

# CS541: Applied Machine Learning

## Introduction

Instructor: Deepti Ghadiyaram



# Meet the crew!



**Prof. Deepti Ghadiyaram  
(Instructor)**



**Dahye Kim  
(TF)**



**Ifreen  
(TA)**

# Instructor's research journey



AUSTIN



TEXAS  
The University of Texas at Austin



facebook AI Research

runway



BOSTON  
UNIVERSITY



# Instructor's research journey



- Computer graphics
- Signal processing

AUSTIN



- Large-scale scene understanding
- Responsible AI

facebook AI Research

- Generative models

runway



- Safety, interpretability, robustness

# Today

- Course logistics
- What is machine learning?
- Classification introduction

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# Class website

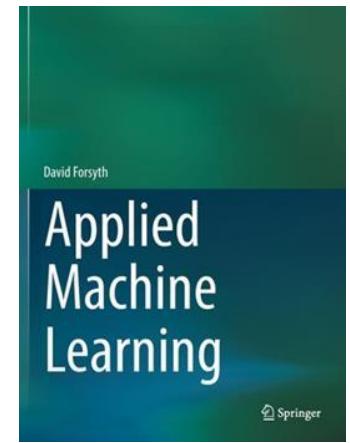


- Main website [CS541 Applied Machine Learning Spring 2025](#)
- **Sign up today!** Piazza (Q&A, homework)  
<https://piazza.com/bu/spring2025/cs541>
  - Course staff may take up to 2 business days to respond to questions.
  - Come to office hours for quick responses!
  - All communication only via Piazza (private messages / public discussions).

# Office Hours

- Deepti's office hours:
  - Tues/ Thurs : 12:45 – 1:45 PM in **CDS 362A**
  - **Sign up [here](#) first before you show up during your slot**
- Dahye's office hours:
  - Wed 4:00PM-6:00PM in **CDS 362B**

# Textbook



- Required textbook:
  - David Forsyth. [Applied Machine Learning](#); 1st ed.
- Other suggested textbooks
  - Bishop, C. M. [Pattern Recognition and Machine Learning](#). Springer. 2007
  - Ian Goodfellow, Yoshua Bengio, Aaron Courville. [Deep Learning](#).

# Problem Sets (35% of grade)

- Bi-weekly problems sets
  - Python coding problems
  - Written math problems
  - Prepare to spend 15+ hours on each
  - Schedule on the website.
- **Late Policy:** 20% penalty per day up to **two days**
  - No special requests
  - Submissions not accepted after two days
- Regrade requests must be submitted 1 week after grades are released

# Quizzes (20% of grade)

- Five quizzes
  - (Primarily) multiple choice
  - Timed: must be completed once started with a limited time window for completion
  - 1 week to complete (no late submissions accepted)

# Participation (10% of grade)

- Should make a contribution every week such as
  - In-class participation (e.g., via **slido**)
  - Ask questions in class.
  - Respond to other student's questions on Piazza



**What word comes to mind  
when you hear "machine  
learning"?**

ⓘ Start presenting to display the poll results on this slide.

# Example Piazza question/responses

<b>One way to ask</b>	What is concept “X”?  A: Look at section Z of book and/or lecture Y.
<b>Better way to ask</b>	Regarding concept “X”, what is the role of “Y”?  A: Explanation..  <u>Key: Be as specific as possible.</u>
<b>One way to ask</b>	Here are my results (a dump of numbers), do they look right?  A: ...
<b>Better way to ask</b>	Here are my loss plot, I suspect overfitting. Did anyone in the class see this trend?  A: Yes - did you try any regularizers?  <u>Key: Be as specific as possible.</u>

# What questions won't get answered by course staff?

- If they are too vague
- If they have already been asked before.
  - Search before you ask.
- **Be respectful of the course staff's time.**

# Labs (20% of grade)

- Hands-on exercises graded based on participation
  - **Bring your laptop** so you can participate
  - Pay attention, be respectful of the instructor
  - Lab materials will be posted to piazza
  - Graded based on **active participation**
- What if you can't make a lab?
  - **Can miss up to 2 labs** throughout the semester.
    - Half credit will be awarded for completing the lab exercise and submitted on gradescope (deadline is the Friday immediately following the lab)

# Class Challenge (15% of grade)



Individual end-of-term project

- Based on a real-world problem, hosted as a private challenge on Kaggle or Gradescope
- Goal is to implement a machine learning approach and apply it to the problem
- **Deliverables:**
  - Original code implementing solutions
  - In class presentations.

# Summary: Grade Breakdown

Problem sets	35%
Quizzes	20%
Participation	10%
Labs	20%
Class Challenge	15%

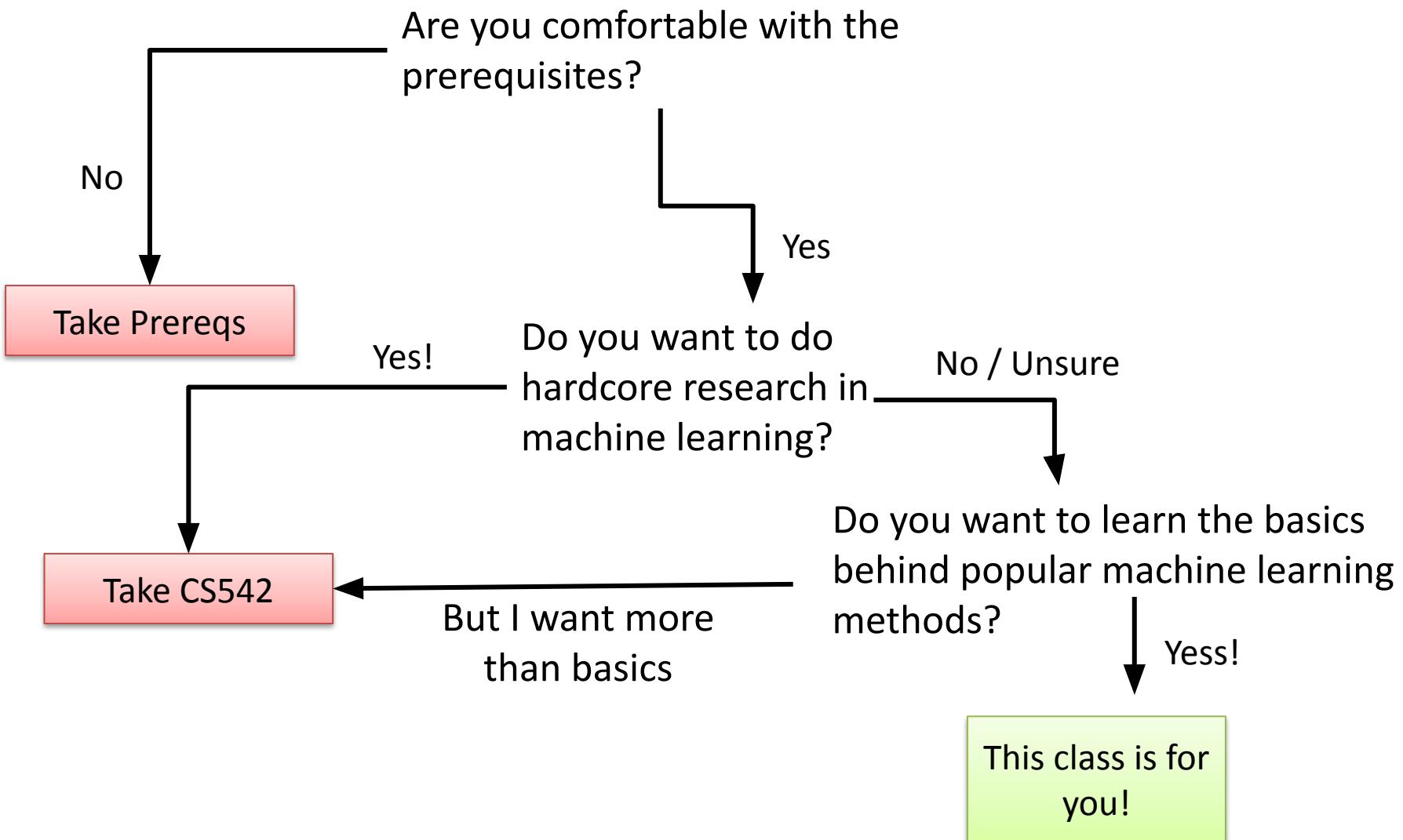
# Is this class for you?

