

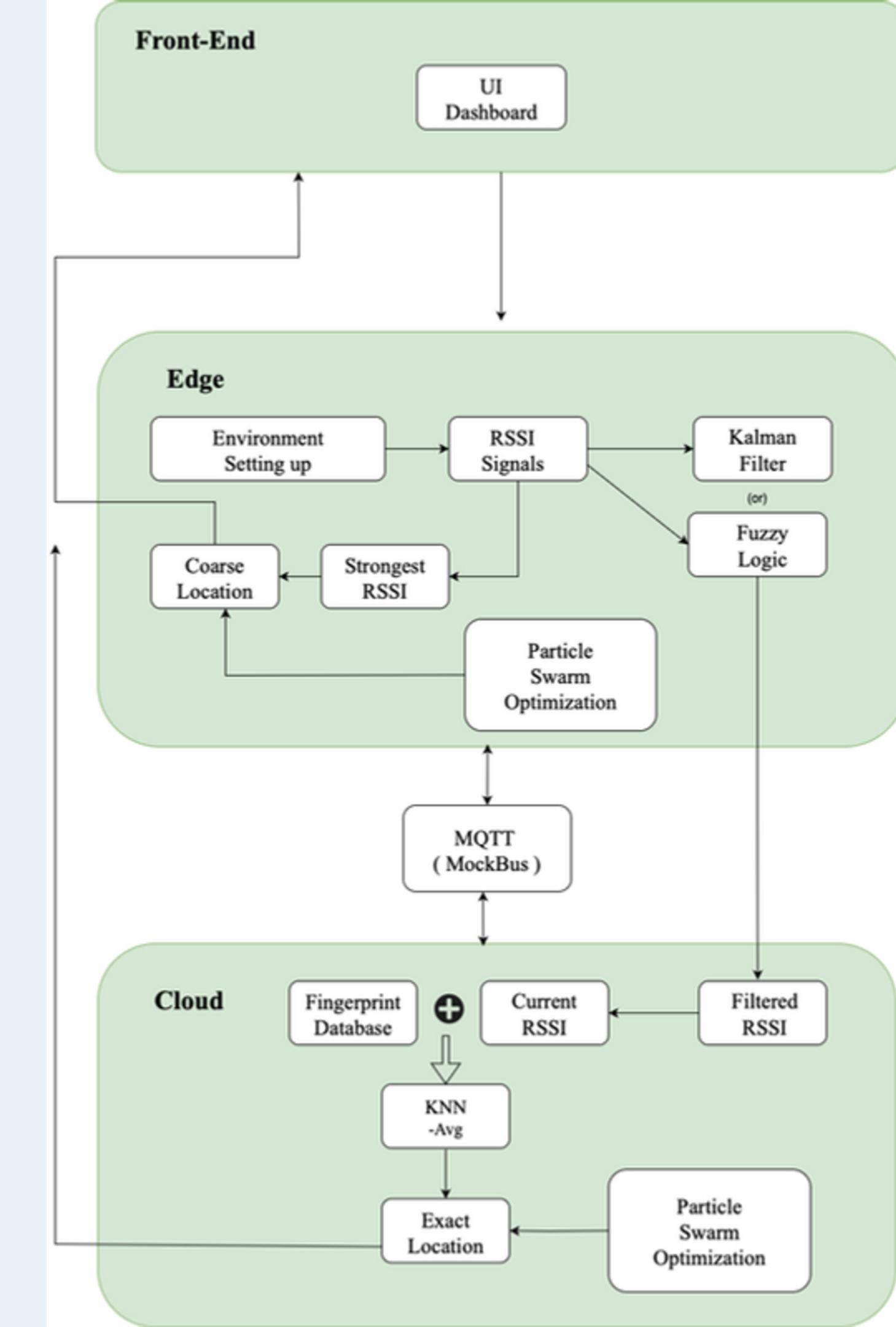
SIMULATED EDGE-IOT SYSTEM FOR INDUSTRIAL RESOURCE TRACKING

- Zhiheng Zhao, Peng Lin,
Leidi Shen, Mengdi Zhang

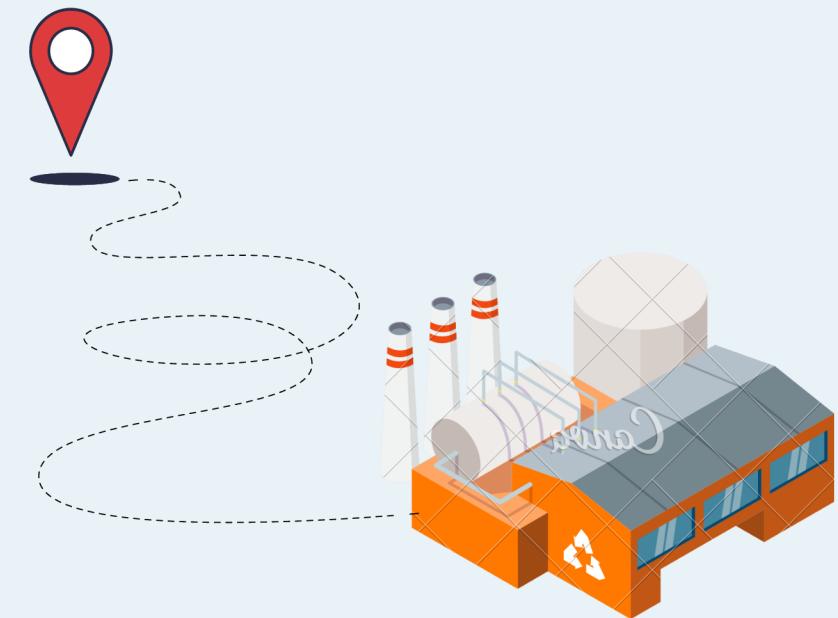
Group No : 3

PROBLEM STATEMENT

Large manufacturing floors hold thousands of items, making it hard for workers to quickly find materials. Edge computing processes data locally to give instant coarse location, ensuring reliability for production. This reduces dependency on cloud, saving computation and storage while speeding up search efficiency.



Environment & Signal Simulation



- Platform: Matplotlib, NumPy, SciPy
- Create 2D grid layout with gateways & WIFI tags.
- Implement RSSI generation (log-normal path loss + Gaussian noise).
- Integrate Kalman filtering (filterpy).
- Deliver a function that, given tag positions, returns filtered RSSI values.

Processing Flow

Implementation(using simpy):

- **Edge Processing (Coarse Location):** RSSI from gateways via log-distance model with noise; strongest gateway chosen.
- **Cloud Processing (Refined Location):** Built fingerprinting dataset; applied k-NN on tag RSSI; refined location computed by weighted averaging of neighbors.
