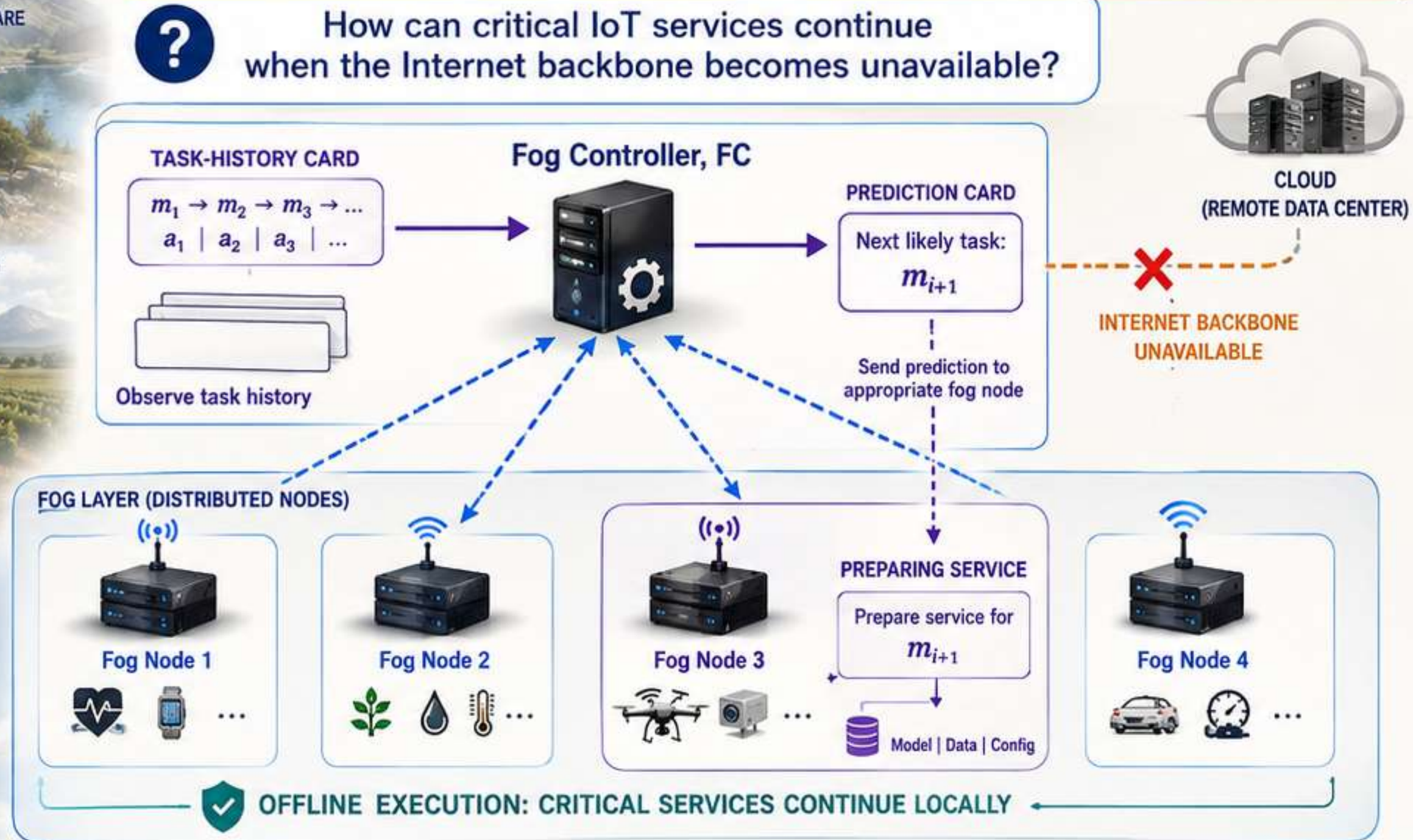


ioFog: Keeping Critical IoT Services Available Offline

Markov-predicted task preparation across a fog-enabled wireless backbone



How can critical IoT services continue when the Internet backbone becomes unavailable?



