

# Confidence-Based Maintenance Optimization in Digital Twins for Aging Sensor Networks

Eita Kobayashi, Nattaon Techasarntikul, Yuichi Ohsita, Hideyuki Shimonishi

The University of Osaka, Osaka, Japan



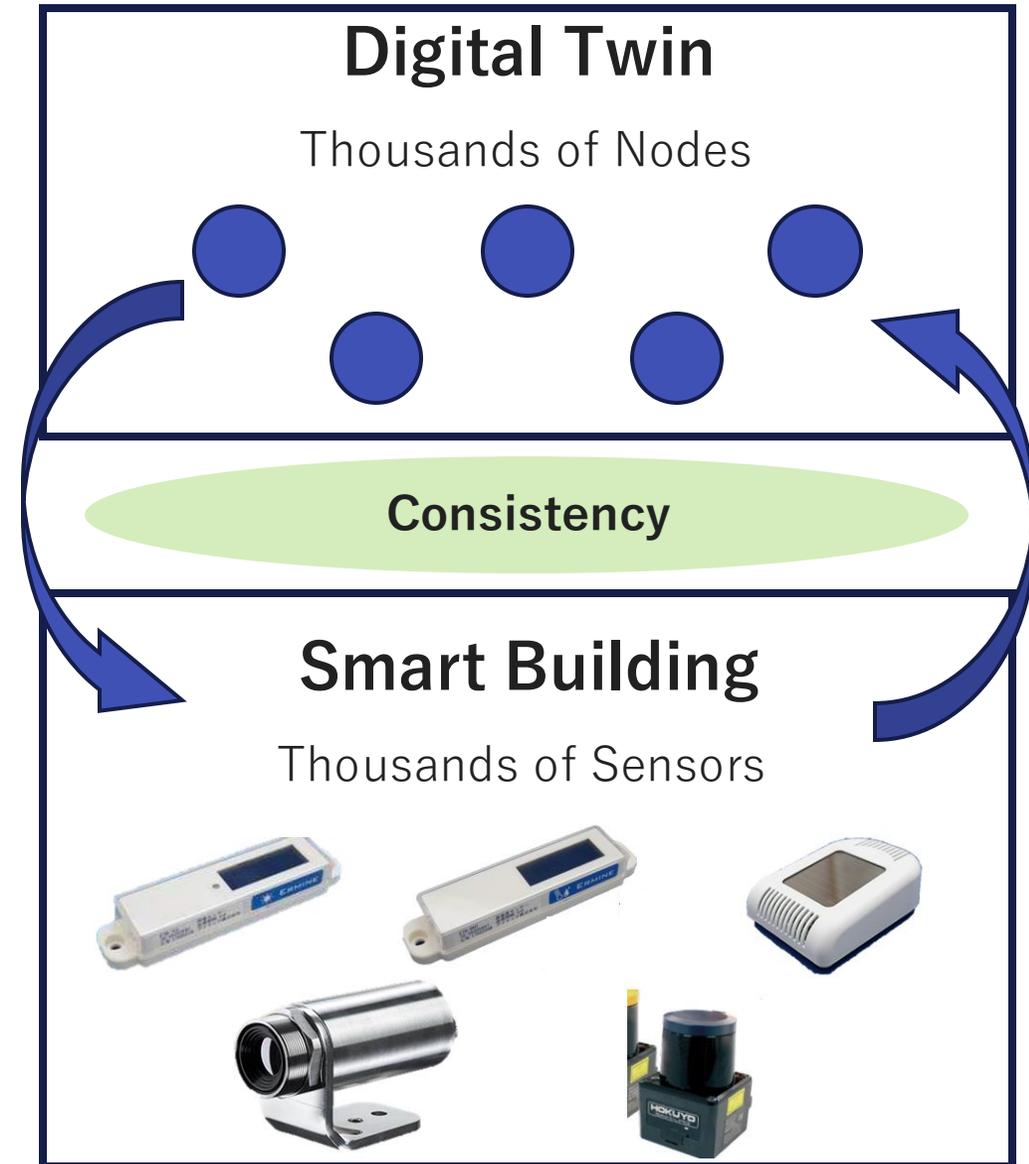
Shimonishi Lab.  
*Osaka Univ.*

## Massive Sensor Deployments

- Hundreds to thousands of IoT sensors in smart buildings.
- Extremely limited maintenance resources.

## Essential for Digital Twins

- **Spatial and operational consistency** of these massive networks.



## Network Aging over Time

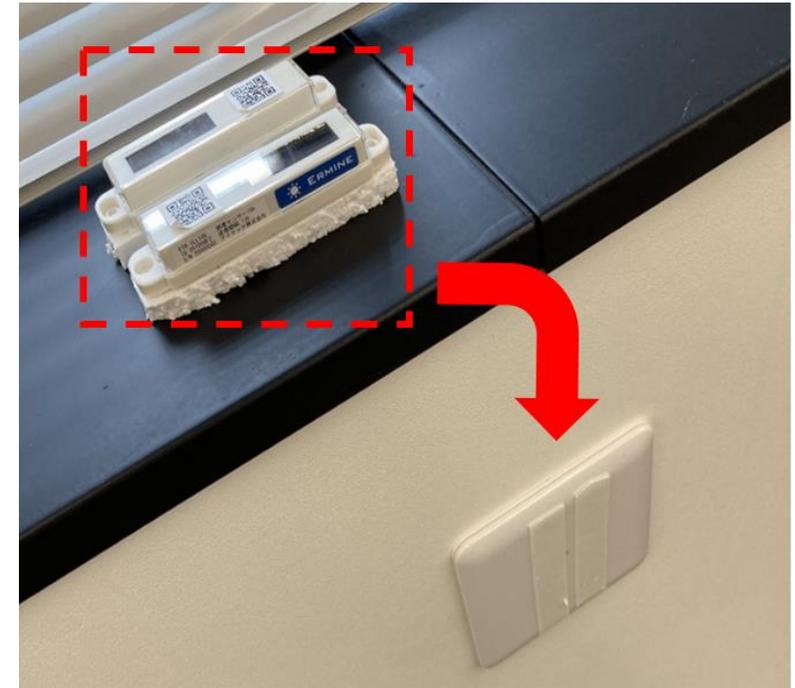
Long-term operation leads to:

- **Sensor degradation**
  - Performance decline
- **Relocation errors**
  - Misplaced or detached due to human/environmental interference



## The Consistency Gap:

A silent mismatch where the Digital Twin assumes a sensor is correct, while it has physically detached in the real world.



e.g., Sensor detached due to environmental interference

## Limitations of Current Approaches

**Time-based (TBM):** Unrealistic cost and time for large-scale deployments.

**Condition-based (CBM):** Misses physical degradation if data appears normal.



## Key Concept

We assign each sensor property a **Confidence Score**:

**Confidence = Probability that the sensor is still correct**

- **Probabilistic Decay Modeling:** Quantifies hidden physical and operational decay as a Confidence Score.
- **Route Optimization:** Minimizes costs by planning paths only for sensors with low scores.

