

Data LossLess streaming processing

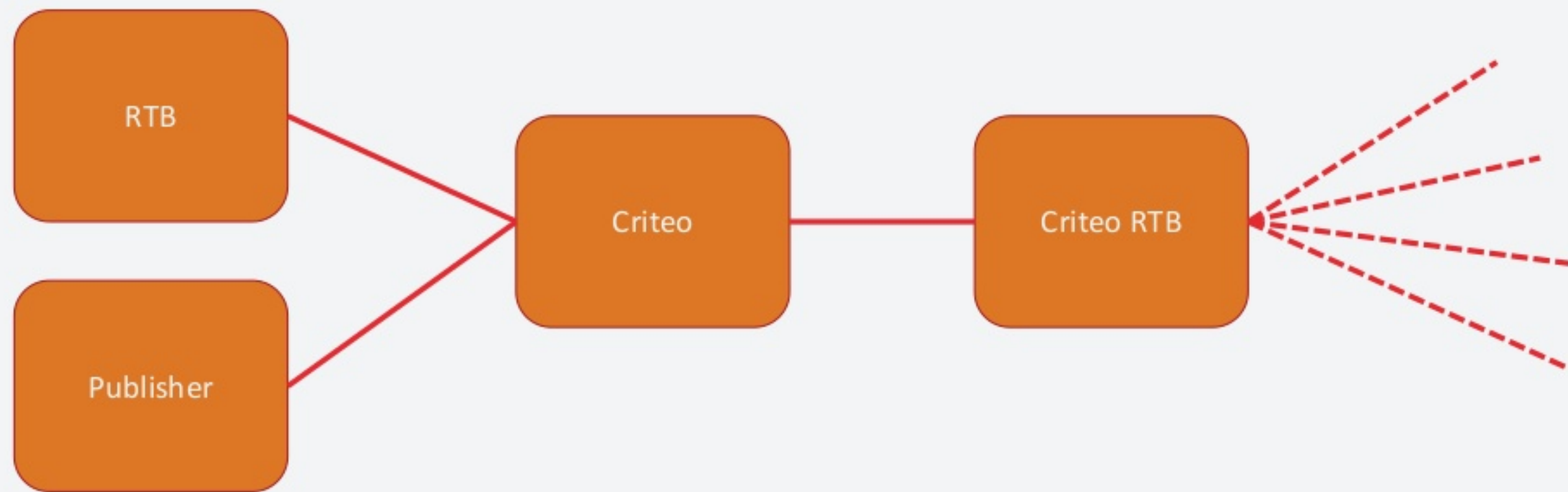
Oleksandr Nitavskyi, Criteo

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“Connecting shoppers to the things they need and love.”



