

Lecture Slides for  
**INTRODUCTION  
TO  
MACHINE  
LEARNING**  
3RD EDITION

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CHAPTER 1:

# INTRODUCTION

# Big Data

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- Widespread use of personal computers and wireless communication leads to “big data”
- We are both producers and consumers of data
- Data is not random, it has structure, e.g., customer behavior
- We need “big theory” to extract that structure from data for
  - (a) Understanding the process
  - (b) Making predictions for the future

# Why “Learn” ?

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- 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)

# What We Talk About When We Talk About “Learning”

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- 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 “Blink” also bought “Outliers”  
([www.amazon.com](http://www.amazon.com))*
- Build a model that is *a good and useful approximation* to the data.

# Data Mining

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- Retail: Market basket analysis, Customer relationship management (CRM)
- Finance: Credit scoring, fraud detection
- Manufacturing: Control, robotics, troubleshooting
- Medicine: Medical diagnosis
- Telecommunications: Spam filters, intrusion detection
- Bioinformatics: Motifs, alignment
- Web mining: Search engines
- ...

# What is Machine Learning?

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- 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

# Applications

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- Association
- Supervised Learning
  - ▣ Classification
  - ▣ Regression
- Unsupervised Learning
- Reinforcement Learning



# Learning Associations

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- Basket analysis:

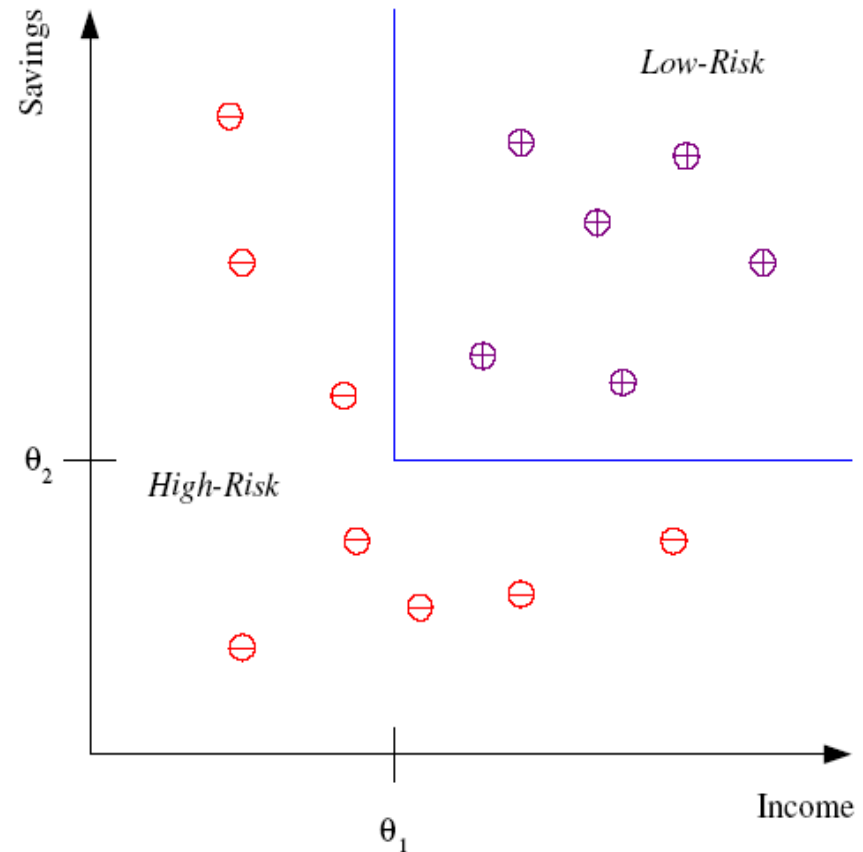
$P ( Y | X )$  probability that somebody who buys  $X$  also buys  $Y$  where  $X$  and  $Y$  are products/services.

Example:  $P ( \text{chips} | \text{beer} ) = 0.7$

# Classification

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- 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

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- Aka Pattern recognition
- Face recognition: Pose, lighting, occlusion (glasses, beard), make-up, hair style
- Character recognition: Different handwriting styles.
- Speech recognition: Temporal dependency.
- Medical diagnosis: From symptoms to illnesses
- Biometrics: Recognition/authentication using physical and/or behavioral characteristics: Face, iris, signature, etc
- Outlier/novelty detection:

# Face Recognition

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Training examples of a person



