Streaming Machine Learning (SML)

Alessio Bernardo & Emanuele Della Valle 05-06-2021

About me



Alessio Bernardo

Ph.D. Student in Data Science:

- Politecnico di Milano
- Research on Streaming Machine Learning

M.Sc. & B.Sc. Computer Engineering:

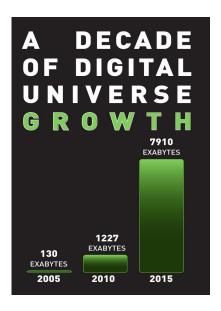
Politecnico di Milano

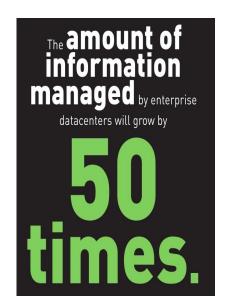
Part I

Introduction

Credits

- Albert Bifet DATA STREAM MINING 2020-2021 course at Telecom Paris
- Alessio Bernardo & Emanuele Della Valle



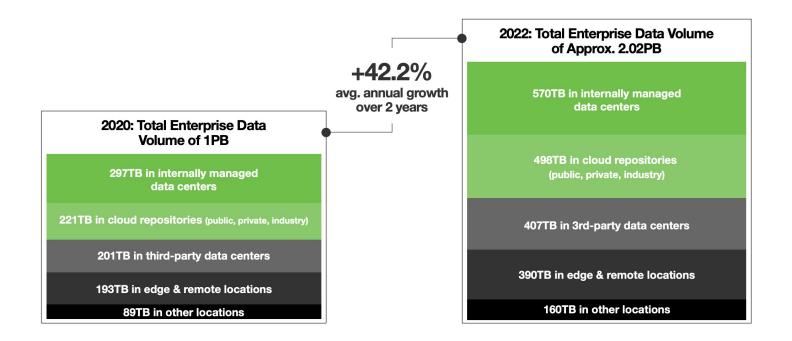




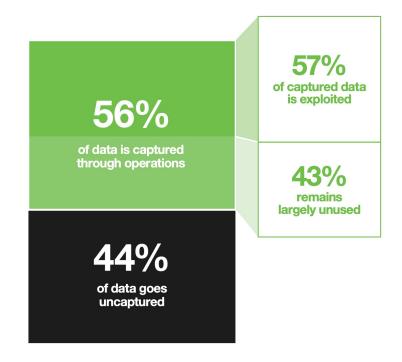
Source: IDC's Digital Universe Study (EMC), June 2011



Source: @LoriLewis and @OfficiallyChadd



Source: SEAGATE TECHNOLOGY, June 2020



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Data is growing, and the rate of growth is accelerating. The sum of data generated by **2025** is set to accelerate exponentially to **175 zettabytes**, an **order of magnitude** bigger than the **storage** production **capability**.

Dave Mosley,
CEO of SEAGATE TECHNOLOGY

Innovation is not driven by trends, but by the need to create more value under constraints. This exponential inflation will thus require analysing almost 30% of global data in real-time.

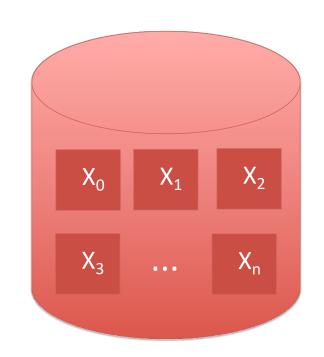
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Batch vs Data Stream