- Integrated with Dataset(<https://github.com/jinyi-yoon/CampusRSSI/tree/main>)
- Selected Environment - Medium-Obs (Office)

```
./debugpy/launcher 54002 -- /Users/kanishkadhanasekar/Downloads/capstone-project-03_cloud9/src/proce  
ssing/material_locator.py
```

```
Enter Material IDs separated by commas : mat115,mat009
```

```
Simulating requests for: MAT115, MAT009
```

```
=====
```

```
[0.00s] NEW REQUEST: MAT115
```

```
=====
```

```
[0.00s] Starting coarse location for MAT115...
```

```
=====
```

```
[0.00s] NEW REQUEST: MAT009
```

```
=====
```

```
[0.00s] Starting coarse location for MAT009...
```

```
=====
```

```
[0.50s] Material ID: MAT115
```

```
RSSI values:
```

```
AP0: -69.76 dBm
```

```
AP1: -70.88 dBm
```

```
AP2: -75.40 dBm
```

```
AP3: -67.20 dBm
```

```
AP4: -68.80 dBm
```

```
AP5: -67.88 dBm
```

```
AP6: -65.56 dBm
```

```
AP7: -60.56 dBm
```

```
AP8: -68.40 dBm
```

```
AP9: -59.08 dBm
```

```
AP10: -70.68 dBm
```

```
AP11: -63.28 dBm
```

```
AP12: -54.76 dBm
```

```
AP13: -63.76 dBm
```

```
AP14: -65.28 dBm
```

```
AP15: -52.48 dBm
```

```
Max RSSI: -52.48 dBm
```

```
Coarse Location Zone: AP15
```

