
The Dataflow Model

A Practical Approach to Balancing Correctness, Latency, and Cost in
Massive-Scale, Unbounded, Out-of-Order Data Processing

Tyler Akidau et al.

Christopher Little

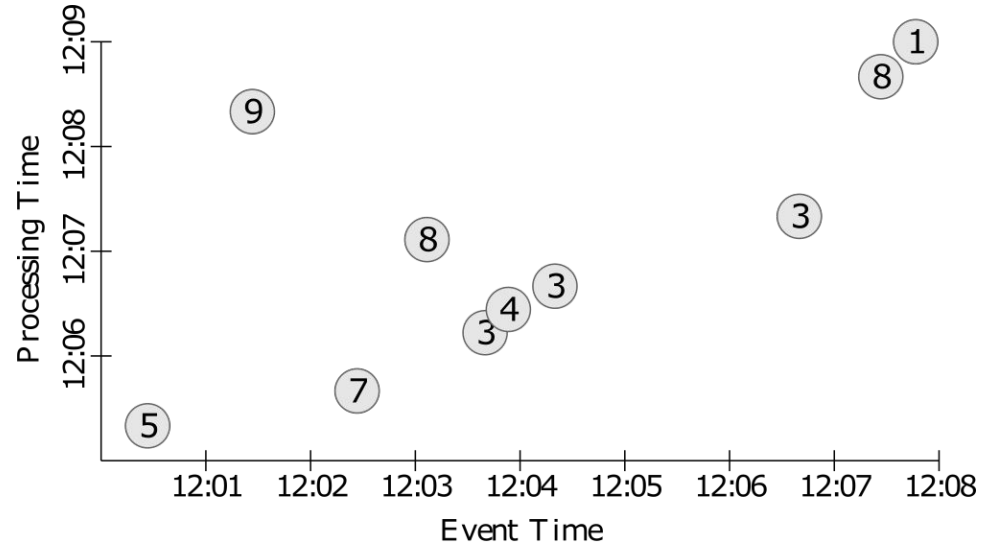
Outline

Prerequisites
Problem
System
Evaluation

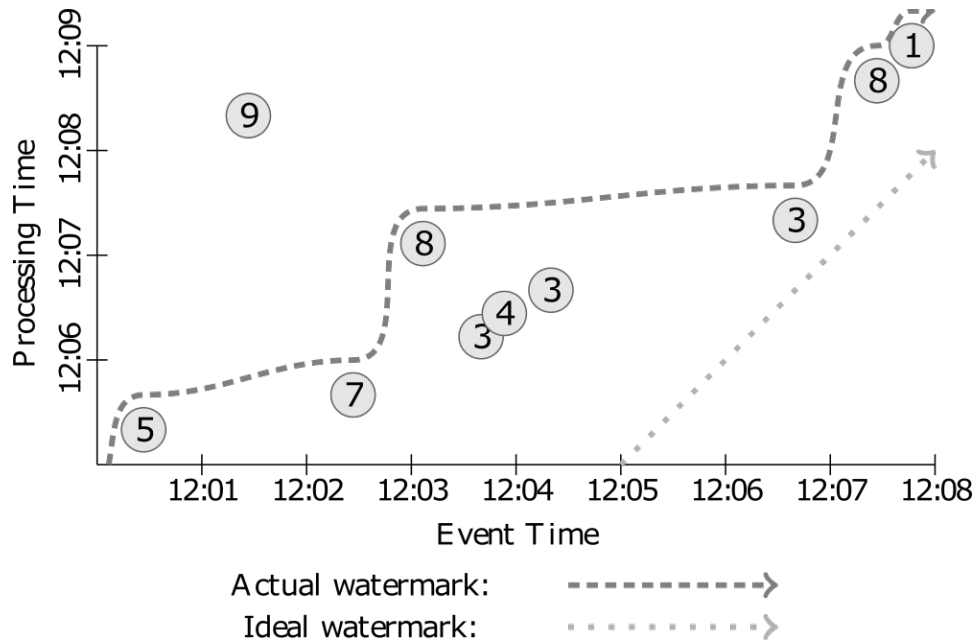


Prerequisites

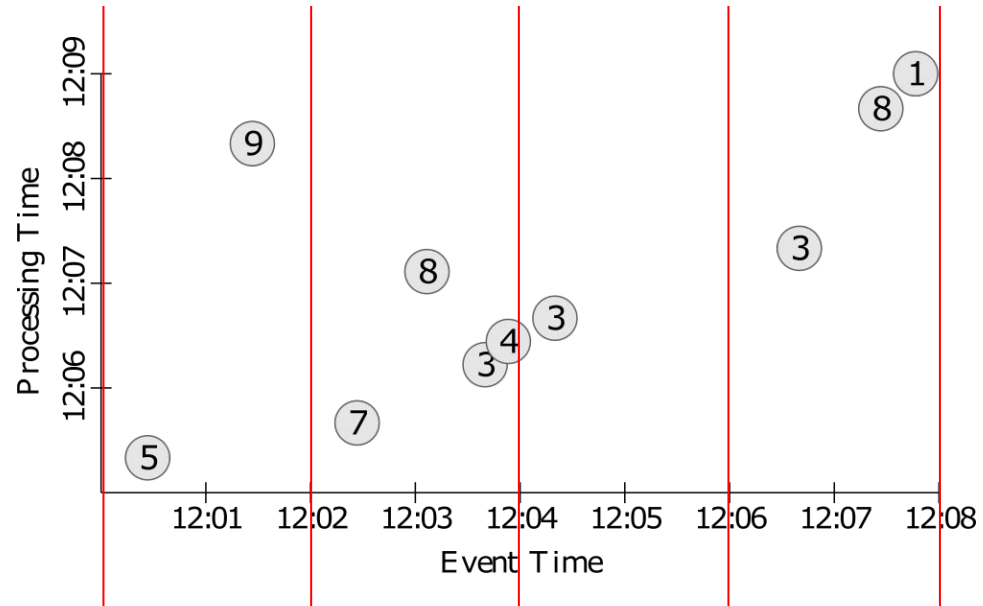
Event vs Processing Time



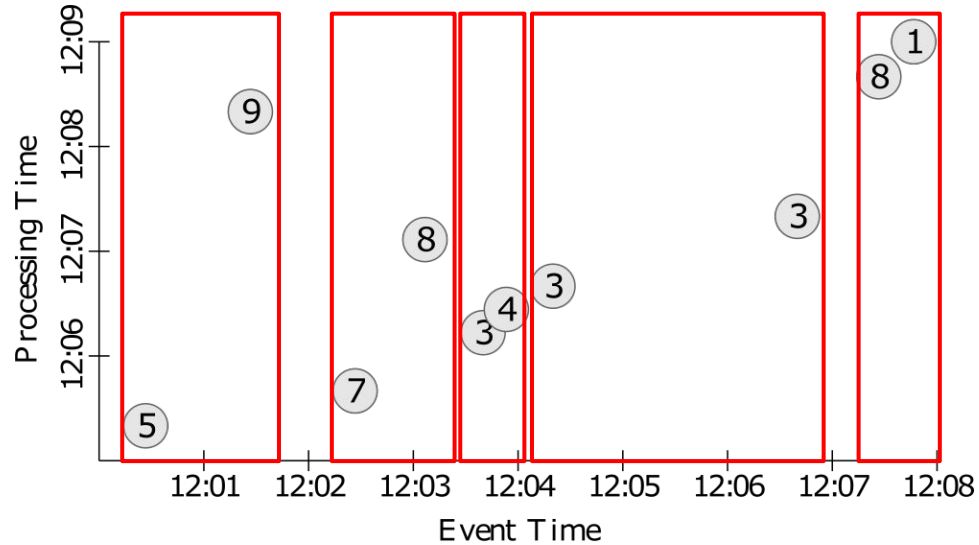
Low Watermark



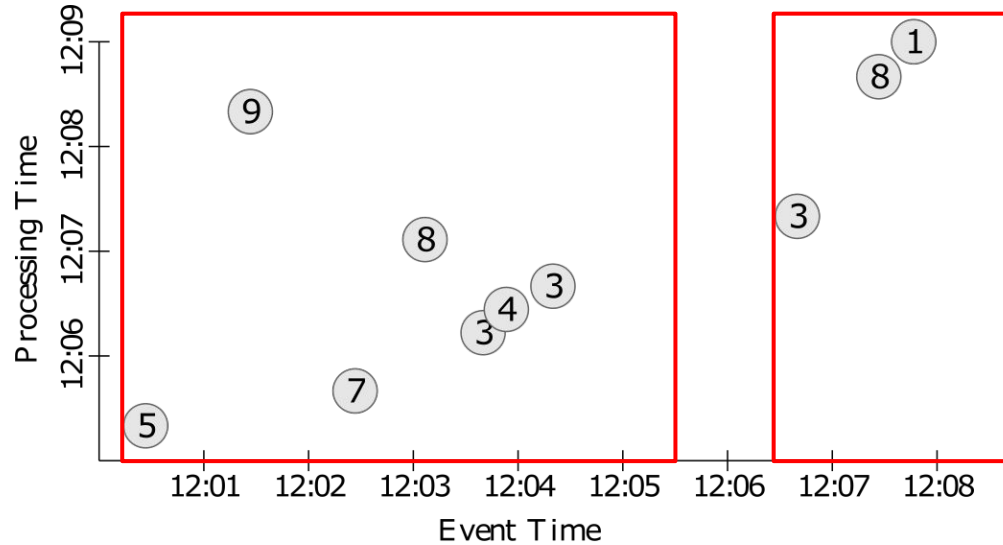
Fixed Windowing



Unaligned Windowing (Tuples)



Unaligned Windowing (Sessions)



Problem

Tracking Video Sessions

- Online/**Offline** video platform
