Advanced Statistical Inference Introduction

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Why do we call it Bayesian?



- ► Reverend Thomas Bayes (London 1701 Kent 1761)
- ► Logic and theology degree from University of Edinburgh in 1722

Why do we call it Bayesian?

- Published works
 - "Divine Benevolence, or an Attempt to Prove That the Principal End of the Divine Providence and Government is the Happiness of His Creature" in 1731
 - "An Introduction to the Doctrine of Fluxions, and a Defence of the Mathematicians Against the Objections of the Author of The Analyst" in 1736

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 - "An Introduction to the Doctrine of Fluxions, and a Defence of the Mathematicians Against the Objections of the Author of The Analyst" in 1736
- ► Thanks to Richard Price: "An Essay towards solving a Problem in the Doctrine of Chances" read to the Royal Society in 1763

Probabilities before Bayes

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- ► "Liber de ludo aleae" (1564 published in 1663) by Gerolamo Cardano (1501–1576)

Some Historical Context - Renaissance

- ▶ Leonardo da Vinci (1452 1519)
- Nicolaus Copernicus (1473 1543)
- Niccolò Fontana Tartaglia (1499 1557)
- ► Galileo Galilei (1564 1642)
- ▶ Johannes Kepler (1571 1630)
- ► Blaise Pascal (1623 1662)
- ► Isaac Newton (1642 1726)
- ► Gottfried Wilhelm von Leibniz (1646 1716)

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- Age of Enlightenment brought impact on sciences and societal changes
 - ► Birth of Economics and Chemistry
 - Steam engine by James Watt (Commercialized in 1776)
 - Industrial Revolution
 - French and American Revolutions
 - U.S. Constitution (1789) influenced by James Maddison, Benjamin Franklin, and Thomas Jefferson. George Washington as President.

- ► A.o.E. in parallel with Baroque and Neoclassicism
- Arts
 - ► "The History of Art in Antiquity" (1764) by Johann Joachim Winckelmann (1717 1768)
 - "Oath of the Horatii" (1784) Jacques-Louis David (1748 1825)
 - "Cupid's Kiss" (1787) by Antonio Canova (1757 1822)





- Age of Enlightenment in parallel with Baroque and Neoclassicism
- Music
 - ▶ Johann Sebastian Bach (1685 1750)
 - ▶ Well Tempered Clavier (Book 1 & 2 in 1722 & 1742)
 - ► Wolfgang Amadeus Mozart (1756 1791)
 - ► Ludwig van Beethoven (1770 1827)
 - Fryderyk Chopin (1810 1849) romantic period

Some Historical Context

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 - ► First professor of Statistics Karl Pearson (1857 1936) PCA
 - ▶ Following the death of Francis Galton (1822 1911) Eugenics

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- ▶ World War I (1914 1918)

Some Historical Context

- Other key statisticians
 - ► Charles Spearman (1863 1945) Rank test
 - Ronald Fisher (1890 1962) Fisher Information, F and Von Mises distributions, LDA
 - ▶ Bruno de Finetti (1906 1985) Philosophy of probabilities, exchangeability
 - John Tukey (1915 2000) FFT
 - Calyampudi Radhakrishna Rao (1920) Cramér-Rao bound
 - ▶ David Cox (1924) Cox processes, Box-Cox transform

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 - First abstraction of a machine that can do any computations (1936)
 - ► The Turing test (1950)
- ► World War II (1939 1945)

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- Computers become a reality
 - ▶ John von Neumann (1903 1957) inspired the design of modern computers

The First Neural Networks

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 - Frank Rosenblatt (1958)
 - ► Minsky & Papert (1969)

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- High expectations that this would develop into models of an actual brain

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 - Risk minimization
 - Regularization
 - VC dimension

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- "Support Vector Machines" Cristianini & Shawe-Taylor, Schölkopf & Smola

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- "Gaussian processes" O'Hagan 1978, Neal 1996, Williams & Rasmussen 1996, Williams & Barber 1998

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 - Statistical Learning Theory
- ► Function estimation

Bayesian Machine Learning

What will you learn in this course?

- Function estimation using the philosophy of Bayes
- Conditioning on data and modeling assumption
- Offers quantification of uncertainty (due to the lack of data and imprecise knowledge of the environment)

Companies with lots of data for which traditional models don't exist:

Google, Microsoft, Amazon, etc

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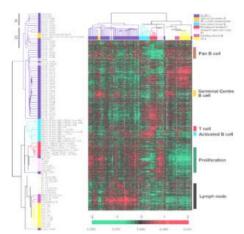
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- ▶ e.g. Recommendations
- Can't write down an equation that describes what I like
- But we can look for patterns in what I buy....
- ...and in what others buy.



Biotech companies who want to diagnose patients and discover biomarkers.



Some examples within EURECOM

Life and Environmental Sciences

- ▶ Diagnosis and progression of neurological disorders
- ► Expensive simulators (climate, tsunami)
- Medical imaging

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Industrial applications

- Fraud detection
- Finance
- Automotive

Course overview

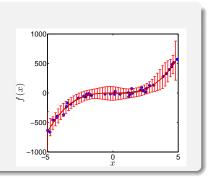
Supervised Learning

Unsupervised Learning

Supervised Learning

Regression

Learning a continuous function from a set of examples.



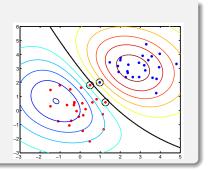
Example

Predicting stock prices (x might be time or some other variable of interest).

Supervised Learning

Classification

Learning a rule that can separate objects of different types from one another.



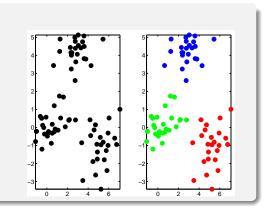
Examples

Disease diagnosis, spam email detection.

Unsupervised Learning



Finding groups of similar objects.



Examples

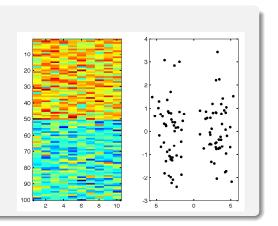
People with similar 'taste', genes with similar function.

Unsupervised Learning

Unsupervised Learning

Projection

Reducing the number of variables – e.g. from 10 to 2.



Examples

Visualizing complex data.

Maths

- ▶ We represent objects as vectors/matrices (arrays of numbers), so we have to do maths.
- Being familiar with calculus (function analysis)
- Good understanding probabilities
- Good understanding of linear algebra

ASI schedule - Thursdays 9am-12pm

Lectures

× 8

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Lectures

 \times 8

Labs

× 5

Assessment

- ► Inverse Class Participation (30%)
- ► Exam (70%)

Contacts

- Moodle
- Virtually
 - Maurizio.Filippone@eurecom.fr
- ► In person
 - ► Office 419
 - https://mauriziofilippone.youcanbook.me

Aside note

▶ I do not write recommendation letters to ASI students

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- ... unless they work on projects under my supervision

Suggested readings

A First Course in Machine Learning

S. Rogers and M. Girolami

Pattern Recognition and Machine Learning

C. Bishop

Information Theory, Inference, and Learning Algorithms

D. MacKay

Machine Learning: A Probabilistic Perspective

K. P. Murphy

Suggested readings

Bayesian Data Analysis

Andrew Gelman

Bayesian Reasoning and Machine Learning

David Barber

Machine Learning

Peter Flach