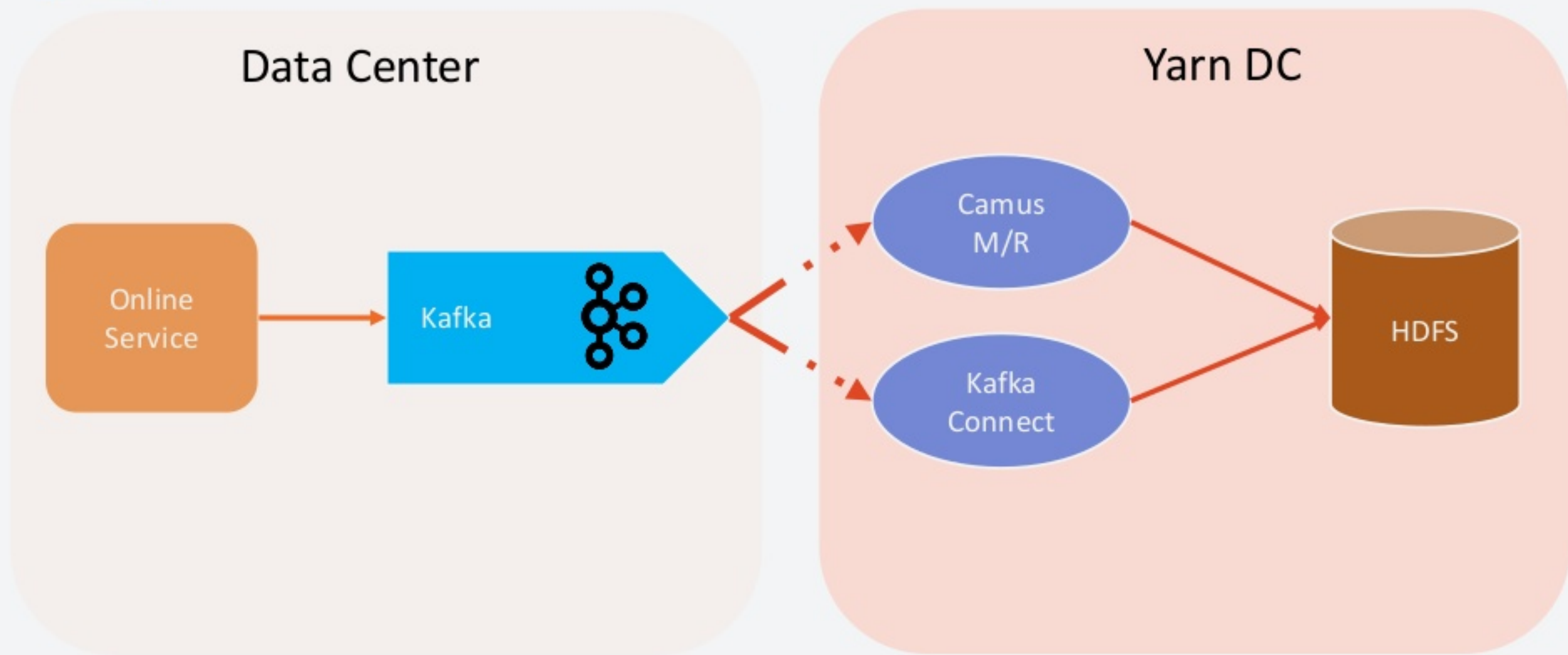
Criteo: business model



Criteo: in numbers



Gathering all data together

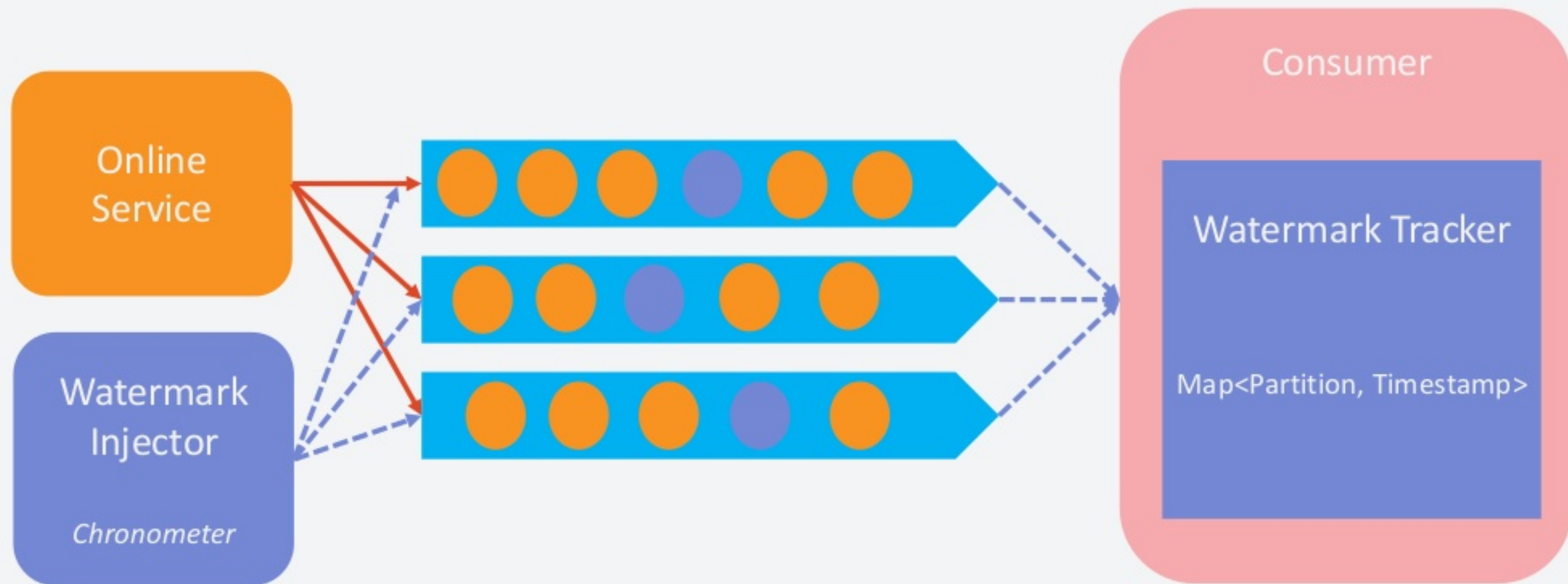