[0.50s] Sending data to cloud for fine location...

[0.50s] Material ID: MAT009

RSSI values:

AP0: -47.64 dBm
AP1: -55.55 dBm
AP2: -57.77 dBm
AP3: -51.73 dBm
AP4: -56.64 dBm
AP5: -60.05 dBm
AP6: -61.18 dBm
AP7: -65.00 dBm
AP8: -50.41 dBm
AP9: -52.09 dBm
AP10: -59.18 dBm
AP11: -55.00 dBm
AP12: -60.64 dBm
AP13: -55.86 dBm
AP14: -59.32 dBm
AP15: -64.59 dBm

Max RSSI: -47.64 dBm

Coarse Location Zone: AP0

[0.50s] Sending data to cloud for fine location...

Computing the exact location....waiting for cloud response...

KNN → Predicted: [0.9 9.9], Error: 0.00 meters

Random Forest → Predicted: [1.39 9.68], Error: 0.54 meters

MLP (NN) → Predicted: [1.5 9.67], Error: 0.64 meters

==> Best Model: KNN with error 0.00 meters

[2.50s] Refined location received: {'KNN': array([0.9, 9.9])}

```
ssing/material_locator.py
Computing the exact location....waiting for cloud response...

KNN → Predicted: [0.9 9.9], Error: 0.00 meters

Random Forest → Predicted: [1.39 9.68], Error: 0.54 meters

MLP (NN) → Predicted: [1.5 9.67], Error: 0.64 meters

    ==>> Best Model: KNN with error 0.00 meters
[2.50s] Refined location received: {'KNN': array([0.9, 9.9])}

-----
```

```
[2.50s] Computing optimal path...

-----
```

Computing the exact location....waiting for cloud response...

```
KNN → Predicted: [0. 2.4], Error: 0.00 meters

Random Forest → Predicted: [0.81 1.87], Error: 0.97 meters

MLP (NN) → Predicted: [0.71 1.03], Error: 1.54 meters

    ==>> Best Model: KNN with error 0.00 meters
[2.50s] Refined location received: {'KNN': array([0. , 2.4])}
```

```
-----
```

[2.50s] Computing optimal path...

```
-----
```

A* Path --> Nodes Explored: 411, Path Length: 6.30, Time: 1.28ms
Dijkstra Path --> Nodes Explored: 5664, Path Length: 7.20, Time: 14.92ms

A* Path --> Nodes Explored: 159, Path Length: 3.10, Time: 0.47ms
Dijkstra Path --> Nodes Explored: 1677, Path Length: 3.10, Time: 4.28ms

```
[5.90s] Has material MAT009 been picked up? (yes/no): █
```

