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Generating Reproducible Out-of-Order Data Streams

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Experiments with Out-of-Order Streams

We provide a <u>scalable</u> data stream generator, which introduces <u>configurable</u> out-of-orderness in real-world data streams. This enables <u>reproducible</u> and <u>realistic</u> experiments.

- Handling out-of-order data streams is a key feature of modern stream processing systems [1].
- Research on the support of out-of-order stream



Adding out-of-orderness to Data Streams

Source Data Stream:



processing requires <u>reproducible</u>, <u>scalable</u>, and <u>configurable</u> experiments on out-of-order data streams. For example, research on efficient window aggregation [2,3,4].

_ Apache Flink



The evaluation of out-of-order capabilities is hard:

- Public real-world datasets do not reflect all aspects of out-of-order streams (e.g., different delays and fractions of out-of-order tuples).
- Experiments without real-world data can lead to unrealistic results.

Architecture Overview



A general out-of-order experiment data generator:

- Introduce out-of-order data to real-world input data sets.
- Generic configuration of out-of-orderness to enable full parameter exploration.
- Reproducible generation of experimental data with configurable out-of-orderness.

Generator Configuration

Goal:

• The generation of out-of-order tuples has to be deterministic and must not change query results.

Naive Solution:

 Generating out-of-order data by adding a random delay to the event time of certain tuples → changes temporal data distribution.

Our Approach:

• Shift ingestion times of source tuples and keep original event times.

Generator Algorithm

Step 1. Preprocessing:

• Analyze the out-of-orderness of the source data stream to take this knowledge into account for data generation.

 $maxTs \leftarrow 0;$ **for** record in sourceFile **do** \parallel **if** record ts < marTs **then**

$1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \\ 7 \\ 8 \\ 9 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 16 \\ 11 \\ 12 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 11 \\ 12 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 10 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 10 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 10 \\ 10 \\ 10 \\ 10 \\ 10 \\ 10 \\ 10 \\ 10 \\ 10 \\ 10 \\ 10 \\ 11 \\ 12 \\ 15 \\ 16 \\ 10 \\$

```
"dataSource": {
    "file": $path$,
    "seperator": ","|";"|"\t",
    "time": {
        "timeIndex": $field$,
        "sourceTimeUnit": "ps"|"ns"|"ms"|"s"
    }
},
"experimentDataConfigurations": [
    {
        "targetOutOfOrderFactor": [0-100],
        "minDelay": 0,
        "maxDelay": 2000,
        "delaySeed": $seed$
    }
}
```

• Configurable aspects of out-of-order streams:

- Fraction of out-of-order tuples (How many tuples are out-of-order?).
- Minimal/maximal event delay (How late are out-of-order tuples?).
- Out-of-order delay distribution (How are delays distributed?).

Step 2. Generation of out-of-order ingestion time:

- If an in-order tuple becomes an out-of-order tuple, we add a random delay to its ingestion time (based on the configured distribution).
- Finally, we sort the data set by ingestion time.

References

[1] Tyler Akidau, Robert Bradshaw, et al. VLDB 2015.

The dataflow model: a practical approach to balancing correctness, latency, and cost in massive-scale, unbounded, out-of-order data processing.

[2] Jonas Traub, Philipp M. Grulich, Alejandro R. Cuéllar, Sebastian Breß, Asterios Katsifodimos, Tilmann Rabl, and Volker Markl. EDBT 2019. *Efficient Window Aggregation with General Stream Slicing*.

[3] Jonas Traub, Philipp M. Grulich, Alejandro R. Cuéllar, Sebastian Breß, Asterios Katsifodimos, Tilmann Rabl, and Volker Markl. ICDE 2018. Scotty: Efficient Window Aggregation for Out-of-Order Stream Processing.

[4] Kanat Tangwongsan, Martin Hirzel, and Scott Schneider. arXiv preprint 2018. Sub-O (log n) Out-of-Order Sliding-Window Aggregation.

Step 3. Ingestion to the stream processing system:

• Only ingest records if ingestion time is reached.

for record in recordBuffer do
 if record.ingestionTime > now then
 wait(record.ingestionTime-now);
 end
 emit(record)
end

