THE DATA SCIENCE LAB Introduction to Data Stream Processing

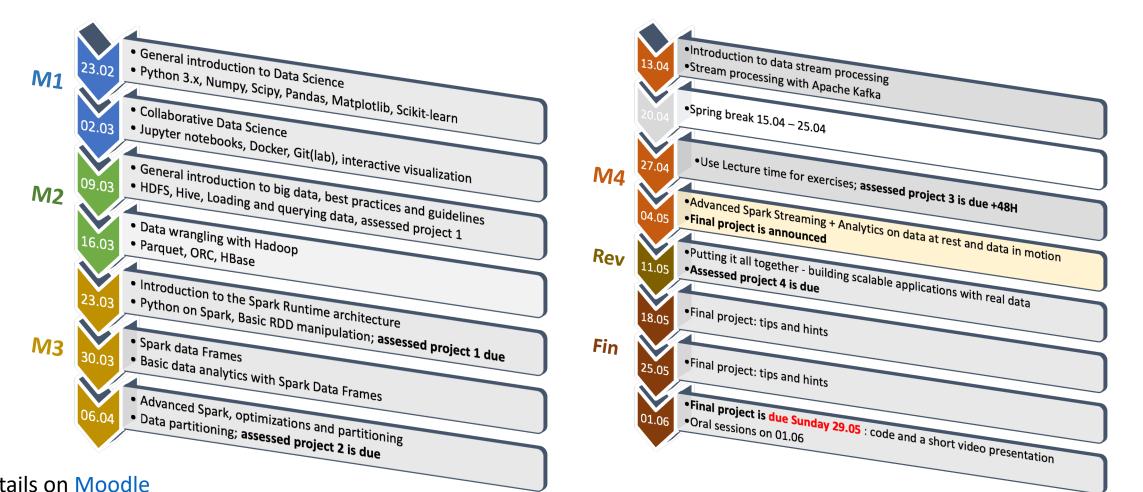
COM 490 – Spring 2022

Week 10

THIS CLASS WILL BE RECORDED



Agenda Spring 2022



^{*}Details on Moodle



Today

- Stream Processing
 - Advanced topics
 - Operations on streaming data (joins)
 - Time constraints
- Exercises (demonstration)
 - https://dslab2022-renku.epfl.ch/projects/com490/lab-course
 - Go to week 9
- Homework 4
 - https://dslab2022-renku.epfl.ch/projects/com490/homework-4
 - due before 10.05 23h59
- Final project instructions



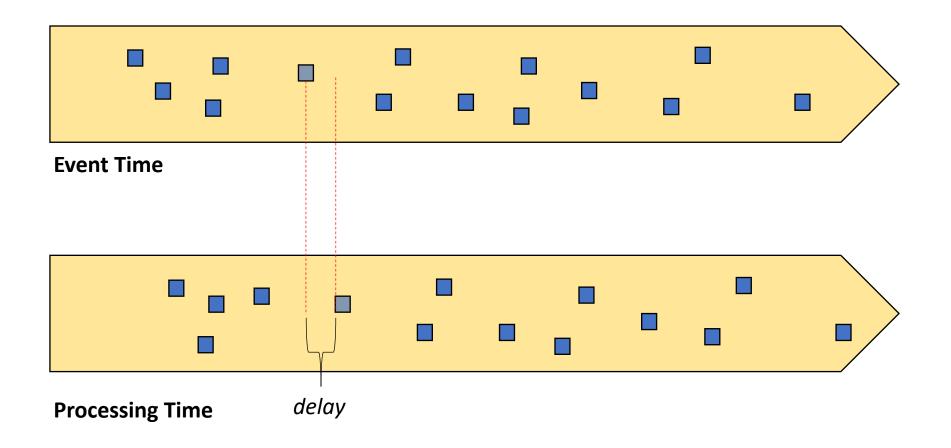
Stream Processing Module

- Objectives
 - Review concepts of stream processing
 - Experiment with typical tools for
 - Data ingestion and processing
- Week 8
 - Concepts
 - Experiments
- Week 9
 - Advanced topics
 - Operations on streaming data (joins)
 - Time constraints
- Week 10
 - Analytics on data at rest and data in motion
 - Graded homework