Time series management

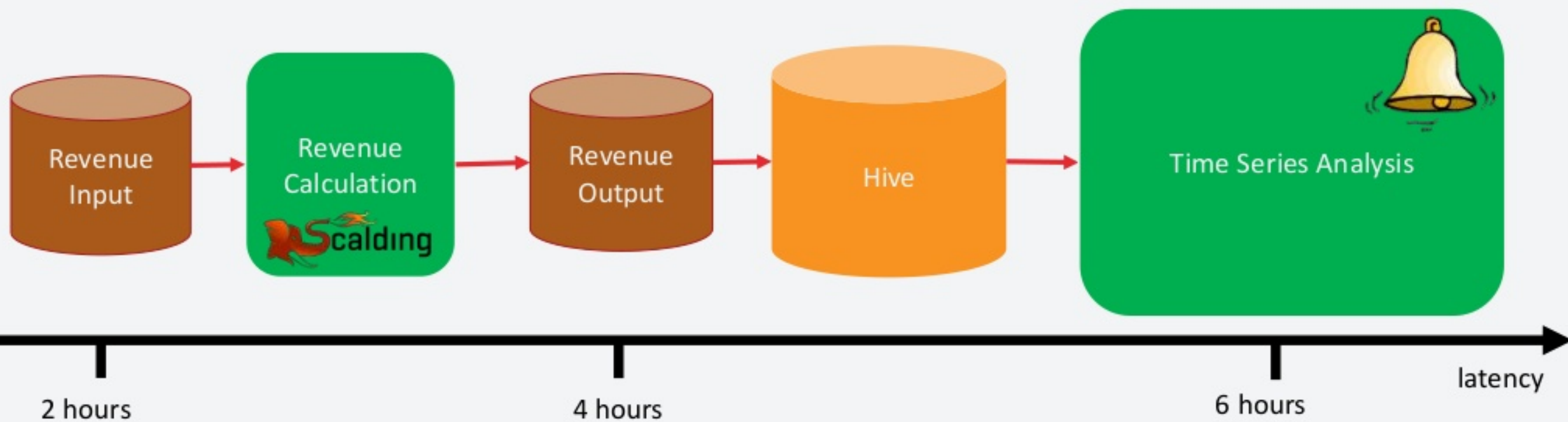


- Single Kafka partition is ordered
- Time of partition **max(watermark)**
- Global time **min(partitions)**

