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0. Introduction

Data Analysis for Networks - NDA (2020-2021) Anastasios Giovanidis

Sorbonne-LIP6







Course (main) Bibliography

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B.1 Gareth James, Daniela Witten, Trevor Hastie, Robert Tibshirani. "An introduction to statistical learning: with applications in R". Springer Texts in Statistics.

ISBN 978-1-4614-7137-0 (DOI 10.1007/978-1-4614-7138-7)

B.2 C. Bishop, "Pattern Recognition and Machine Learning", Springer 2006.

ISBN 978-0387-31073-2

B.3 H. Pishro-Nik, "Introduction to probability, statistics, and random processes", available at https://www.probabilitycourse.com, Kappa Research LLC, 2014.

Surveys - Overview

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S.1 Raouf Boutaba et al. - "A comprehensive survey on machine learning for networking: evolution, applications and research opportunities", Journal of Internet Services and Applications, Springer (2018) 9:16 DOI 10.1186/s13174-018-0087-2

Stats VS Machine Learning

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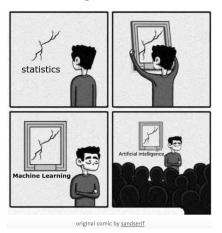


Figure: "When you're fundraising, it's Al. When you're hiring, it's ML. When you're implementing, it's logistic regression."

Intro

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Data Analysis and Machine Learning (ML) revolutionise our world!

- ► Computer Vision (CV) and Natural Language Processing (NLP): classifying images, facial recognition, automatic translation.
- Recommendation engines: Amazon, Netflix, or Youtube.

Been around since a very long time...

- Statistics is a branch of mathematics dealing with the collection, analysis, interpretation, and presentation of masses of numerical data.
- "Machine Learning, is the field of study that gives computers the ability to learn without being explicitly programmed" (Arthur Samuel, 1959)

Why now? Sufficient and cheap computational power & lots, lots of (labeled) data available e.g. Facebook and Google photos, WWW...

Highlights I

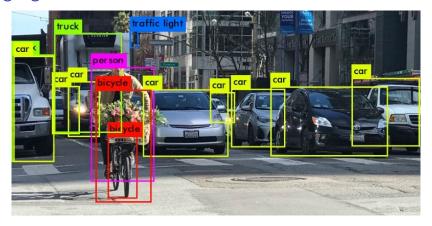


Figure: Object detection and recognition for driverless cars.

Highlights II

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Behind Hey Siri: How Apple's AI-Powered Personal Assistant Uses DNN

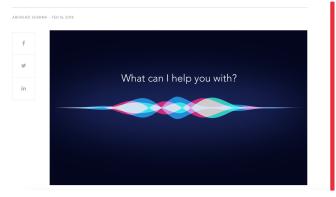


Figure: Speech recognition.

Highlights III



Figure: Useful recommendations.

Taxonomy of ML methods

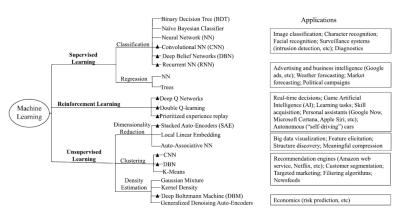


Figure: Taxonomy and applications (Fadlullah, et al (IEEE, 2017)).

Method differences

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All three methods require a common element to work:

DATA!!!

The difference is the type of data available or collected:

- Supervised: Labelled data, model learning.
- Unsupervised: Unlabelled data (majority of telecom data).
- Reinforcement: Exploration-exploitation. Data is the rewards collected by application of an action.

Labeling is a non-trivial process to establish the ground-truth. Often hand-made by experts.

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Make a distinction between static and dynamic environments: Data from the first are n-dimensional points, from the second **time-series**.

History of Data Analysis and ML methods A. Giovanidis 2020

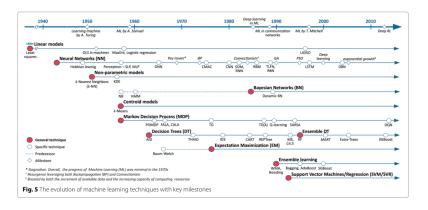


Figure: ML historical evolution (from [S.1]).

Main tasks to perform

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What can we do with all these methods?

- Estimation: quantify unknown parameters from observations.
- ▶ Inference: guess the unknown underlying statistics.
- Regression: guess an underlying model and predict possible outcomes of an experiment.
- Classification: decide on the class of an object.
- Dimensionality Reduction: compress the information contained in several features to easier describe an object.
- Clustering: group objects based on affinity.