UAV
MONITORING

CONNECTED
VEHICLE

HIGHLIGHTS OF THE PAPER

01



CONTROLLER-BASED FOG ARCHITECTURE:

A central Fog Controller maintains a global view of fog-node resources and coordinates task allocation.

02



MARKOV TASK PREDICTION:

The FC predicts the next likely task type using each fog node's task history and arrival-time pattern.

03



OFFLINE SERVICE PREPARATION:

Predicted services are prepared at fog nodes so critical tasks can continue during unreliable Internet connectivity.

04



EXPERIMENTAL FINDINGS:

The Markov predictor achieves up to 68% accuracy. ioFog substantially reduces service time and provides a service-delivery ratio close to 99%.



The evaluation compares ioFog with conventional Fog and Cloud architectures.



Predict the next task before the backbone connection disappears.



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When the Internet Fails, the IoT Service Must Continue

Critical decisions cannot wait for a distant cloud

1 CRITICAL SERVICE



Remote healthcare vital signs



Refinery gas alarm



UAV navigation



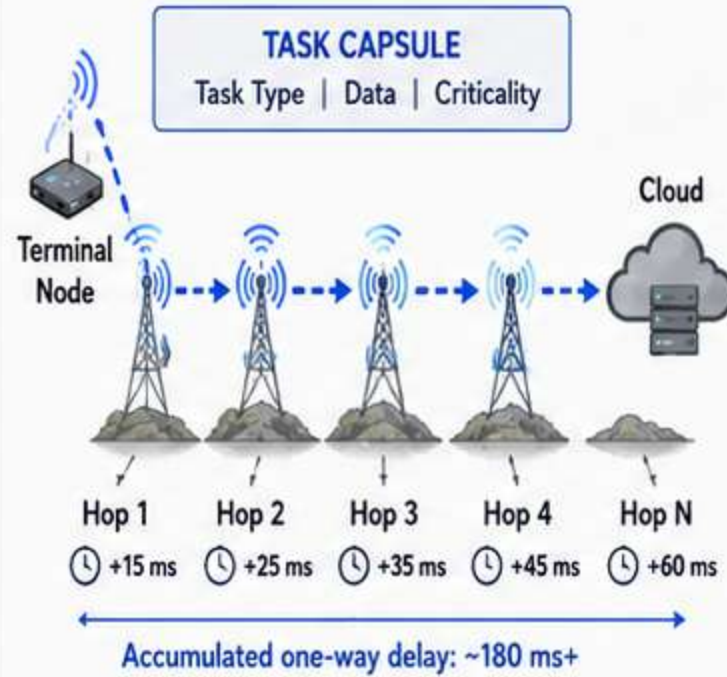
Precision agriculture



Autonomous vehicle

Some IoT decisions are safety-critical and latency-sensitive.

