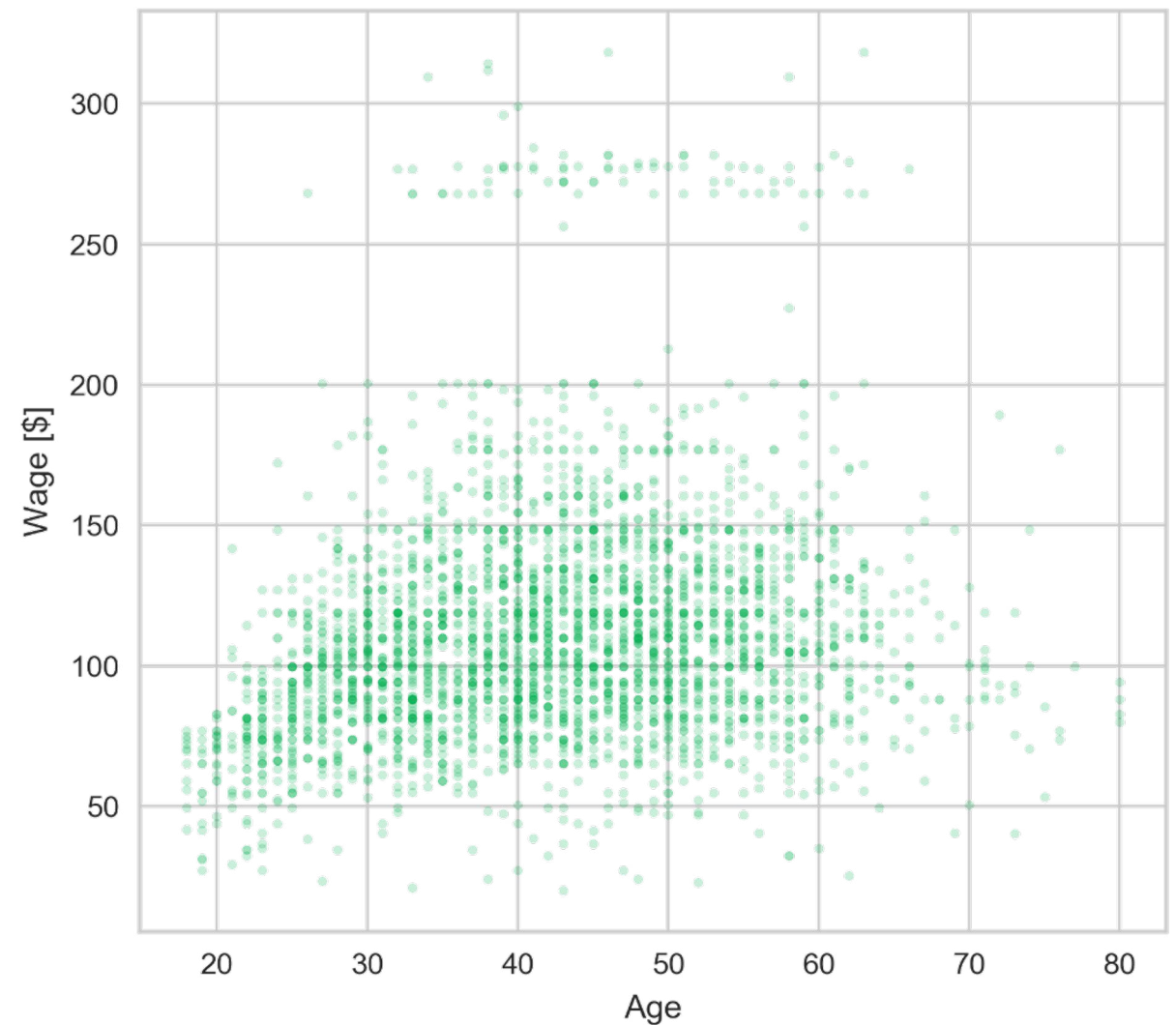
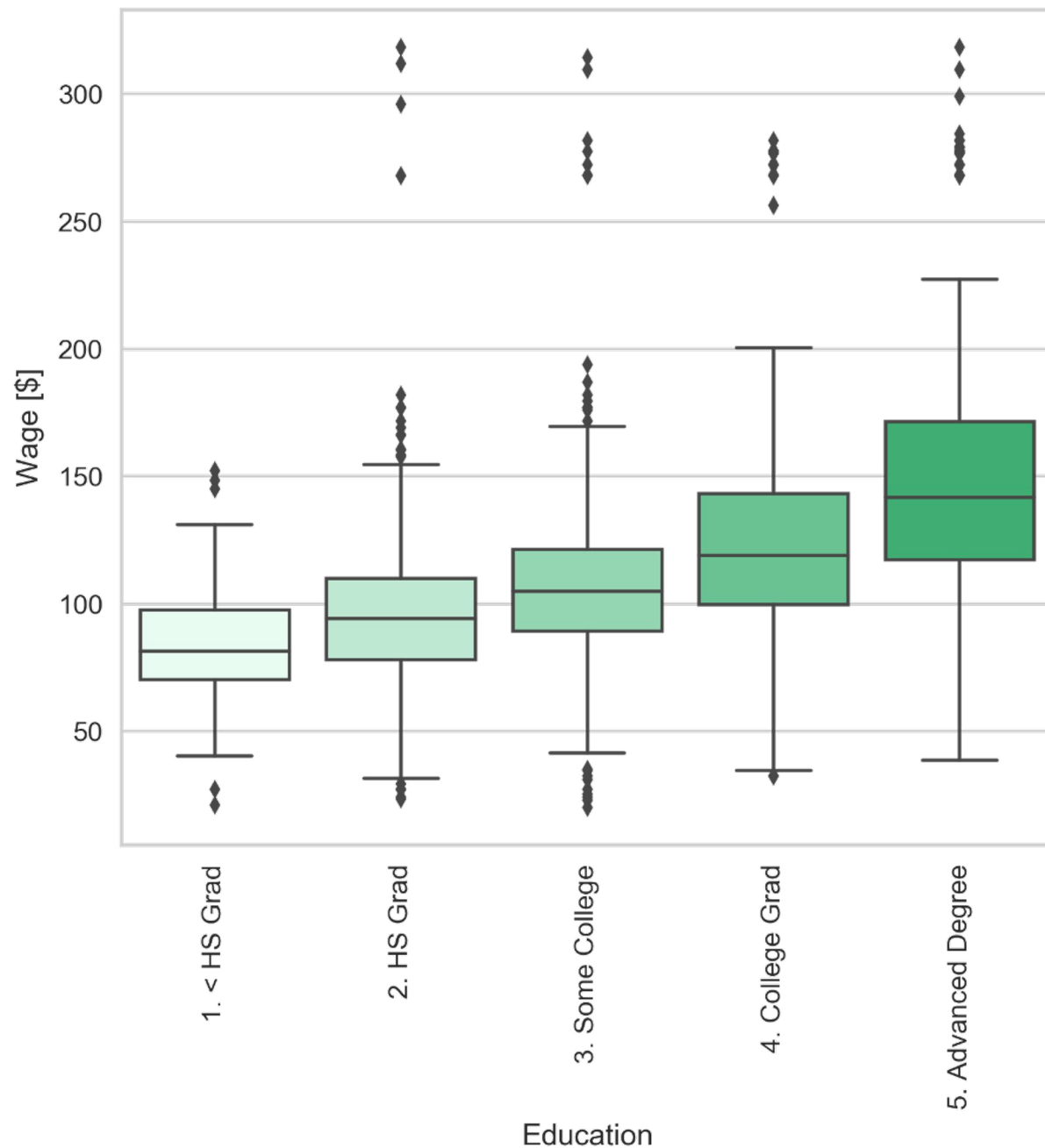


