



Machine Learning

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Textbooks

3

- **Introduction to Machine Learning - Ethem Alpaydin**
- **Pattern Recognition and Machine Learning, Bishop.**
- **Machine Learning, Mitchell, Tom.**
- The Elements of Statistical Learning, Hastie, T., R. Tibshirani, and J. H. Friedman.
- *Foundations of Machine Learning* by Mehryar Mohri, Afshin Rostamizadeh and Ameet Talwalkar
- Machine Learning: A Probabilistic Perspective, by Kevin P. Murphy.
- Introduction to Data Mining by Tan, Steinbach and Kumar
- Pattern Classification (2nd ed.) by Richard O. Duda, Peter E. Hart and David G. Stork
- Pattern Recognition, 4th Ed., Theodoridis and Koutroumbas

Grading Criteria

4

- Midterm Exam \approx 25%
- HW, Comp. Assignments and projects: \approx 30%
- Final exam \approx 45%

- **Course Website:**
- <http://ivut.iut.ac.ir> or <http://elearning.iut.ac.ir/>
- **Email: Ahmadzadeh@cc.iut.ac.ir**

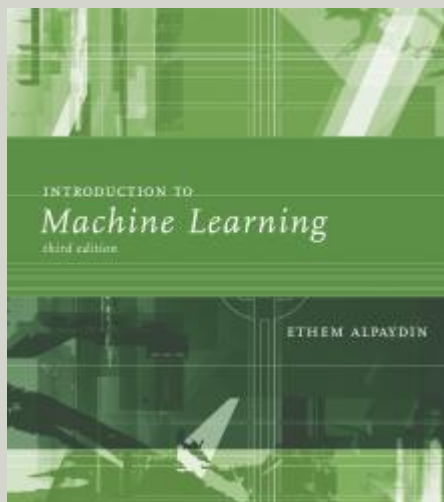
- EBooks ...

Contents

- *1 Introduction 1*
- *2 Supervised Learning 21*
- *3 Bayesian Decision Theory 49*
- *4 Parametric Methods 65*
- *5 Multivariate Methods 93*
- *6 Dimensionality Reduction 115*
- *7 Clustering 161*
- *8 Nonparametric Methods 185*
- *9 Decision Trees 213*

- *10 Linear Discrimination 239*
- *11 Multilayer Perceptrons 267*

- *12 Local Models 317*
- *13 Kernel Machines 349*
- *14 Graphical Models 387*
- *15 Hidden Markov Models 417*
- *16 Bayesian Estimation 445*
- *17 Combining Multiple Learners 487*
- *18 Reinforcement Learning 517*
- *19 Design and Analysis of ML Experiments 547*
- *A Probability 593*



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

ETHEM ALPAYDIN

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<http://www.cmpe.boun.edu.tr/~ethem/i2ml3e>

CHAPTER 1:

INTRODUCTION

Big Data

8

- 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”?

9

- 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”

10

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

Data Mining

11

- **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?

12

- 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

Machine Learning vs Pattern Recognition

13

- **Pattern Recognition:** automatic discovery of regularities in data and the use of these regularities to take actions – classifying the data into different categories. Example: handwritten recognition. Input: a vector \mathbf{x} of pixel values. Output: A digit from 0 to 9.
- **Machine Learning:** a large set of input vectors $\mathbf{x}_1, \dots, \mathbf{x}_N$, or a training set is used to tune the parameters of an adaptive model. The category of an input vector is expressed using a target vector \mathbf{t} . The result of a machine learning algorithm: $y(\mathbf{x})$ where the output y is encoded as the target vectors.

Applications

14

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

Learning Associations

15

- Basket analysis:

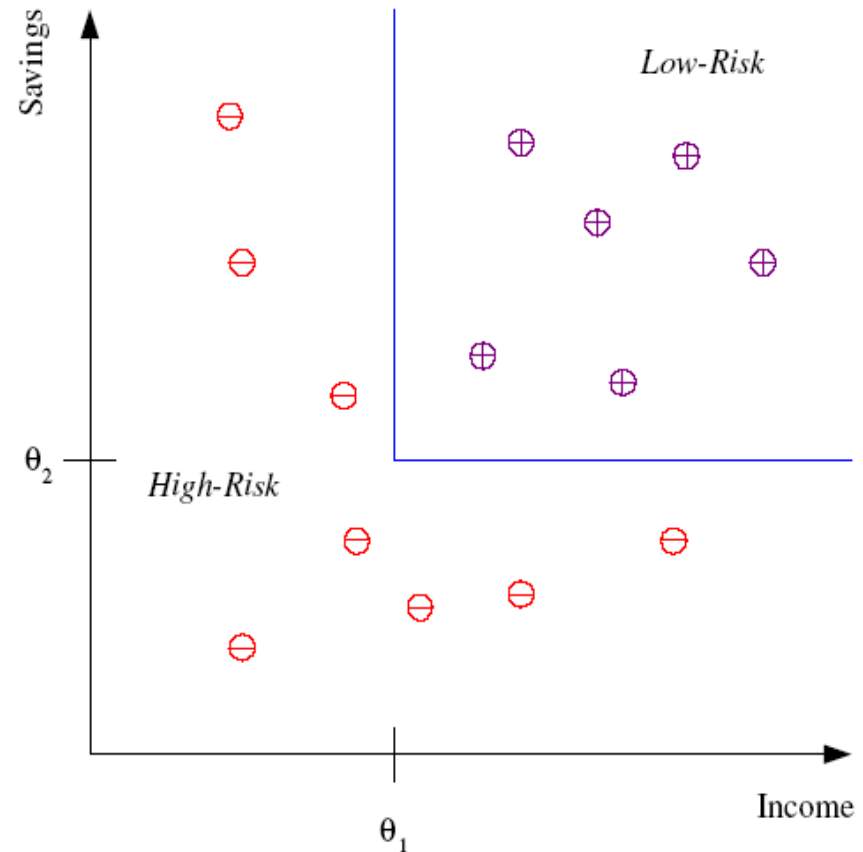
$P (Y | X)$ probability that somebody who buys X also buys Y where X and Y are products/services.