 - Want aggregate stats per user: track **sessions**
 - Pay advertisers per view: must be **correct**
 - Want to adjust bids fast: **low latency**
 - Must scale: **distributed system**
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“A major shortcoming of all the models and systems mentioned above, is that they focus on input data as something which will at some point become complete.”

System

-
-
- **What** results are being computed.
 - **Where** in event time they are being computed.
 - **When** in processing time they are materialized.
 - **How** earlier results relate to later refinements.
-

-
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Two Primitive Transforms

(fix, 1)

(fit, 2)

ParDo(ExpandPrefixes)



(f, 1)

(fi, 1)

(fix, 1)

(f, 2)

(fi, 2)

(fit, 2)

GroupByKey



(f, [1, 2])

(fi, [1, 2])

(fix, [1])

(fit, [2])



Session Windowing Example

(k1, (v1, 13:02))
(k2, (v2, 13:14))
(k1, (v3, 13:57))
(k1, (v4, 13:20))

ParDo
AssignWindows

(k1, (v1, [13:02, 13:32]))
(k2, (v2, [13:14, 13:44]))
(k1, (v3, [13:57, 14:27]))
(k1, (v4, [13:20, 13:50]))

GroupByKey
MergeWindows

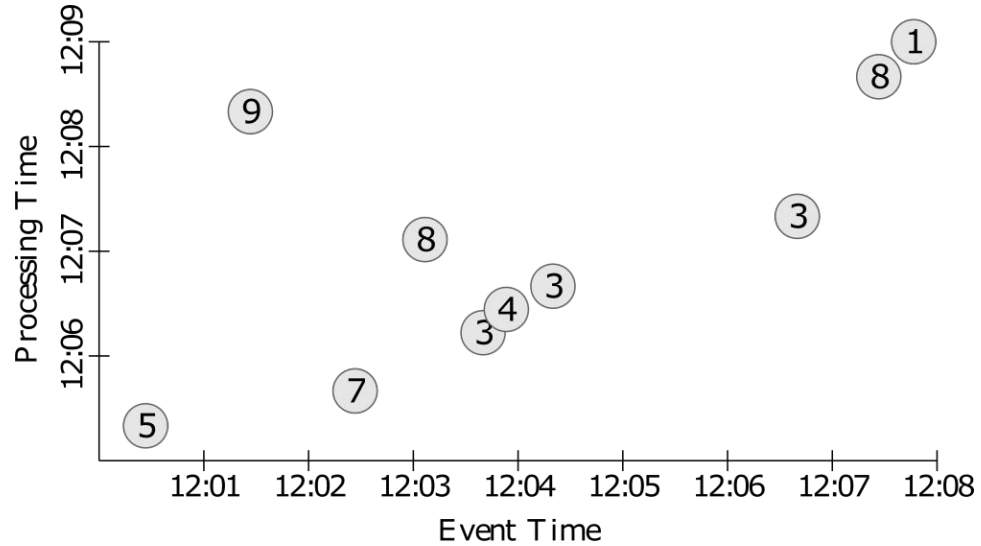
(k1, ([[v1, [13:02, 13:32]]
 , [v3, [13:57, 14:27]]
 , [v4, [13:20, 13:50]]]))
(k2, ([[v2, [13:14, 13:44]]]))