## Workflow of the Proposed Maintenance Process

### 1. Assign Confidence Score:

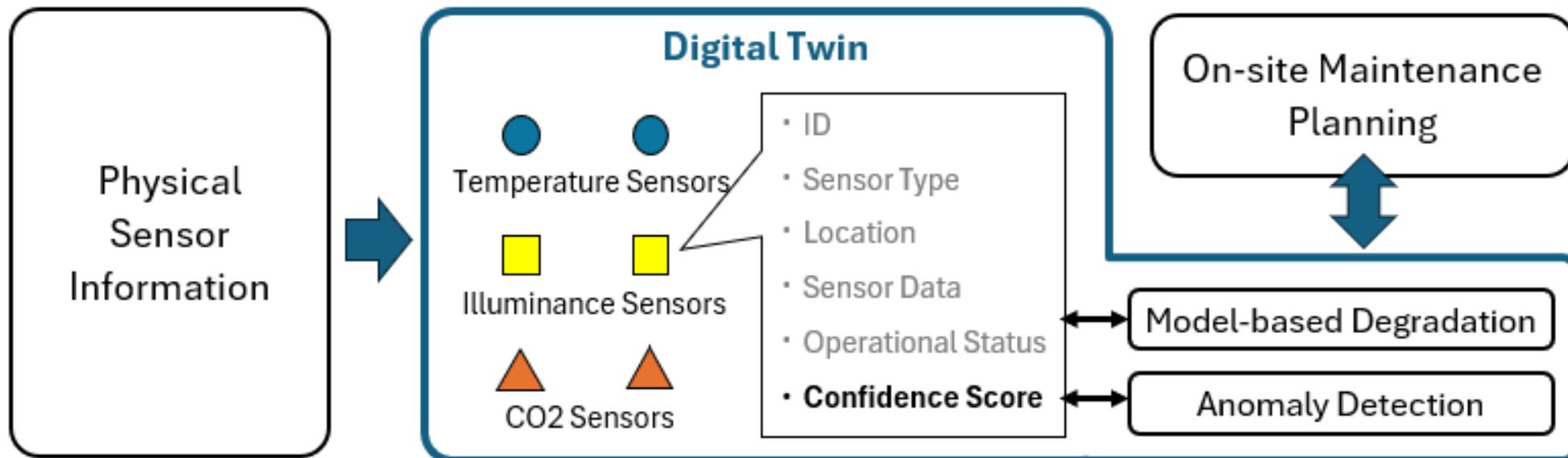
Each sensor maintains a dynamic confidence value.

### 2. Dynamic Update:

Score decreases over time or upon anomaly detection.

### 3. Maintenance Optimization:

Prioritizes low-score sensors to plan cost-effective inspection routes.



## Overall Confidence ( $S^d$ ):

Joint probability that all  $n$  properties are consistent.

(In this study,  $n=2$ : **Placement** ( $s_{loc}^d$ ) & **Operation**( $s_{op}^d$ ) )

$$S^d = \prod_{i=1}^n s_i^d$$

### 1. Placement Consistency ( $s_{loc}$ )

- Models rare, independent misplacement events.
- Modeled as a **Poisson process** (rate  $\lambda$ ).

$$s_{loc}^d = \exp(-\lambda (t - t_0))$$

### 2. Operation Consistency ( $s_{op}$ )

- Models hardware degradation and failure over time.
- Modeled via **Weibull distribution** (scale  $\eta$ , shape  $k$ )

