



# Application for the Doctoral Contract 2025

## Indoor Localization System for Patient Behavior Analysis in the Context of Connected Healthcare

Ph.D. Candidate:

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**Dr. Huaqiong LI** UCAS

# Outline

## Personal Background

- Academic Background
- Research Experience
- Internship
- Skills

## Research Context & Approach

- Research Context
- Research Objectives
- Challenges & Roadmap

## Summary & Motivation

- PhD Timeline
- Achievements & Suitability
- Strengths & Motivation

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# Academic Background



- **Engineering Diploma in Wireless Communication & IoT Systems** 2022-2025

ISEP – École d'ingénieurs du numérique, Paris, France

- Embedded Systems & Sensors, IoT Networks & Protocols, Neural Networks, 4G/5G/6G

- **Bachelor's Degree in Computer Science** 2018-2022

- *With Honors (Mention Très Bien)*

Université Clermont Auvergne, Clermont-Ferrand, France

- Machine Learning, Path Recognition, Sensor Integration, Signal Processing, Programming

# Research Experience

## 1. Performance Evaluation of Wi-Fi Networks with TDMA Coexisting with CSMA/CA

*Wireless Personal Communications, Springer, 2020*

- Introduced TDMA into Wi-Fi to enhance real-time performance and reliability.
- Results: Reduced delay, improved reliability with minimal impact on CSMA/CA throughput.

## 2. Indoor Localization in IoT Networks Based on Graph Neural Networks

*Article under preparation, 2025*

- Compared DNN vs GNN for Wi-Fi RSSI fingerprint localization.
- Results: GNN achieved lower localization error and shorter runtime, showing higher robustness.

# Graduation Internship



## Energy Optimization in 5G Core Networks – Orange Innovation (24 Mar – 22 Sep 2025)

### ➤ Objective:

- Analyze and optimize the energy consumption of open-source 5G Core (free5GC vs OAI).

### ➤ Method:

- Deployed an energy monitoring framework (Kepler + Prometheus + Grafana on Kubernetes).

### ➤ Key Finding:

- Theoretically, Free5GC consumes more energy than OAI, but observations show the opposite.
- Identified frequent heartbeat signaling as the main cause of OAI's higher energy usage.

### ➤ Result:

1. Extending heartbeat cycle reduced OAI's energy consumption without affecting stability.
2. Integrated findings into Orange's internal AI assistant for future deployments.

# Skills



**Programming:** Python, Matlab



**AI/ML:** DNN, GNN, Computer Vision



**Signal Processing:** RSSI, feature extraction



**IoT Systems:** sensors, data collection



**Communication:** English, French, Chinese

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# Research Context

## 1. Smart hospitals require indoor localization to:

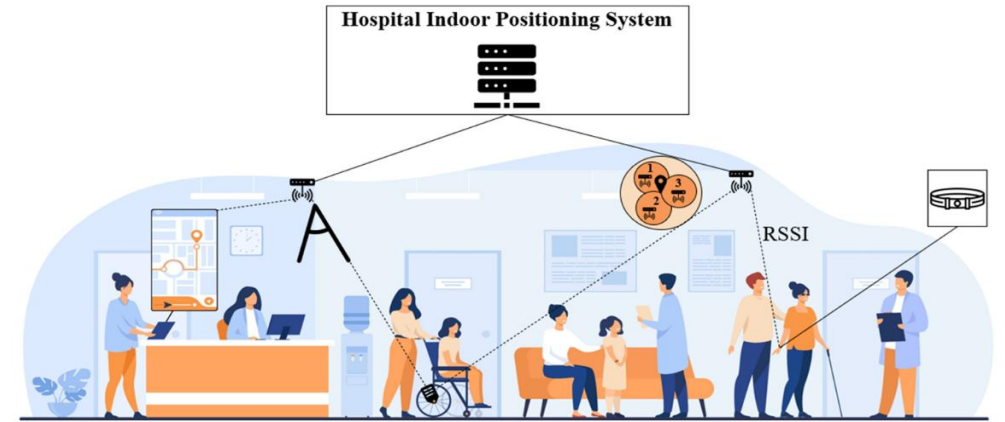
- **Reduce safety risks** : wandering, fall detection
- **Improve operational efficiency** : equipment & staff management
- **Support real-time** patient monitoring
- **Behavior analysis**