Some Tasks

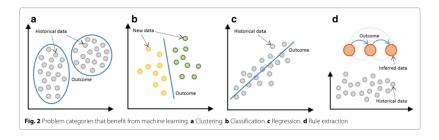
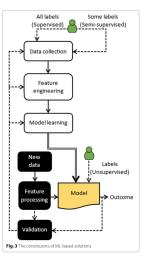


Figure: Task examples (from [S.1]).

General methodology



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Telecom Network science and Data

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Telecommunication networks offer the infrastructure for ML.

But! Their design and functionality can profit from data analysis and ML, through Telemetry: massive data availability about QoS, QoE, KPIs...

Main possibilities:

- 1. Traffic: prediction, classification, routing.
- 2. Performance: congestion control, resource management, fault management, QoS/QoE management.
- 3. Anomaly detection: hardware/software failure.
- 4. Security: Intrusion detection, DoS or DDoS Attacks.

Traffic IP

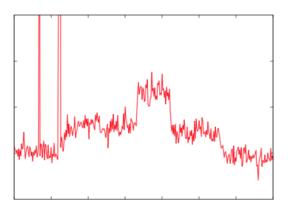


Figure: image from thesis Audrey Wilmet.

Traffic

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Prediction

Forecast future traffic from previously observed data: Time series forecasting through ARMA models (auto-regressive moving average)

Classification

Associate network traffic to pre-defined classes, e.g. HTTP, FTP, WWW, DNS, P2P

or applications, e.g. Skype, YouTube, Netflix...

Features: port number, packet payload, host behaviour, flow features, QoS requirements.

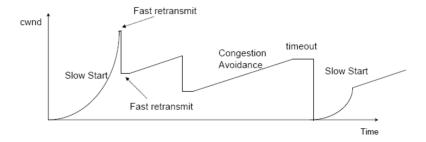
Traffic can be encrypted! Rely on stochastic characteristics.

Routing

Select a path for packet transmission with an objective: cost minimisation, link utilisation, QoS provisioning, etc.

Use of Reinforcement Learning techniques, to explore the environment without supervision (trial-and-error learning).

TCP



TCP congestion control

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TCP protocol limits the packet sending rate when congestion is detected.

But! TCP recognizes and handles all packet losses as network congestion (buffer overflow).

A packet loss can be due to other reasons:

- Packet reordering.
- Fading and shadowing in wireless.
- Wavelength contention in optical.

Solution: Classify the cause of packet loss and reduce TCP transmission rate only when congestion.

Features: inter-arrival time, round-trip time, one-way delay.

Also, learn the appropriate window reduction per congestion event!

Network security

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Protect the network against cyber-threats.

Attacks can compromise the network's availability and resources.

 \blacksquare Businesses are under security threats \to cost billions in damage and recovery, may have impact on their reputation.

Current Security measures include :

- Encryption of network traffic, Anti-viruses, Firewalls, etc.
- ⇒ Extra protection:
 - ► Intrusion Detection/Prevention: phishing, DoS, DDoS, ...

Monitor the network for malicious / anomalous activities, find patterns (=attack signatures) in big datasets that deviate from normal behaviour.

What is normal? Unsupervised learning, clustering methods.

Structure of the course I

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Methods from statistics, machine-learning and stochastic processes.

Each course on Wednesdays: 2 hours Theory + 2 hours Python Lab

Part I: Statistics

- C1. Intro to NDA / Probability basics (30 September 2020)
- C2. Frequentist Estimation (07 October 2020)
- C3. Hypothesis Tests (14 October 2020)
- C4. Bayes Rule (21 October 2020)

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Structure of the course II

Part II: Machine Learning

a. Supervised

- C5. Regression pt.1 (28 October 2020)
- C6. Regression pt.2 (04 November 2020)
- C7. Cross-Validation (18 November 2020)
- C8. Classification (25 November 2020)
- C9. Trees-Forests (09 December 2020)
- ► C10. Regularisation or SVM (13 January 2021)

b. Unsupervised

- C11. Clustering (16 December 2020)
- ► C12. PCA and Anomaly Detection(06 January 2021)

Structure of the course III

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Part III: Time-series

- ► C13. Time-Series pt.1 (20 January 2021)
- ► C14. Time-Series pt.2 (27 January 2021)

■ Beginning February 2021 final exam.

Final Note:

40% Python code from all TPs

60% Final exam.

Teaching material

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Course Material (slides):

https://github.com/yokaiAG/DataNets-Course

People:

- Anastasios Giovanidis (responsible)
- Contributors: Maximilien Danisch (clustering), Lionel Tabourier (time-series).

Contact / Questions:

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END