What is machine learning?

Lecture 01



Wage data from workers in the mid-Atlantic region

How do you predict how much someone will make?

How can you tell these flowers apart?

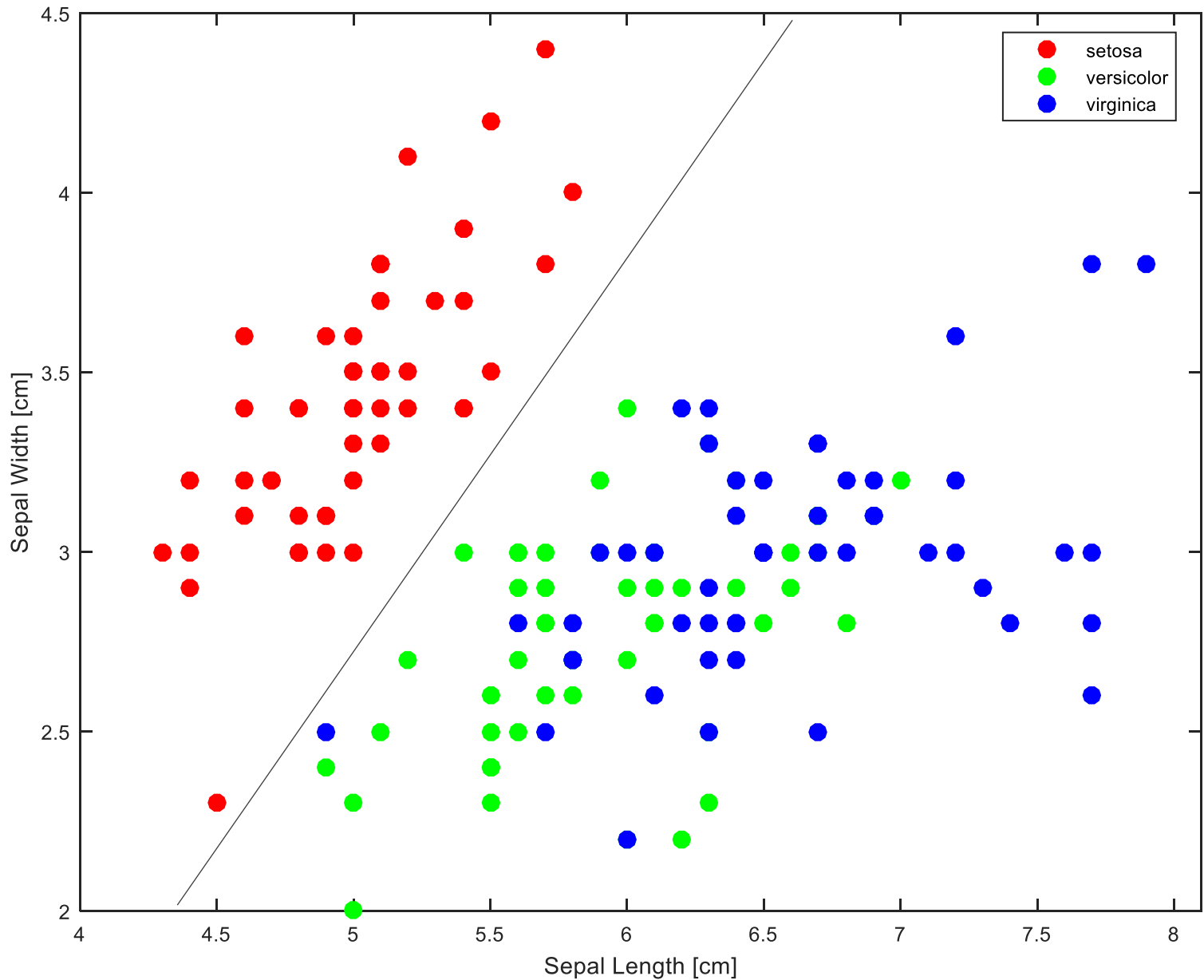


Iris setosa



Iris virginica

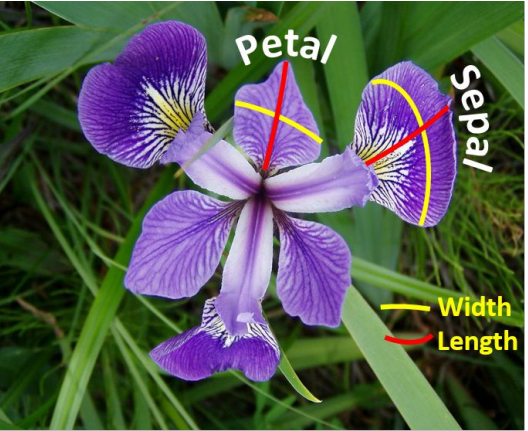
Image Sources: Srishti Sawla (setosa) and Ivo Dinov, University of Michigan SOCR (versicolor and verginica)



setosa

versicolor

virginica



Data Source: Fisher Iris Data
Image Sources: Srishti Sawla (*setosa*) and Ivo Dinov, University of Michigan SOCR (*versicolor* and *virginica*)

Challenges



What
is
this?

We **generalize** from past experiences



Image: "It's not what it seems" by artist Hikaru Cho

our data must be
representative

How about this one?