Example: $P (\text{Chips} | \text{Yogurt}) = 0.7$

Classification

16

- 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

17

- **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

18

Training examples of a person



Test images

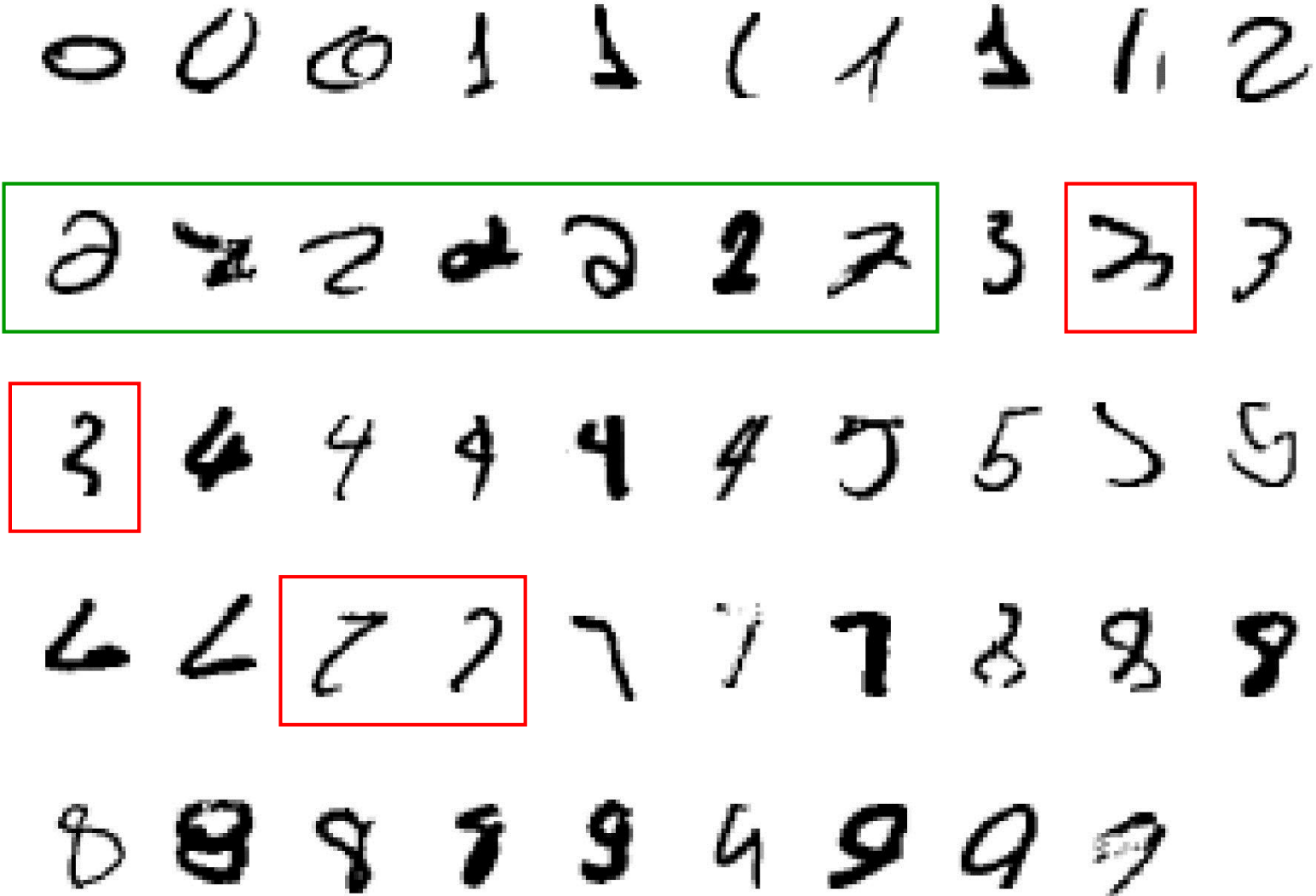


ORL dataset,
AT&T Laboratories, Cambridge UK

A classic example of a task that requires machine learning:

It is very hard to say what makes a 2

19



Regression

□ Example: Price of a used car

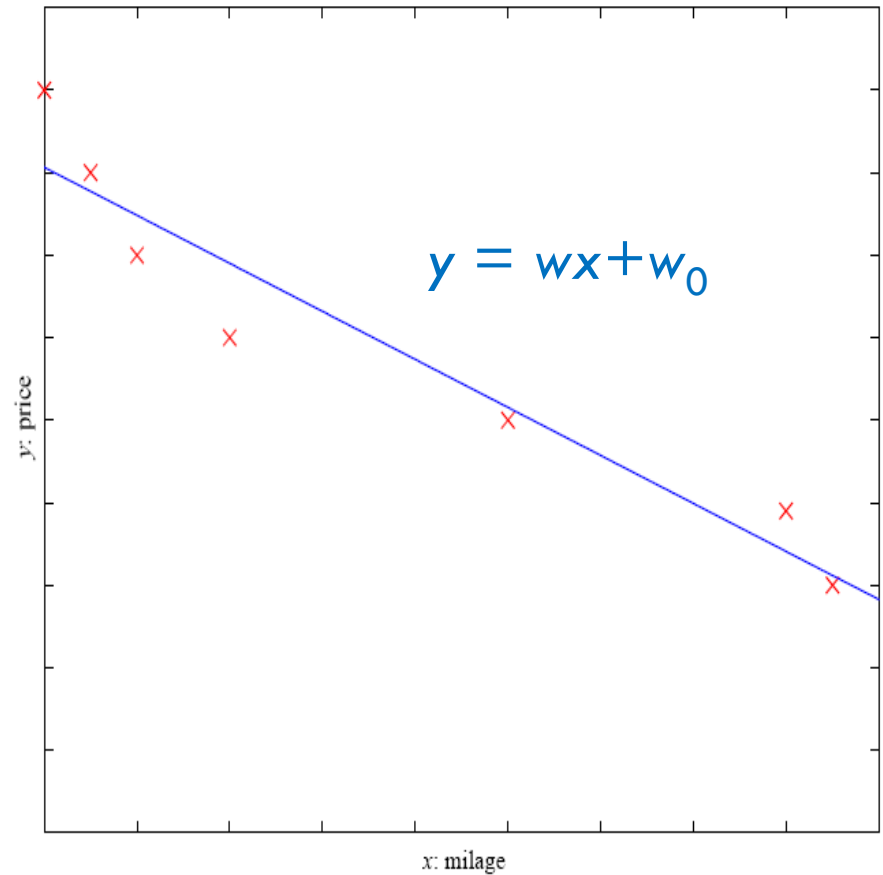
□ x : car attributes

y : price

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

$g(\cdot)$ model,

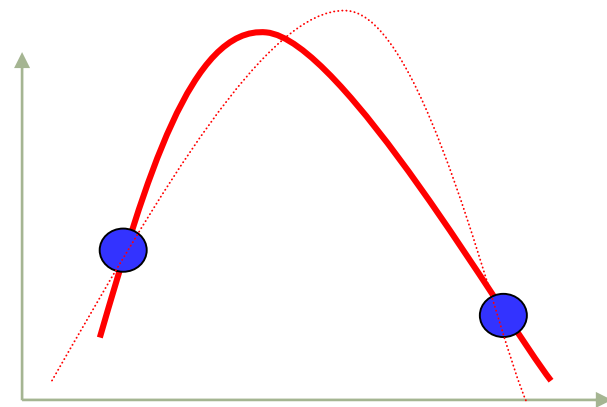
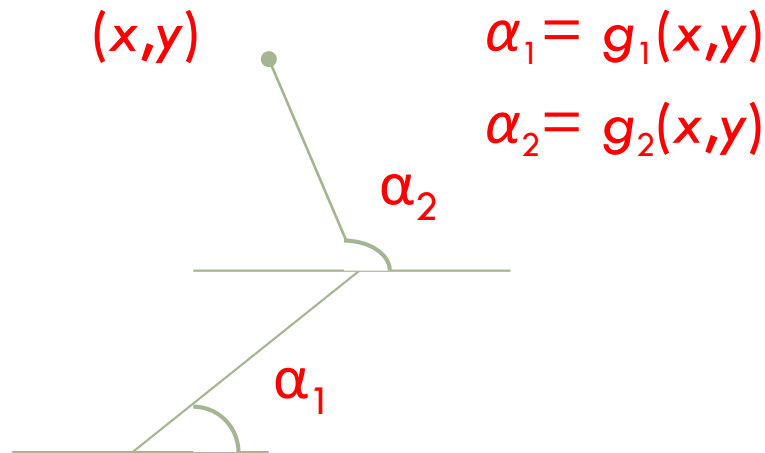
θ parameters



Regression Applications

21

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



- Response surface design

Supervised Learning: Uses

22

- **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

23

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

Reinforcement Learning

24

- 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 - Journals

25

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

- ❑ Journal of Machine Learning Research 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

26

- 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)
- ...