	Yes	No
1. Do you want to do research in machine learning or artificial intelligence?	<input type="radio"/>	<input type="radio"/>
2. Do you want pursue a PhD (especially in machine learning)?	<input type="radio"/>	<input type="radio"/>
3. Do you want to learn <i>both</i> the math and the code behind popular machine learning methods?	<input type="radio"/>	<input type="radio"/>
4. Are you comfortable with the prerequisites?	<input type="radio"/>	<input type="radio"/>

## Alternative Machine Learning Courses

- [CAS CS 542 Principles of Machine Learning](#) designed to provide students the background to conduct research in machine learning
  - Math intensive, theoretical.

# Is this class for you?



# Summary: Is this class for you?

*CS542 Principles of Machine Learning* and *541 Applied Machine Learning* are alternatives for substantially similar material.

- Your program may only accept one of them for credit towards your degree
- Take CS542 if you want to do machine learning research (i.e., create new machine learning tools)
- Take this course if you want to obtain the background to use machine learning tools effectively.

# Course climate

- Some may be more familiar with some topics, some less, and that's ok!
- Please respect each other and listen to each other carefully.
- **It is important to actively participate.**
- No screens (laptops, tablets, phones) during the class.

# Course goals

Understand fundamental and popular approaches of machine learning

- Get hands-on experience in coding ML algorithms.
- Maximize your learning experience.
- Have fun!

# Academic Policy

- Zero tolerance policy for academic dishonesty.
  - Copying from fellow students
  - Using an LLM to generate code or solving assignments.
- Will be reported directly to the academic conduct committee.
- Detailed academic [policy](#)

# Today

- Course logistics
- **What is machine learning?**
- Classification introduction

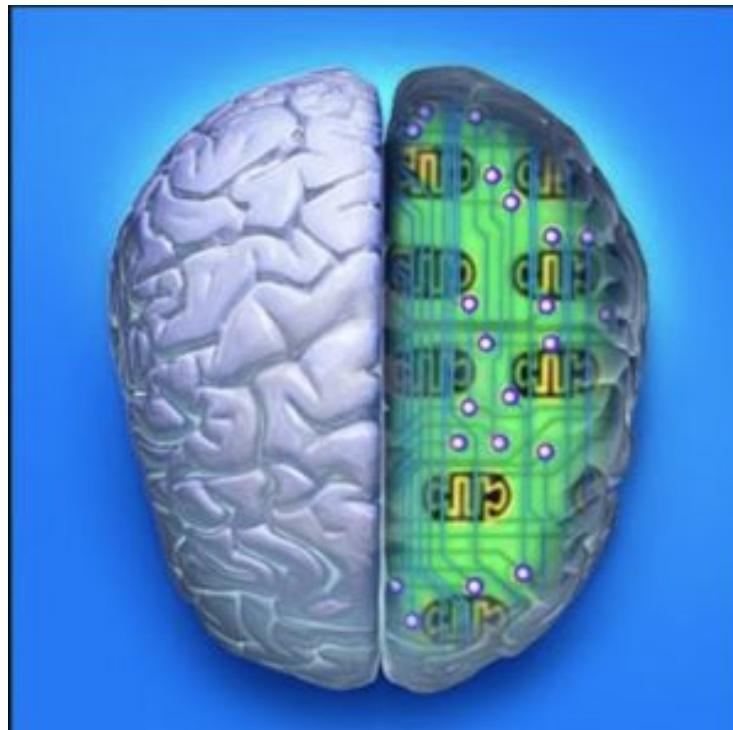


# What is Machine Learning?

- ⓘ Click **Present with Slido** or install our [Chrome extension](#) to activate this poll while presenting.