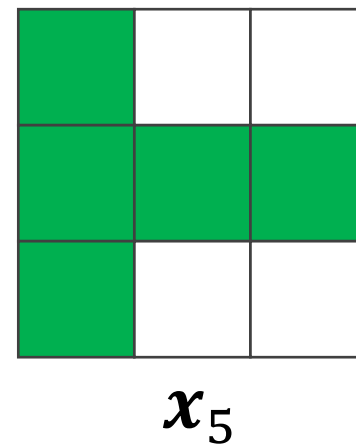
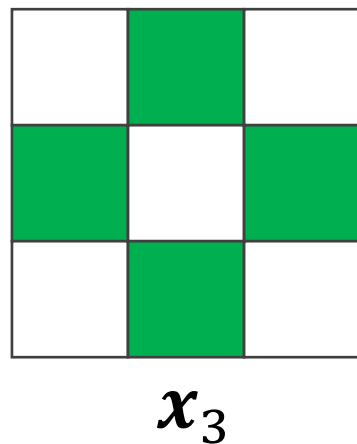
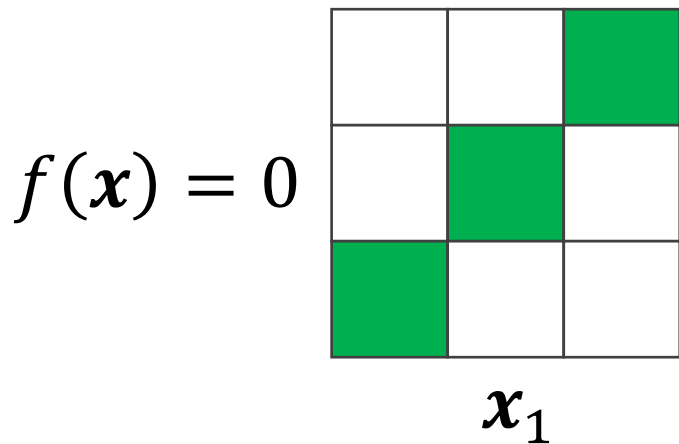
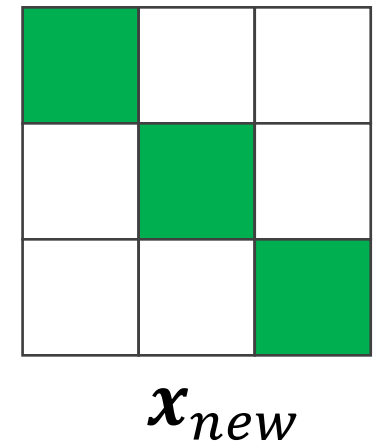
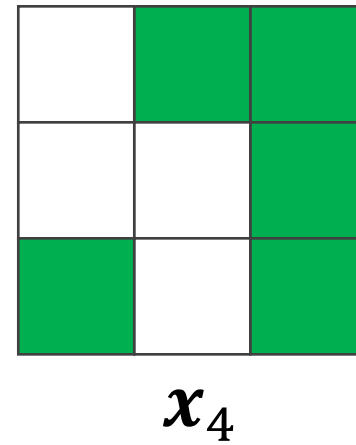
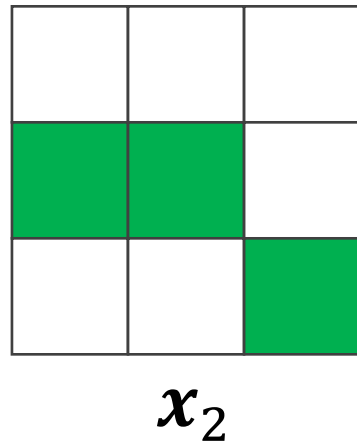
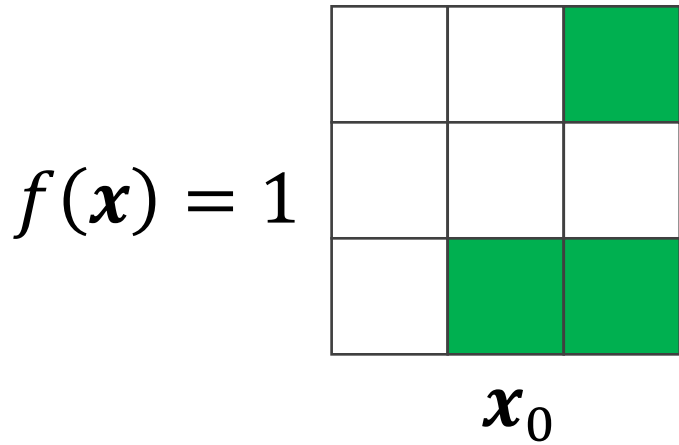


Image by artist Hikaru Cho



Image: "It's not what it seems" by artist Hikaru Cho

Predict which class x_{new} belongs to...