ParDo
MergeWindows

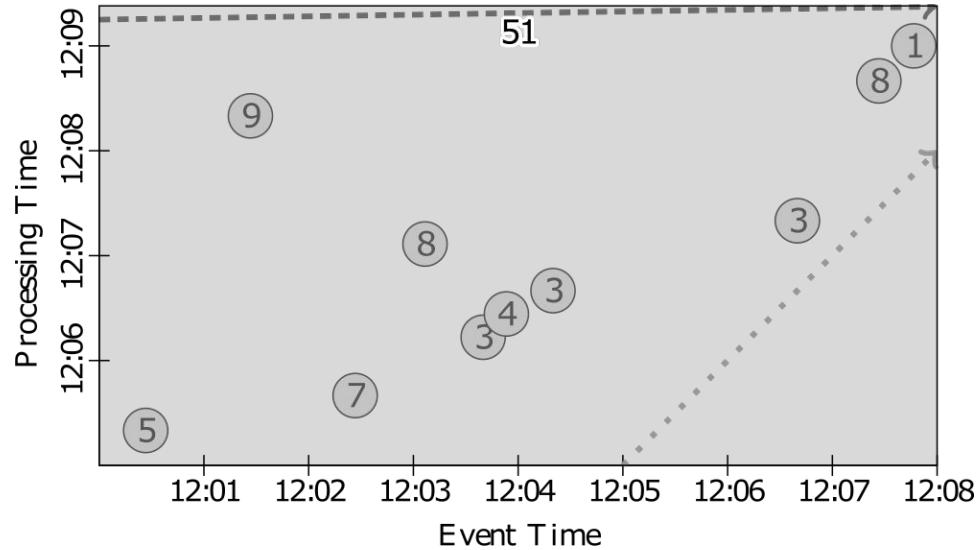
(k1, ([v1, v4], [13:02, 13:50]))
(k1, ([v3], [13:57, 14:27]))
(k2, ([v2], [13:14, 13:44]))

-
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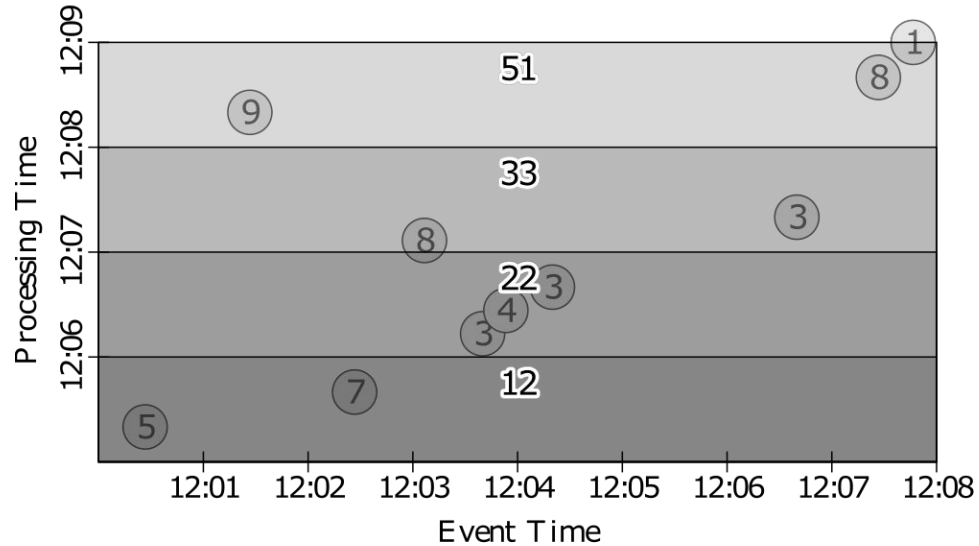
Triggering