2 LONG-DISTANCE BACKBONE

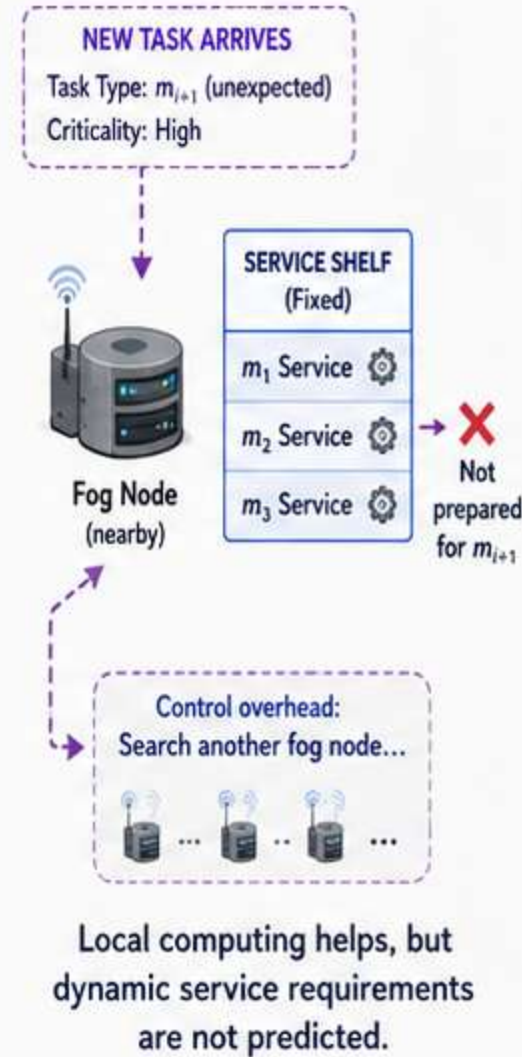


Multi-hop wireless backbones introduce latency and unreliable Internet access.

