CS 60050 Machine Learning

Introduction

Some slides taken from course materials of Andrew Ng and P. Domingos

Course Logistics

- Instructor: Saptarshi Ghosh
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 - Office: CSE Department, room 207
- Class hours (attendance is compulsory)
 - Wednesday 11:00 11:55
 - Thursday 12:00 12:55
 - Friday 08:00 08:55
- Course Website:

http://cse.iitkgp.ac.in/~saptarshi/courses/ml2020s/

Evaluation

- Mid-semester exam (20%)
- End-semester exam (40%)
- Assignments (40%)
 - 4-5 in number
 - Programming in C / C++ / Java / Python

Text & Reference Materials

- R. Duda, et al., Pattern Classification, Wiley
- T. Mitchell, Machine Learning, McGraw-Hill
- C. Bishop, Pattern Recognition and Machine Learning, Springer
- Tan, Steinbach, Kumar, Introduction to Data Mining, Pearson
- Many resources available freely on Web

A Few Quotes

- "A breakthrough in machine learning would be worth ten Microsofts" (Bill Gates, Chairman, Microsoft)
- "Machine learning is the next Internet" (Tony Tether, Director, DARPA)
- Machine learning is the hot new thing" (John Hennessy, President, Stanford)
- "Web rankings today are mostly a matter of machine learning" (Prabhakar Raghavan, Dir. Research, Yahoo)

Motivation for ML

- It is hard to write programs for certain tasks
 - Human face or handwriting recognition
 - Playing complex games like chess
 - Recommending movies that a person will like
- Why?
 - We do not ourselves know how to solve
 - Algorithm will be too complicated
 - Too many instances of the program needed (e.g., one for every user)

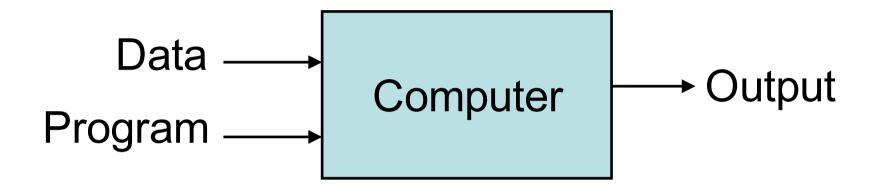
A classic example of a task that requires machine learning: It is very hard to say what makes a 2

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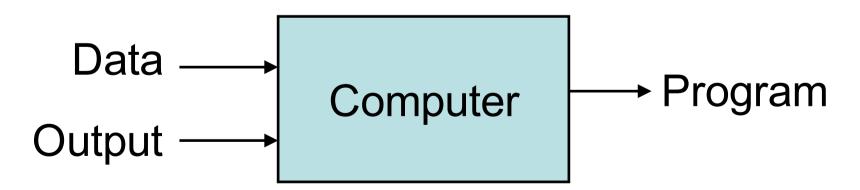
Motivation for ML

- Instead of writing a program by hand, collect lots of examples that specify the correct output for a given input.
- A machine learning algorithm takes these examples and produces a program that does the job.
- If done right, the program works for new cases as well as the ones we trained it on.

Traditional Programming



Machine Learning



Human learning vs. ML

Humans prefer to learn how to design the algorithm / program

• But for some very complex tasks, humans also learn from example inputs and outputs

Human learning vs. ML

- Some aspects common to both
- Desired: Generalization to new or unknown examples
 - E.g., teach a child to look out for vehicles while crossing a road
- Challenge: Learning a wrong model due to specific training examples used
 - E.g., teach a child to distinguish between "safe" (cow, cat) and "dangerous" animals (lion, tiger)

Machine Learning definition

• Arthur Samuel (1959):

Field of study that gives computers the ability to learn without being explicitly programmed.