$$s_{op}^d = \exp(-((t - t_0)/\eta)^k)$$

## 1. Target Selection (Thresholding)

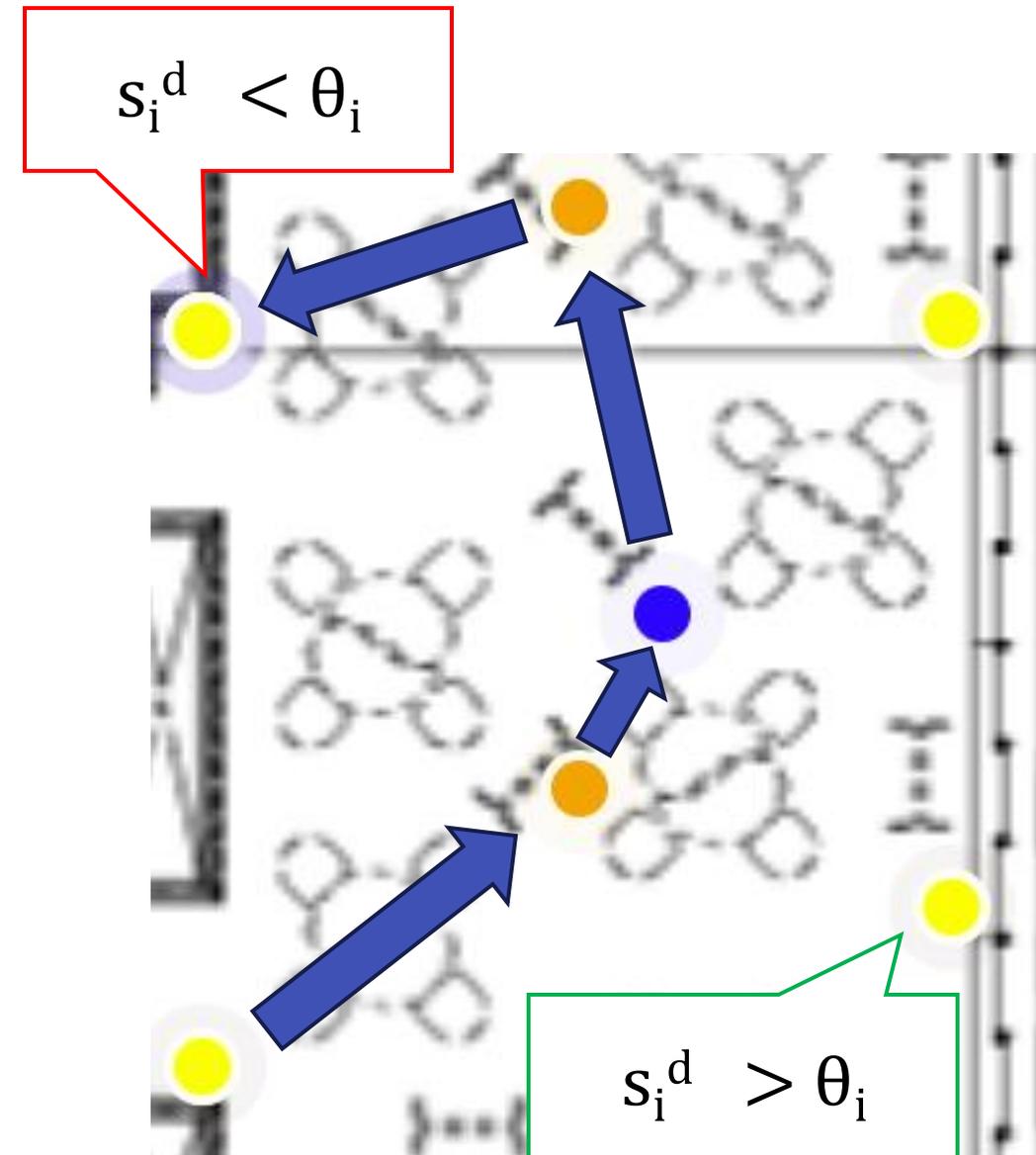
- Select property  $i$  of device  $d$  if  $s_i^d < \theta_i$ .
- Eliminates unnecessary inspections.

## 2. Route Optimization

- Minimizes movement cost between selected devices.
- Approximated using a greedy algorithm.

## 3. Confidence Score Update

- If property is verified/fixed:  
 $s_i^d \leftarrow 1.0$  (Reset  $t_0$  to  $t_{\text{now}}$ ).
- Ensures the digital registry accurately reflects physical reality.



## Target Environment

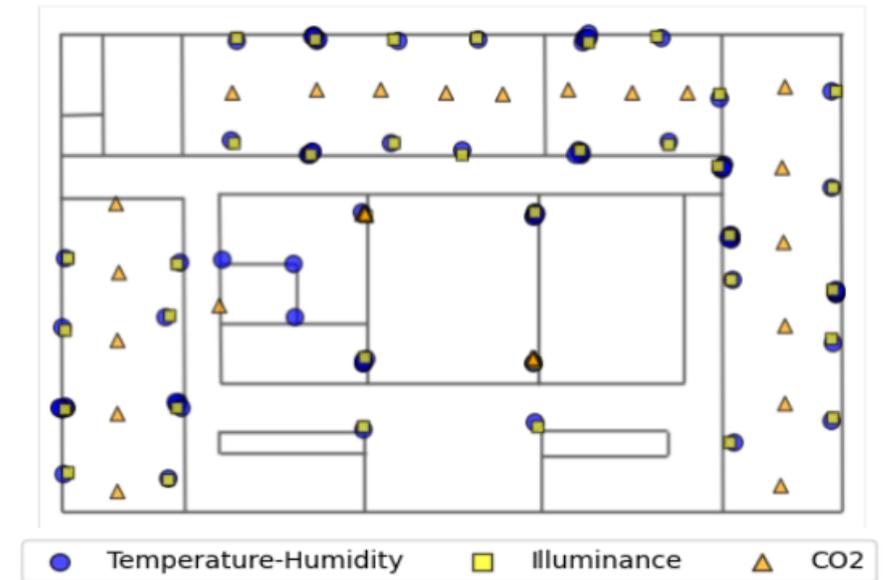
- Smart-campus digital twin (54m × 32.4m floor).
- 147 sensors: Temperature/Humidity, Illuminance, and CO<sub>2</sub>.
- 1-year continuous operation simulation.

## Evaluation Metrics

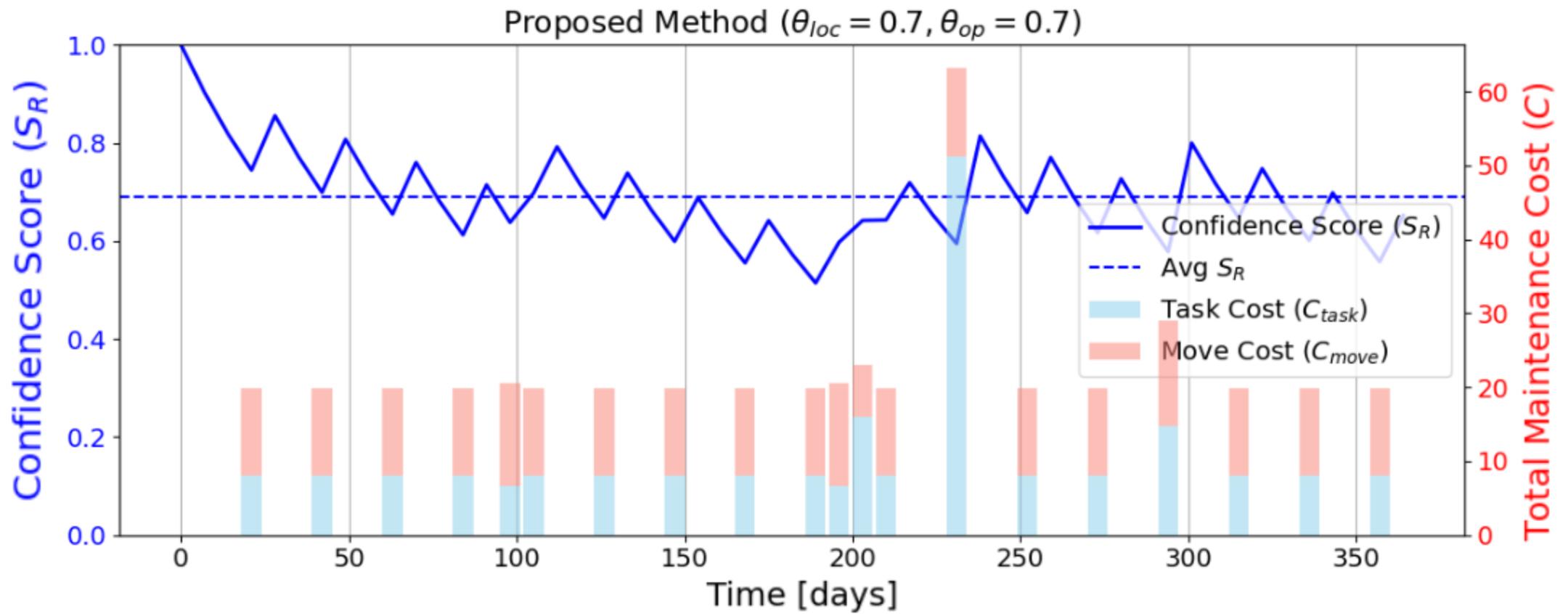
- Average Confidence ( $S_R$ )
- Total Maintenance Cost (USD)
  - Move Cost + Task Cost