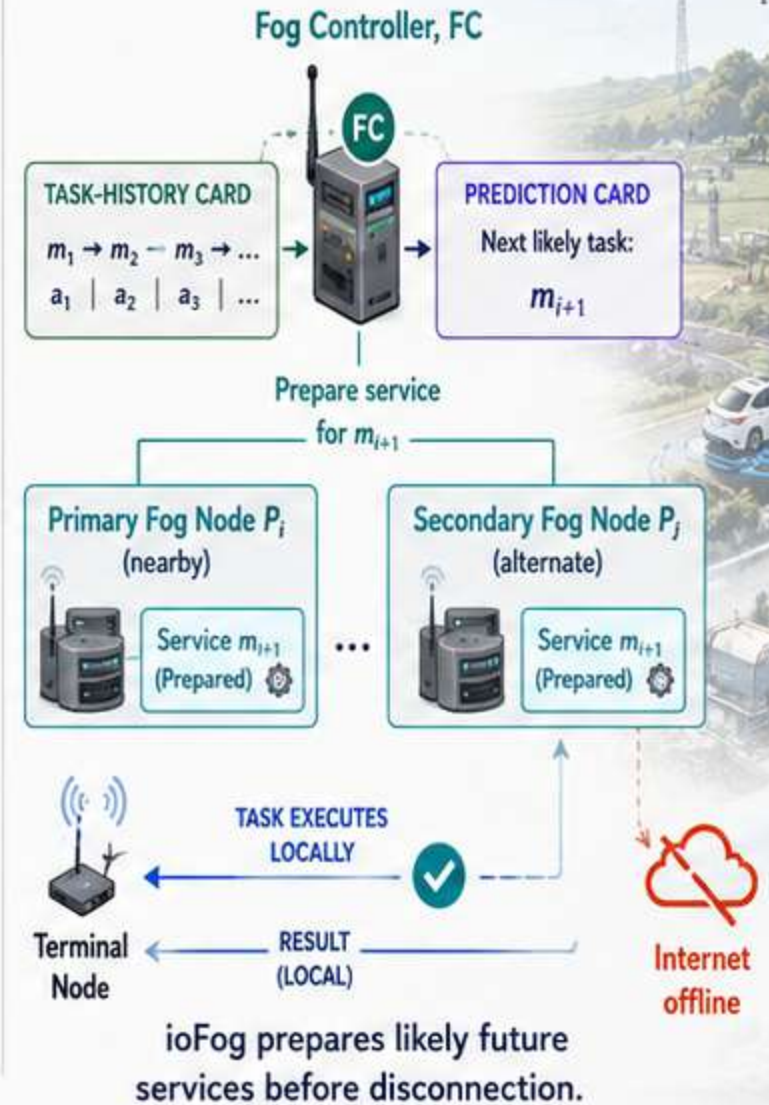
3 CLOUD-DEPENDENT FAILURE



4 CONVENTIONAL FOG LIMITATION



5 IOFOG RESPONSE



RESEARCH GAP PANEL



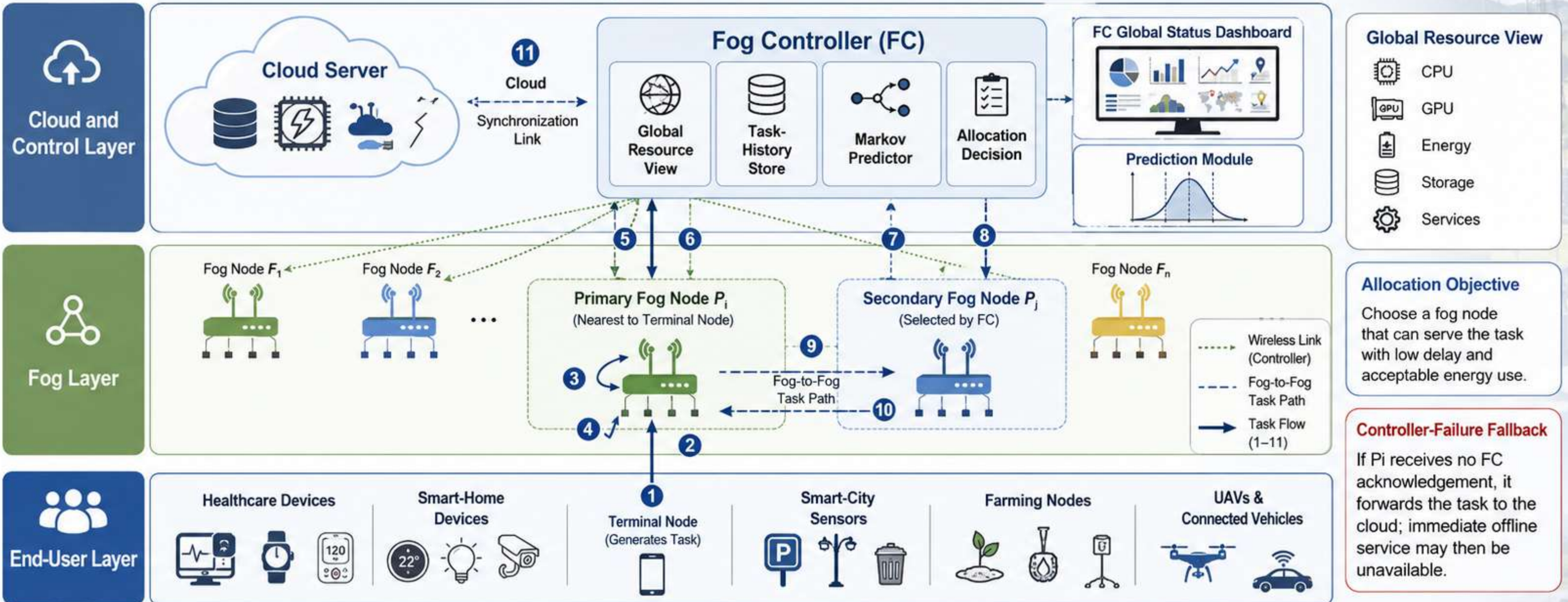
Existing offline-IoT systems provide local connectivity but lack distributed prediction and global fog-resource coordination.



Offline availability requires both local computing and advance knowledge of likely service demand.