## 2. System requirements:

- High accuracy
- Low latency
- Energy efficiency
- Scalability
- Integration with hospital IT systems



# Research Objectives

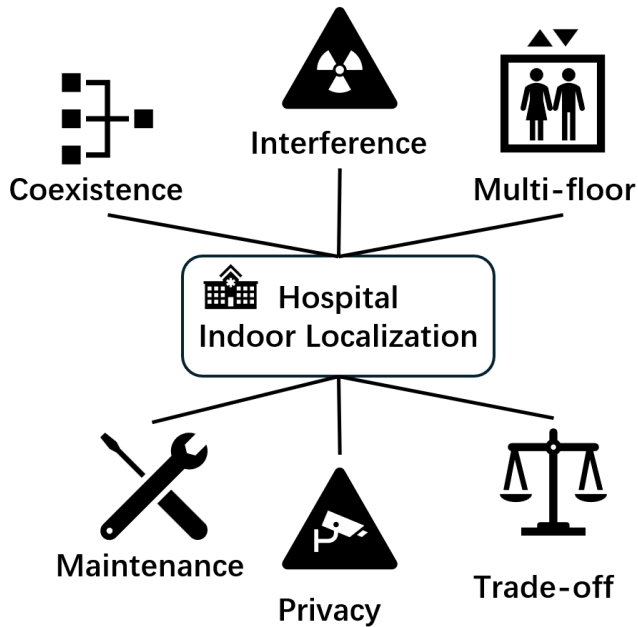


## 1. Objectives:

- Develop a **multimodal localization system** robust in hospital environments
- Enable **real-time tracking** of patients, staff, and assets
- Achieve a balance between **accuracy – latency – energy efficiency**

## 2. Proposed Approach:

- **Data acquisition:** Wi-Fi, BLE, infrared, wearables
- **Multimodal fusion:** from baseline methods to advanced AI/ML
- **Real-time edge inference** with lightweight models
- **Evaluation:** accuracy (2D/3D), floor detection, latency, energy



# Challenges

- **Multi-technology coexistence** : Wi-Fi, BLE, IR, ultrasound, etc.
- **Interference** : medical devices (X-ray, MRI) , moving assets
- **Multi-floor & vertical localization** : stairs, elevators
- **Calibration & maintenance costs** : drift, battery replacement
- **Privacy & IT integration** within hospital infrastructure
- Trade-off between **accuracy – latency – energy efficiency**

# Roadmap

1. **Define research scope** : application scenarios & main goals
2. **Data collection** at Wenzhou Hospital : on-site measurements in representative areas
3. **Sensor & technology selection** : identify best combination
4. **System pipeline** : data capture → synchronization → fusion → feature extraction
5. **Model development** : baselines → advanced multimodal AI/ML
6. **On-site validation** : test accuracy, latency, energy, stability
7. **Optimization & deployment** : improve accuracy & efficiency, enable scalable use

