

# 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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# Outline

## Personal

- Academic Background
- Research Experience
- Graduation Internship

## Research

- Research Context
- Research Objectives
- Challenges
- Roadmap

## Summary

- PhD Timeline
- Key Strengths
- Research Fit

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# Academic Background



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

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

- IoT & Wireless Networking, AI/ML for communications, Wi-Fi Systems, Embedded Sensors

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

- *With Honors (Mention Très Bien)*

Université Clermont Auvergne, Clermont-Ferrand, France

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

# Research Experience

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

**Xiaofan Guo** et al. *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

**Xiaofan Guo**, Wafa Njima. *Article under preparation*, 2025

- Comparative study : Graph Neural Networks (GNNs) vs Deep Neural Networks (DNNs) for fingerprinting-based indoor localization.
- Results: GNN achieved lower localization error and faster runtime, showing greater 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 cores (free5GC vs OAI).

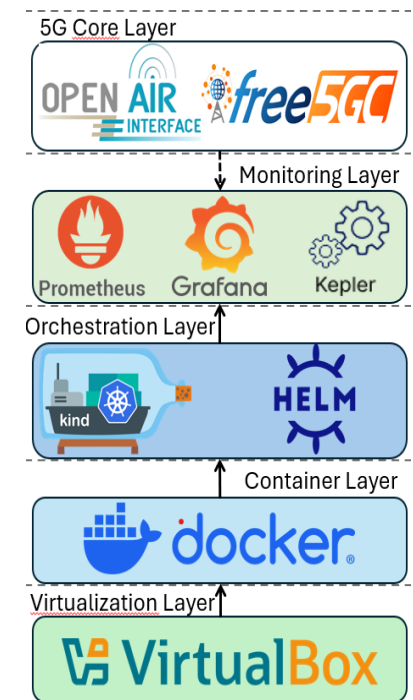
### Key Findings

- Contrary to theoretical expectations, OAI consumed more energy than free5GC.
- Frequent **heartbeat signaling** identified as the main cause of high energy consumption.

### Results

- Extending heartbeat **reduced OAI's energy consumption** without affecting stability.
- Findings were integrated into Orange's AI assistant to guide future deployments.

Figure1 Testbed Architecture for 5GC



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# Research Context

## Hospital Needs

- Accurate Indoor Localization
- Behavior Analysis
- Operational Efficiency
- Patient Safety & Quality of Care

## Limitations Current Methods

- Low accuracy in indoor positioning
- Limited behavior analysis
- Lack of integration



Figure2: Indoor Tracking of Patients in a Hospital



# Research Objective

## Objectives

Enable real-time localization, monitoring, and behavior analysis

## Solution

Develop a multimodal indoor localization system for hospitals with integrated behavior analysis

## Hight Constraint

Reliable behavior analysis; Achieve a balance between accuracy, latency, and energy efficiency