$f(x_{\text{new}}) = ?$

Machine learning is an **ill-posed problem**

There are often **many** models that fit
your **training data** similarly well

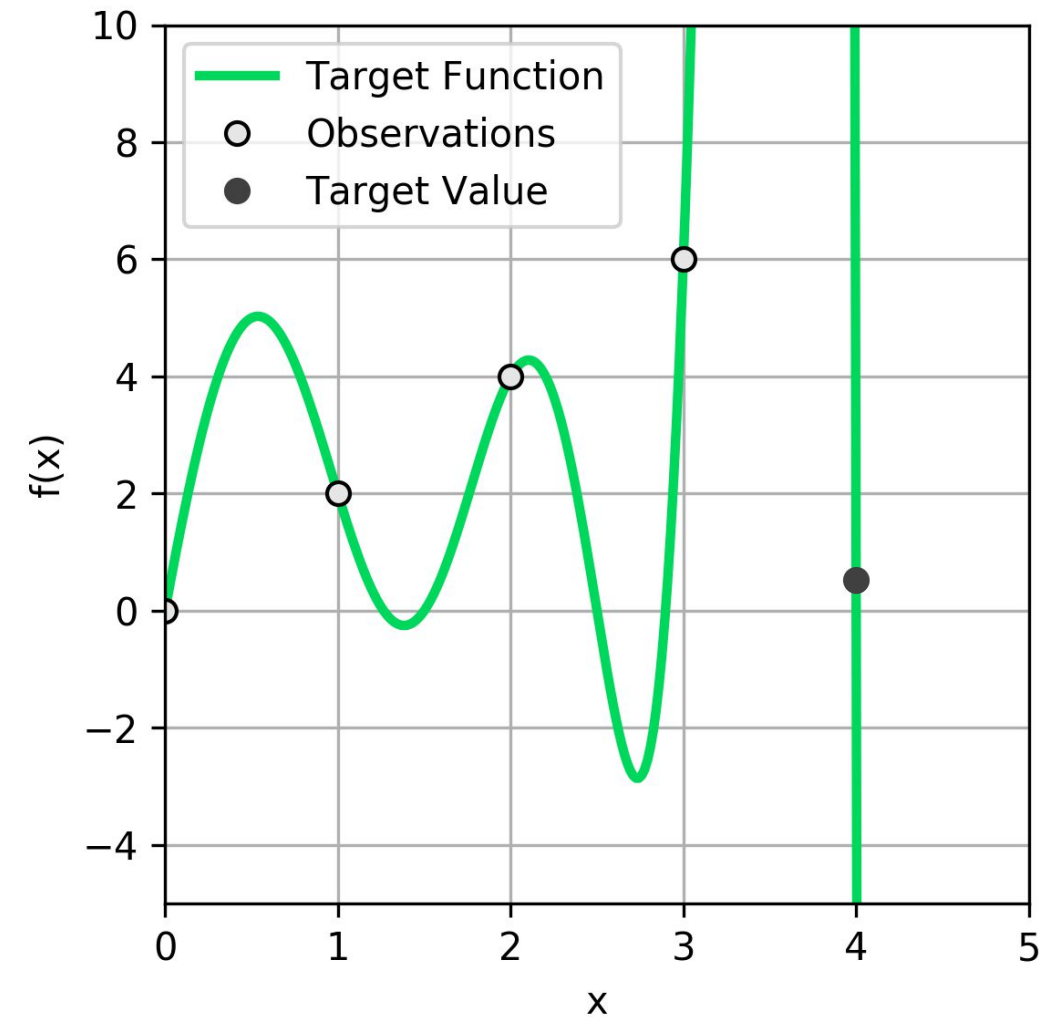
So how do we choose which to use?

the best models
generalize well

Predict the next value in the sequence...

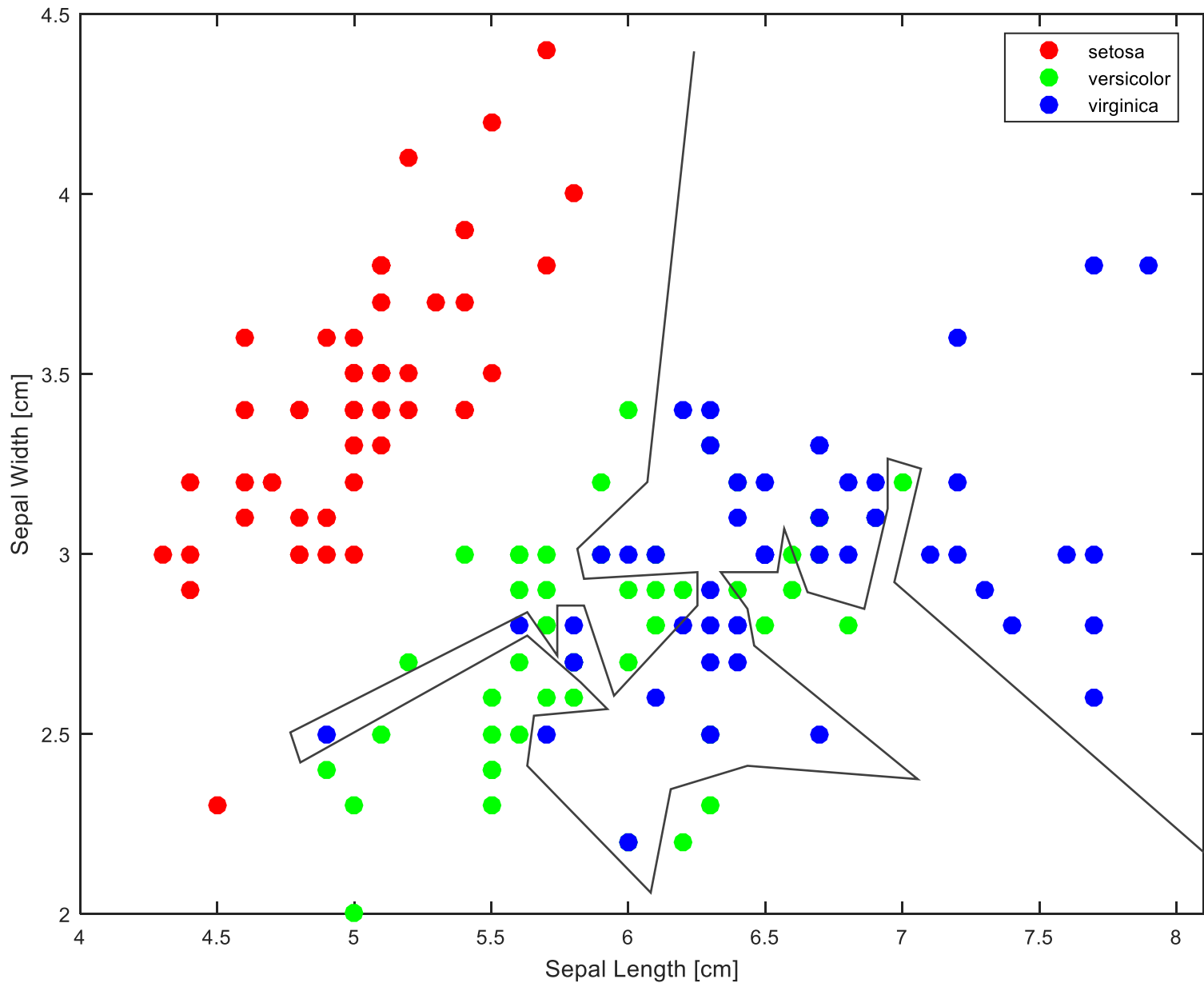
x	0	1	2	3	4
$f(x)$	0	2	4	6	?