# What is Machine Learning?

*Build algorithms capable of imitating human intelligence.*



*learn from data*



# Why do we need Machine Learning?

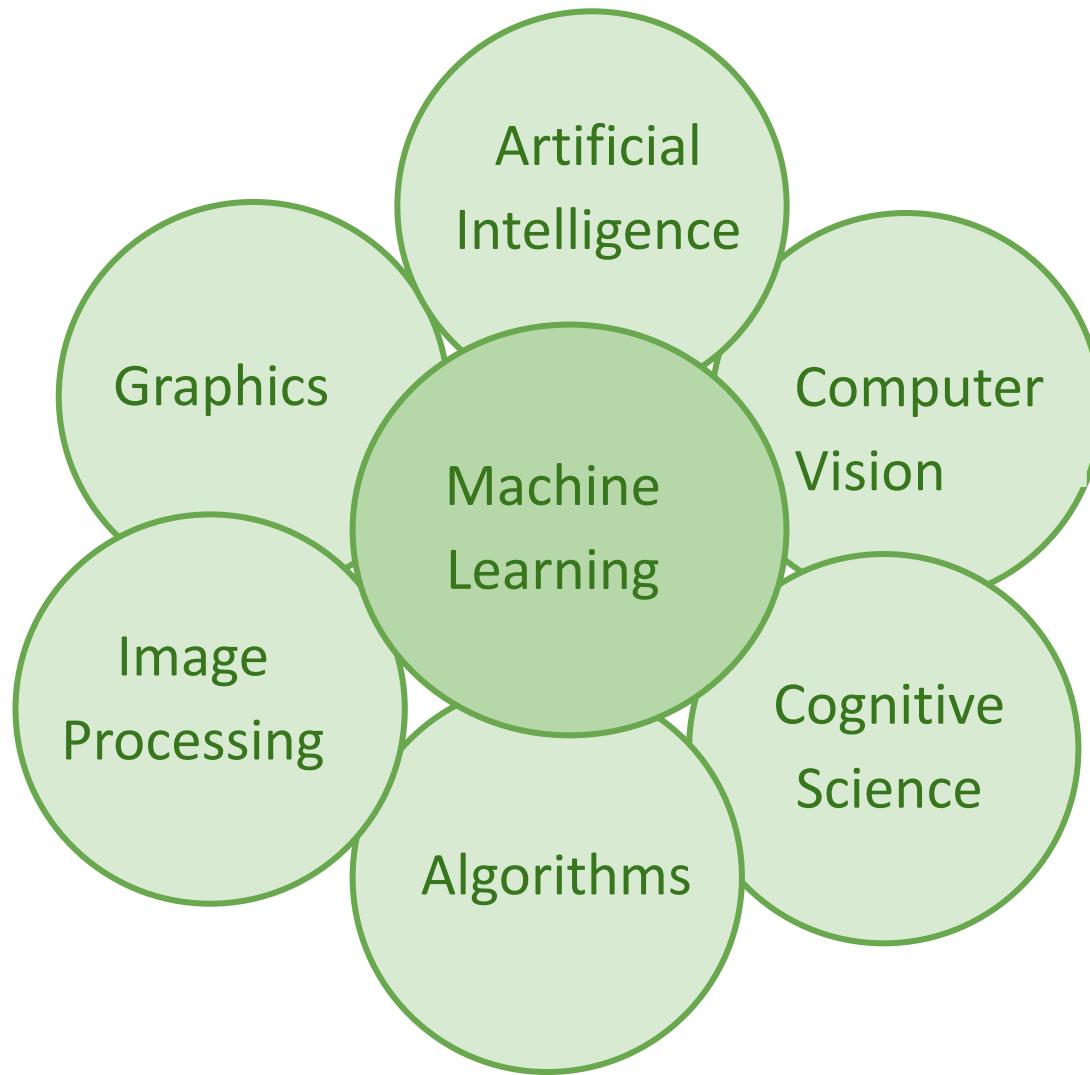
- ⓘ Click **Present with Slido** or install our [Chrome extension](#) to activate this poll while presenting.

# Machine Learning:

## Why do we need it?

- Help automate boring, hard tasks
- Hard to program computer directly to do the task
- Instead, program a computer to **learn** from examples
- Often use “big data” examples.

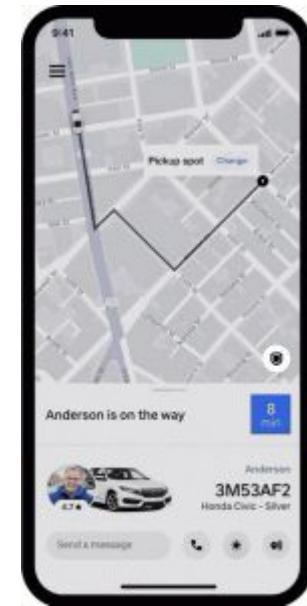
# Relation to other fields





# Machine Learning:

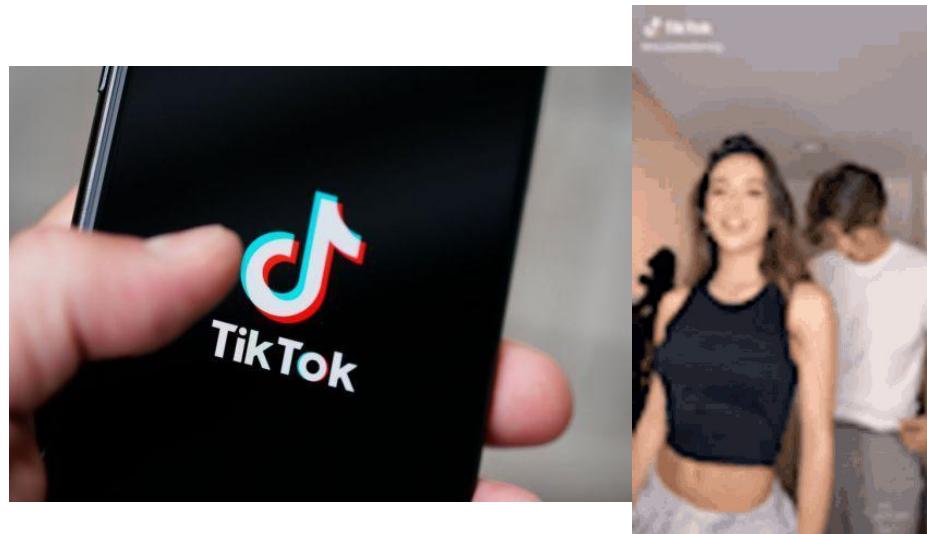
## name three ways you used it today!



# Machine Learning in Real Life: social media & entertainment



AI Speakers & assistants

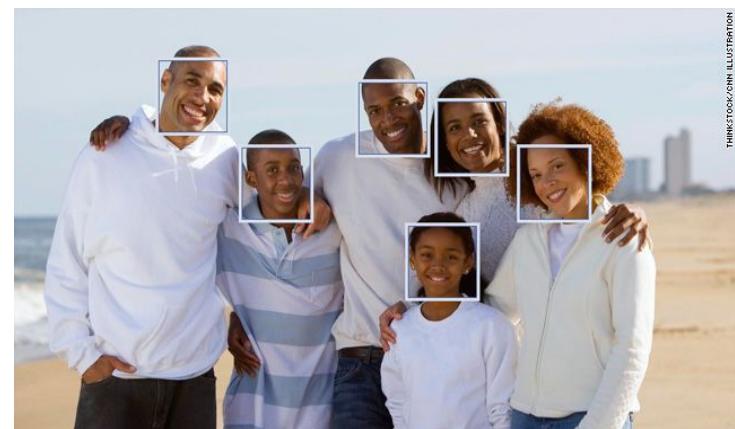


Recommendation algorithm on Tiktok

**Other Movies You Might Enjoy**

[Amelie](#)  
  
[Add](#)  


[Y Tu Mama Tambien](#)  
  
[Add](#)  