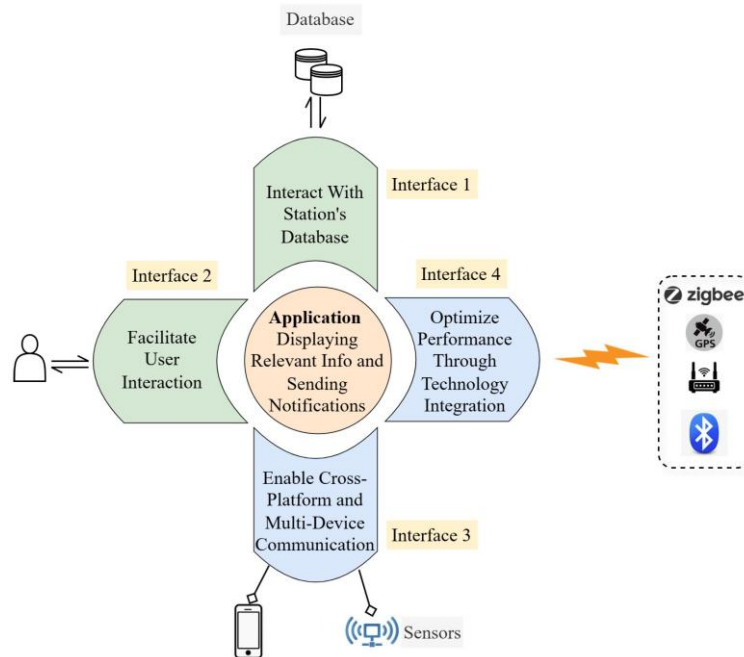


Figure3: System architecture of BLE-based indoor localization

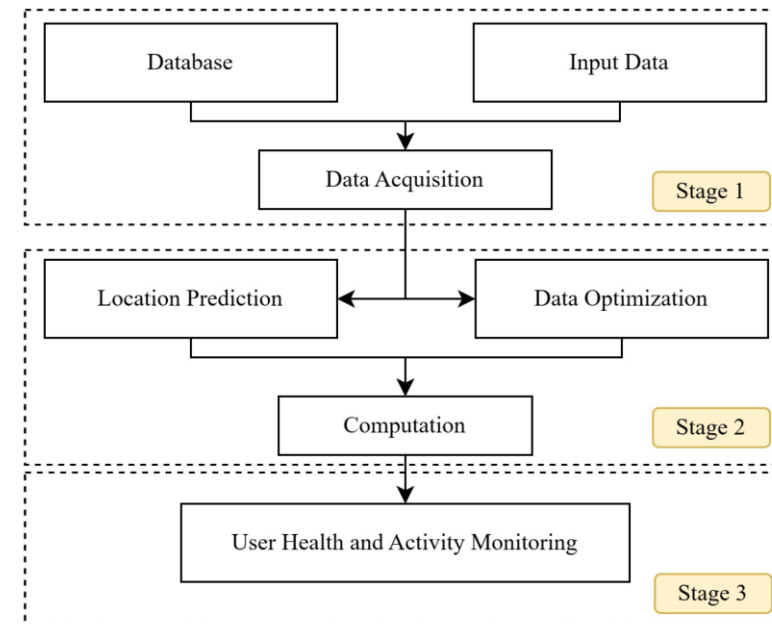


Figure4: Tracking workflow for patients and staff in healthcare settings

# Challenges

## Technical Challenges

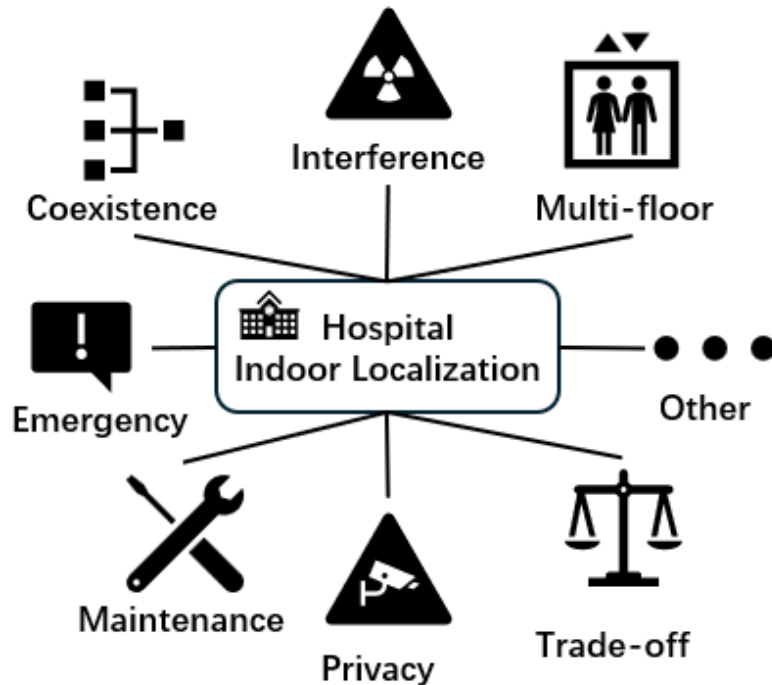
- Multi-technology coexistence
- Dynamic and complex hospital environments
- Multi-floor & 3D localization
- Accuracy–latency–energy trade-off
- Reliable behavior recognition across heterogeneous sensor data

## Operational Challenges

- Frequent calibration and database updates
- Scalability & robustness in large hospitals
- Seamless integration with existing IT systems

## Security & Safety Challenges

- Protecting patient privacy
- Ensuring reliability in emergencies
- Ethical management of behavior-related data