# Outline

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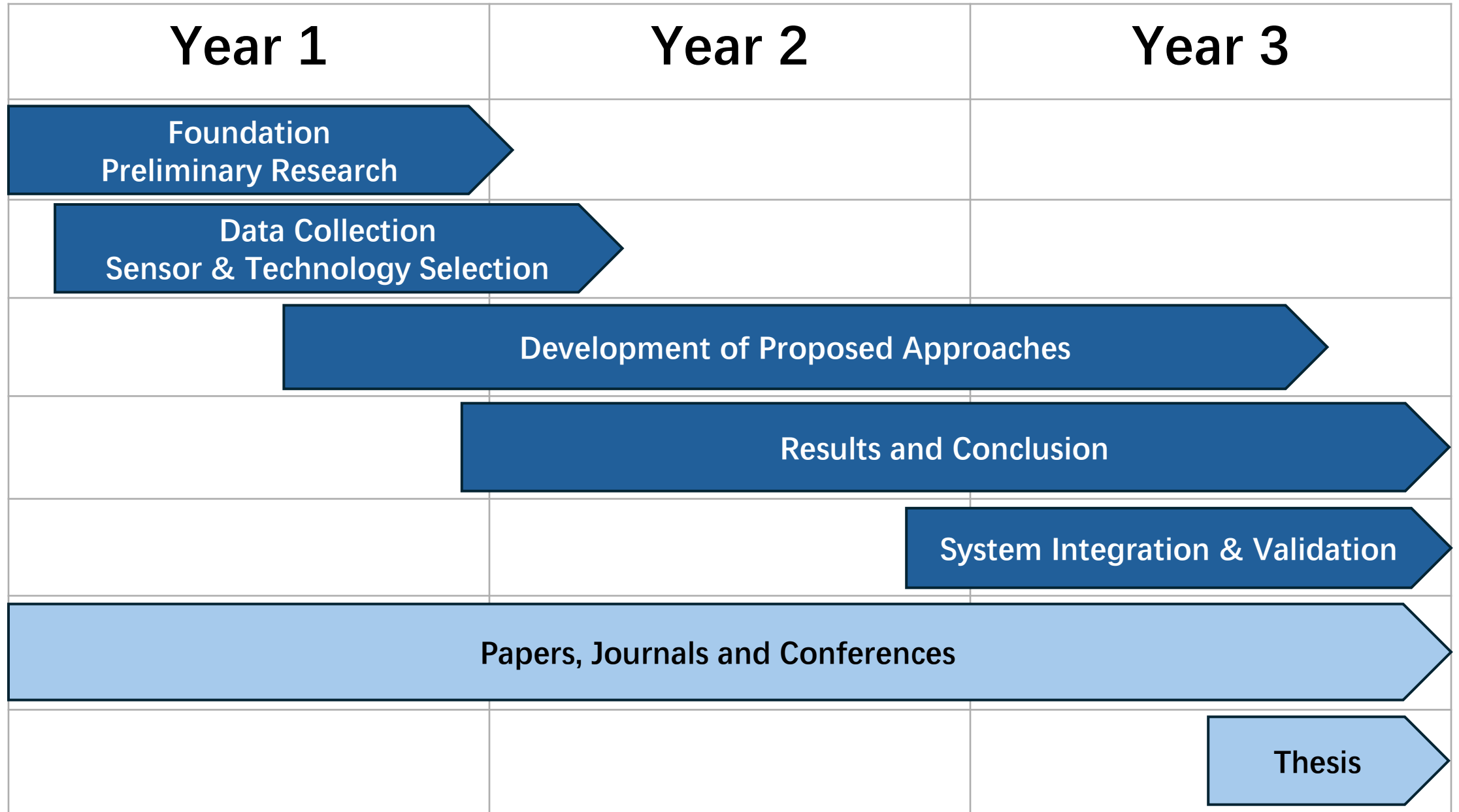
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# PhD Timeline



# Achievements

- **Wi-Fi networks:** Introduced TDMA into CSMA/CA, reduced latency, improved reliability
- **Indoor localization:** Applied GNN for fast and accurate predictions
- **Network energy efficiency:** Optimized 5G Core heartbeat, reduced energy & stable operation
- **IoT & health data:** Collected multimodal signals (temperature, fall detection)

# Suitability

- Expertise in **indoor localization & energy-efficient networking**
- Experience across **data collection, AI modeling, and system deployment**
- Ability to **integrate AI + wireless + IoT** into healthcare applications
- Strong potential to contribute to **hospital-oriented indoor tracking research**

# Strengths

- **Hands-on ability:** from algorithm design to system validation
- **Interdisciplinary profile:** AI, wireless, IoT, healthcare, energy efficiency
- **Communication & collaboration:** familiar with both academic & industrial contexts
- **Multilingual:** Fluent in English, French, Chinese

# Motivation

- Strong interest in wireless communications and the Internet of Things
- Committed to advancing wireless communications, AI, and IoT in healthcare applications
- Intend to integrate experience in sensors, network energy efficiency, and indoor localization
- Passionate about research: motivated to identify problems and find solutions