Figure 3

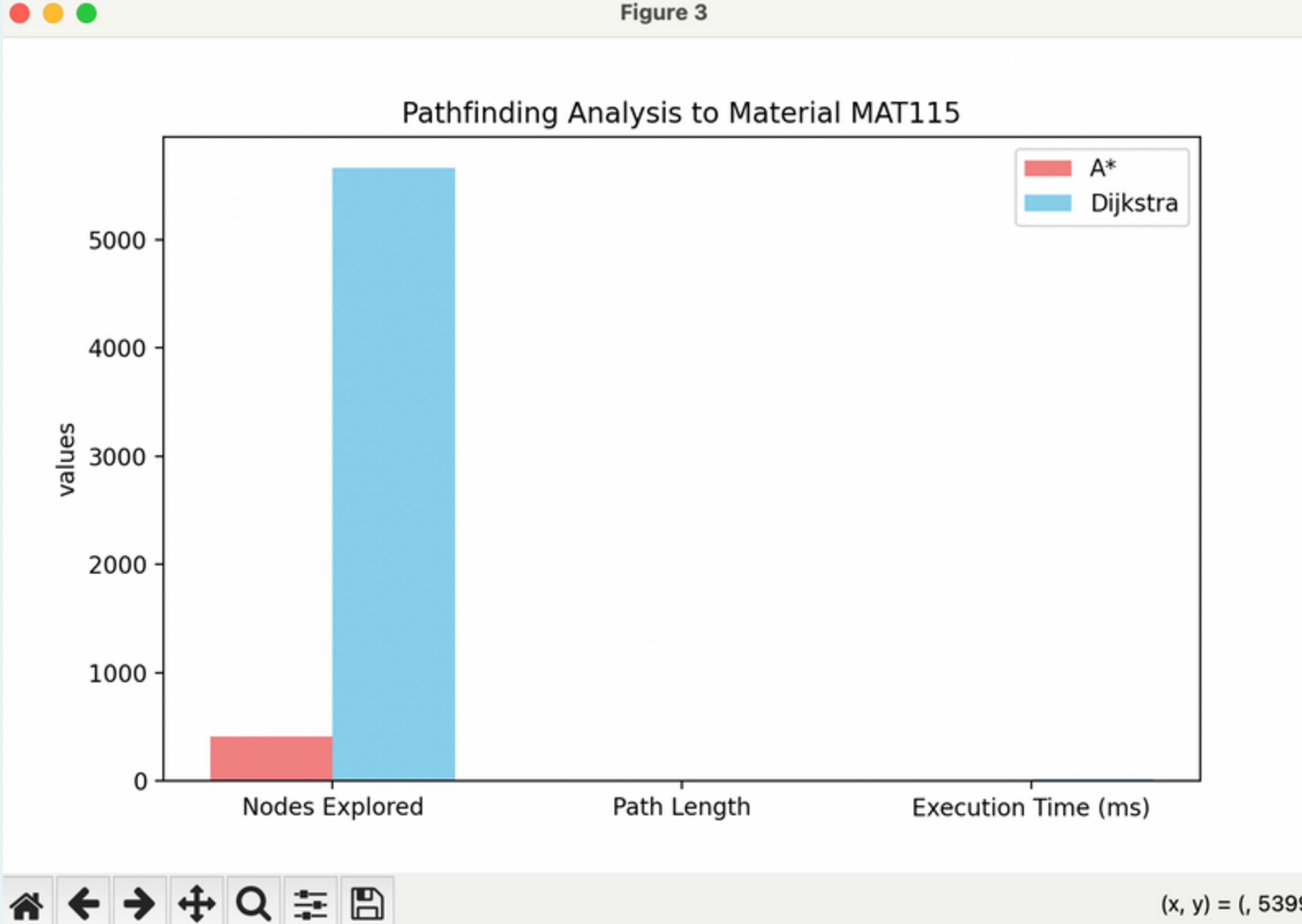
