CHAPTER 1:

Introduction

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Data Mining

Applications:

- Retail: Market basket analysis, Customer relationship management (CRM)
- Finance: Credit scoring, fraud detection
- Manufacturing: Optimization, troubleshooting
- Medicine: Medical diagnosis
- Telecommunications: Quality of service optimization
- Bioinformatics: Motifs, alignment
- Web mining: Search engines



What We Talk About When We Talk About "Learning"

- Learning general models from a data of particular examples
- Data is cheap and abundant (data warehouses, data marts); knowledge is expensive and scarce.
- Example in retail: Customer transactions to consumer behavior:

People who bought "Da Vinci Code" also bought "The Five People You Meet in Heaven"

Build a model that is a good and useful approximation to the data.

Why "Learn"?

- Machine learning is programming computers to optimize a performance criterion using example data or past experience.
- There is no need to "learn" to calculate payroll
- Learning is used when:
 - Human expertise does not exist (navigating on Mars),
 - Humans are unable to explain their expertise (speech recognition)
 - Solution changes in time (routing on a computer network)
 - Solution needs to be adapted to particular cases (user biometrics,loan approvals,parametric classification)



What is Machine Learning?

- Machine Learning
 - Study of algorithms that improve their performance at some task with experience
- Optimize a performance criterion using example data or past experience.
- Role of Statistics: Inference from a sample
- Role of Computer science: Efficient algorithms to
 - Solve the optimization problem
 - Representing and evaluating the model for inference



Growth of Machine Learning

- Machine learning is preferred approach to
 - Speech recognition, Natural language processing
 - Computer vision
 - Medical outcomes analysis
 - Robot control
 - Computational biology
- This trend is accelerating
 - Improved machine learning algorithms
 - Improved data capture, networking, faster computers
 - Software too complex to write by hand
 - New sensors / IO devices
 - Demand for self-customization to user, environment
 - It turns out to be difficult to extract knowledge from human experts | failure of expert systems in the 1980's.

Applications

- Association Analysis
- Supervised Learning
 - Classification
 - Regression/Prediction
- Unsupervised Learning
- Reinforcement Learning

Learning Associations

Basket analysis:

P(Y|X) probability that somebody who buys X also buys Y where X and Y are products/services.

Example: P (chips | beer) = 0.7

Market-Basket transactions

TID	Items
1	Bread, Milk
2	Bread, Diaper, Beer, Eggs
3	Milk, Diaper, Beer, Coke
4	Bread, Milk, Diaper, Beer
5	Bread, Milk, Diaper, Coke



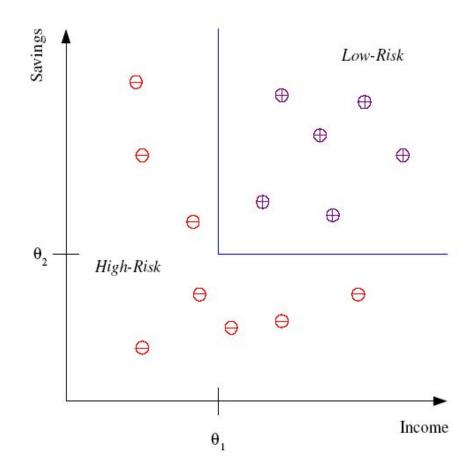
Supervised Learning: Use Cases

Example: decision trees tools that create rules

- Prediction of future cases: Use the rule to predict the output for future inputs
- Knowledge extraction: The rule is easy to understand
- Compression: The rule is simpler than the data it explains
- Outlier detection: Exceptions that are not covered by the rule, e.g., fraud

Classification

- Example: Credit scoring
- Differentiating between low-risk and high-risk customers from their income and savings



Discriminant: IF $income > \theta_1$ AND $savings > \theta_2$ THEN low-risk ELSE high-risk

Classification: Applications

- Aka Pattern recognition
- Face recognition: Pose, lighting, occlusion (glasses, beard), make-up, hair style
- Character recognition: Different handwriting styles.
- Speech recognition: Temporal dependency.
 - Use of a dictionary or the syntax of the language.
 - Sensor fusion: Combine multiple modalities; eg, visual (lip image) and acoustic for speech
- Medical diagnosis: From symptoms to illnesses
- Web Advertizing: Predict if a user clicks on an ad on the Internet.

Face Recognition

Training examples of a person









Test images









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Prediction: Regression

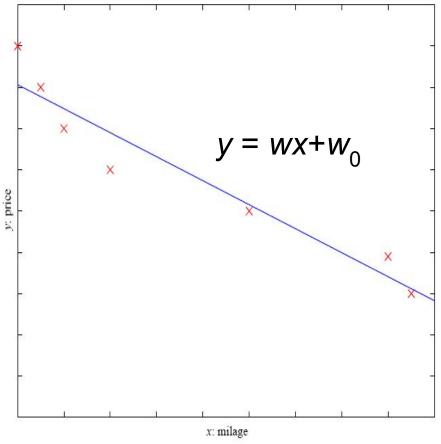
- Example: Price of a used car
- x : car attributes

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y: price
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$$y = g(x \mid \theta)$$

g() model,

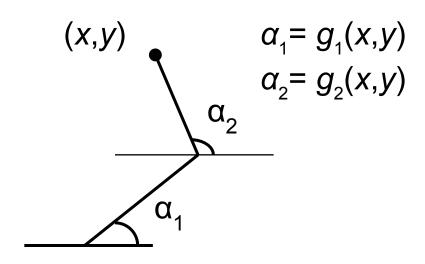
 θ parameters



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Regression Applications

- Navigating a car: Angle of the steering wheel (CMU NavLab)
- Kinematics of a robot arm



Unsupervised Learning

- Learning "what normally happens"
- No output
- Clustering: Grouping similar instances
- Other applications: Summarization, Association Analysis
- Example applications
 - Customer segmentation in CRM
 - Image compression: Color quantization
 - Bioinformatics: Learning motifs

Reinforcement Learning

Topics:

- Policies: what actions should an agent take in a particular situation
- Utility estimation: how good is a state (□used by policy)
- No supervised output but delayed reward
- Credit assignment problem (what was responsible for the outcome)
- Applications:
 - Game playing
 - Robot in a maze
 - Multiple agents, partial observability, ...