• Tom Mitchell (1998):

Well-posed Learning Problem: A computer program is said to *learn* from experience E with respect to some task T and some performance measure P, if its performance on T, as measured by P, improves with experience E. Some more examples of tasks that are best solved by using a learning algorithm

- Recognizing patterns
 - Facial identities or facial expressions
 - Handwritten or spoken words
 - Medical images
- Generating patterns
 - Generating images or motion sequences
- Recognizing anomalies
 - Unusual sequences of credit card transactions
 - Unusual patterns of sensor readings in a nuclear power plant or unusual sound in your car engine.
- Prediction
 - Future stock prices or currency exchange rates

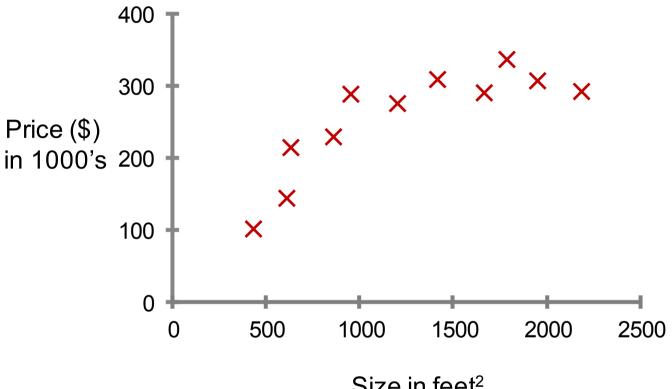
Some web-based examples of machine learning

- The web contains a lot of data. Tasks with very big datasets often use machine learning
- Spam filtering, fraud detection
 - The enemy adapts so we must adapt too.
- Recommendation systems
 - Need to adapt to millions of individuals. E.g., Youtube video, Netflix movies recommendation
- Information retrieval
 - Find documents or images with similar content.
- Data Visualization
 - Display a huge database in a revealing way

Types of Learning

- Supervised (inductive) learning
 - Training data includes desired / correct outputs or labels
- Unsupervised learning
 - Training data does not include desired outputs
- Semi-supervised learning (various forms)
 - Training data includes a few desired outputs
 - Training data has desired outputs, but for a different (related) task
- Reinforcement learning
 - Rewards from sequence of actions

Housing price prediction.



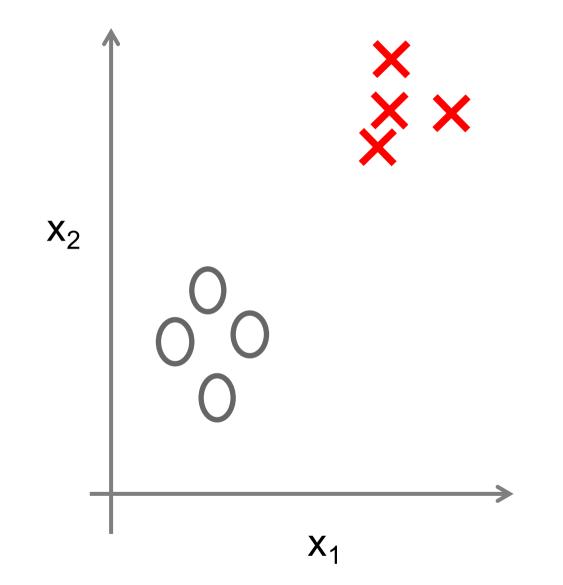
Size in feet²

Supervised Learning

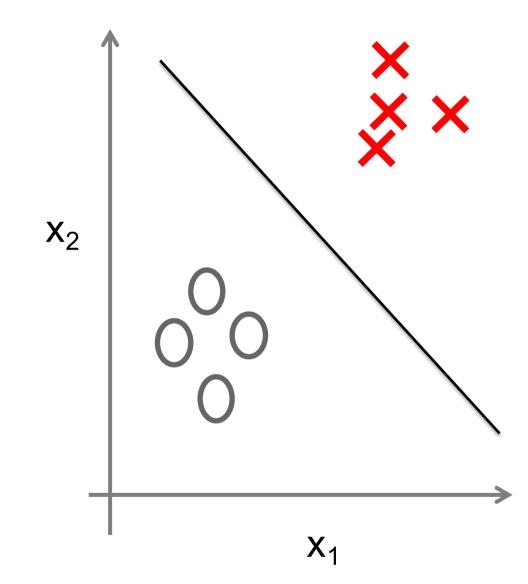
"right answers" given

Regression: Predict continuous valued output (price)