## Comparison Baselines (TBM)

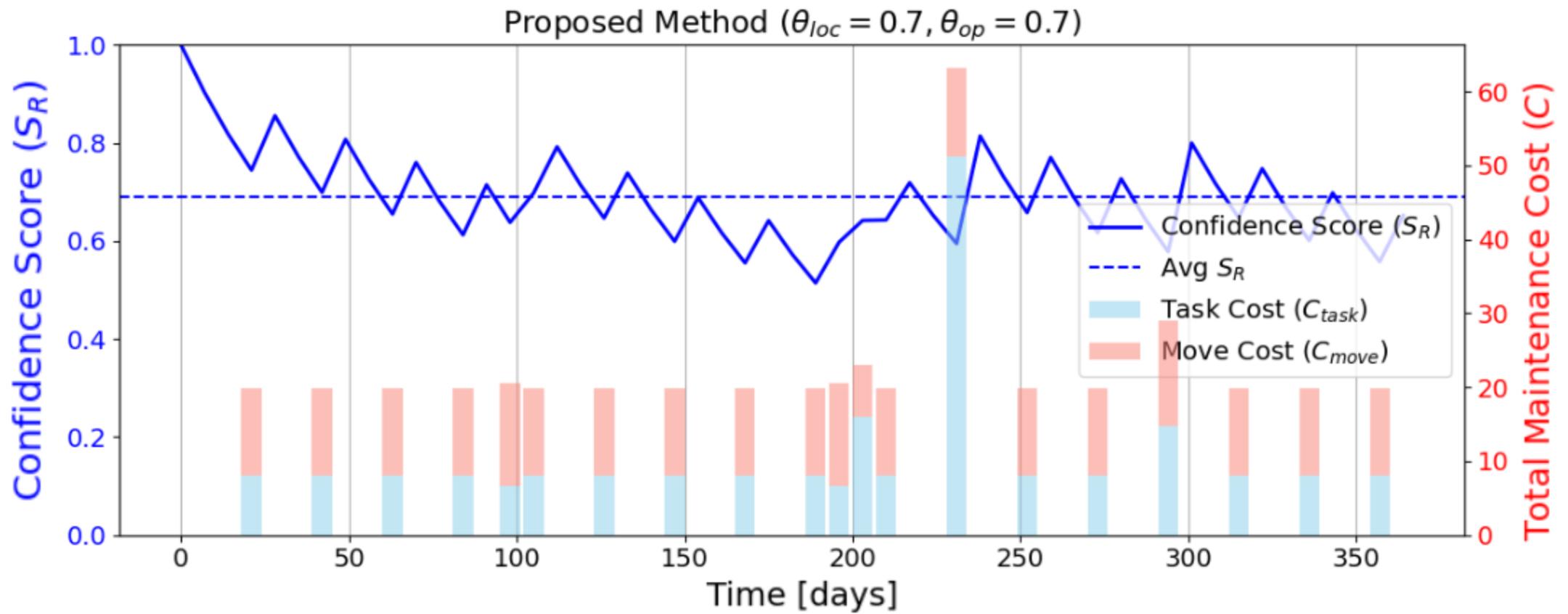
- **Random Maintenance:** Inspects 25% or 50% of devices randomly.
- **Full Maintenance:** Inspects 100% of devices at fixed intervals.