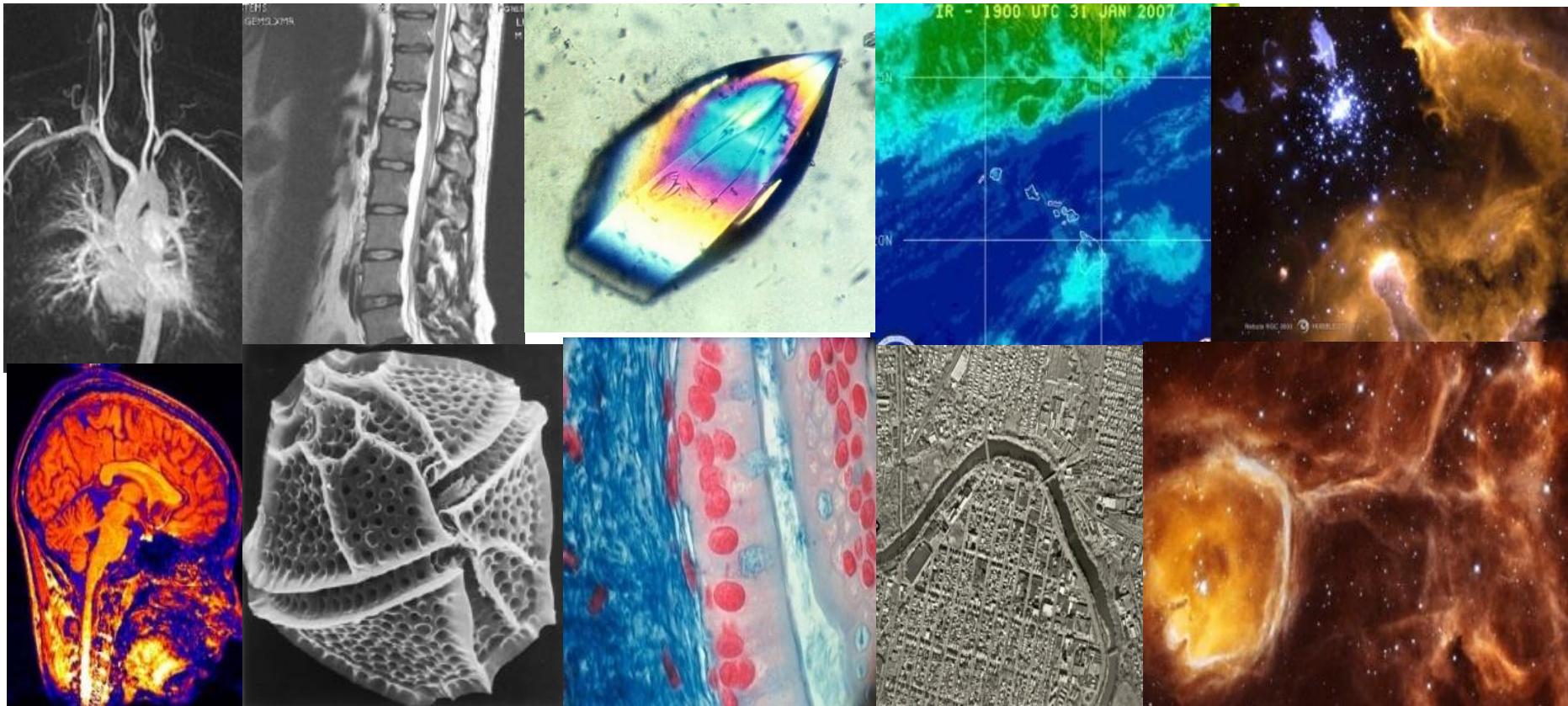
face tagging on social media

# Machine Learning in Real Life: Smart Cars



- Stanford/Google one of the first to develop self-driving cars
- Cars “see” using many sensors: radar, laser, cameras

# Machine Learning in Real Life: Medical and Scientific Data

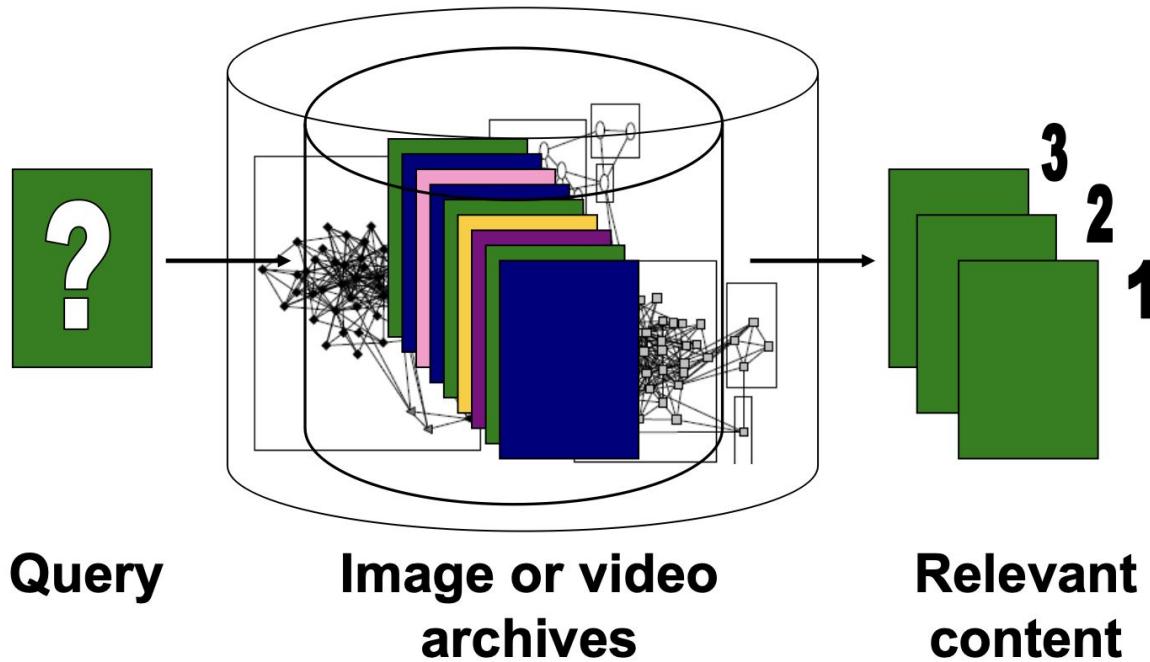


- Healthcare, medical image analysis
- Climate and geography
- Biology
- Astronomy.