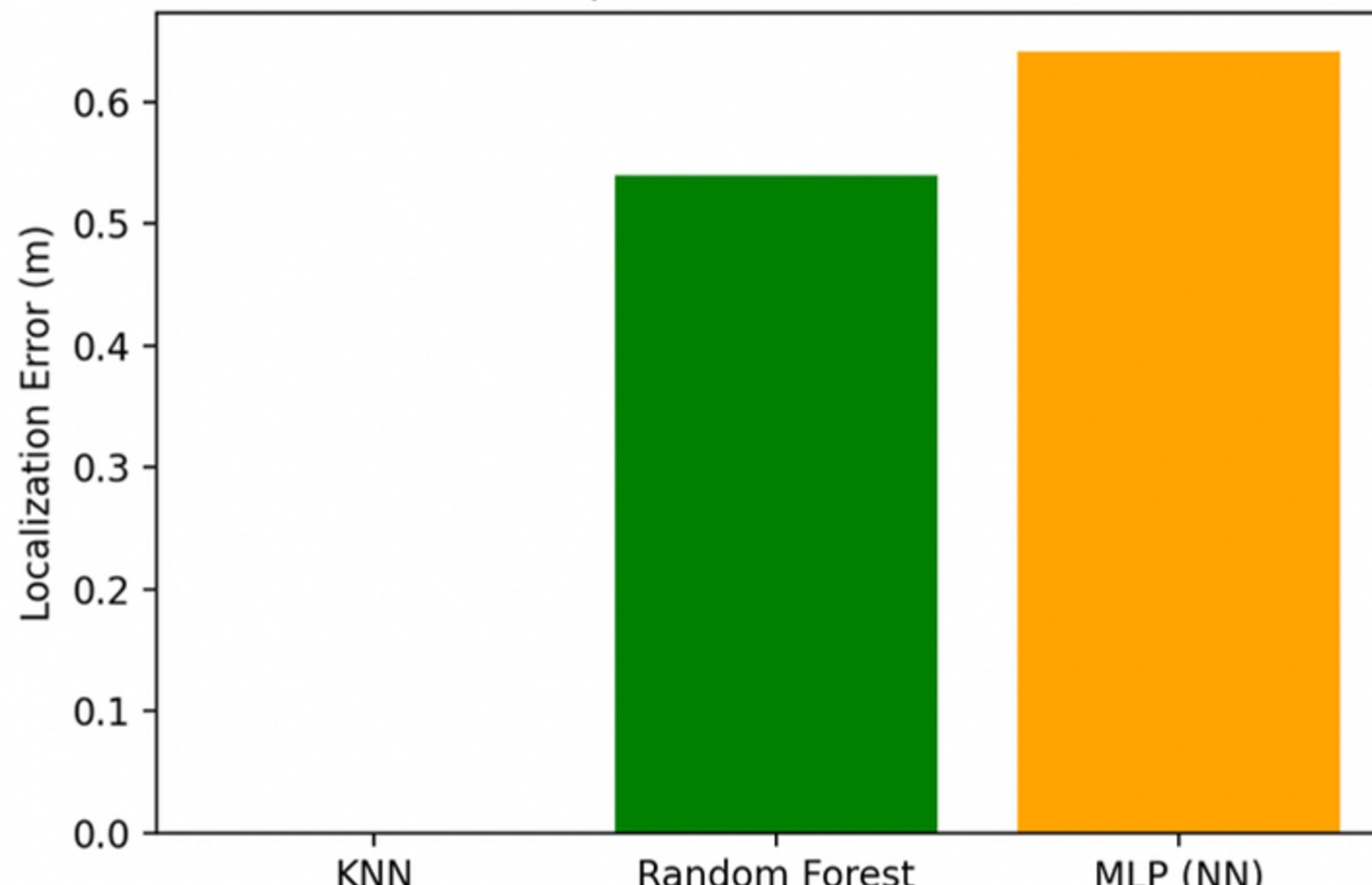