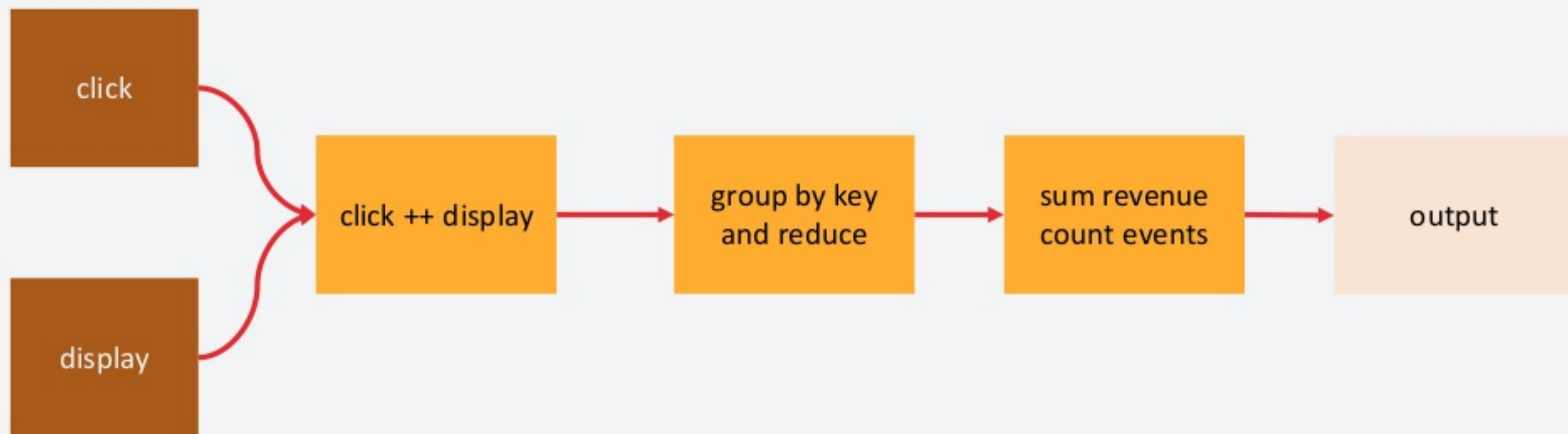
Time series management



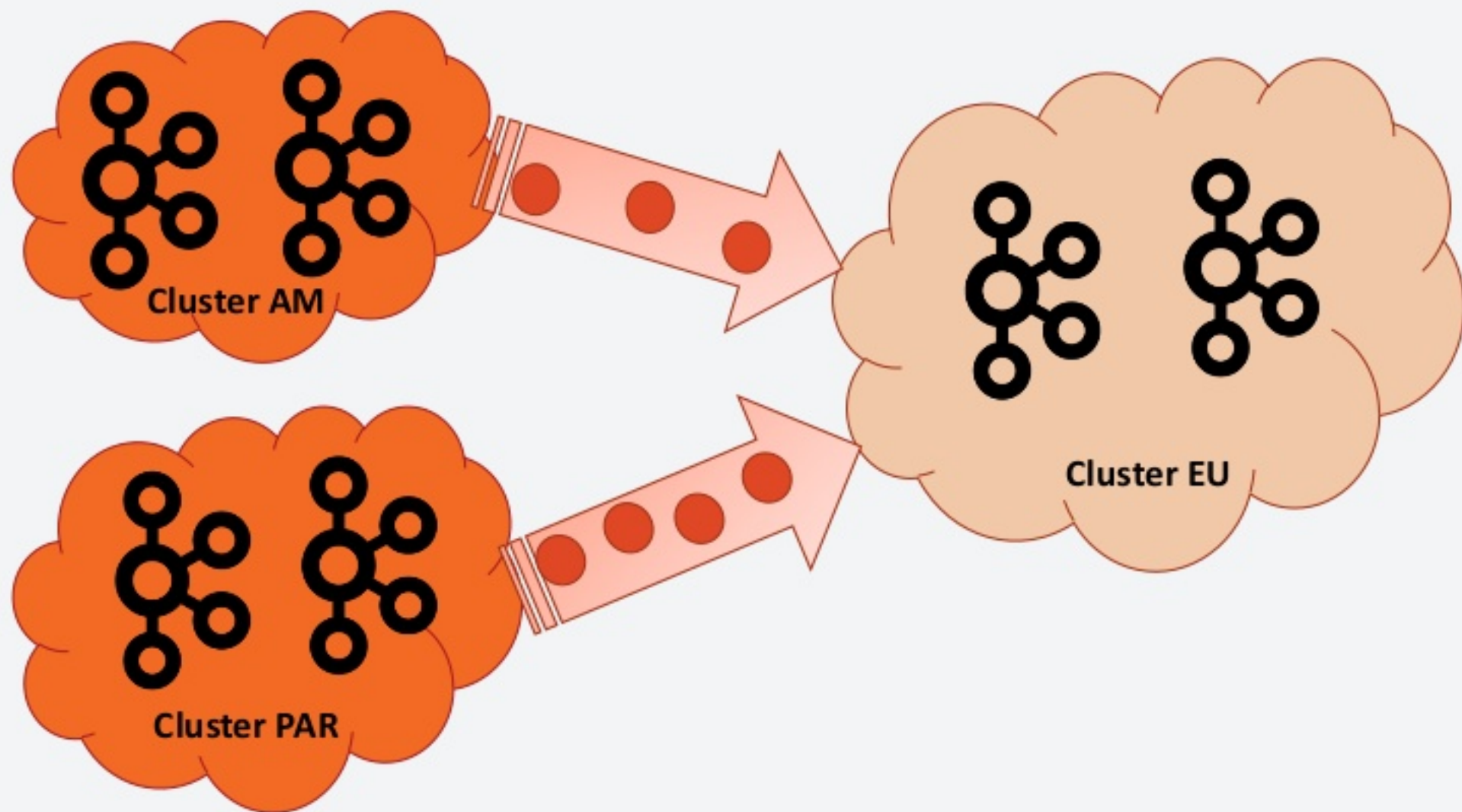
Revenue anomalies offline



Business logic



Stream processing infrastructure



Stream processing infrastructure

- A rogue Kafka client can significantly reduce Kafka cluster performance
- We need to shuffle the data uniformly
- A streaming pipeline needs to have platform level (EU, AS, ...) data