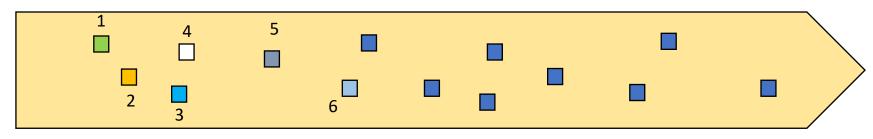


Today's Agenda

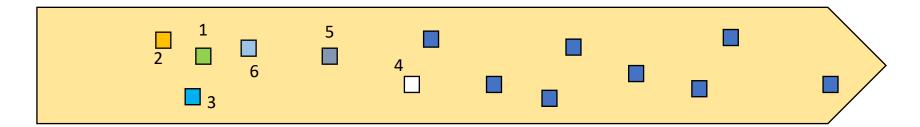
- Week 10
 - Advanced topics
 - Solution to exercises
- Introduction of the final project
- Your questions
 - exercises
 - homework 4





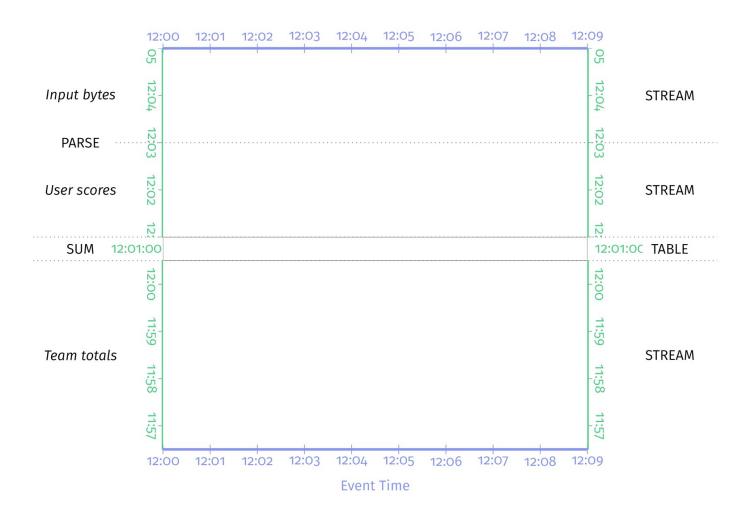


Event Time 1. 2. 3. 4.



