scientist, medical diagnosis, translation, self-driving....

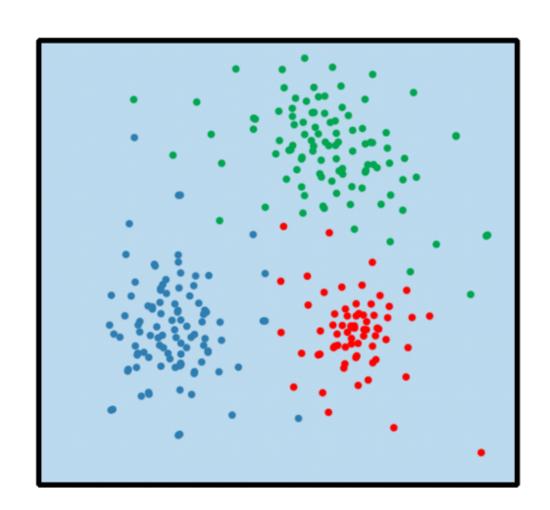
- Foundation of artificial intelligence one of the most important technology for the society in the next 10s of years
  - ChatGPT, large language model, large multimodal model

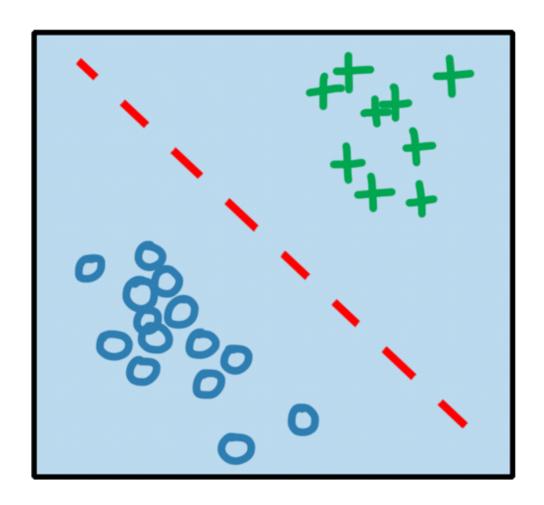
## Taxonomy of Machine Learning

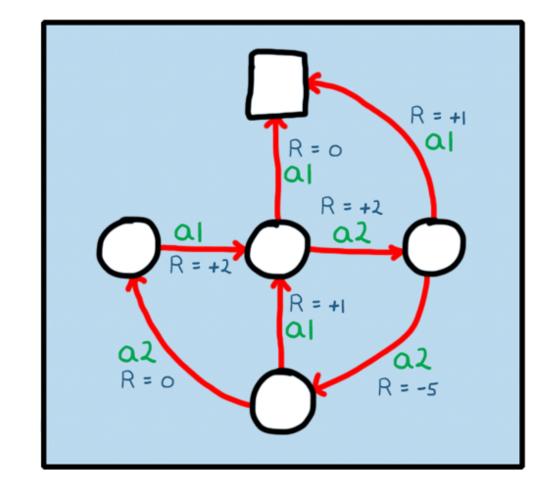
# machine learning

unsupervised learning supervised learning