Stream processing first try



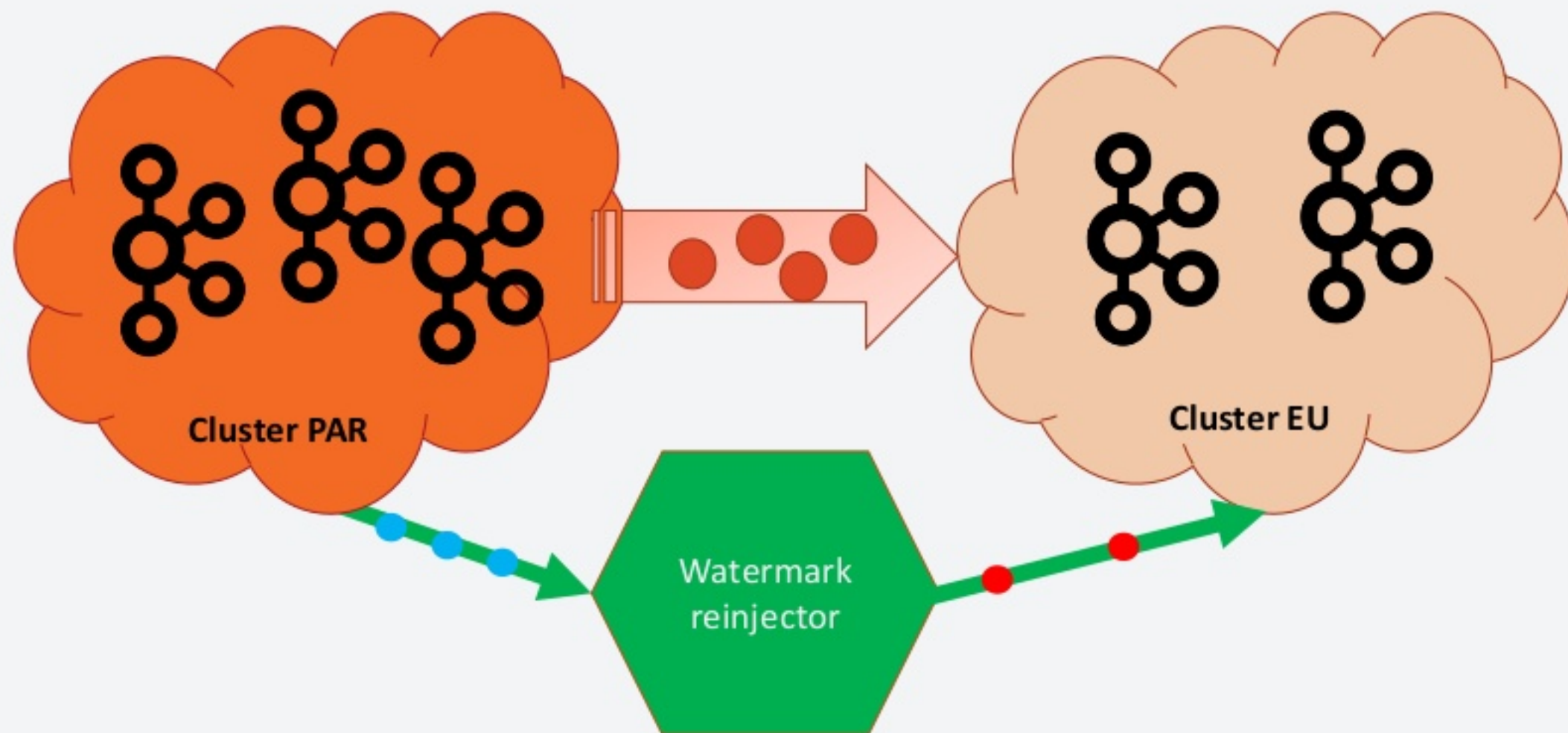
Streaming done right



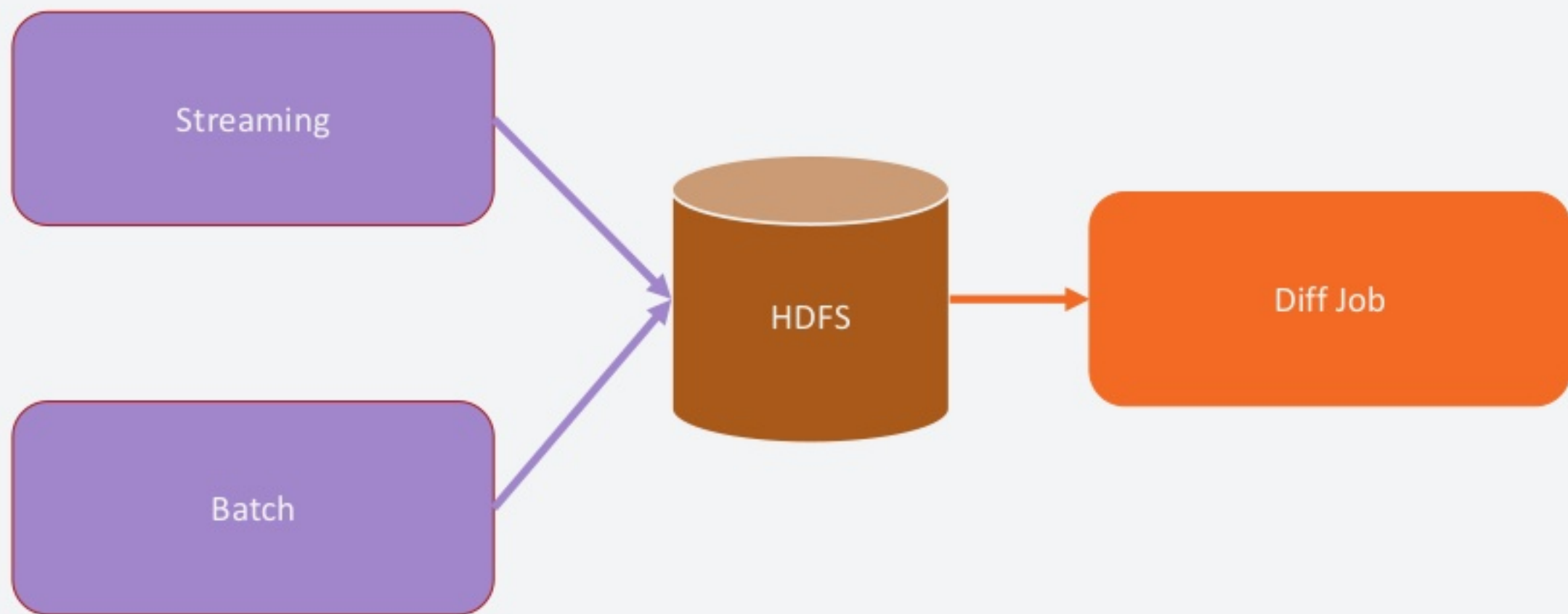
Streaming done right

- Flexible event time support
- Robust state management
- Short development cycle