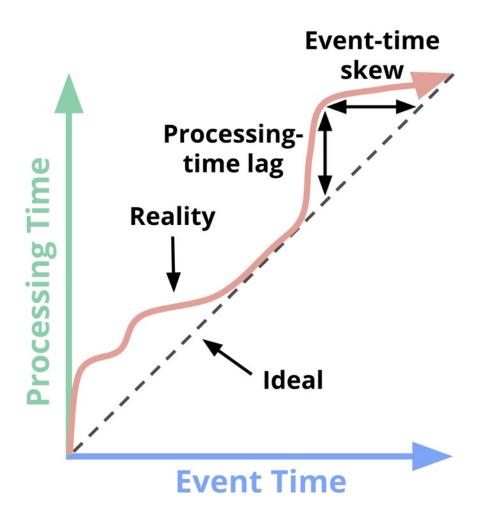
Processing Time





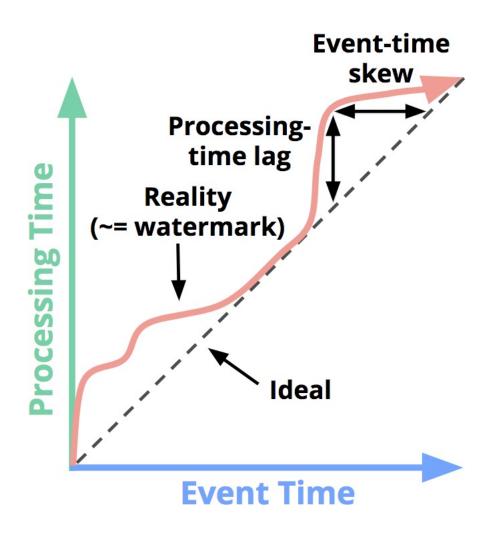
Credits: Tyler Akidau (et al.), Streaming Systems, O'Reilly Media, 2018.





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Watermarking in Spark Streaming

Count on a sliding window of 10 minutes long with 5 minutes sliding interval with 10 minutes watermarking (lateness).

```
from pyspark.sql.functions import *

windowedAvgDF = \
   eventsDF \
        .withWatermark("eventTime", "10 minutes") \
        .groupBy(window("eventTime", "10 minutes", "5 minutes")) \
        .count()
```

Operations

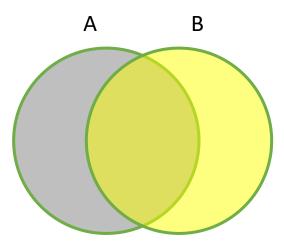
How it works



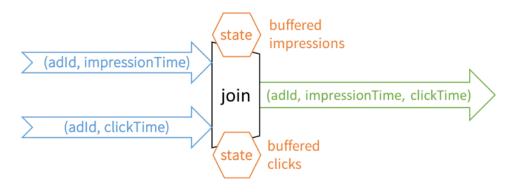
- DStream: continuous stream of data
 - Created from inputs (e.g. Kafka) or derived from other DStreams
 - Supports transformations like RDDs
 - (map, count, join, etc)

- Joins
 - Combining different sets of data (tables) to get a set of results based on some criteria

- Streaming Streaming
 - Stream-Static joins
 - Stream-Stream joins



- Stream-Stream Joins:
 - 2 streams over a common key
- Example: Ad monetization (*)

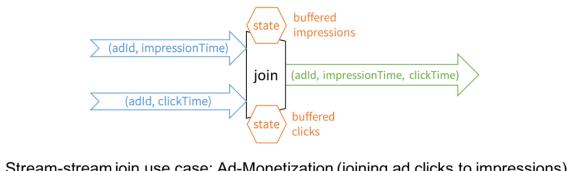


Stream-stream join use case: Ad-Monetization (joining ad clicks to impressions)

(*) Databricks Engineering Blog, Introducing Stream-Stream Joins in Apache Spark 2.3 by <u>Tathagata Das</u> and <u>Joseph Torres</u>, <u>https://bit.ly/2I58Ve7</u>

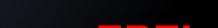


- Streaming Constraints (*)
 - Handling of late/out of order data
 - Boundaries



Stream-stream join use case: Ad-Monetization (joining ad clicks to impressions)

- Solution
 - Use watermarks for delays / out of order impressionTime and clickTime
 - Set time range boundaries for the join conditions on event-time windows



```
from pyspark.sql.functions import expr
impressions = spark.readStream. ...
clicks = spark.readStream. ...
# Apply watermarks on event-time columns
impressionsWithWatermark = impressions.withWatermark("impressionTime", "2 hours")
clicksWithWatermark = clicks.withWatermark("clickTime", "3 hours")
# Join with event-time constraints
impressionsWithWatermark.join(
  clicksWithWatermark,
  expr("""
    clickAdId = impressionAdId AND
    clickTime >= impressionTime AND
    clickTime <= impressionTime + interval 1 hour</pre>
    """)
```



Useful references

- [1] https://spark.apache.org/docs/latest/streaming-programming-guide.html
- [2] https://kafka.apache.org/documentation/
- [3] https://www.oreilly.com/radar/the-world-beyond-batch-streaming-101/
- [4] http://www.streamingbook.net/