reinforcement' learning







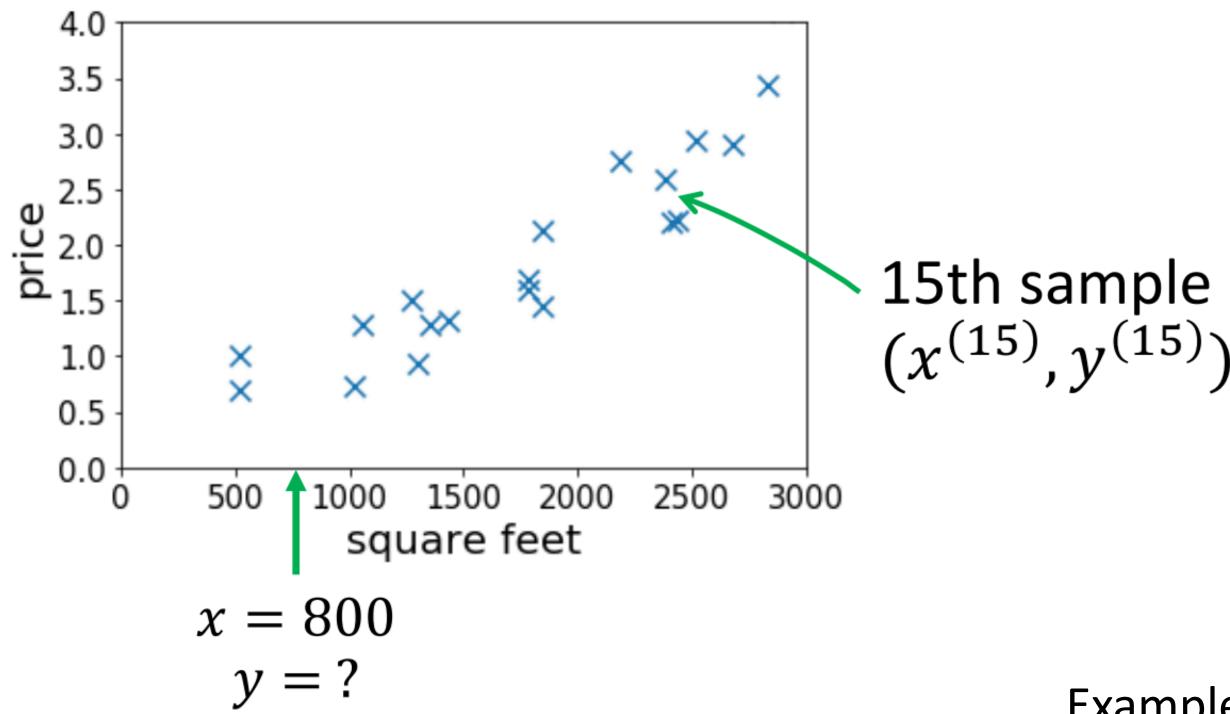
# Supervised Learning

#### Housing Price Prediction

Given: a dataset that contains n samples

$$(x^{(1)}, y^{(1)}), ... (x^{(n)}, y^{(n)})$$

 $\blacksquare$  Task: if a residence has x square feet, predict its price?

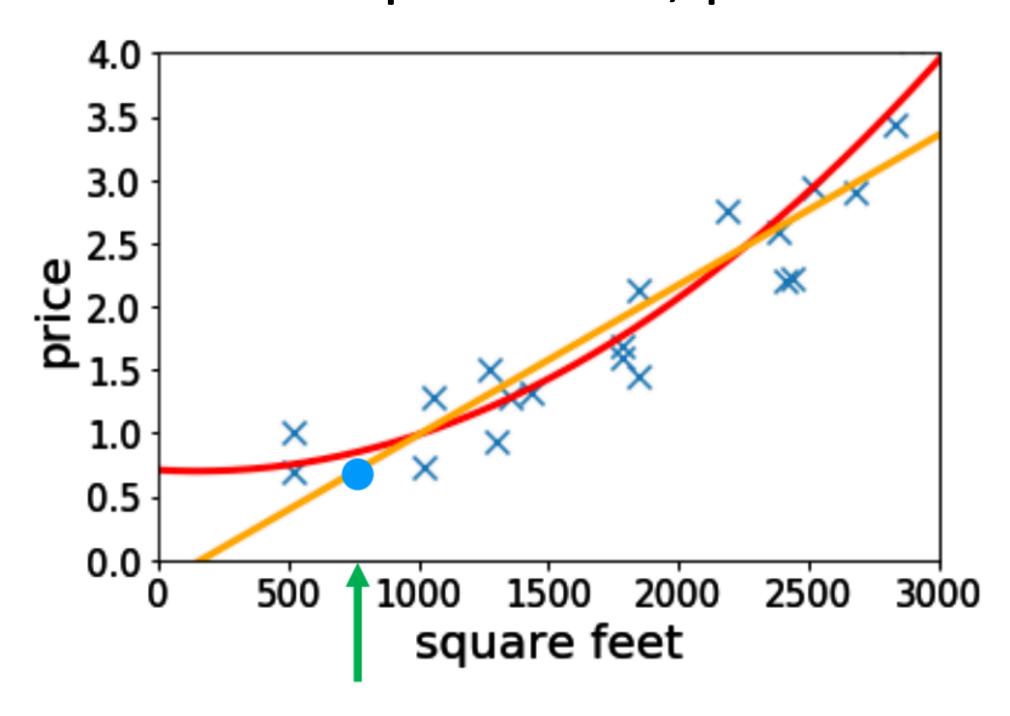


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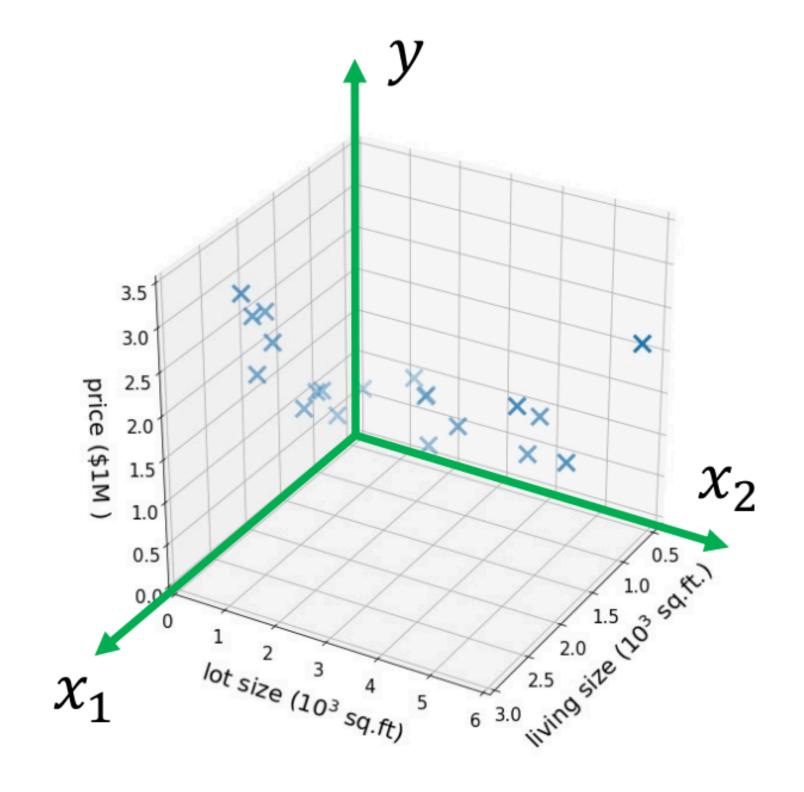
#### More Features

