

OPC UA Industrial Plant HUB

A Distributed Data Collection System for Industrial Automation

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Distributed Systems Course - A.A. 2025-2026
University of Bologna

The Problem

Industrial Data Collection Challenges

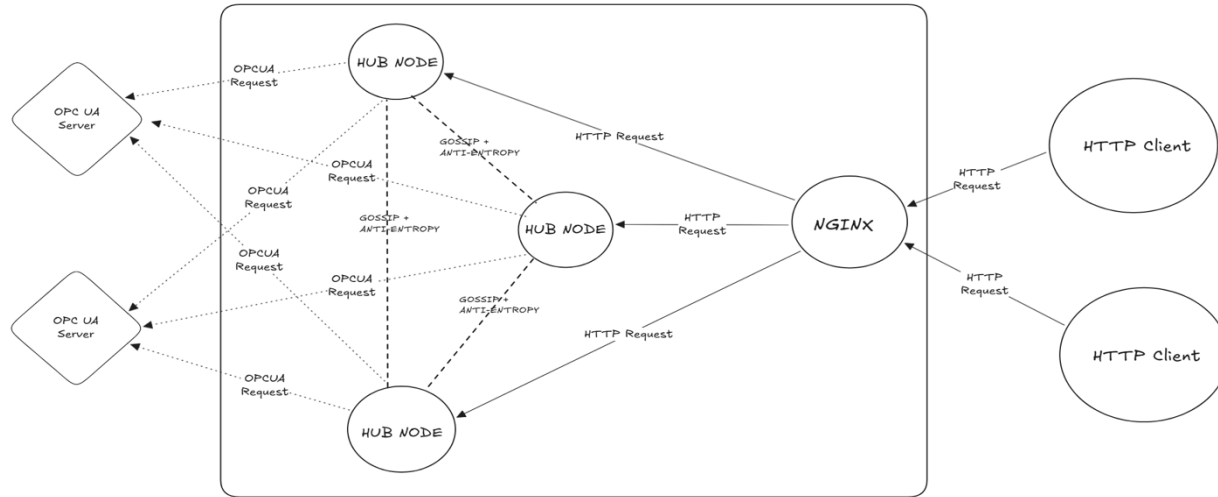
- **24/7 Operations**
Plants require continuous monitoring
- **Single Point of Failure**
Traditional architectures are risky
- **Geographic Distribution**
Plants span multiple locations
- **Dynamic Reconfiguration**
Add sources without downtime

Distributed Multi-HUB Solution

- **3+ HUB Nodes**
Parallel OPC UA data collection
- **No Central Coordinator**
Peer-to-peer synchronization
- **nginx Load Balancer**
Client transparency + health checks
- **Dynamic Addition**
Runtime OPC UA server config

System Architecture

Hybrid Client-Server / Peer-to-Peer Design



Key Components:

Client-Server Layer: REST API with location transparency

P2P Layer: Gossip + Anti-Entropy protocols

HUB Nodes: OPC UA collection + Local SQLite + REST server

CAP Theorem Positioning

Why AP (Availability + Partition Tolerance)?

Property	Choice	Rationale
Consistency	Eventual	Industrial monitoring tolerates 30s lag
Availability	✓ High	24/7 operations cannot tolerate downtime
Partition Tolerance	✓ Full	Nodes must operate independently during network failures

Design Decision:

Clients prefer "stale data" over "no data"

Strong consistency → synchronous coordination → higher latency + lower availability

Eventual consistency with convergence guarantee: **<60 seconds**

Gossip Protocol

Heartbeat-Based Distributed Failure Detection

Protocol Flow:

1. Every 5 seconds

Each HUB → POST /internal/gossip/ping to peers

2. Message includes Lamport Clock

(sender's logical time)

3. Peer responds

ACK + its own LC

4. API Timeout 3s

→ node marked as not responsive

5. 3 missed heartbeats (15s)

→ **node marked DEAD**

Properties

- **Strong Completeness**

All survivors detect failure

- **Weak Accuracy**

False positives possible,
automatic recovery

Validation

Detection time measured:

15.6s ± 0.4s

(5 test runs)

Anti-Entropy Protocol

Pull-Based Eventual Consistency

Synchronization Flow:

1. Every 10 seconds

HUB check local Lamport Clock state for each peer

2. Request sync from alive peers

POST /internal/anti-entropy/sync with LC state vector

3. Peer identifies gaps

Queries local data not yet propagated to requester based on LC comparison and respecting casual ordering)

4. Returns missing data points

Delta containing [node_id, LC, tag, value, ...]