Accurate event time processing



Result verification





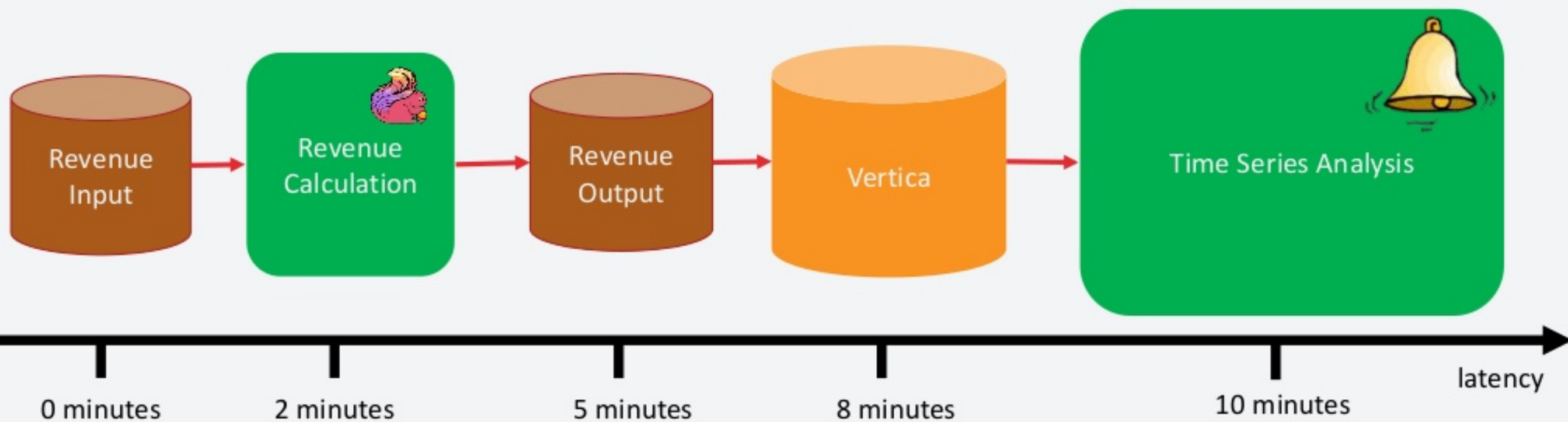
Correctness results

Happy path looked great

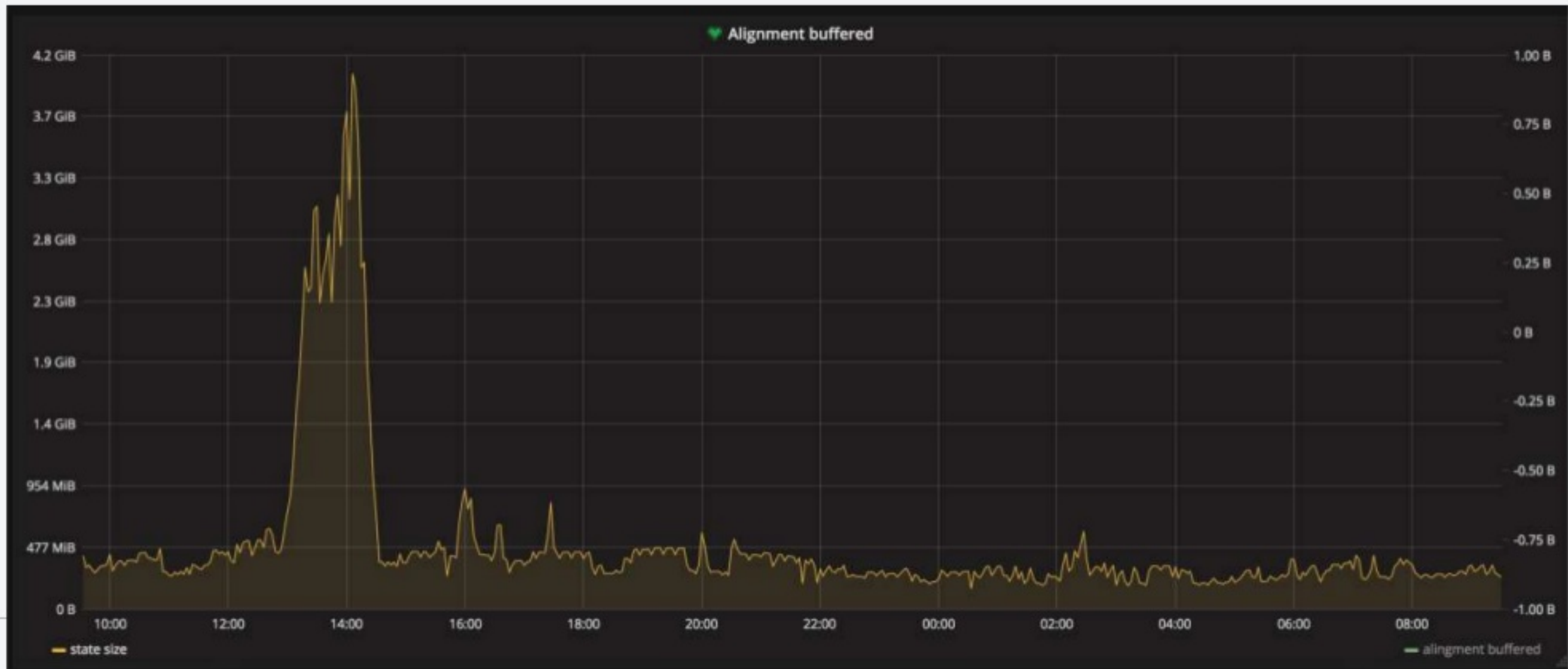
- no watermark issues
- small duplication rate
- no lost state of the Flink
- no big latency
- no spike of late events