Suppose we also know the lot size

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(size, lot size) 
$$\rightarrow$$
 price features/input label/output  $x \in \mathbb{R}^2$   $y \in \mathbb{R}$ 



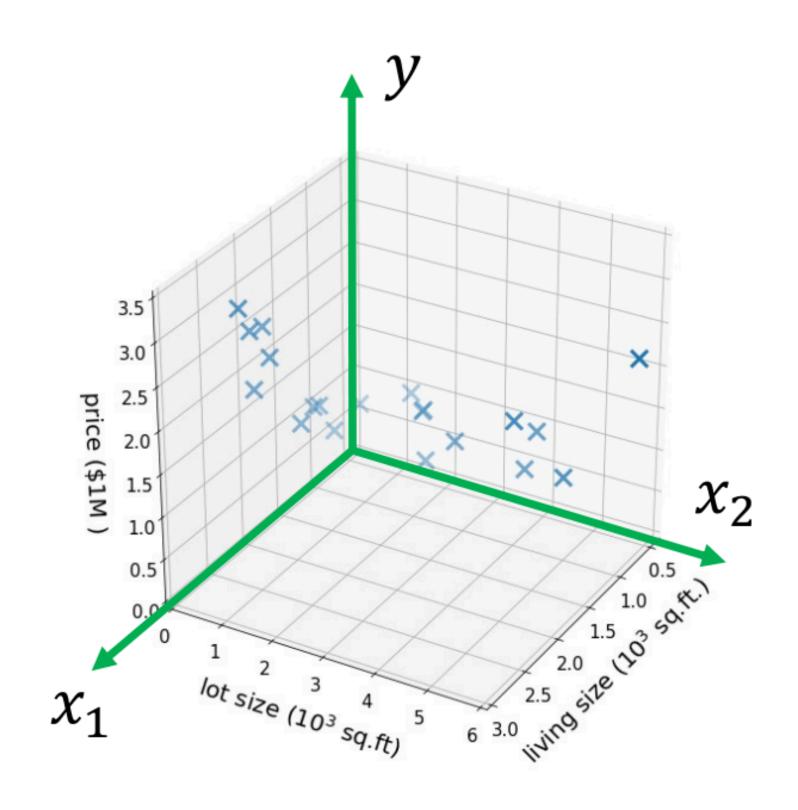
#### **More Features**

- Suppose we also know the lot size
- Task: find a function that maps

(size, lot size) 
$$\rightarrow$$
 price  
features/input label/output  $x \in \mathbb{R}^2$   $y \in \mathbb{R}$ 

