#### A SUPPLEMENTARY

## A.1 Implementations in IoTDB

All algorithms we presented in this paper have been fully implemented in Apache IoTDB, an open-source time series database system. The corresponding Java classes are Data allocation table [N7], Data partition table [N8], Greedy CopySet replication algorithm [N9] and Min cost flow distribution algorithm [N10].

#### A.2 Customers Workload

IoTDB has served over 1000 backbone and industrial leading enterprises [N11]. We will introduce some successful practices in the following.

A.2.1 Changan Automobile. Changan Automobile [N12] is one of the "Big Four" state-owned car manufacturers of China. Changan Automobile built a smart platform for querying massive amounts of connected vehicle data and remote diagnostics based on Apache IoTDB.

A.2.2 Mcc Cisdi. Mcc Cisdi [N13] is one of the core subsidiaries of MCC under China Minmetals, the Fortune Global 500. CISDI provides overall solutions for domestic and foreign steel industry customers. Cisdi built the core part of the Water and Soil Cloud Industrial IoT Platform for data development governance base on Apache IoTDB.

A.2.3 AUTO AI. AUTO AI [N14] is a provider of technologies and integrated hard/software solutions in the field of intelligent cockpit. It uses Apache IoTDB to manage the driving behavior data of all service vehicles, with a total of 1.6 million online vehicles every day.

The real customers' workload is shown at Table 3. Taking AUTO AI as an example, we collected its production cluster metric from May 2, 2024 to May 9, 2024 at Table 4. It can be seen that the Write Throughput and the Used Disk Space of this cluster fluctuate very little (the coefficient of variation is less than 5%), so the cluster is relatively balanced.

### A.3 Experiment Setup

We first conducted the simulated evaluation on a personal Mac computer, which is equipped with 10-core M1 Pro chip and 16 GB memory. Then we deployed an IoTDB cluster consisting of 11-nodes on Tencent Cloud for processing the IoTDB evaluation. Each node is equipped with a 4-core CPU, 8 GB memory, 100 GB SSD storage and 1.5 GB ethernet. Of these 11 nodes, 8 are used to deploy DataNodes, 2 are used to deploy IoT-benchmark, and the remaining 1 is used to deploy ConfigNode.

A.3.1 IoT-Benchmark. IoT-benchmark [N15] is an open-source benchmark for TSDB in IoT scenario, which we used for generating periodic time series data according to the configuration and send the data batch by batch to IoTDB-Server. We configured IoT-benchmark to simulate sensors with different data types, configured batch size to 100, and assembled the data generated by every 100 sensors into one write request, with the average size of each write request being 60 KB. Finally, each DataNode in IoTDB cluster processes about 1000 write requests per second on average. We approximate the write load limit that the experimental environment can bear by

Table 3: Customers workload

Enterprise	Changan Automobile	Mcc Cisdi	AUTO AI
Cluster Size (DataNodes)	3	3	6
CPU core (per DataNode)	64	8	16
Memory (per DataNode)	256 GB	32 GB	128 GB
Disk (per DataNode)	10 TB	2 TB	1.2 TB
Time Series	1.2 billion	2 million	64.8 million
Write Frequency	Per 30 s	70% per 1 s 20% per 10 s 10% per 0.1 s	Per 3 s
Write Throughput	avg 6000 k/s max 10000 k/s	avg 3000 k/s	avg 1550 k/s max 3000 k/s
TTL	3 months	7 days	5 days
Time Partition	7 days	7 days	7 days

Table 4: IoTDB cluster in AUTO AI

Write Throughput	Used Disk Space
516113/s	231 GB
516792/s	235 GB
521633/s	238 GB
523417/s	229 GB
508919/s	244 GB
510220/s	252 GB
516182.33	238.17
5843.49	8.61
1.13%	3.62%
	516113/s 516792/s 521633/s 523417/s 508919/s 510220/s 516182.33 5843.49

the IoT-benchmark to simulate the high write load in the real production environment.

A.3.2 Alternative Algorithms. We implemented the following DataRegion placement algorithms [N16] to compare with our GCR algorithm. The first one is a Greedy algorithm which always place the new DataRegion to the DataNode with the smallest number of DataRegion since it's a intuitive solution. We then implemented the Copyset [4] and Tiered Replication [3] algorithms. Because the optimization of GCR algorithm in disaster recovery capability is inspired by these two papers. For all DataRegion placement algorithms, we have them generate the same number of DataRegion placement schemes so that each DataNode ends up holding an average of 6 DataRegions, where the load factor 6 is the empirical parameter, we conducted some experiments beforehand to determine the appropriate number of DataRegions hosted by each DataNode.

We implemented the following primary replica selection algorithms [N17] to compare with our CFD algorithm. The first one is a Greedy algorithm which always select the DataRegion on the

DataNode that holds the fewest number of primary replica currently as the primary one to act as an intuitive solution. The second one is a Random algorithm which select a random DataRegion as the primary one. We want to simulate the situation where the DataRegionGroup elects the primary replica by itself, and each DataRegion has the same weight to be elected as the primary one by using this algorithm.

# A.4 Supplementary experiments

We will redo the IoTDB evaluation, which we hope can answer the following questions that reviewers are concerned about:

- R1D6: Un-balanced partitions. We have configured IoT-benchmark to generate sensors with different load. And we will set the same configuration when redo experiments.
- R2D1: Add the comparison of time series dimension. The number of time series will be increased proportionally after the cluster is expanded in expansion experiment.
- R3O2: broader deployment and evaluation. As shown in Table 3, most of our customers we serve can meet their business requirement through deploying an IoTDB cluster with not exceeding six DataNodes. Thus, to better demonstrate the expansion capabilities that benefit from our algorithms implemented in IoTDB. We will

set the number of DataNodes to 16 and deploy a 19-nodes IoTDB cluster for expansion experiment.

• R3D1: Using real world dataset. We will use IoT-benchmark to simulate a real workload from one of our customers for both expansion and disaster recovery experiment.

To sum up, we will use IoT-benchmark to simulate a customer's real workload and genererate sensors with different load to redo 6.2 IoTDB evaluation. Furthermore, we'll start from 8-DataNodes IoTDB cluster and extends it to 16 DataNodes, exponentially increase the number of time series after expansion for the expansion experiment.

[N7] Data allocation table. https://github.com/apache/iotdb/blob/master/iotdb-co

 $[N8]\ Data\ partition\ table.\ https://github.com/apache/iotdb/blob/master/iotdb-coresponded to the partition of the partit$ 

 $[N10]\,Min\,cost\,flow\,distribution\,algorithm.\,https://github.com/apache/iotdb/blob/algorithm.$ 

 $[N11]\ User\ cases.\ https://www.timecho-global.com/product$ 

[N12] Changan Automobile. https://en.wikipedia.org/wiki/Changan\_Automobile

[N13] Mcc Cisdi. http://www.cisdigroup.com.cn/html/2/abou/

[N14] AUTO AI. https://www.autoai.com/en

[N15] IoT-benchmark. https://github.com/thulab/iot-benchmark

[N16] Alternatives of replica placement algorithms. https://github.com/CRZbulab

 $[N17]\ Alternatives\ of\ primary\ replica\ selection\ algorithms.\ https://github.com/CR$