# Proposed Roadmap

**Step 1 – Define Scenarios, metrics and sensors**

**Step 2 – Collect Data**

**Data Fusion and IT Integration**

**Step 3 – Transfer / Modeling**

**AI and ML advanced methods**

**Step 4 –Data Analysis**

**Evaluation and Insights**

**Step 5 – Validation and Deployment**

**Build an intelligent healthcare indoor localization system with behavior analysis**

# Outline

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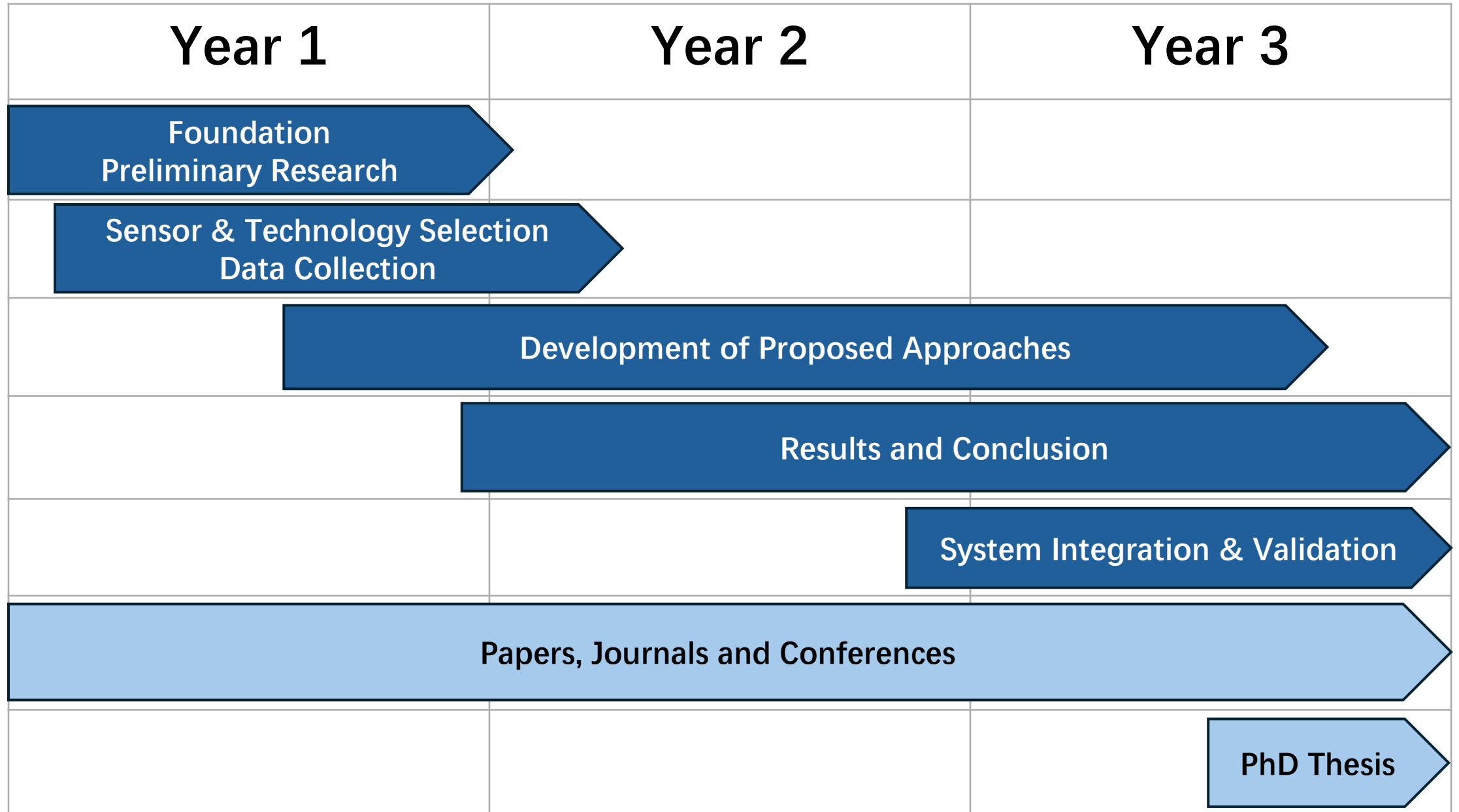
## Research

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- Research Objectives
- Challenges
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- Key Strengths
- Research Fit