• Dataset:  $(x^{(1)}, y^{(1)}), ..., (x^{(n)}, y^{(n)})$ 

where 
$$x^{(i)} = (x_1^{(i)}, x_2^{(i)})$$



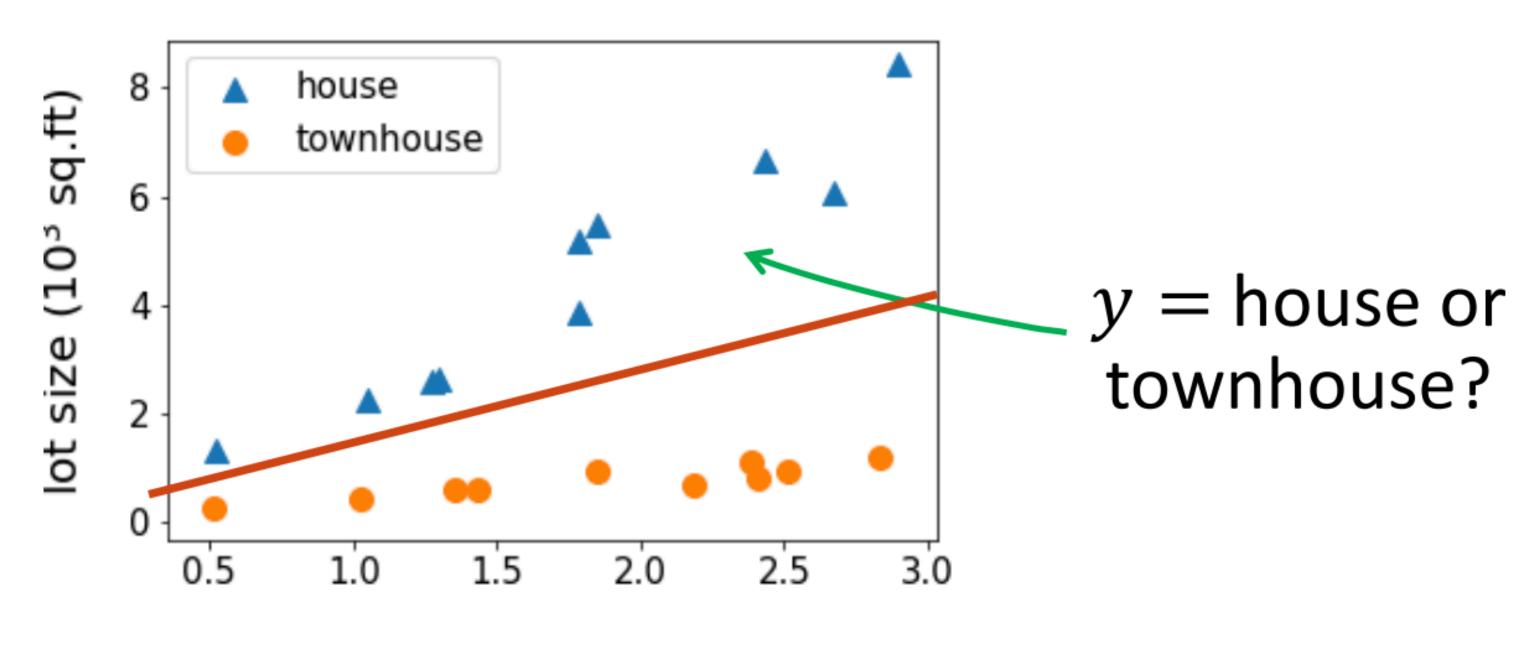
#### High-dimensional Features

High-dimensional Features
$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ \vdots \\ x_d \end{bmatrix} \xrightarrow{\text{---- living size}} --- \text{lot size}$$
--- # floors
--- condition
--- zip code
$$\vdots$$