Test images



ORL dataset,  
AT&T Laboratories, Cambridge UK

# Regression

□ Example: Price of a used car

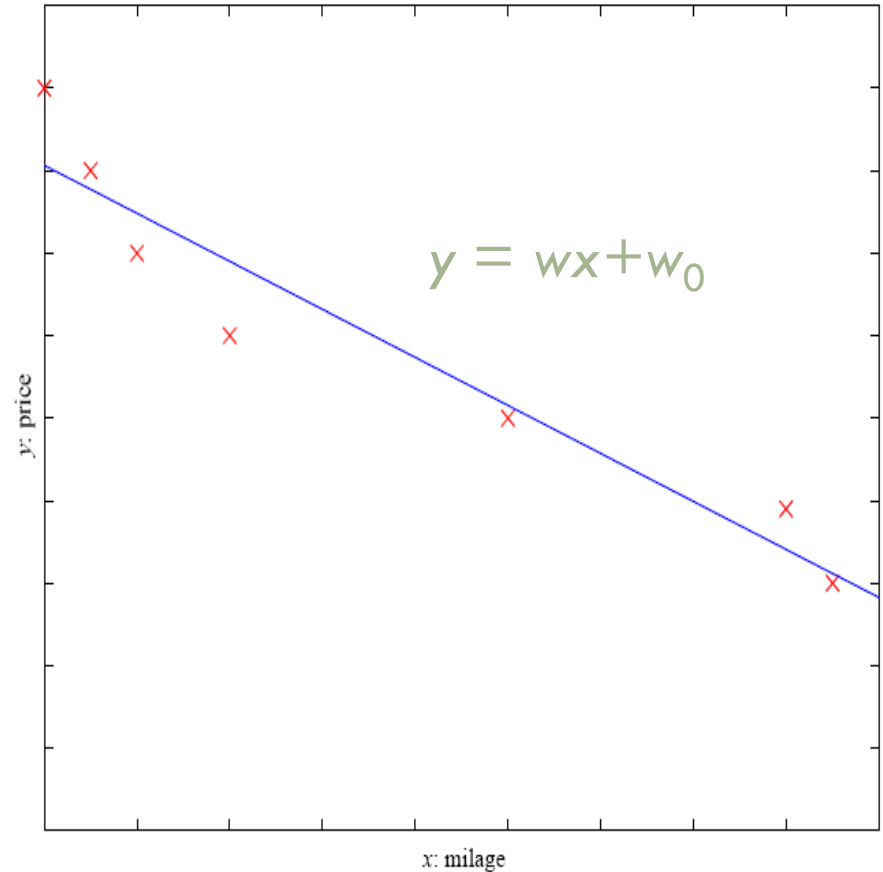
□  $x$  : car attributes

$y$  : price

$$y = g(x | \theta)$$

$g(\cdot)$  model,

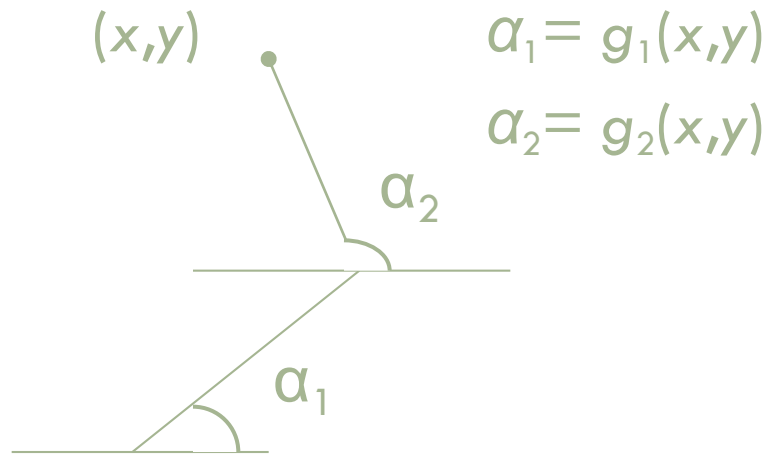
$\theta$  parameters



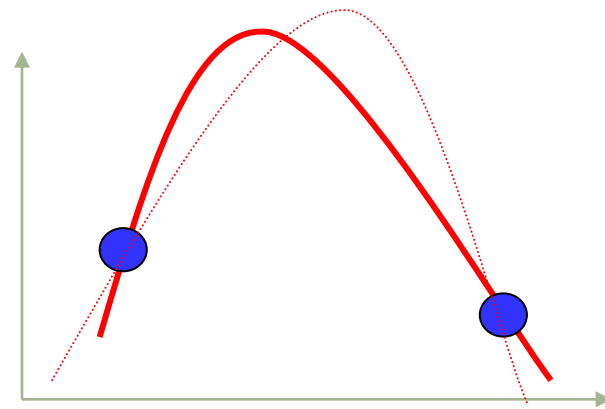
# Regression Applications

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- Navigating a car: Angle of the steering
- Kinematics of a robot arm



- Response surface design



# Supervised Learning: Uses

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- **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

# Unsupervised Learning

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- Learning “what normally happens”
- No output
- Clustering: Grouping similar instances
- Example applications
  - Customer segmentation in CRM
  - Image compression: Color quantization
  - Bioinformatics: Learning motifs



# Reinforcement Learning

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- Learning a policy: A **sequence** of outputs
- No supervised output but delayed reward
- Credit assignment problem
- Game playing
- Robot in a maze
- Multiple agents, partial observability, ...

# Resources: Datasets

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- UCI Repository: <http://www.ics.uci.edu/~mlearn/MLRepository.html>
- Statlib: <http://lib.stat.cmu.edu/>

# Resources: Journals

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- Journal of Machine Learning Research [www.jmlr.org](http://www.jmlr.org)
- Machine Learning
- Neural Computation
- Neural Networks
- IEEE Trans on Neural Networks and Learning Systems
- IEEE Trans on Pattern Analysis and Machine Intelligence
- Journals on Statistics/Data Mining/Signal Processing/Natural Language Processing/Bioinformatics/...

# Resources: Conferences

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- International Conference on Machine Learning (ICML)
- European Conference on Machine Learning (ECML)
- Neural Information Processing Systems (NIPS)
- Uncertainty in Artificial Intelligence (UAI)
- Computational Learning Theory (COLT)
- International Conference on Artificial Neural Networks (ICANN)
- International Conference on AI & Statistics (AISTATS)
- International Conference on Pattern Recognition (ICPR)
- ...