# PhD Timeline



# Key Strengths

<b>Wireless &amp; IoT</b>	· Wi-Fi, 5G Core; Indoor Localization, Energy-efficient Systems
<b>AI &amp; Data</b>	· DNN / GNN; Data modelling; Matlab-Signal processing; Python
<b>Hardware &amp; Sensors</b>	· IoT devices; Health Data (temperature, fall detection)
<b>Research &amp; Innovation</b>	· Publications; Problem-solving; Experimentation Skills
<b>Collaboration</b>	· Multilingual (EN/FR/中文); Academic & Industrial Teamwork
<b>Soft Skills</b>	· Analysis; Taste for Innovation, Research and Challenges

# Research Fit



## Orange Internship

- Reduce 5GC energy



## Final Year Project

- Indoor localization



## Wearable Sensors

- Health monitoring system



## Multilingual

- EN, FR, 中

Energy-Efficiency

Localization

Healthcare

Collaboration

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

Behavior Analysis

Integrated

Incentive



## Undergraduate Projects

- Path & image recognition



## Interdisciplinary

- AI, Wireless, IoT, Healthcare, Energy Efficiency, Signal



## Motivation

- Strong interest in Wireless & IoT and Research

**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 (Time Division Multiple Access) into Wi-Fi to improve real-time performance and reliability when coexisting with CSMA/CA (Carrier Sense Multiple Access with Collision Avoidance)