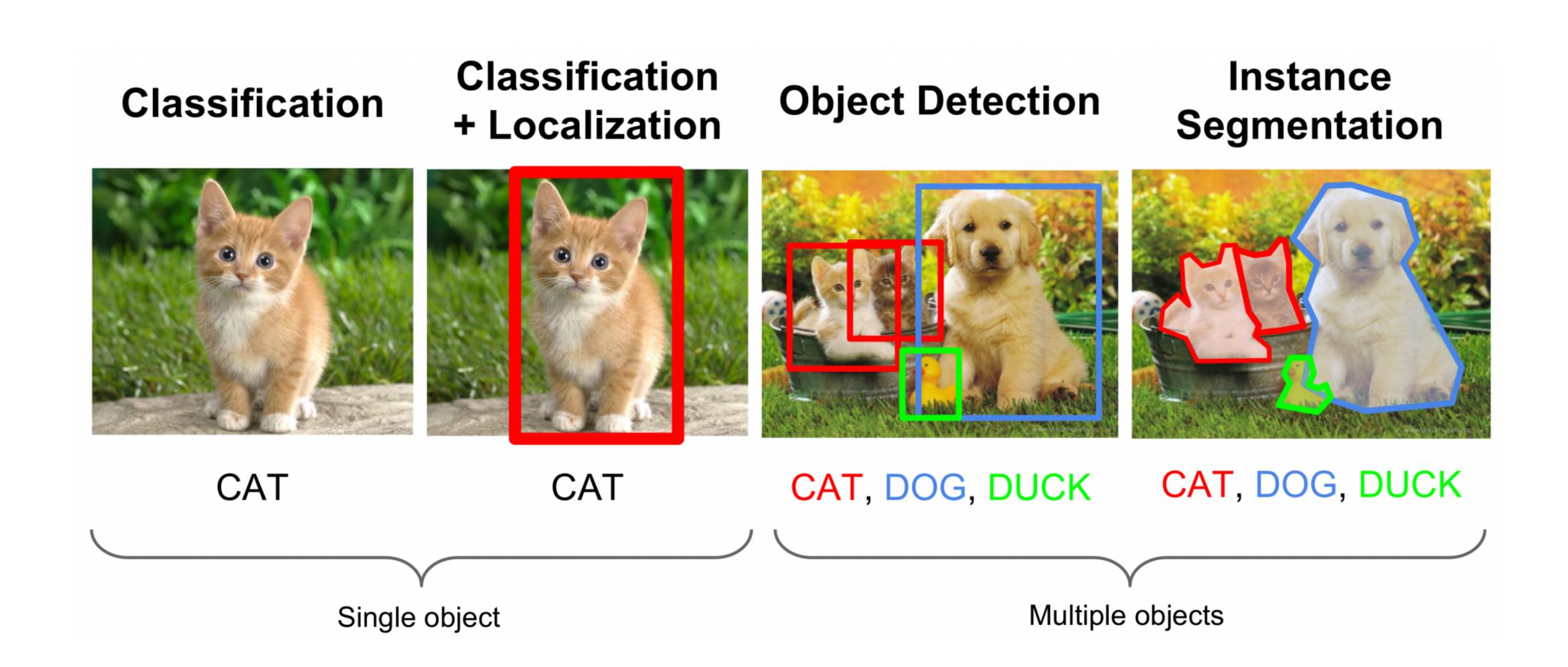
$$\vdots$$

#### Regression vs Classification

- Regression: if  $y \in \mathbb{R}$  is a continuous variable
  - E.g., price prediction
- Classification: the label is a discrete variable
  - E.g., predicting the types of residence



## Supervised Learning in Computer Vision



# Supervised Learning in Natural Language Processing



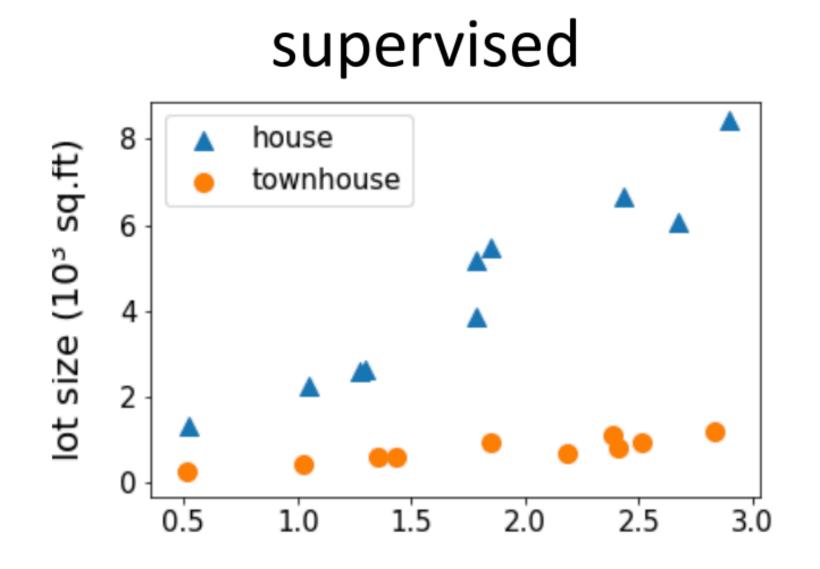


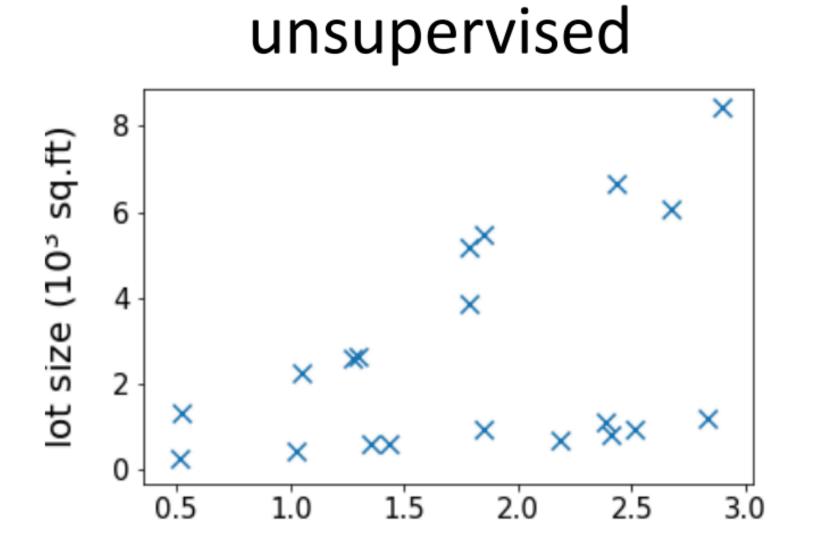
This course will only cover basic and fundamental things about ML