# Machine Learning in Real Life: Robotics



# Machine Learning in Real Life: Search and data organization



**Sample applications:** Shop for products visually.

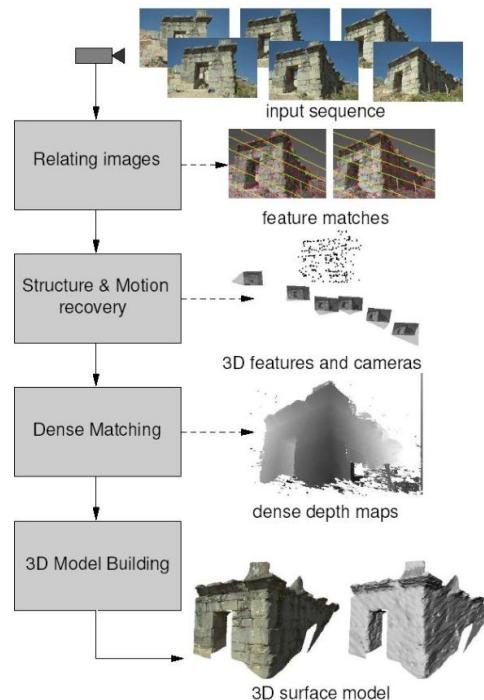
# Machine Learning in Real Life: Measurement

Real-time stereo

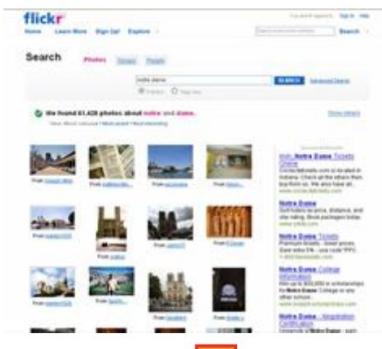


Pollefeys et al.

Structure from motion



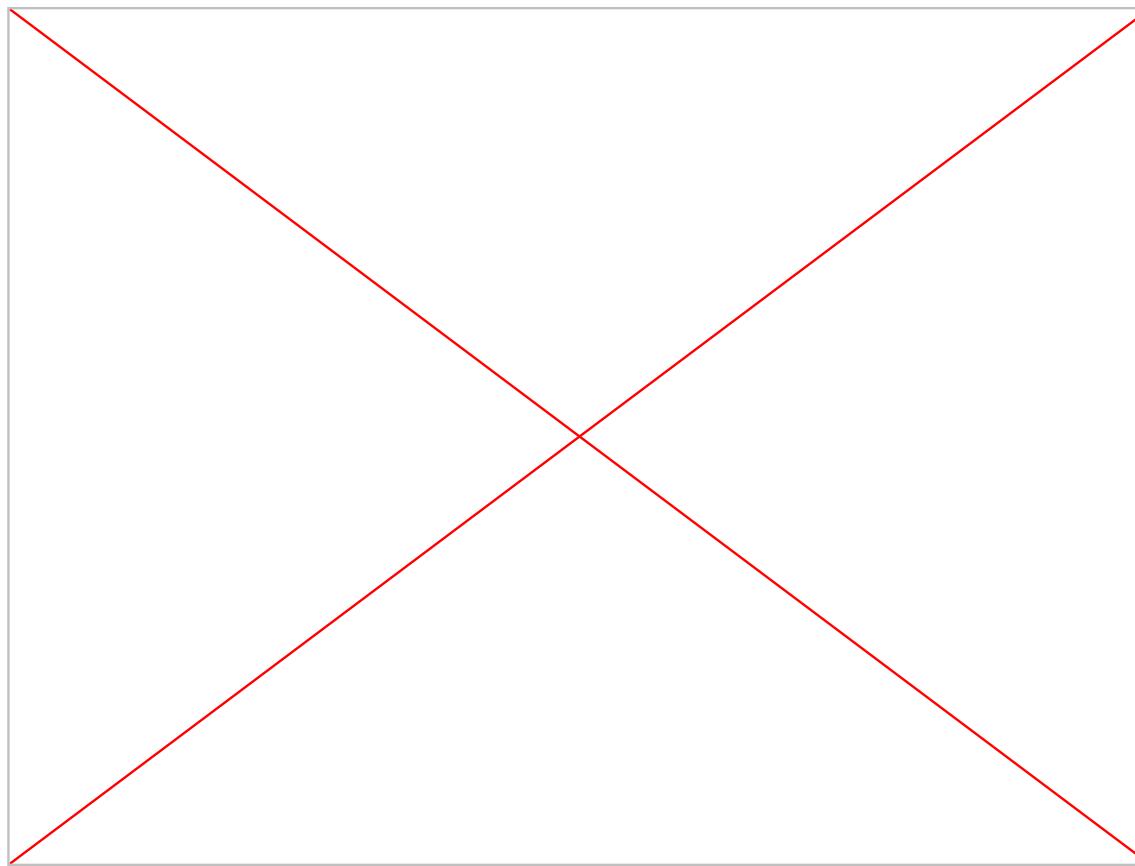
Multi-view stereo for  
community photo collections



Goesele et al.

**Sample applications:** Build virtual immersive worlds, self-driving cars

# Machine Learning in Real Life: Generation



**Sample applications:** automatic media creation, generative art, etc.

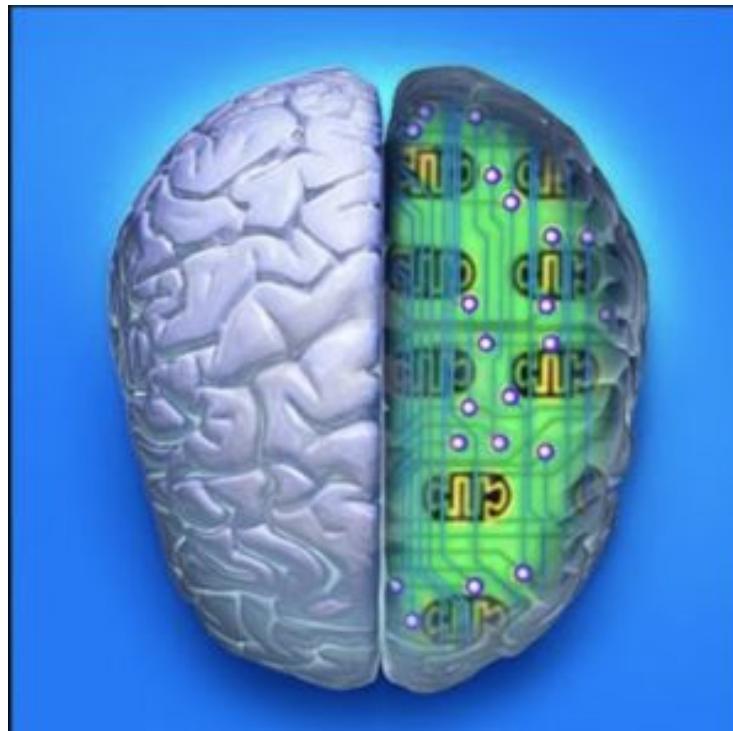
# Machine Learning in Real Life: Computational Finance

- Predict stock price, cash flows, financing programs, optimize pricing for merchants, fraud detection, etc.