Inside the ioFog Architecture

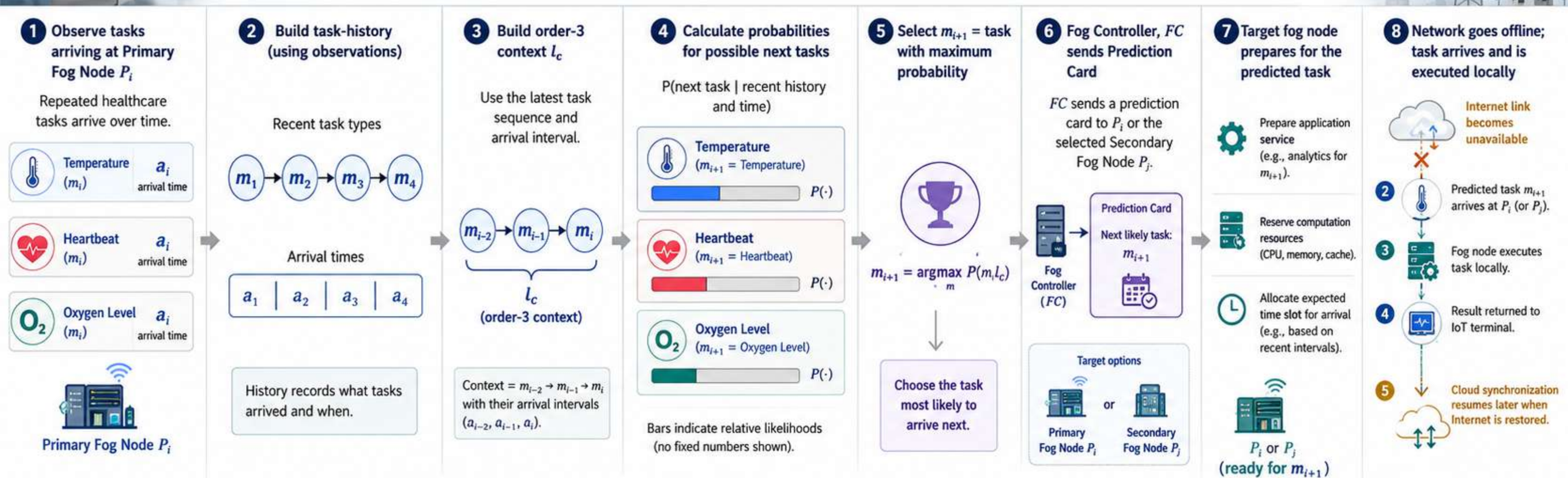
Global fog awareness with local offline execution



FC coordinates globally, while fog nodes execute locally.

Predict the Next Task Before the Network Goes Offline

An order- x Markov model learns task sequences and arrival patterns.



Scientific panel

History $H_i = \text{task types } M + \text{arrival times } A$

M $M = \{m_1, m_2, \dots, m_i\}$ (task types)

A $A = \{a_1, a_2, \dots, a_i\}$ (arrival times)

Output:
predicted immediate future task m_{i+1}



Research annotation

The model predicts **task demand**, not the Internet-disconnection time.

- Task / Observed path
- Successful fog execution
- Prediction / Planning path
- Offline / Disconnected



Recent task history becomes a service-preparation plan for offline operation.