**Thank you for your attention.**

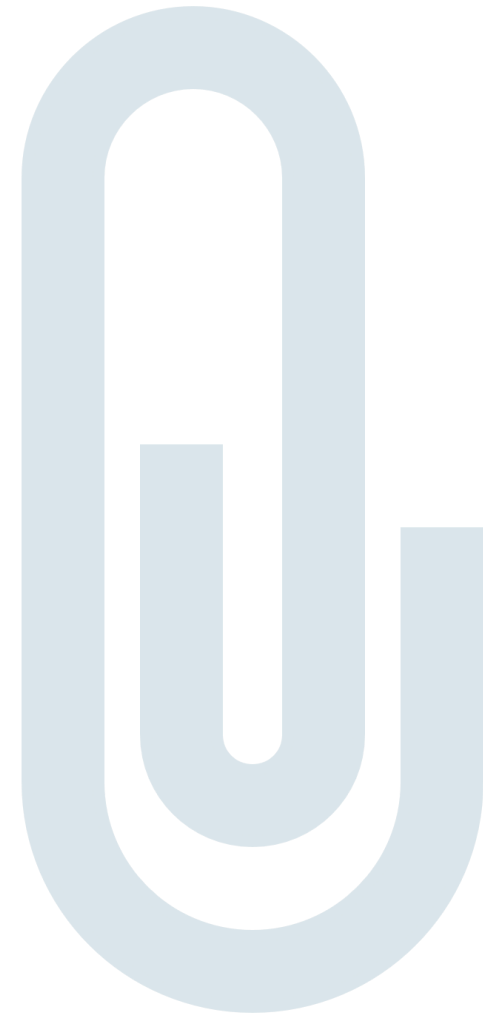
**I would be happy to answer any questions.**





# Appendix

1. Research Experience-Published Article
2. Research Experience-Article Under Preparation
3. Graduation Internship
4. Methods & Contributions
5. Future Goals



# Research Experience-Published Article

## Performance Evaluation of the Networks with Wi-Fi based TDMA Coexisting with CSMA/CA

*Wireless Personal Communications, Springer, 2020*

➤ **Objective:**

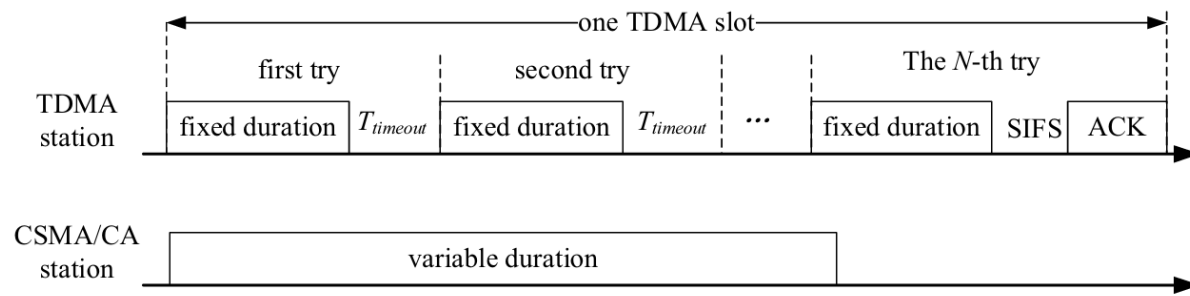
- Introduce TDMA into Wi-Fi to improve real-time performance and reliability when coexisting with CSMA/CA.

➤ **Method:**

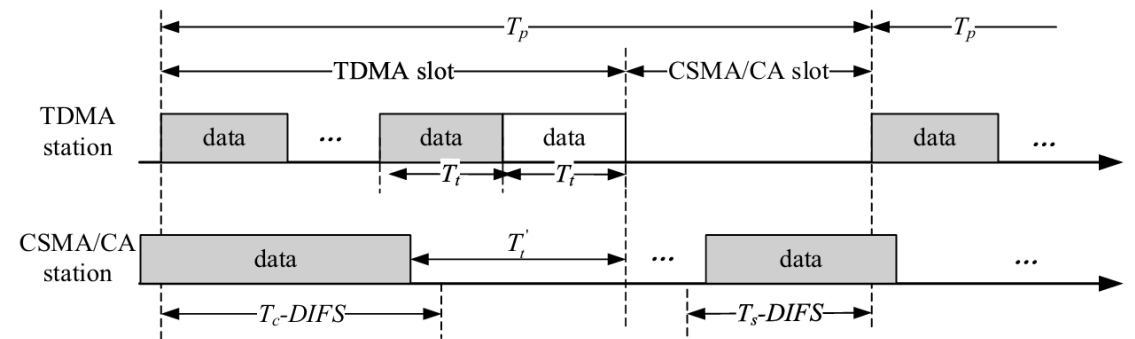
1. **Canceling Carrier Sensing:** TDMA devices transmit directly in allocated slots without checking channel availability.
2. **Priority Mechanism:** TDMA devices gain priority over CSMA/CA devices on the shared channel.
3. **Collision Handling:** TDMA devices immediately retransmit data until an ACK is received, ensuring reliability.