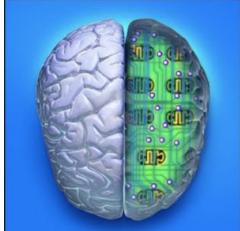
# What is Machine Learning?

*Build algorithms capable of imitating human intelligence.*

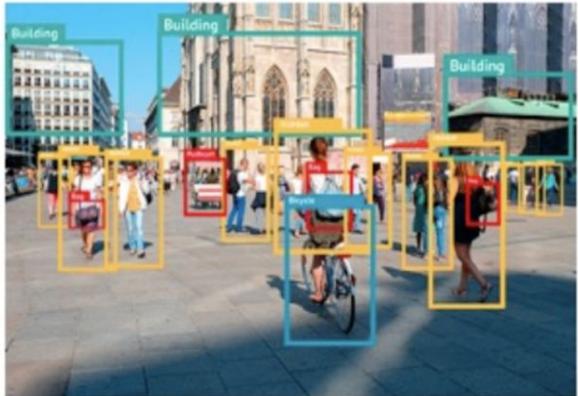


*learn from data*

# So much progress!



**Scene planning**



**Robot-assisted surgery**



**Drug discovery**



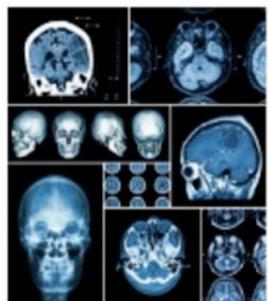
**Autonomous vehicles**



**Facial recognition**



**Robotics**



**Diagnostics**

# What helped us get here?



# What helped us get here?



Slide credit: Justin Johnson

# Types of learning



Supervised



Unsupervised



Reinforcement



# Supervised Learning

- Given a **training set** consisting of **inputs** and **outputs**, learn to map novel, unseen inputs to outputs
- The novel inputs are called a **test set**

Training set	Used to learn an algorithm $f$	Set of images
Input	Some attributes of underlying data	Image features
Output	A property we want to predict.	Cat / not cat?

Categorical



# Supervised Learning

- Given a **training set** consisting of **inputs** and **outputs**, learn to map novel, unseen inputs to outputs
- The novel inputs are called a **test set**

Training set	Used to learn an algorithm $f$	Financial data
Input	Some attributes of underlying data	Market info
Output	A property we want to predict.	Stock price

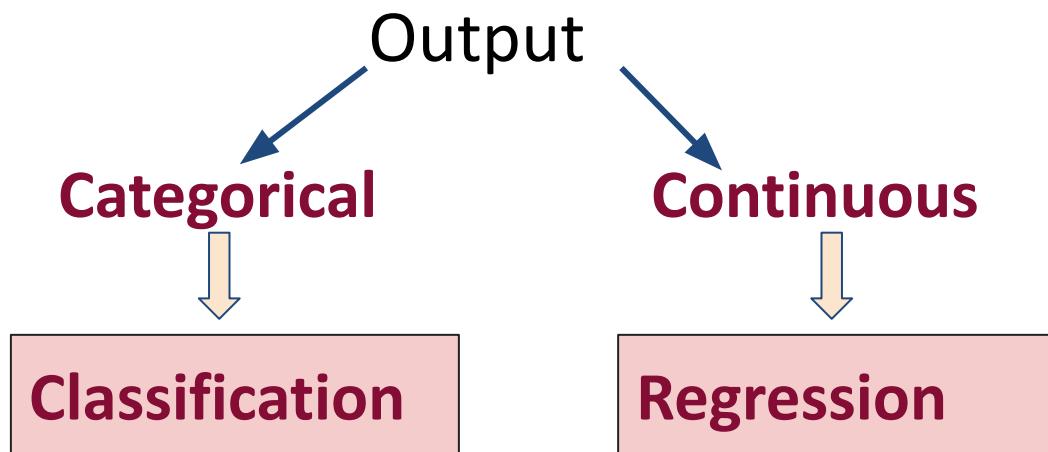
Continuous





# Supervised Learning

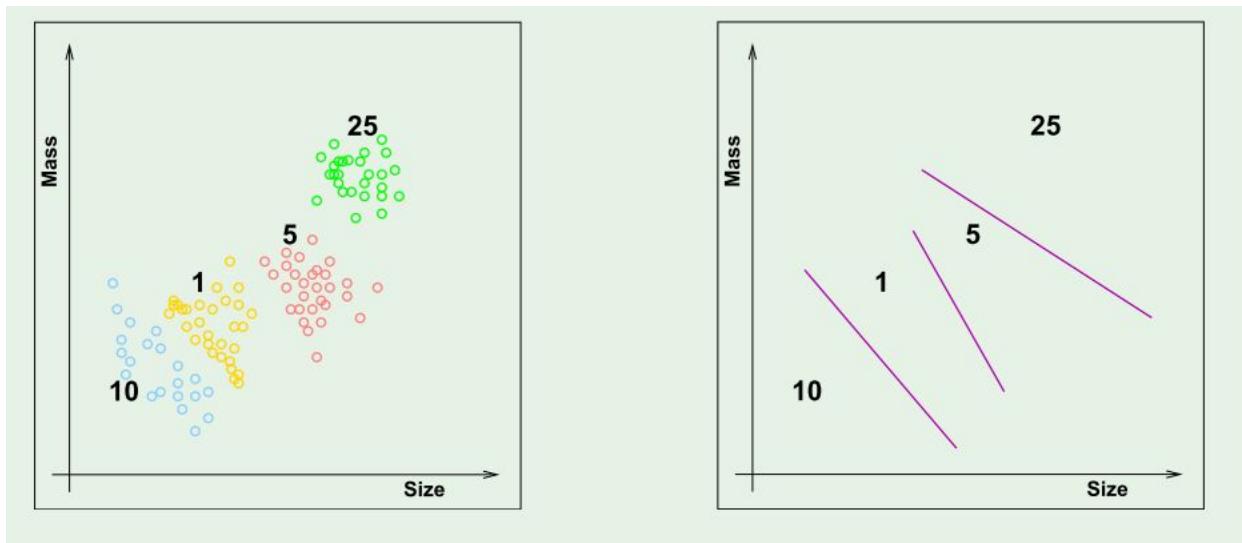
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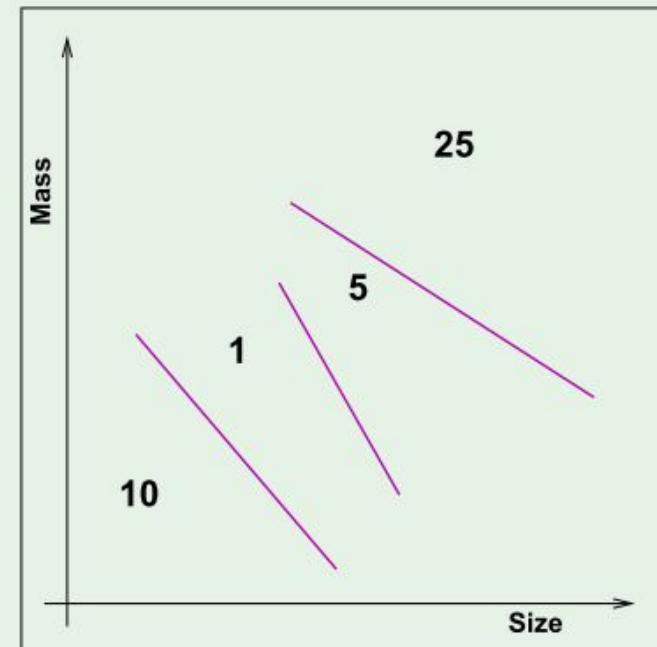
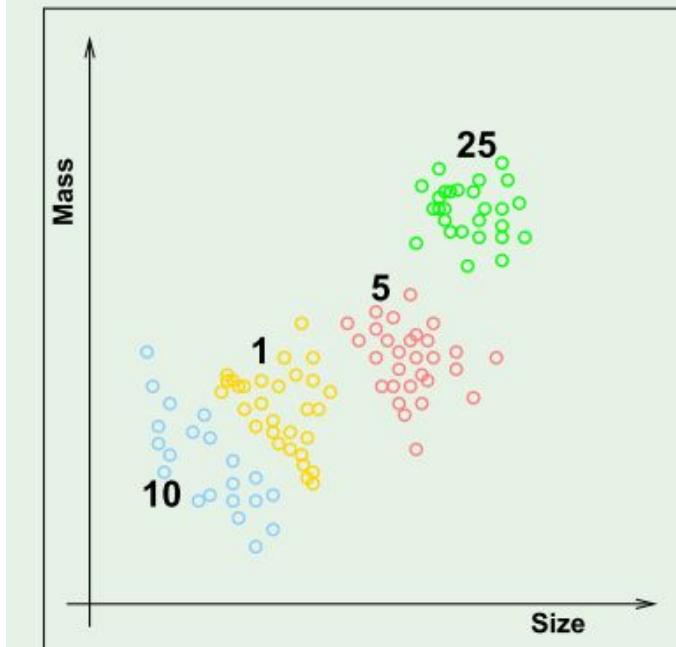
# Example of Supervised Learning