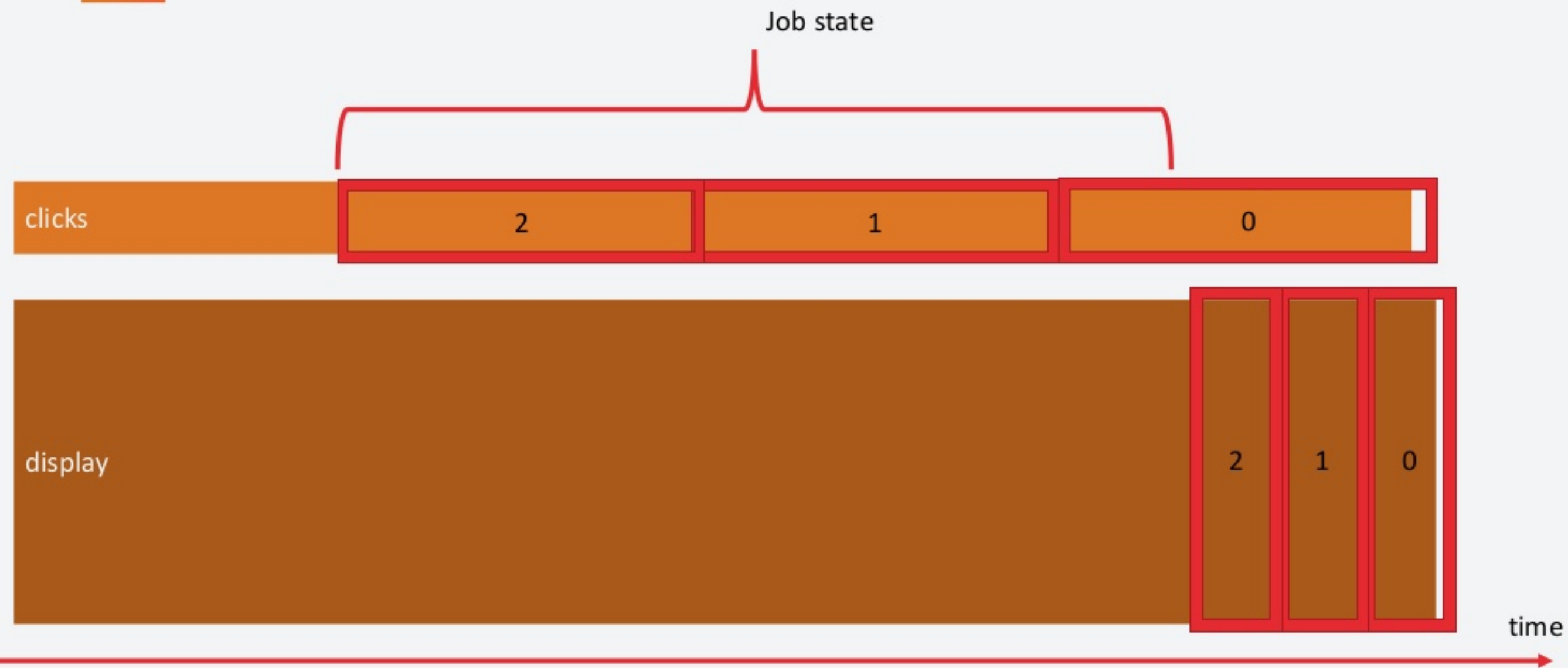
Revenue anomalies online



Job state



Problems with a Kappa I



More generic problems with a Kappa

partition 0

partition 1

partition 2

partition 3

partition 4

Streaming enables use-cases

- Our TSE can monitor any ads campaign changes in "realtime"
- We have enabled a short term campaigns which takes only several hours to run
- We efficiently cut campaigns spending when campaign is over

Notes to take

- Simple business use-case can allow to focus on the infrastructure
- Store watermarks within your data for catching up
- Flink is an enabler

“

Thank you

we are hiring

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