➤ **Result:**

- The TDMA mechanism significantly reduced delay for real-time applications and improved reliability through retransmissions.
- As frame length and duty cycle increased, CSMA/CA throughput slightly decreased but the impact remained manageable.



**Fig. 2** Preemptive TDMA scheme



**Fig. 4** The interference a CSMA/CA frame suffered from the Wi-Fi TDMA station

# Research Experience-Article Under Preparation

## *Indoor Localization in IoT Networks Based on Graph Neural Networks*

2025

### ➤ Objective:

- Compare the performance of DNN and GNN in Wi-Fi RSSI fingerprint-based localization, Including predictions for longitude, latitude, floor and predicted time.

### ➤ Method:

1. **DNN Encoder-Decoder:** Built an encoder-decoder baseline model with multi-task regression to jointly predict.
2. **GNN with k-NN Graphs:** Constructed graphs using k-NN and applied GCN layers with message passing to capture spatial topology.
3. **Evaluation Metrics:** valuated models by comparing hidden layer configurations, localization error, and runtime performance.

### ➤ Result:

- GNN achieved lower localization error and shorter runtime with fewer hidden layers, improving accuracy and robustness.

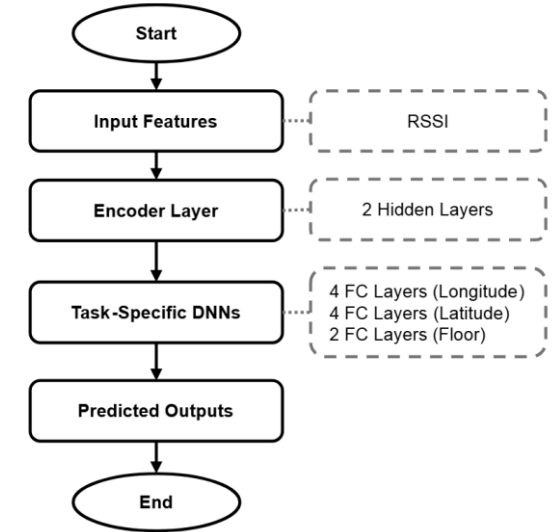


Fig. 1. Overall data preprocessing and prediction workflow of the DNN-based localization model. (FC: Fully Connected).

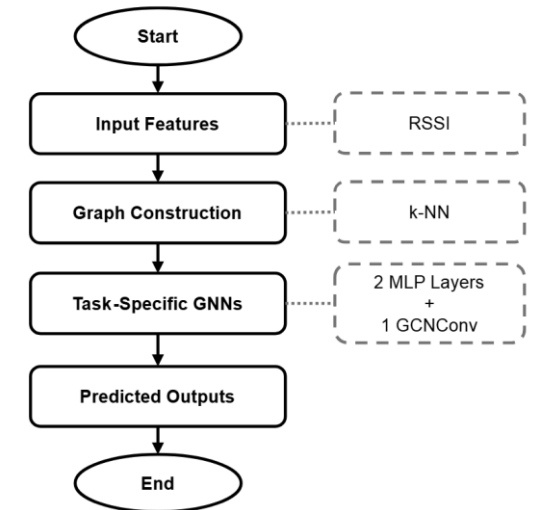


Fig. 2. Overall data preprocessing and prediction workflow of the GNN-based localization model. (MLP: Multi-Layer Perceptron, GCNConv: Graph Convolution layer).