**Consistent Confidence:** Maintains high overall confidence ( $S_R$ ) above the threshold throughout the year.

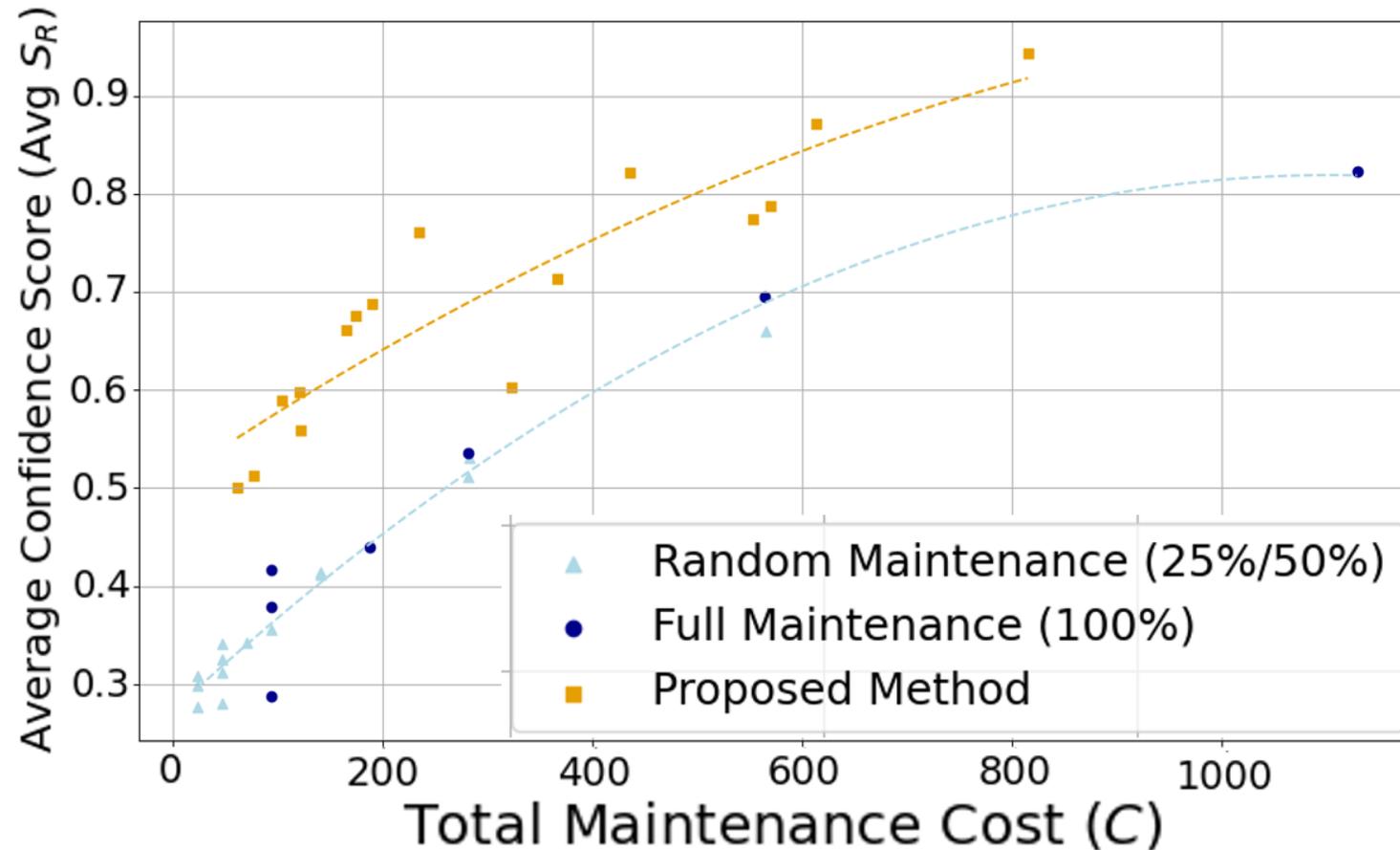


**On-Demand Maintenance:** Task and movement costs are incurred only when confidence drops, preventing unnecessary inspections.



## Higher Efficiency

- Proposed method (orange squares) consistently lies above the baselines.
- Achieves higher average confidence ( $S_R$ ) for similar or lower costs ( $C$ ).

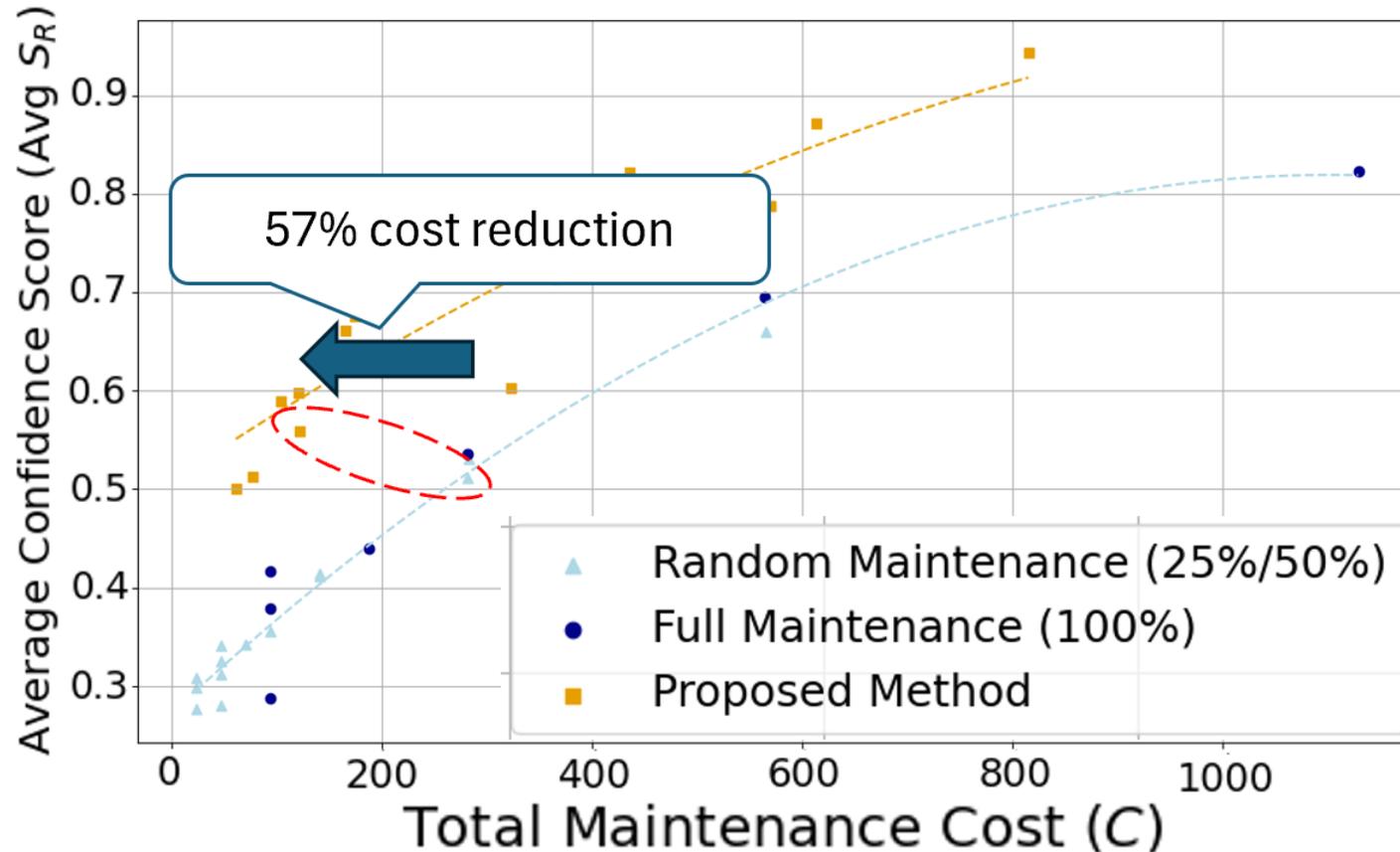


## 1. Significant Cost Reduction (vs. Random)

Proposed: Mean  $S_R = 0.559$ , Cost = 122 USD

Random (50% / 60 days): Mean  $S_R = 0.531$ , Cost = 283 USD

→ Achieved a **57% cost reduction** with a comparable confidence level.