recognize coins



- Given training set consisting of coin denomination (penny, nickel, dime, quarter), mass and size
- Learn to predict denomination

Input	Mass, size
Output	Denomination



# From the plots, what are the inputs to the model?

- ⓘ Click Present with Slido or install our [Chrome extension](#) to activate this poll while presenting.



## From the plots, what are the inputs to the model?

Multiple Choice Poll  56 votes  56 participants

Size, Mass - 48 votes



Size, Mass, Denomination - 6 votes



Mass, Denomination - 1 vote

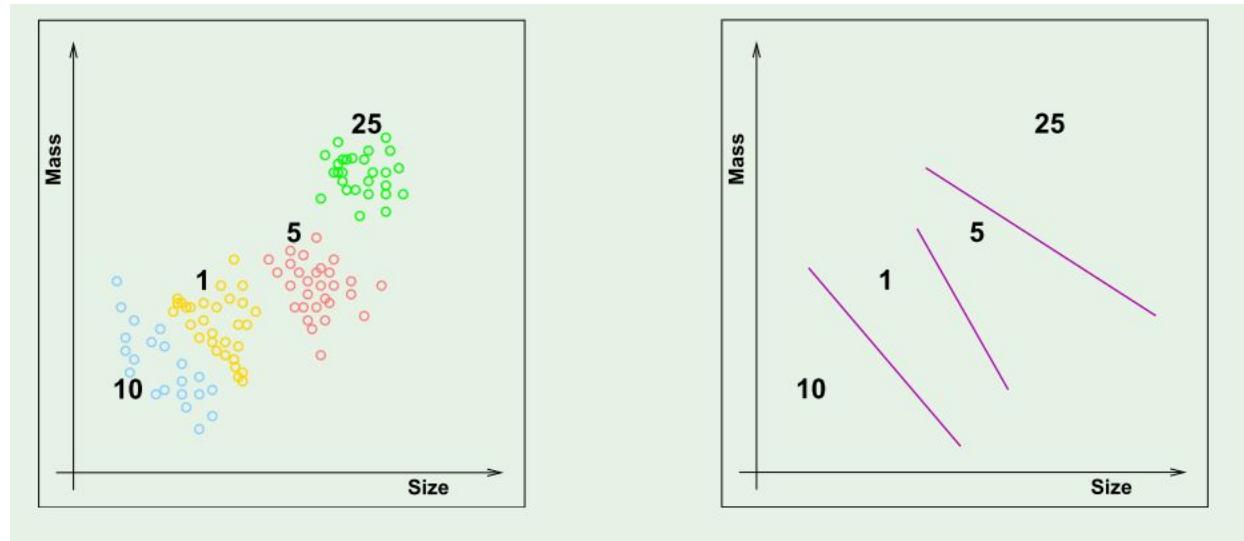


Size, Denomination - 1 vote



# Example of Supervised Learning

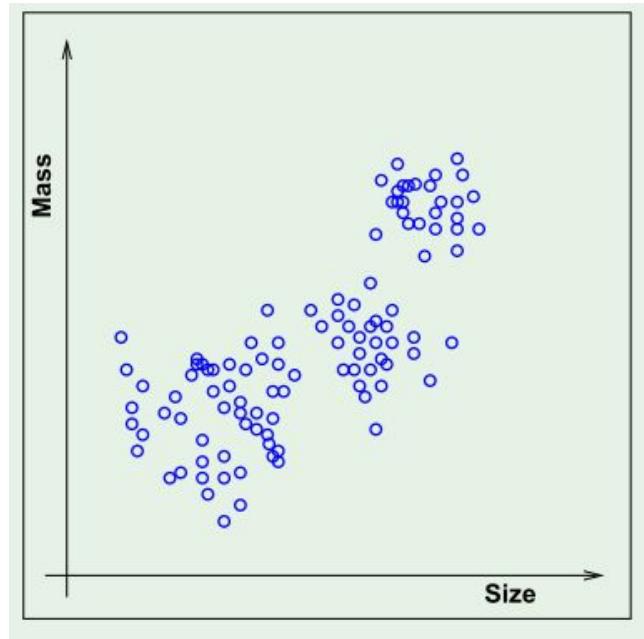
recognize coins



- Given training set consisting of coin denomination (penny, nickel, dime, quarter), mass and size
- Learn to predict denomination