# Graduation Internship (24 Mar – 22 Sep 2025)

## Energy Optimization in 5G Core Networks – Orange Innovation



### ➤ Objective:

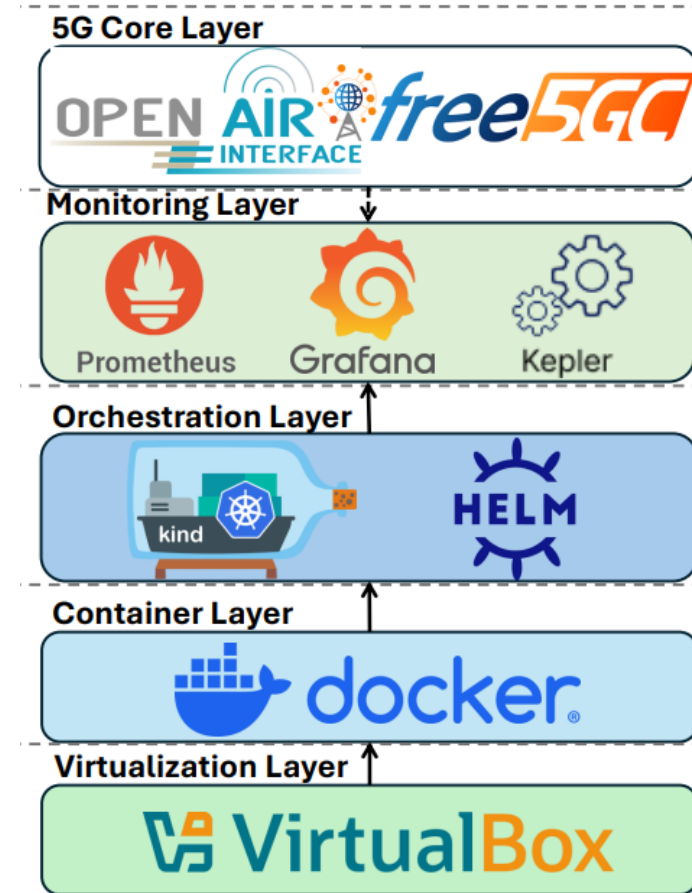
- Analyze the energy consumption of open-source 5G Core (free5GC and OAI) and explore methods to reduce it.

### ➤ Method:

1. **Deployed** an energy monitoring framework (Kepler + Prometheus + Grafana on Kubernetes).
2. **Installed and compared** free5GC and OAI; experiments revealed OAI consumed significantly more energy.
3. **Proposed three hypotheses:** database implementation differences, residual container processes, and frequent inter-Pod signaling.
4. Identified **frequent AMF heartbeat** signaling as the main cause, **modified** the source code, and extended the cycle from 10s to 100s.

### ➤ Result:

1. Extending the heartbeat cycle significantly reduced OAI's energy consumption while maintaining system stability.
2. The achievements were integrated into an AI assistant at Orange to support rapid future deployment.



# Methods & Contributions

Multi-floor

## *Multi-floor · Fixed Localization*

### Possible Approaches

- Multimodal input: combine Wi-Fi, BLE, UWB.
- Add a floor classification module in DNN.

### Possible Challenges

- Complex data synchronization.
- Strong floor-to-floor signal variation.

## *Multi-floor · Mobile Localization*

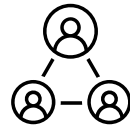
### Possible Approaches

- Fuse Wi-Fi, BLE, UWB, and sensor data.
- Use GNN combined with temporal models for cross-floor trajectory modeling.

### Possible Challenges

- Cross-floor attenuation & dynamic interference.
- High computational and scalability requirements.

Fixed



Mobile



## *Single-floor · Fixed Localization*

### Possible Approaches

- Use Wi-Fi or BLE RSSI fingerprinting.
- DNN regression model for (x, y) prediction.

### Possible Challenges

- Signal interference from walls and devices.

## *Single-floor · Mobile Localization*

### Possible Approaches

- Use BLE/UWB with IMU sensors.
- GNN (k-NN graph) for trajectory modeling.

### Possible Challenges

- Dynamic interference sources.
- High real-time computational cost.

Single-floor

# Future Goals

## Short-term (3 years)

Develop a multimodal AI indoor localization system with high accuracy, low energy cost, and scalability in smart hospitals.

## Mid-term (5 years)

Promote the deployment and standardization of AI-based indoor localization and behavior analysis in healthcare.

## Long-term (10 years)

Explore the integration of 6G networks and smart healthcare.