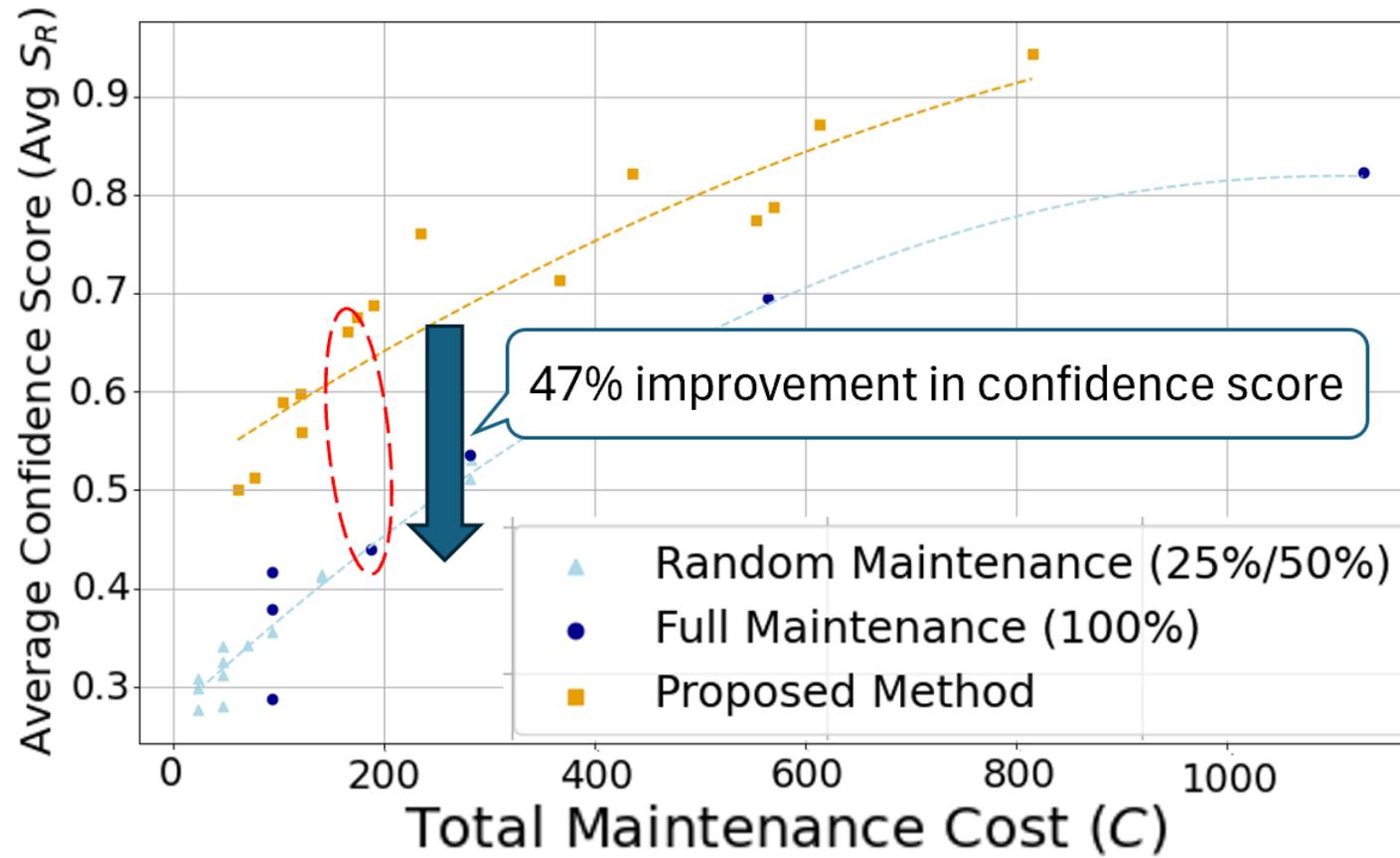


## 2. Significant Confidence Improvement (vs. Full)

Proposed: Mean  $S_R = 0.661$ , Cost = 165 USD

Full (100% / 180 days): Mean  $S_R = 0.440$ , Cost = 188 USD

→ Achieved a **47% confidence improvement** with a maintenance cost.