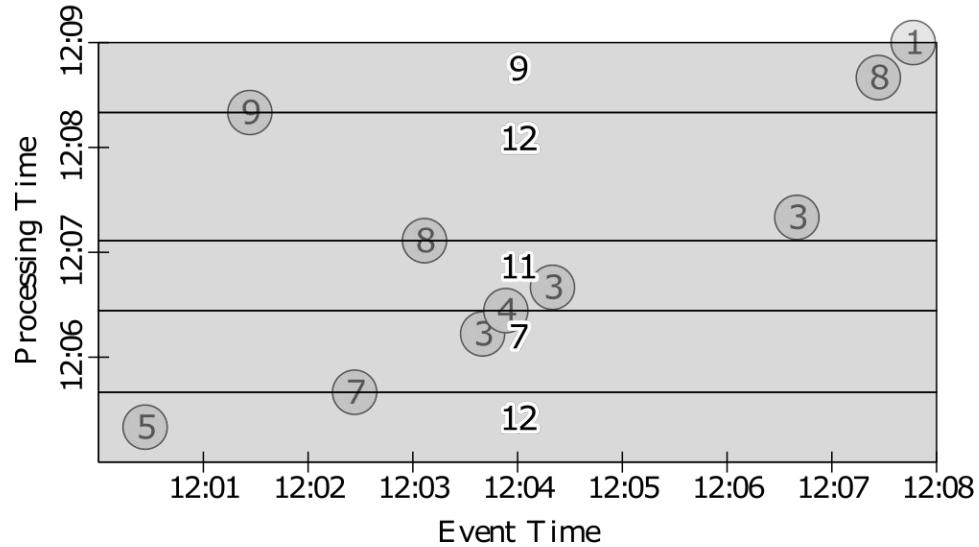
Triggering (end of time)



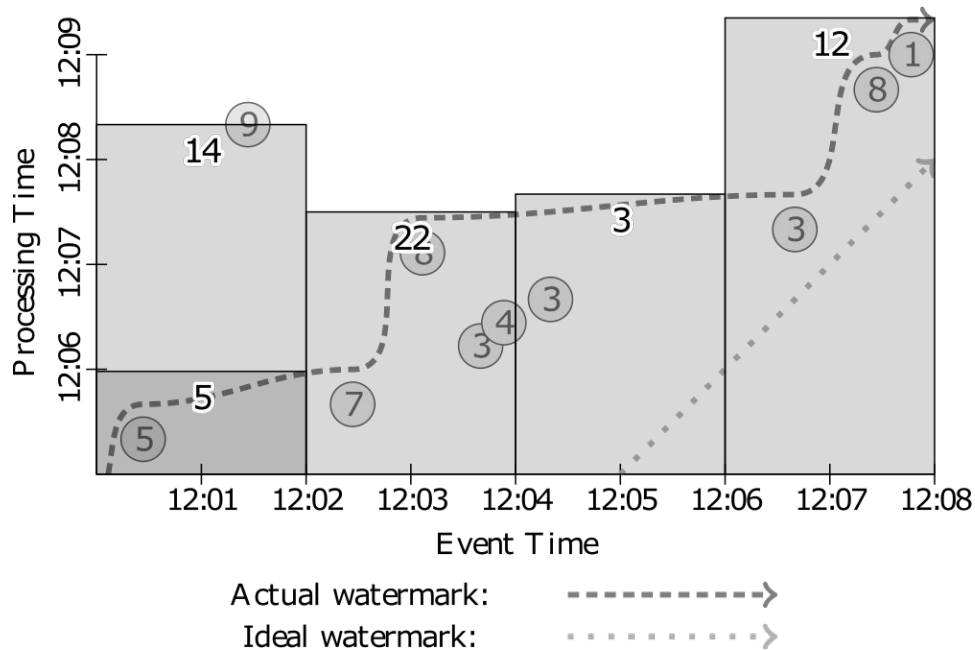
Triggering (periodically)



Triggering (on input, tuples)

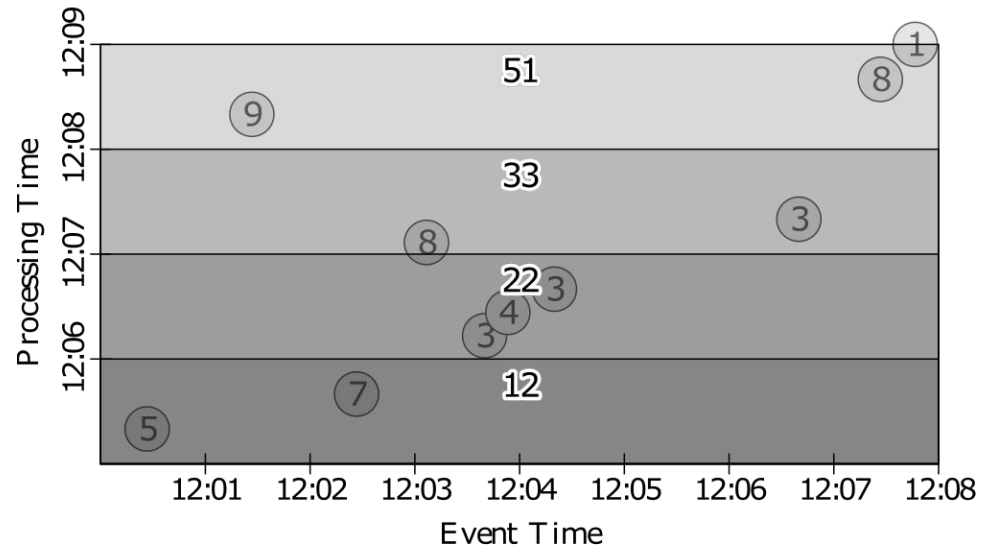


Triggering (on watermark+input)