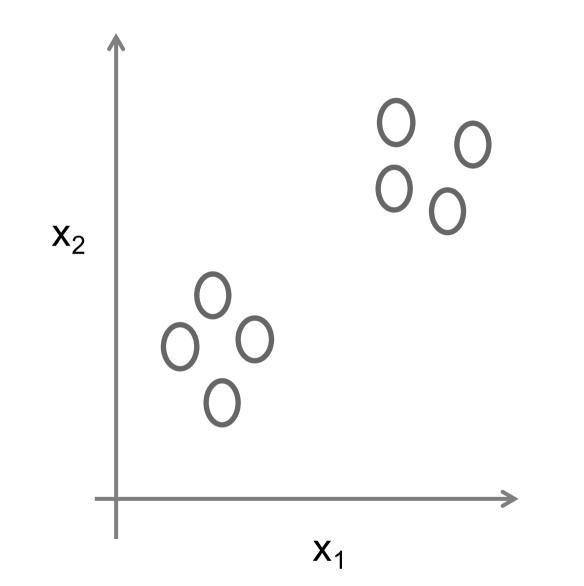
Supervised Learning: Classification



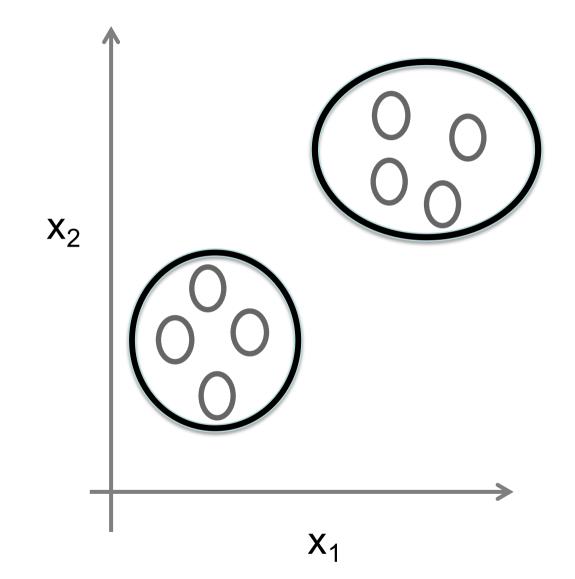
Supervised Learning: Classification



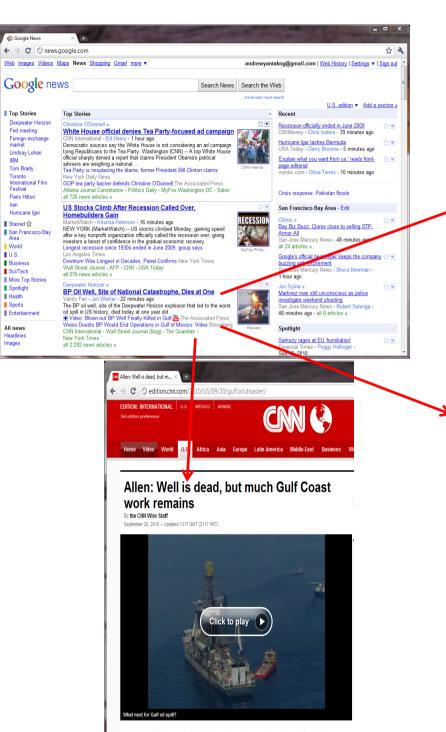
Unsupervised Learning: Clustering

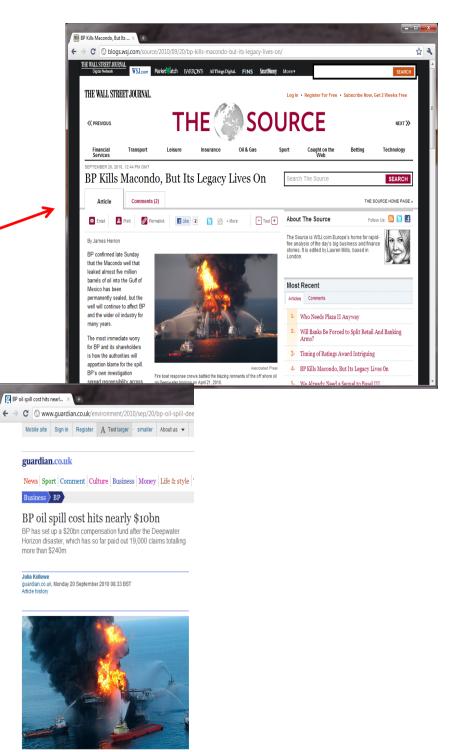


Unsupervised Learning: Clustering



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market Lindsay Lohan	Democratic sources say the White House is not considering an ad campaigr tying Republicans to the Tea Party. Washington (CNN) A top White House	e USA Today - Gerry Broome - 5 minutes ago	
IBM Tom Brady Toronto	official sharply denied a report that claims President Obama's political advisers are weighing a national Tea Party is misplacing the blame, former President Bill Clinton claims New York Daily News	CNN Interna 'Explain what you want from us,' reads front-page editorial msnbc.com - Olivia Torres - 10 minutes ago	
International Film Festival Paris Hilton	GOP tea party backer defends Christine O'Donnell The Associated Press Atlanta Journal Constitution - Politics Daily - MyFox Washington DC - Salon all 726 news articles »	n Crisis response: Pakistan floods	
Iran Hurrisons Isor	US Stocks Climb After Recession Called Over,	San Francisco Bay Area - Edit	
Hurricane Igor Starred 😭 San Francisco Bay Area World	Homebuilders Gain MarketWatch - Kristina Peterson - 16 minutes ago NEW YORK (MarketWatch) US stocks climbed Monday, gaining speed after a key nonprofit organization officially called the recession over, giving investors a boost of confidence in the gradual economic recovery. Longest recession since 1930s ended in June 2009, group says	RECESSION Clorox » Bay Biz Buzz: Clorox close to selling STP. Armor All San Jose Mercury News - 48 minutes ago - all 24 articles »	