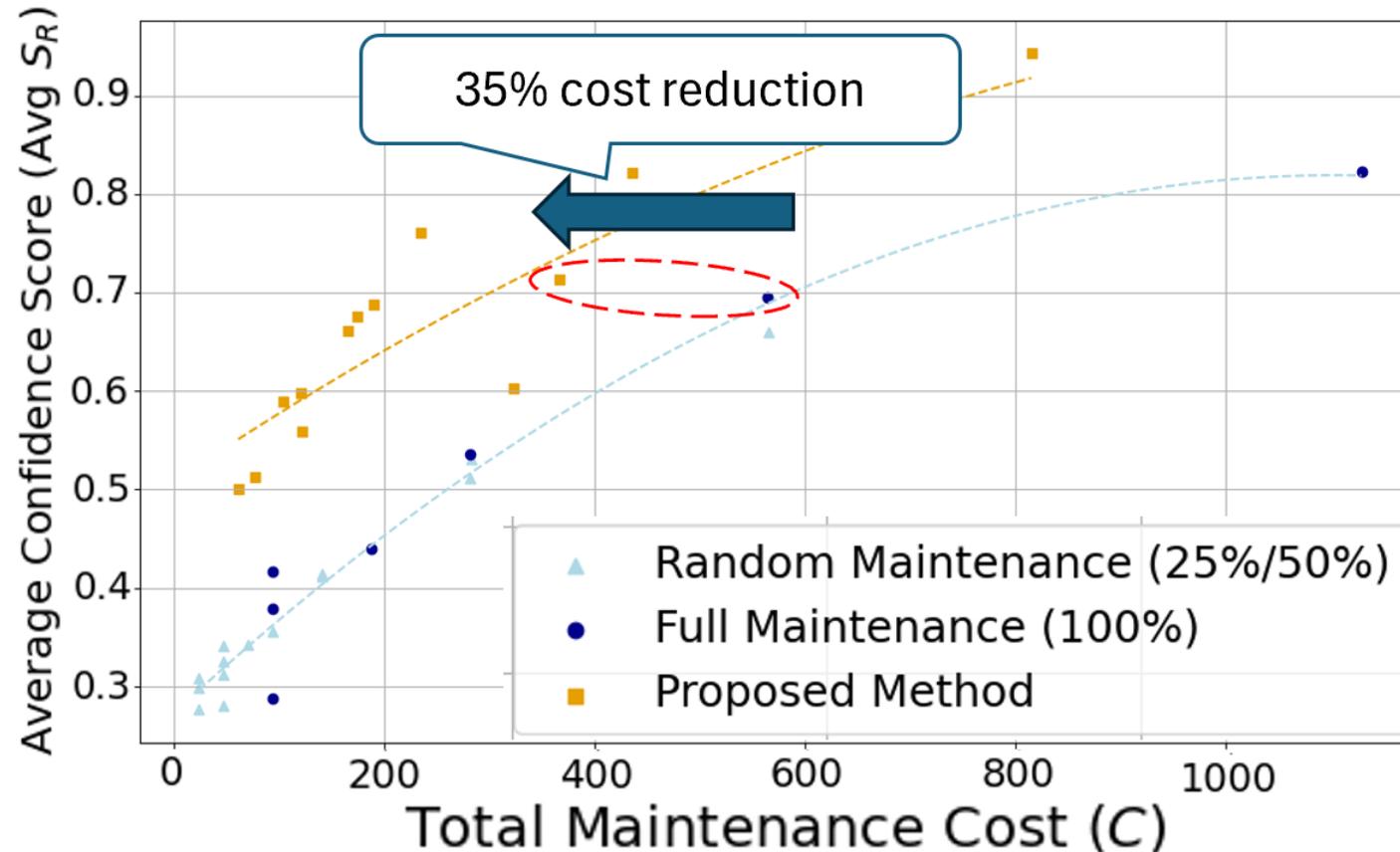


## 3. Significant Cost Reduction (vs. Full)

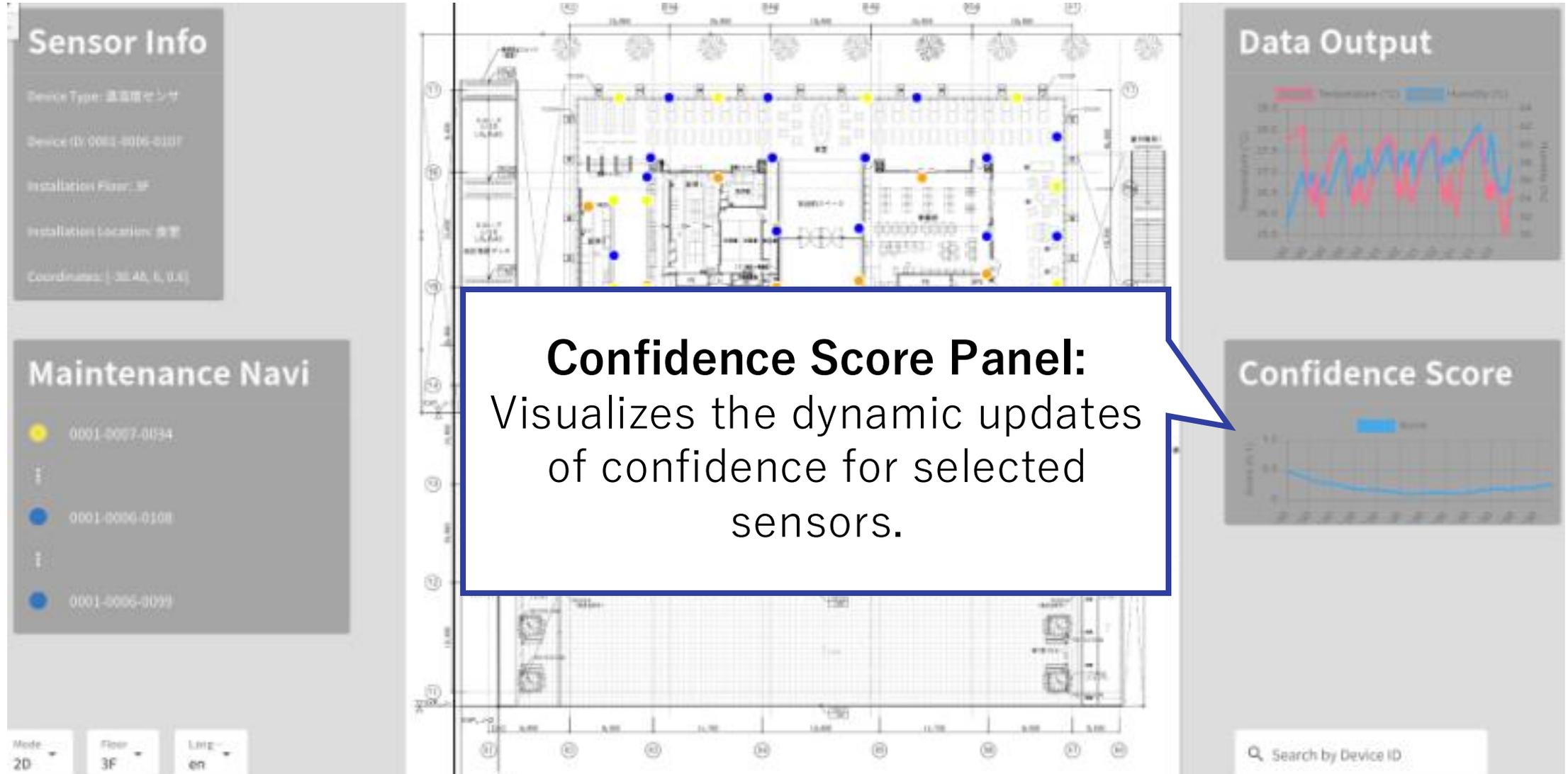
Proposed: Mean  $S_R = 0.714$ , Cost = 367 USD

Full (100% / 60 days): Mean  $S_R = 0.696$ , Cost = 565 USD

→ Achieved a **35% cost reduction** with a comparable confidence level.