# Unsupervised Learning

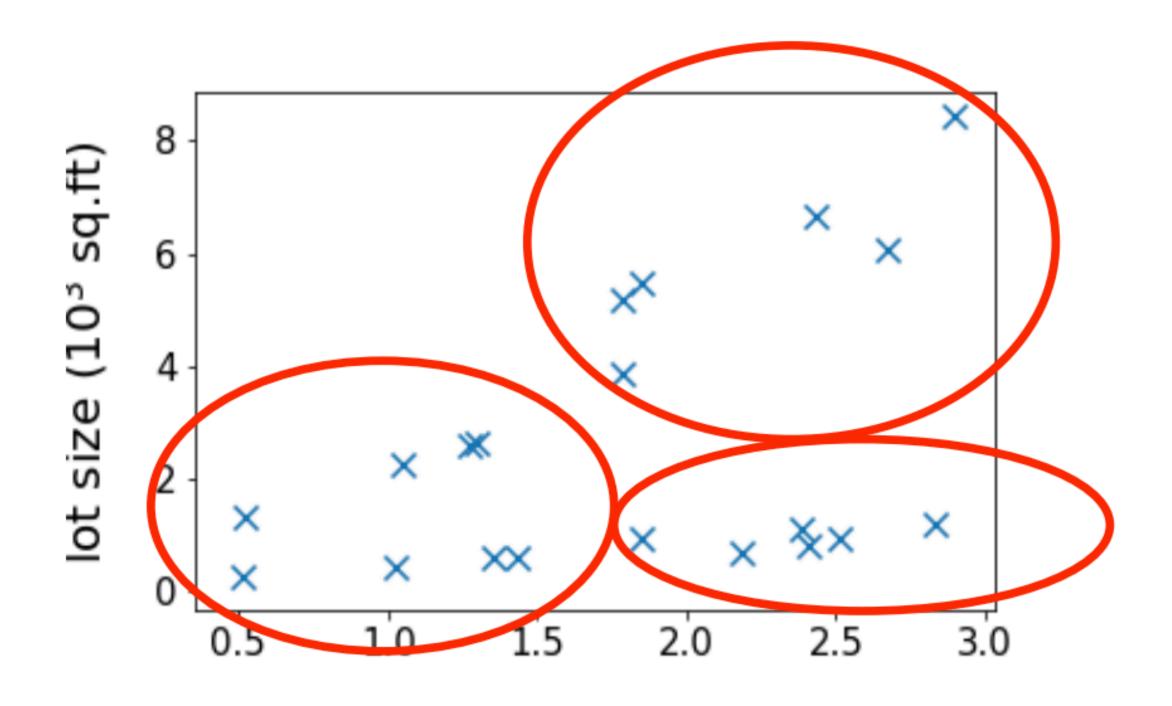
### Unsupervised Learning

- Dataset contains no labels  $x^{(1)}, ..., x^{(n)}$
- Typically very vague goal: to find interesting structures in the data