$$f(4) = \boxed{0.530}$$



A model:

$$f(x) = 16.2x - 6.36x^2 - 11.9x^3 - 4.77x^4 + 7.03x^5 + 8.32x^6 - 9.01x^7 + 2.75x^8 - 0.275x^9$$



setosa



versicolor



virginica



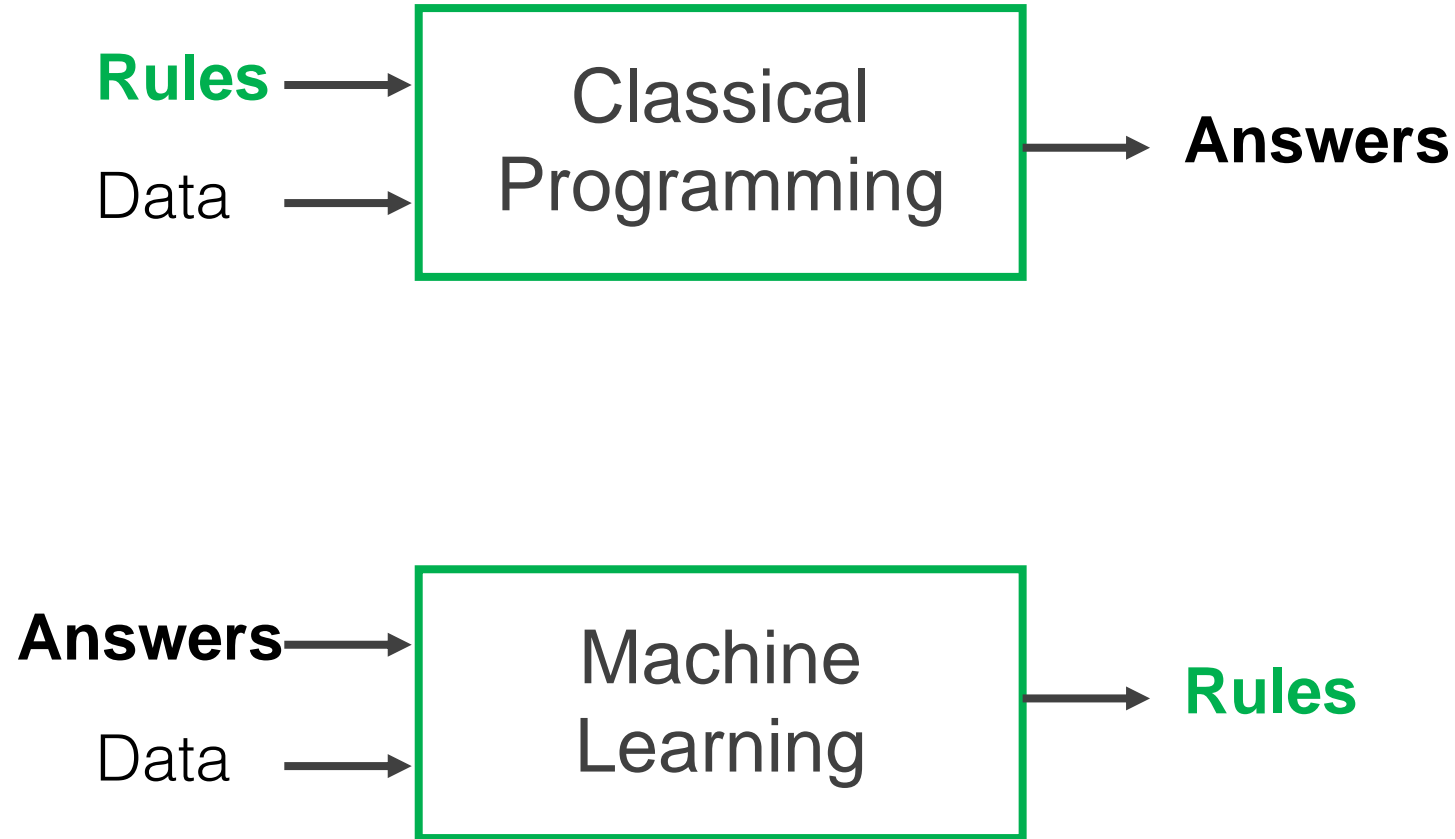
Complex models overfit to the data

overfit works against
generalization



**Learning representations of data
enables learning**

Machine learning suggests an alternative programming paradigm



François Chollet, *Deep Learning with Python*, 2017

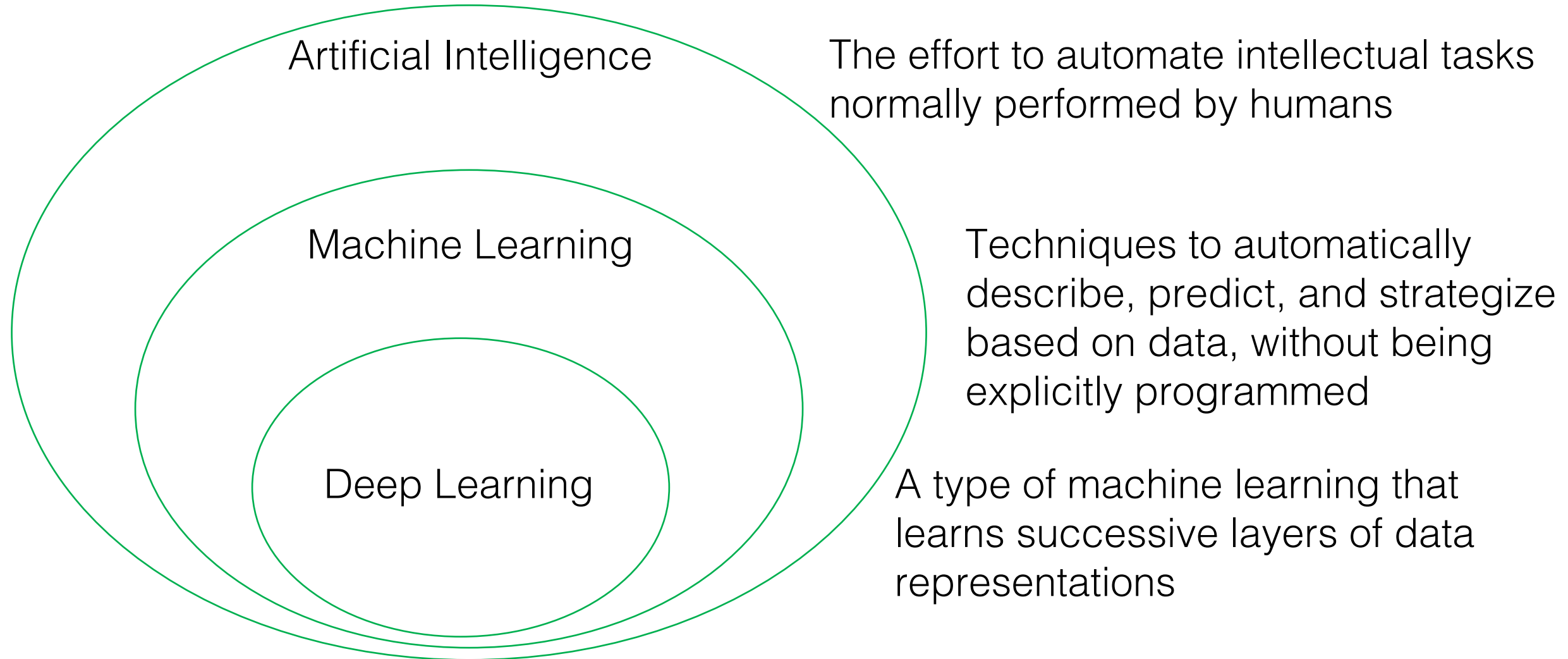
What is machine learning?