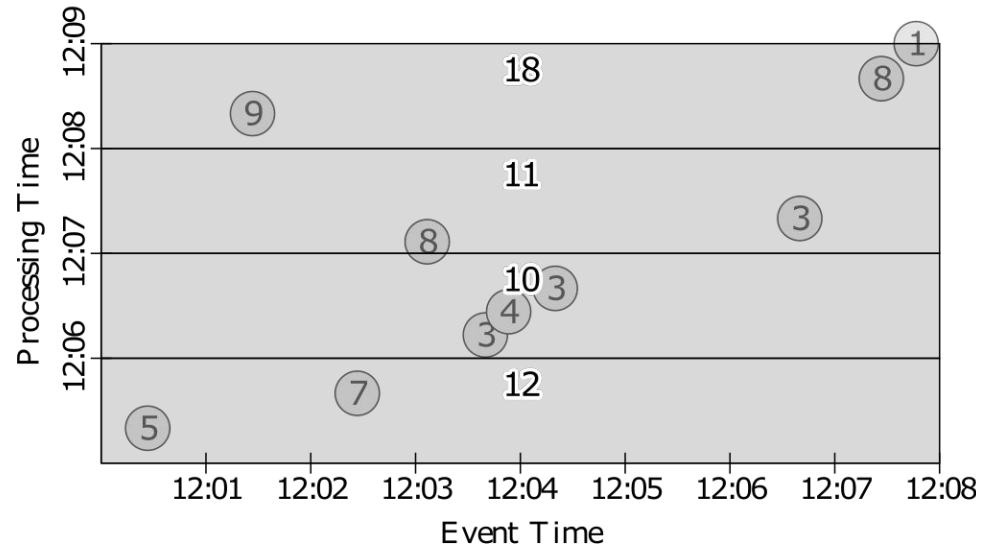


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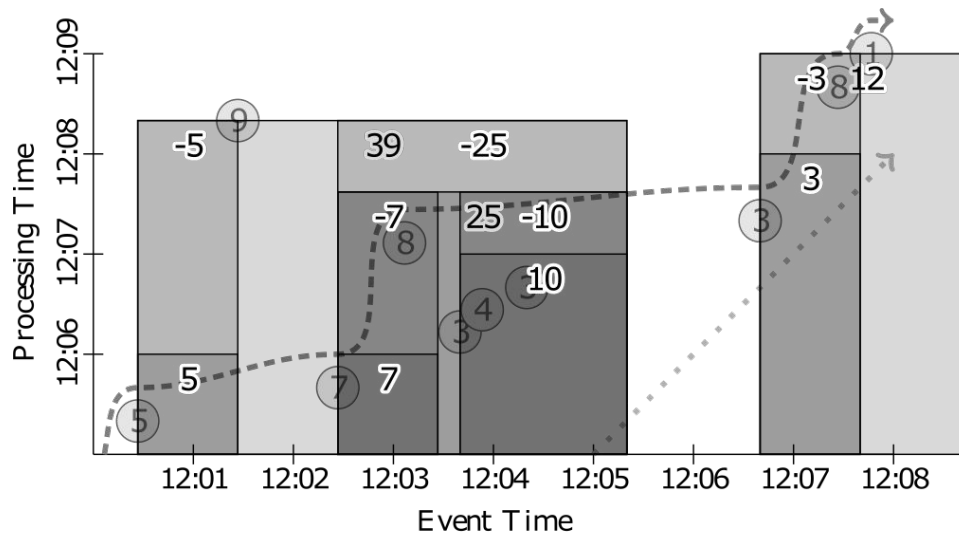
Accumulating



Discarding



Accumulating + Retracting



Actual watermark: ----->
Ideal watermark:>

-
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-

Evaluation

Evaluation

- Name
 - Concepts
 - Necessity
 - Clarity
-

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