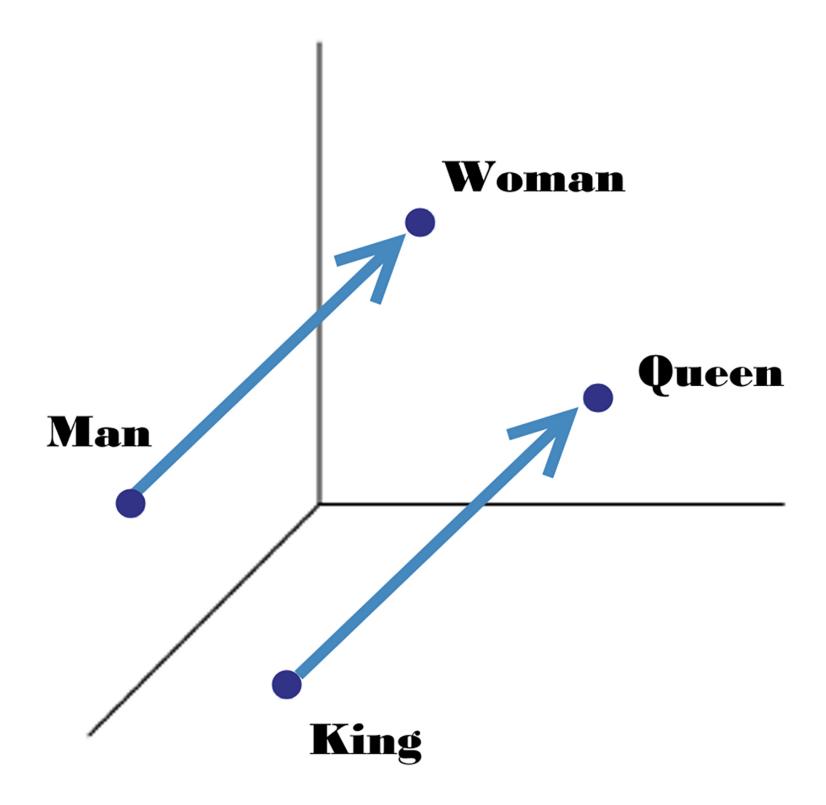


# Clustering



# Word Embeddings





#### Topic Models

#### TOPIC 1

#### TOPIC 2

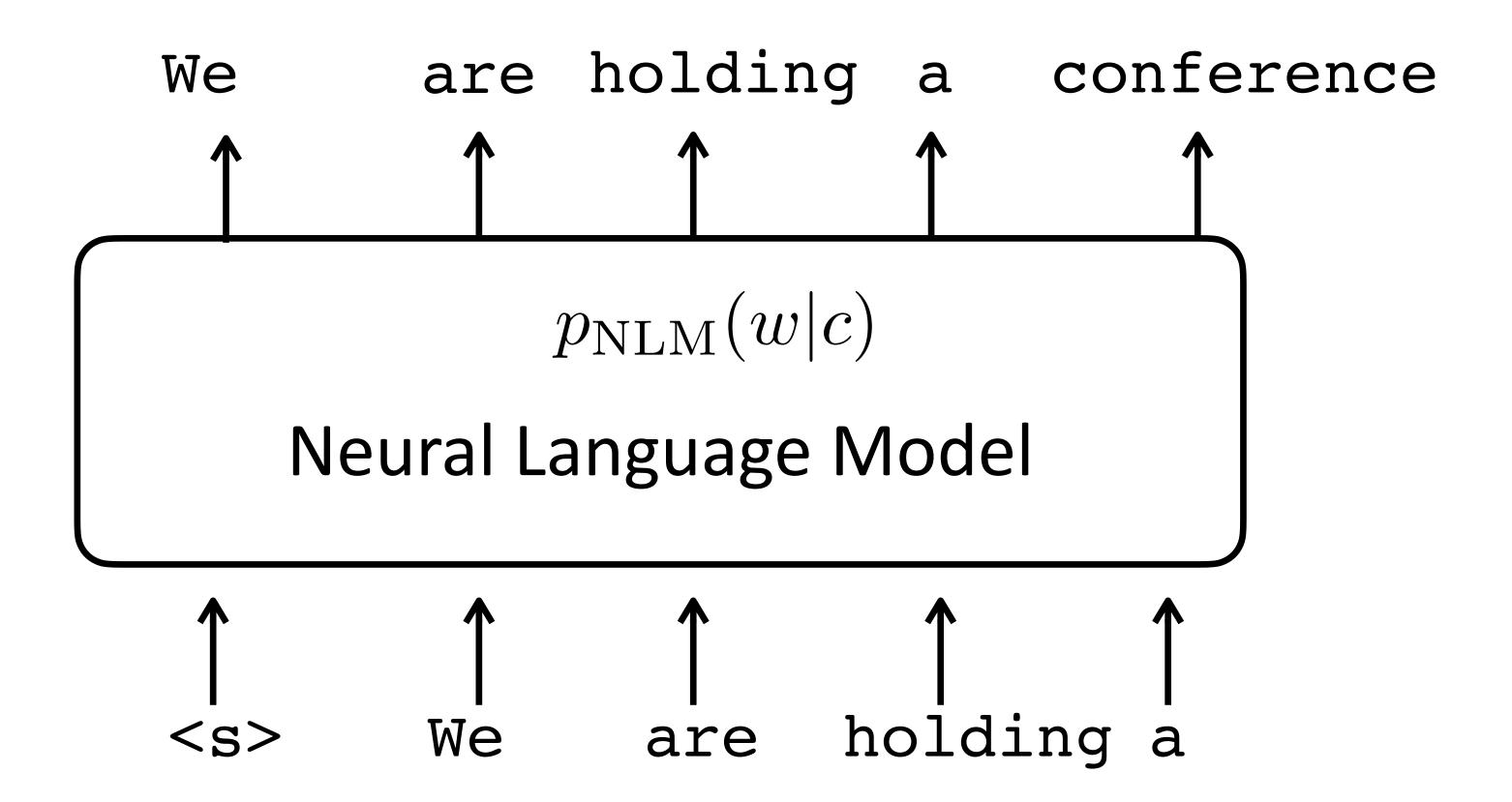
#### TOPIC 3

computer, technology, system, service, site, phone, internet, machine

sell, sale, store, product, business, advertising, market, consumer

play, film, movie, theater, production, star, director, stage

### Language Models



#### Large Language Models

#### SYSTEM PROMPT (HUMAN-WRITTEN)

In a shocking finding, scientist discovered a herd of unicorns living in a remote, previously unexplored valley, in the Andes Mountains. Even more surprising to the researchers was the fact that the unicorns spoke perfect English.

#### MODEL COMPLETION (MACHINE-WRITTEN, 10 TRIES)

The scientist named the population, after their distinctive horn, Ovid's Unicorn. These four-horned, silver-white unicorns were previously unknown to science.

Now, after almost two centuries, the mystery of what sparked this odd phenomenon is finally solved.

Dr. Jorge Pérez, an evolutionary biologist from the University of La Paz, and several companions, were exploring the Andes Mountains when they found a small valley, with no other animals or humans. Pérez noticed that the valley had what appeared to be a natural fountain, surrounded by two peaks of rock and silver snow.

Pérez and the others then ventured further into the valley. "By the time we reached the top of one peak, the water looked blue, with some crystals on top," said Pérez.

Pérez and his friends were astonished to see the unicorn herd. These creatures could be seen from the air without having to move too much to see them — they were so close they could touch their horns.