5. Requester converges state

Merge received data points

UpdateLC = max(local_lc, received_lc) per node

Conflict resolution: Last-Write-Wins

Why Pull-Based?

- ✓ Natural backpressure
(receiver controls rate)
- ✓ No tracking overhead
(sender doesn't track)
- ✓ Partition resilient
(auto-resumes)

Validation

Convergence time:

10-25s
(typical 15s)

Lamport Clocks

Casual Ordering Without Synchronized Time

Implementation:

- Each HUB maintains **local LC** (integer counter)
- Increment on every event:
 - Data point collected → LC++
 - Server config added → LC++
 - Message sent/received →
 $LC = \max(\text{local}, \text{received}) + 1$
- Data stored with metadata:
(node_id, Lamport_clock, tag, value, ...)

Conflict Resolution: Last-Write-Wins

Example:

- Node A adds server config
→ assigns **LC = 1000**
- Node B adds same server
(during partition)
→ assigns **LC = 1001**

After sync:

Conflict resolved by highest LC

→ Node B config “wins” (deterministic)

Key Property: Provides casual ordering across distributed events without physical clock sync.

Key Design Decisions

1. Leaderless Multi-Master

- Every node = source + replica
- No master/coordinator
- No SPOF at application layer

2. Local Storage + Replication

- Each HUB: SQLite database
- No central DB dependency
- Trade-off: Eventual consistency

3. Stateless HTTP API Layer

- JWT-based authentication
- Horizontal scalability
- nginx round-robin distribution

4. Dynamic Configuration

- Add servers via POST API
- Config propagates <30s
- No system restart required

Technology Stack

Backend

- Python 3.12+
- FastAPI + uvicorn
- SQLite (local DB)
- asyncua (OPC UA)

Infrastructure

- Docker Compose
- nginx (TLS + LB)
- TLS 1.2/1.3 encryption
- Nodes health checks

Distributed Protocols

- Gossip Protocol
- Anti-Entropy
- Lamport Clocks
- JWT Auth

Testing Strategy

3-Level Validation Approach

1. Functional Testing

- OPC UA connection + polling
- Anti-Entropy data synchronization
- Dynamic server addition
- Gossip failure detection
- Nginx traffic redirection
- JWT auth + RBAC

2. Non-Functional Testing

- Eventual consistency
- Network partition (iptables)

3. Deployment Testing

- Docker Compose from scratch
- Persistence across restarts

Performance Results

Gossip Failure Detection

Target: Detect within 15s

Measured: 15.6s $\pm$ 0.4s

 **PASSED**

Anti-Entropy Convergence

Target: Converge <60s

Measured: 10-25s (typically 15s)

 **PASSED**

Load Balancer Failover

Target: Redirect <30s

Measured: 0 errors, seamless

 **PASSED**

Dynamic Server Addition

Target: Propagate <30s

Measured: Node-B: 12s, C: 18s

 **PASSED**

Limitations & Future Updates

1. OPC UA Full Tree Browsing • Startup 5-10 min with >10K nodes

Future: Selective and dynamic OPC UA node configuration

2. SQLite Scalability • ~200 writes/s single-writer limit – possible constant lock contention

Future: Migrate to PostgreSQL/TimescaleDB/InfluxDB

3. Stateless JWT • No immediate revocation (24h validity)

Future: Token blacklist in Redis

4. Single nginx Instance • Infrastructure SPOF

Future: nginx replication + DNS LB

5. No Observability • Missing Prometheus metrics/tracing

Future: /metrics endpoints + Grafana

Learning Outcomes

Hands-On Experience with Distributed Systems Concepts

- ✓ **CAP Theorem:** Practical tradeoffs between C/A/P through AP implementation
- ✓ **Logical Clocks:** Lamport clocks for causal ordering without synchronized time
- ✓ **Eventual Consistency:** Convergence guarantees with measurable times (10-60s)
- ✓ **Failure Detection:** Gossip Protocol with completeness/accuracy analysis
- ✓ **Replication Strategies:** Pull-based vs push-based (backpressure advantages)

Key Insight: *Distributed systems principles apply to industrial automation, bridging “theoretical” aspects and real-world manufacturing requirements*

Thanks for the attention

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