Batch

Random access to data

No restrictions on memory/time for training

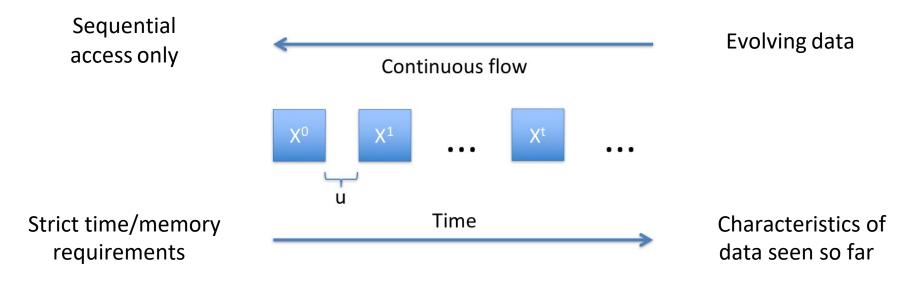


Well defined training phase

Access to all labeled data used for training

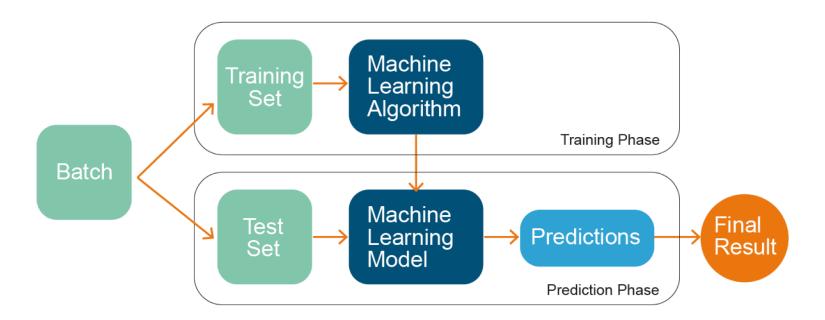
Data Stream

Continuous flow of data generated at high-speed in dynamic, time-changing environments.

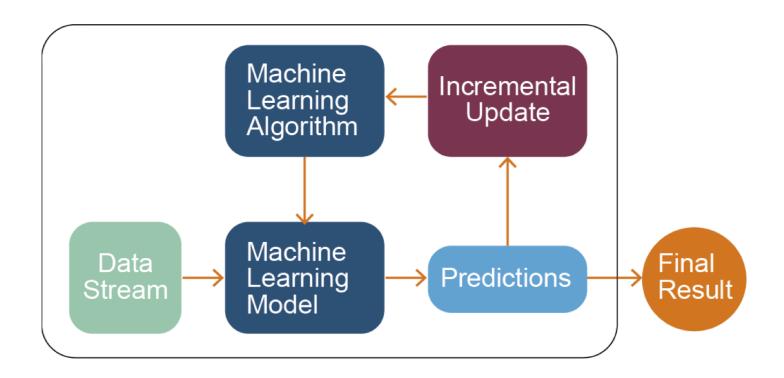


ML vs SML

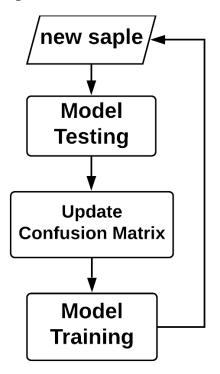
ML Models



SML Models



Prequential Evaluation



Estimate prequential error (PE):

Sliding window of size w

Fading factor

$$PE_i = \frac{\sum_{k=1}^{i} \alpha^{i-k} * e_k}{\sum_{k=1}^{i} \alpha^{i-k}} \quad \text{with } 0 \ll \alpha \le 1$$

Gama, J., Sebastião, R. and Rodrigues, P.P.: Issues in evaluation of stream learning algorithms. In ACM KDD, 2009.

SML Models

- Incorporate data on the fly
- Unbounded training sets
- Resource efficient
- Dynamic models



Benefits

- One sample at a time
- Incremental models
- Time and Memory management

Challenges

Non-stationarity (Concept drift)

Class imbalance

Hyper-parameter Tuning

QUIZ

- 1. What are the data streams characteristics?
 - a. All data are available, non-stationary, bounded
 - b. One sample available at a time, non-stationary, unbounded
 - c. One sample available at a time, unbounded, access to old data
- 2. How do the SML models address the time and memory problem?
 - a. Updating the model with the new sample and then discarding it
 - b. Updating the model with the new sample and then saving it
 - c. Saving every time the new sample and retraining anew the model

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EXERCISE 1: From batch to stream learning LAB 1: Prequential error