# Large Language Models

extstyle  ext	Please unscramble the letters into a word, and write that word: taefed =
${\tt Target \ Completion} \ \rightarrow$	defeat
${\tt Context}  \to $	L'analyse de la distribution de fréquence des stades larvaires d'I. verticalis dans une série d'étangs a également démontré que les larves mâles étaient à des stades plus avancés que les larves femelles. =
Target Completion $ ightarrow$	Analysis of instar distributions of larval I. verticalis collected from a series of ponds also indicated that males were in more advanced instars than females.
${\tt Context}  \to $	Q: What is 95 times 45? A:
${\tt Target~Completion}  \to $	4275

# Reinforcement Learning

# AlphaGo

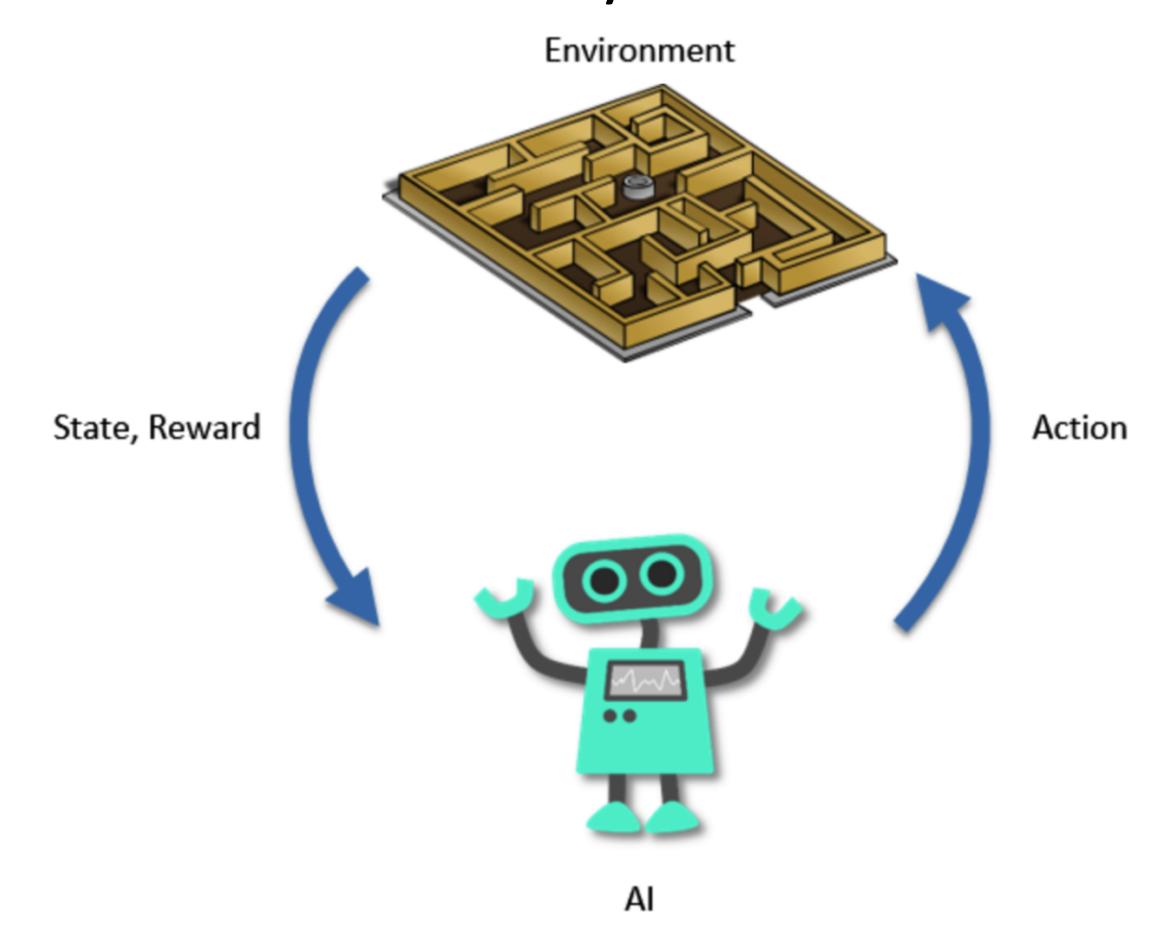


#### Atari Breakout Game



# Reinforcement Learning

RL can collect data interactively



# Other Topics to Cover

Introduction		
Math basics		
Supervised learning basics		
Logistic regression		
Generalized linear models, classification		
Kernel methods		
SVM		
Naive Bayes		
MLE, MAP		
Gradient descent, SGD, Newton's method		
Generalization, bias-variance tradeoff		
Clustering		
Expectation Maximization		
PCA/ICA		
mid-term exam		
Probabilistic Graphical Models		
HMM		
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Neural Networks, backprop	
Neural Networks, architectures	
Neural architectures	
Variational autoencoder	
Generative adversarial networks	
Reinforcement Learning	
Labor day	
Languge models	
Pretraining	
Large language models	

Thank You!
Questions?