A class of techniques where the **goal** is to **describe**, **predict**, and **strategize**...

...**based on** data and past experience...

...and do so **automatically**, with minimal human intervention.

What is machine learning?



François Chollet, *Deep Learning with Python*, 2017

Types of machine learning tools

Types of learning

Unsupervised learning

Supervised learning

Reinforcement learning

Common use case

Describe

Predict

Strategize

Types of machine learning

	Supervised Learning	Unsupervised Learning	Reinforcement Learning
Goal	Predict ...from examples	Describe ...structure in data	Strategize learn by trial and error
Data	(x, y)	x	delayed feedback
Types	<ul style="list-style-type: none">• Classification• Regression	<ul style="list-style-type: none">• Density estimation• Clustering• Dimensionality reduction• Anomaly detection	<ul style="list-style-type: none">• Model-free learning• Model-based learning

Sale Price Prediction

Input Data:

Home characteristics
(Numerical & Categorical)

Target Data:

Price estimate (numerical)

Learning Category:

Supervised Learning
Regression



Kyle Bradbury

27708 Real Estate

1 home for sale

[Homes for You](#) [Newest](#) [Cheapest](#) [More](#)

HOUSE FOR SALE
\$599,900 5 bds · 4 ba · 3,264 sqft
1640 Marion Ave, Durham, NC

FOR SALE
\$599,900
Price cut: -\$79,100 (6/17)
Zestimate®: \$619,585

EST. MORTGAGE
\$2,284/mo

[Get pre-qualified](#)

Zestimate®: \$619,585

1640 Marion Ave, Durham, NC 27705

5 beds · 4 baths · 3,264 sqft

SPACIOUS RANCH W FINISHED LL WALKOUT! 5 BEDROOMS AND 4 BRAND NEW BATHS! RENOVATED WITH CUSTOM FEATURES THRUOUT! CONTEMPORARY HOME WITH MANY HANDICAP ACCESSIBLE REQUIREMENTS ALREADY IN PLACE! VAULTED CEILINGS! SECLUDED TREED LOT! GREAT HOME FOR LIVING AND ENTERTAINING WITH LARGE REAR DECK! WONDERFUL CONTEMPORARY FEEL THAT LIVES LARGE WITH EASY ACCESS TO DUKE UNIVERSITY: SHOPPING; HEALTH CARE; PARKS; R SHOPPING; AND EASY HIGHWAY AC

Spam Filters

From: Internal Revenue Service
[mailto:yourtaxrefund@InternalRevenueService.com]

Sent: Tuesday, July 22, 2008 9:47 AM

Subject: Get your tax refund now

Importance: High

After the last annual calculations of your account activity we have determined that you are eligible to receive a tax refund of \$479.30 .

Please submit the tax refund request and allow us 2-6 days in order to process it.

A refund can be delayed for a variety of reasons. For example submitting invalid records or applying after the deadline.

To access the form for your tax refund, please click here (<http://e-dlogs.rta.mi.th:84/www.irs.gov/>)

Note: Deliberate wrong inputs will be prosecuted by law.

Regards,

Internal Revenue Service

Input Data:

Email text (text)

Target Data :

Spam/not spam
(category)

Learning Category:

Supervised Learning
Classification (binary)

Spam example source: itservices.uchicago.edu

Where's Waldo = Computer Vision Problem



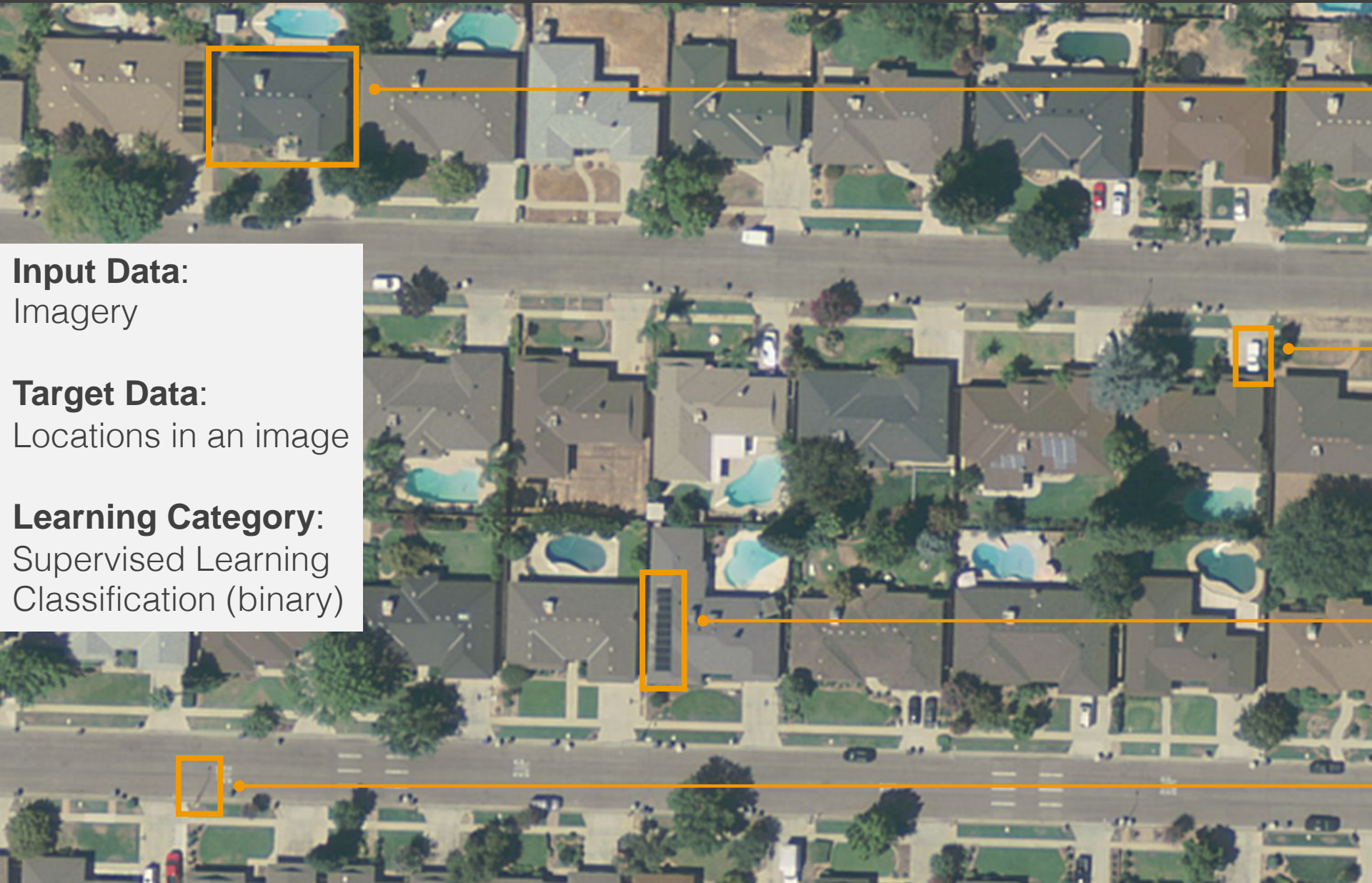
Input Data:
Color Imagery (Image)