U.S. Business Sci/Tech More Top Stories	Los Angeles Times Downturn Was Longest in Decades, Panel Confirms New York Times Wall Street Journal - AFP - CNN - USA Today at 276 news articles »	Google's official beekeeper keeps the company 🏠 💌 buzzing with excitement San Jose Mercury News - Bruce Newman - 1 hour ago	
Spotlight Health Sports Entertainment	Deepwater Horizon » BP Oil Well, Site of National Catastrophe, Dies at One Vanity Fair - Juli Weiner - 22 minutes ago The BP oil well, site of the Deepwater Horizon explosion that led to the worst oil spill in US history, died today at one year old.	48 minutes and - all 6 articles »	
All news	Weiss Doubts BP Would End Operations in Gulf of Mexico: Video Bloomber CNN International - Wall Street Journal (blog) - The Guardian -		
Headlines Images	New York Times all 2,292 news articles »	Sarkozy rages at EU 'humiliation' Financial Times - Peggy Hollinger - Sep 16, 2010	





BP's costs for the Deepwater Horizon disaster have hit \$10bn. Photograph Ho/Reuters

STORY HIGHLIGHTS (CNN

(CNN) -- The ruptured Macondo well, a mile under the Gulf of Mavice off the Louisiana coast, has been propounced dead

What we will cover

- Supervised learning
 - Linear Regression (single and multiple variable)
 - Classification (logistic regression, decision tree, support vector machines, etc.)
 - Neural networks
- Unsupervised learning
 - Clustering
 - Dimensionality reduction
- Ensemble Learning: Bagging, Boosting
- Recent advanced issues fairness, interpretability in ML
- Theory of Generalization the very basics

ML in Practice

- Understanding domain, prior knowledge, and goals
- Data integration, cleaning, pre-processing, selection, etc.
- Feature selection some features may be sensitive, protected, etc.
- Learning models
- Evaluating and interpreting results
- Consolidating and deploying discovered knowledge
- Loop