Input	Mass, size
Output	Denomination

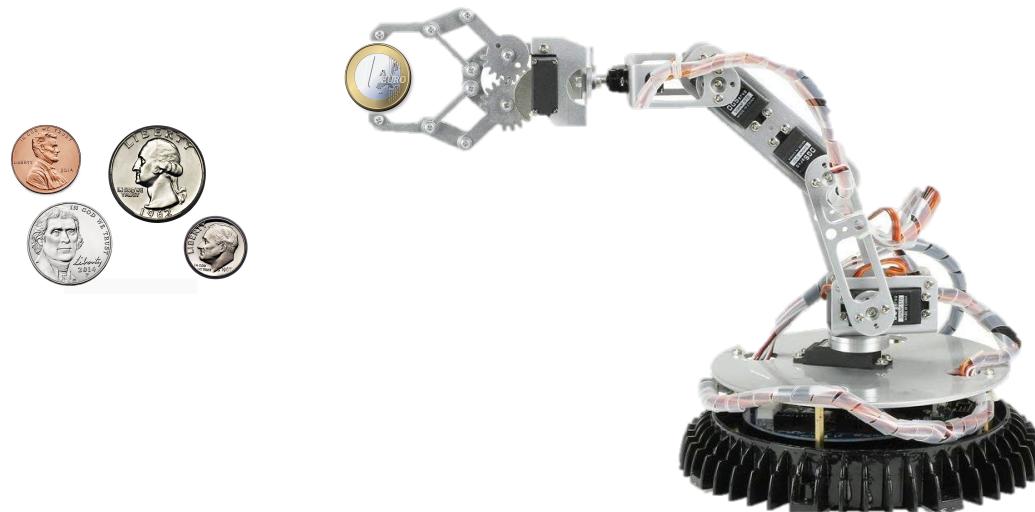
# Unsupervised Learning



- Given training set consisting of ~~coin denomination (penny, nickel, dime, quarter)~~ mass and size
- Learn... something?

# Reinforcement Learning

learn to pick up coins



- Given only input, but can take action
- Predict output (action), get a reward for it

# Lab sessions tomorrow

- Introduction to numpy
- Hands-on exercises graded based on participation
  - **Bring your laptop** so you can participate
  - Lab materials will be posted to piazza
  - Upload your lab solutions on gradescope
  - Graded based on participation

# Next Class

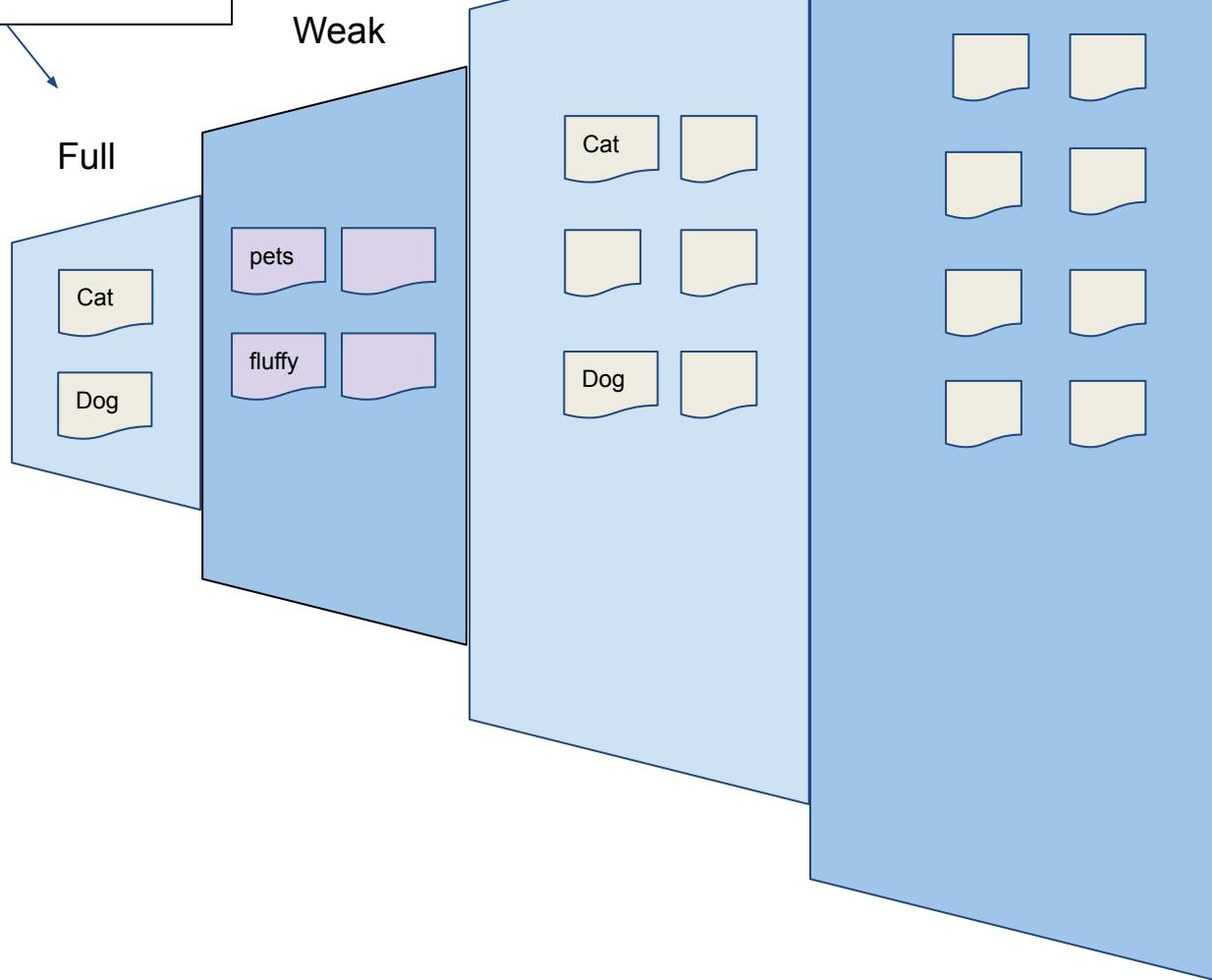
## **Classification I:**

Classification intro, error rates, nearest neighbors;

**Reading:** Forsyth Ch 1.1-1.3

# Types of learning

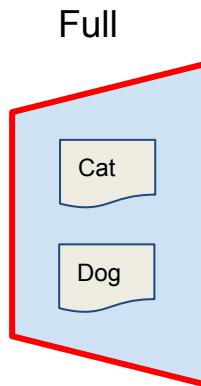
Most expensive!  
Least amount of noise.



Least expensive!  
Large amount of noise.

# Levels of supervision

**Most expensive!**  
Least amount of noise.



Weak

Semi

Self