What the ioFog Experiments Show

Prediction accuracy, service time and offline reliability

1 🎯 Task-Prediction Accuracy

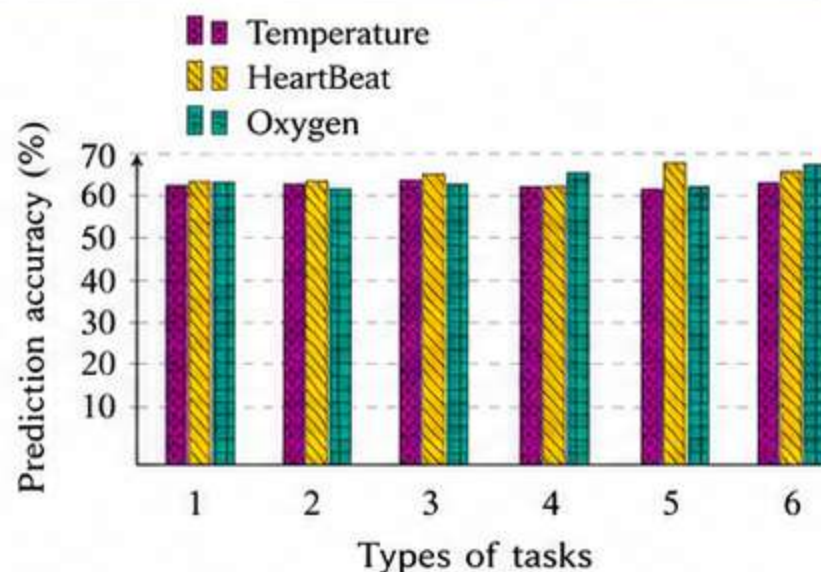


Fig. 4. Prediction Accuracy

✓ Up to 68% prediction accuracy

2 🕒 Average Service Time

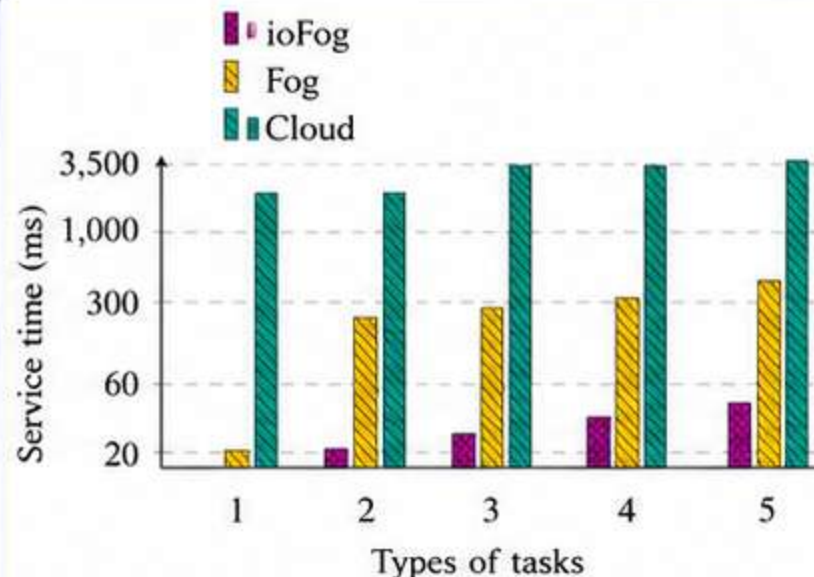


Fig. 5. Average Service Time

✓ ioFog provides the lowest service time

3 🛡️ Offline Service Delivery

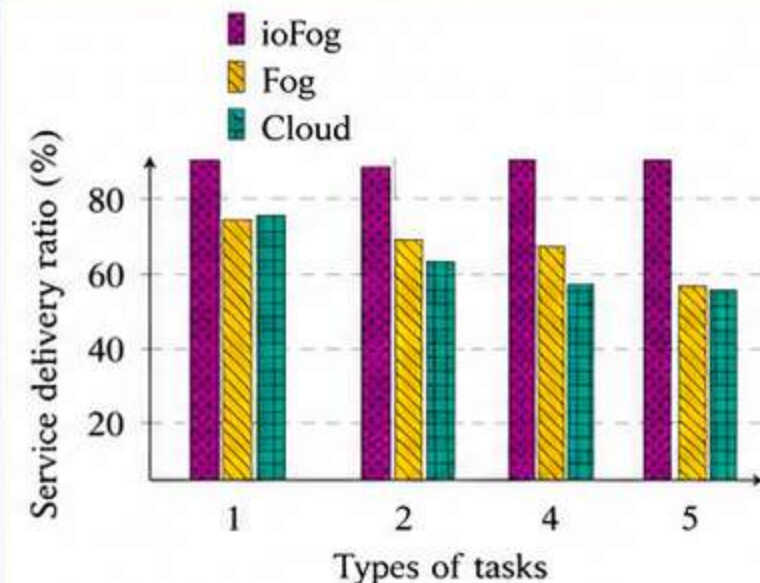


Fig. 6. Average Service Delivery Ratio

✓ Almost 99% service-delivery ratio

Simulation Setup

iFogSim

dynamic service requests

Wi-Fi-based long-distance point-to-point backbone

multiple backbone levels

three healthcare task types

Temperature
Heartbeat
Oxygen Level

Comparison:
ioFog, Fog and Cloud

Conclusion Chain



⚠️ Current Limitations

The centralized FC introduces a single-point-of-failure concern. Prediction accuracy is moderate, and failure to receive an FC acknowledgement may force cloud fallback.

🚀 Future Work

- improve task-prediction accuracy
- use machine-learning methods for prediction and resource optimization
- deploy ioFog on a real IoT testbed

🏆 Research Conclusion

ioFog combines global fog coordination with Markov-based task prediction to maintain critical IoT services during unreliable Internet connectivity.

Observe the demand. Predict the service. Keep the IoT running offline.