### ➤ 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.

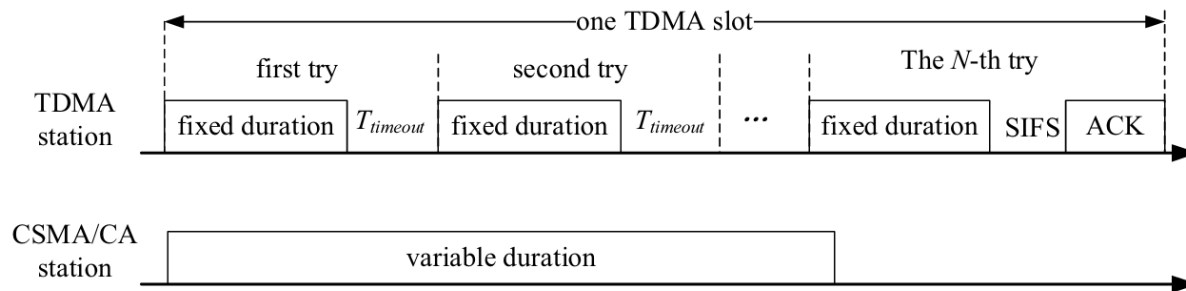


Fig. 2 Preemptive TDMA scheme

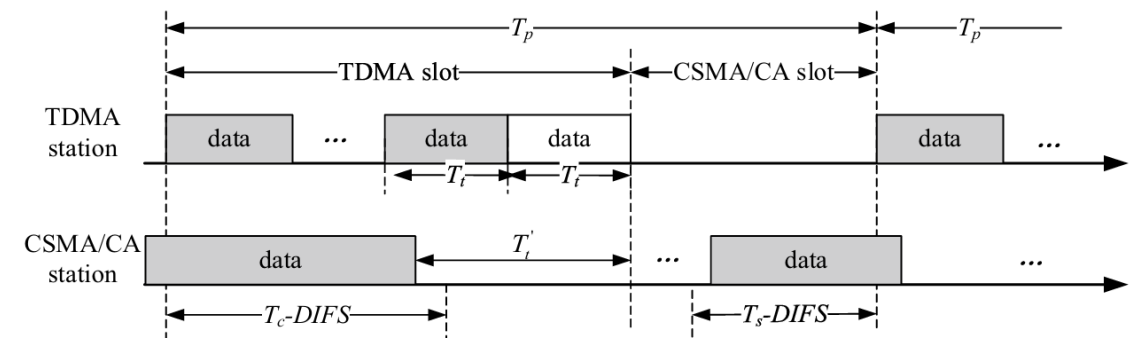


Fig. 4 The interference a CSMA/CA frame suffered from the Wi-Fi TDMA station

# 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.

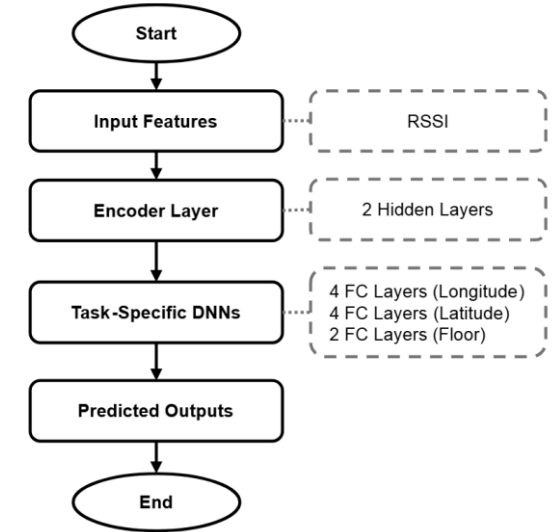


Fig. 1. Overall data preprocessing and prediction workflow of the DNN-based localization model. (FC: Fully Connected).

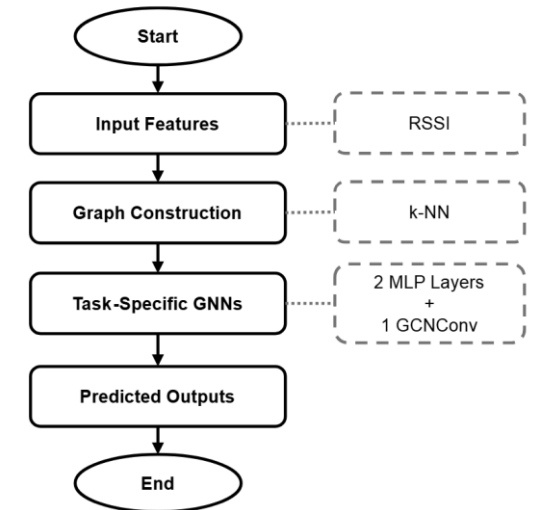


Fig. 2. Overall data preprocessing and prediction workflow of the GNN-based localization model. (MLP: Multi-Layer Perceptron, GCNConv: Graph Convolution layer).

# Graduation Internship (24 Mar – 22 Sep 2025)

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



### ➤ Objective:

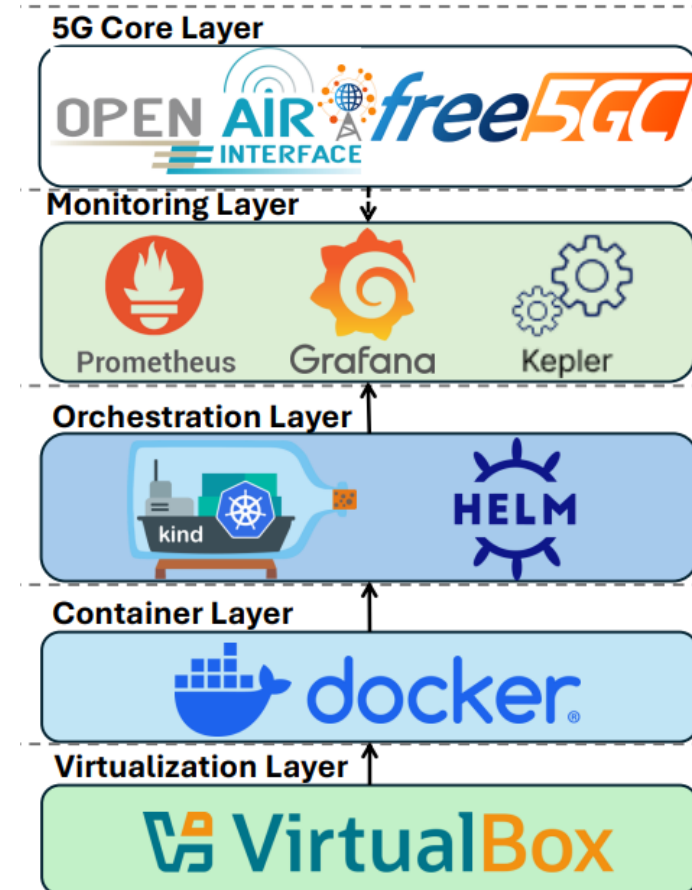
- Analyze the energy consumption of open-source 5G Core (free5GC and OAI) and explore methods to reduce it.

### ➤ 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

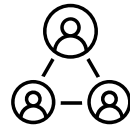
- Fuse Wi-Fi, BLE, UWB, and sensor data.
- Use GNN combined with temporal models for cross-floor trajectory modeling.

### 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.