Target Data:
Locations in an image
(label for each pixel)

Learning Category:
Supervised Learning
Classification (binary)

Image source: www.whereswaldo.com/

Object Recognition: Energy Systems



Input Data:

Imagery

Target Data:

Locations in an image

Learning Category:

Supervised Learning
Classification (binary)

Building

behind-the-meter
energy consumption

Car

transportation
energy consumption

Solar Array

distributed energy
resources

Light Pole

access to electricity

Credit Fraud

Input Data:

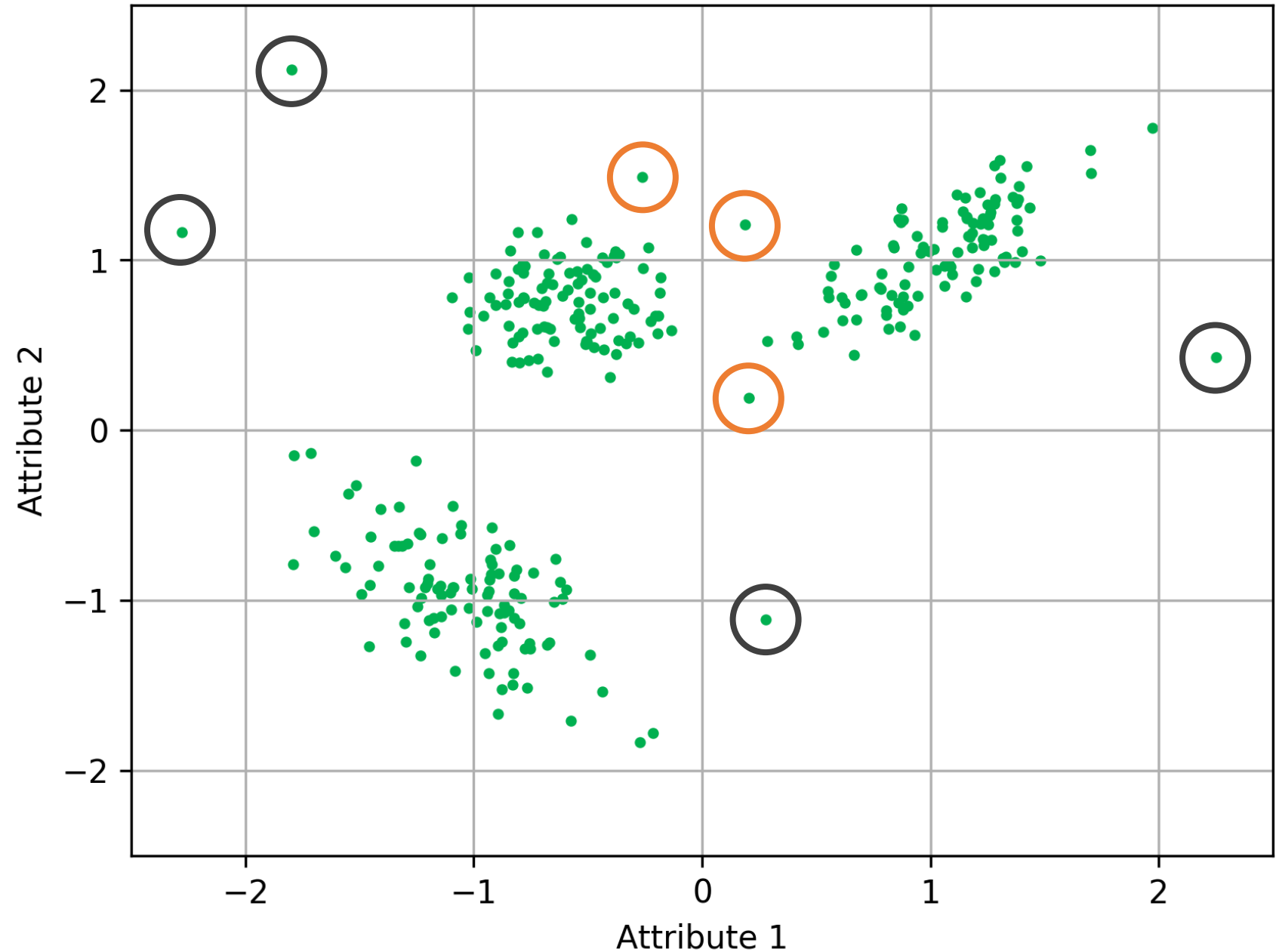
Account transactions, dates,
locations, demographic
information
(Numerical and categorical)

Target Data:

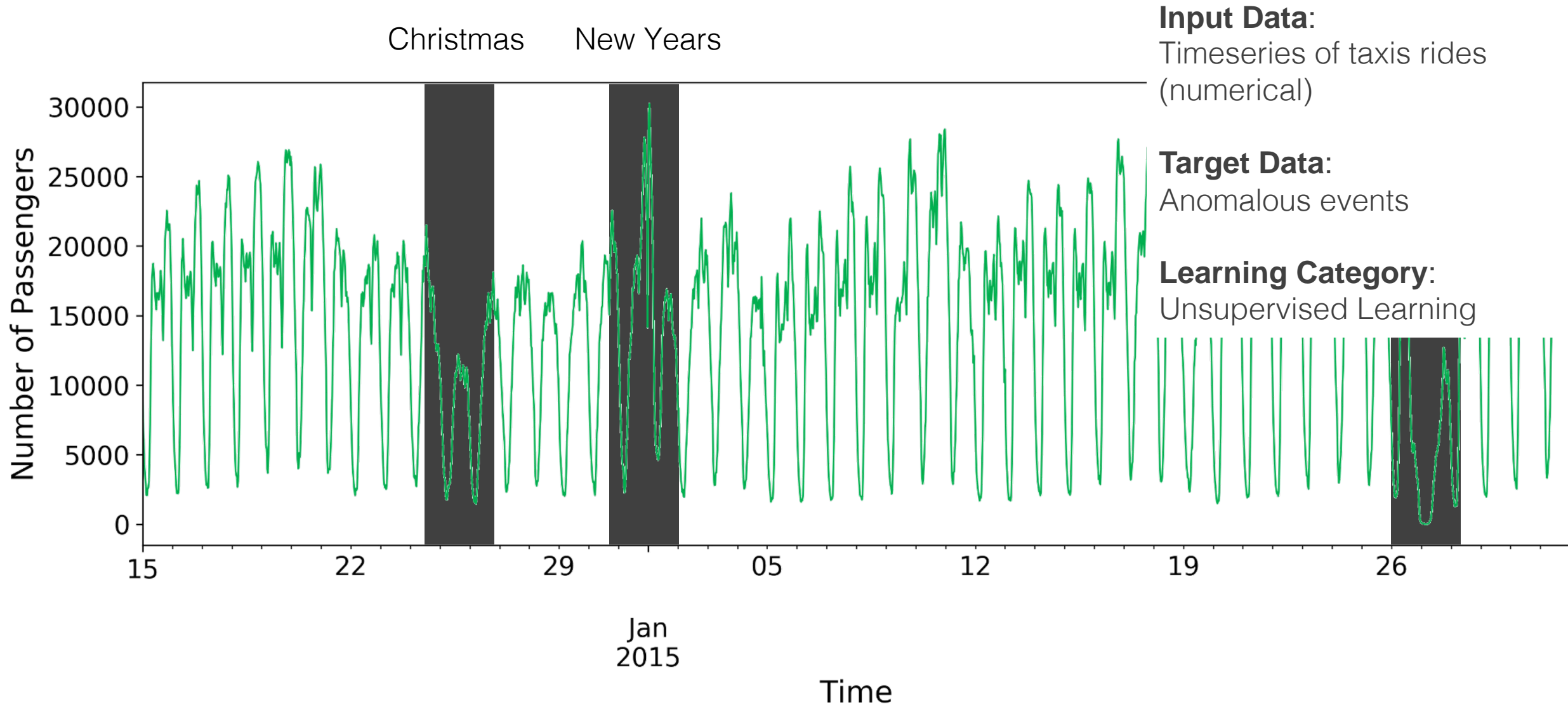
Anomalous transactions

Learning Category:

Unsupervised Learning
Clustering, Density
Estimation



Anomalous Event Detection: NYC Taxis



Data source: Numenta Anomaly Benchmark (NAB), from kaggle.com

Video Recommendations



Sherlock

97% Match 2017 TV-14 4 Series

97% Match



Season 3's episode "The Abominable Bride," which originally aired as a TV movie, won two Emmys.



MY LIST



Input Data:

User video ratings
(numerical and categorical)

Target Data:

User rating of video
(numerical)

Learning Category:

Recommender Systems
~Supervised & Unsupervised



NETFLIX

OVERVIEW

EPISODES

MORE LIKE THIS

DETAILS

Learning a strategy to master games

Input Data:

Moves taken and occasional feedback on win/loss
(Numerical and categorical)

Target Data:

Win/loss (Maximizing rewards)

Learning Category:

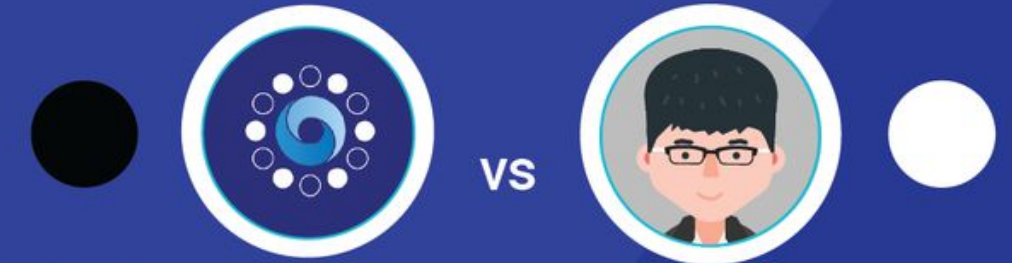
Reinforcement Learning



THE ULTIMATE GO CHALLENGE

GAME 3 OF 3

27 MAY 2017



AlphaGo

Winner of Match 3

Ke Jie

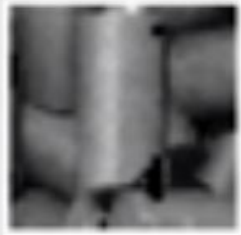
RESULT B + Res



Google DeepMind

Manufacturing – learn to pick up iron cylinders

Success Failure



Input Data:

Actions taken and occasional feedback on success/failure (Numerical and categorical)

Target Data:

Success/failure (Maximizing rewards)

Learning Category:

Reinforcement Learning



Source: MIT Technology Review; Company: **FANUC**

Types of machine learning

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Course logistics

Learning objectives

- Fundamentals of **machine learning**
- **Structure** a machine learning problem
- Automatically make **decisions** from data
- Understand the techniques/algorithms and when to use them
- Communicate and effectively **interpret** machine learning output
- Implement your own **end-to-end machine learning project**

Pedagogy

- **Good learning is active learning**
- **Desirable difficulty leads to meaningful learning**
 - Also helps create mental models for independent learning
 - Enhances abilities to interpret machine learning results
- **Reading, reflection, and recall is a pattern for effective learning**
 - You'll interact with each concept 4 times typically (lectures, readings, quizzes, and assignment)
 - Helps avoid illusion of knowledge

Course website

kylebradbury.github.io/ids705

Graded Components

Assignments	50% (6 assignments, 8.3% each)
Quizzes	20% (One before each class, <1% each)
Final project	30%

Action items

1. Complete the first set of readings
2. Take Lecture 2 Quiz on Sakai (available immediately after class)
3. Log into each of the course sites (Sakai, Piazza, Gradescope)
4. Begin working on Assignment #1
5. [optional] Coding session with TA's: Friday, Jan 22, 3:15-4:15pm