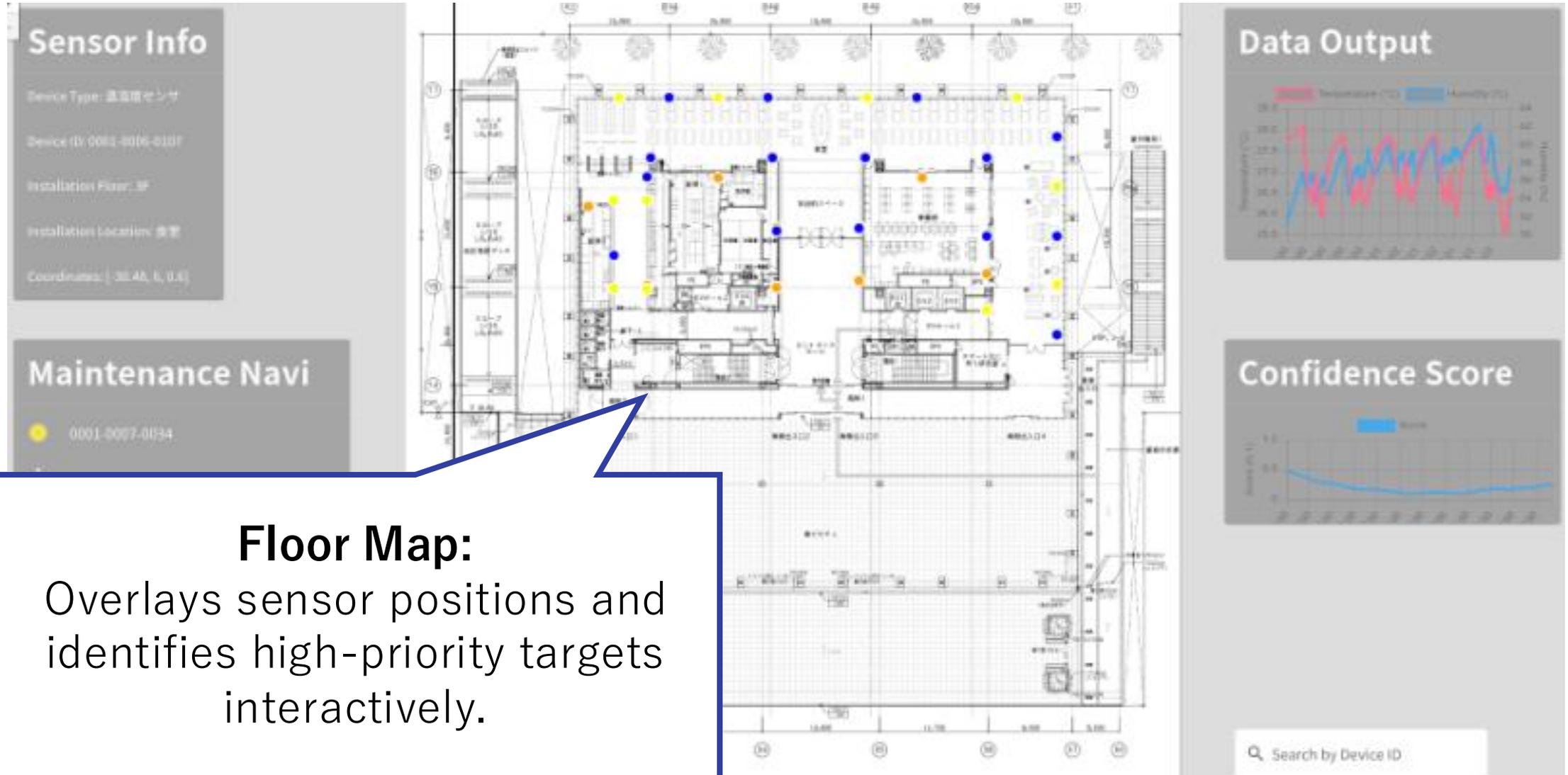




The screenshot displays a software interface for a Digital Twin (DT) application. It features a central 2D floor plan of a building with various colored dots (yellow, blue, orange) representing sensor locations. The interface is divided into several panels:

- Sensor Info:** A panel on the left providing details for a selected sensor:
  - Device Type: 温度検センサ
  - Device ID: 0001-0006-0107
  - Installation Floor: 3F
  - Installation Location: 機室
  - Coordinates: (30.46, 1, 0.6)
- Maintenance Navi:** A panel below Sensor Info showing a list of sensor IDs with corresponding colored dots:
  - 0001-0007-0034 (Yellow)
  - 0001-0006-0108 (Blue)
  - 0001-0006-0099 (Blue)
- Data Output:** A line graph on the right showing two data series: Temperature (°C) in red and Humidity (%) in blue over a 24-hour period.
- Confidence Score:** A line graph below Data Output showing a confidence score over time.
- Search:** A search bar at the bottom right labeled "Search by Device ID".
- Mode/Floor/Lang:** Control buttons at the bottom left for "Mode" (2D), "Floor" (3F), and "Lang" (en).

**Maintenance Navi:**  
Dynamically generates and updates optimized inspection routes.



## Conclusion

- Proposed a confidence-based maintenance framework for DTs.
- Integrated probabilistic modeling and route optimization.
- Achieved up to **57% cost reduction** maintaining high reliability.

## Future Work

- **Extend properties:**  
Incorporate power supply and communication status into the score.
- **Information value:**  
Optimize routing based on uncertainty reduction (Information Gain).