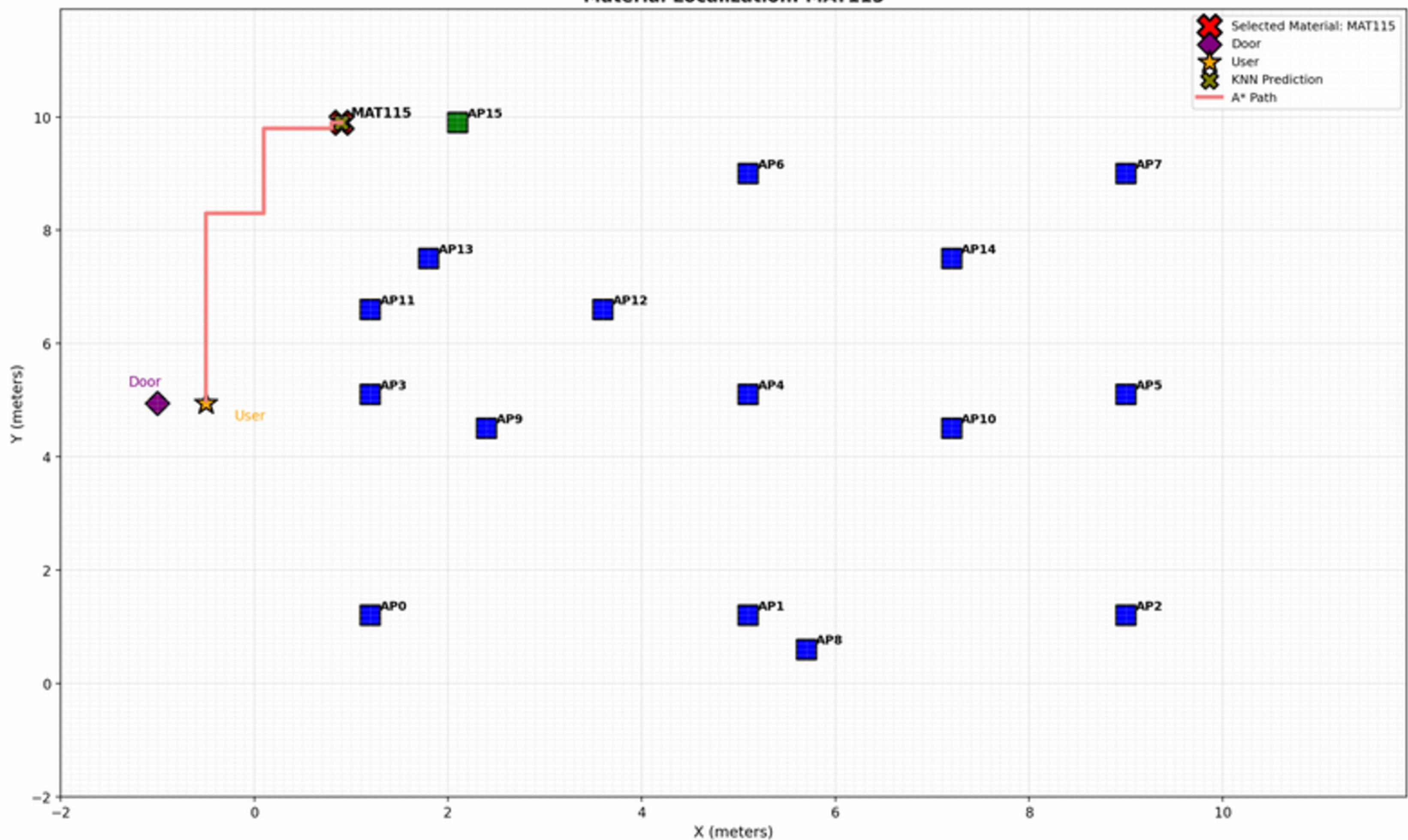
Figure 1

Model Comparison for MAT115 Material



Material Localization: MAT115

Selected Material: MAT115
Door
User
KNN Prediction
A* Path



Resource Allocation & Communication

Node Definition

Edge & cloud with specs: CPU, memory, throughput, uplink/downlink delays,gw_id

Task Modeling

Tasks generated via material_task_source(), not just a simple generator

Scheduling Policies (Scheduler.decide)

FCFS → Always to cloud

Load Balancing (LB) → Based on Shortest estimated finish time

Latency-Aware → based on end-to-end latency

Tracks arrivals, route counts, completed tasks, latencies

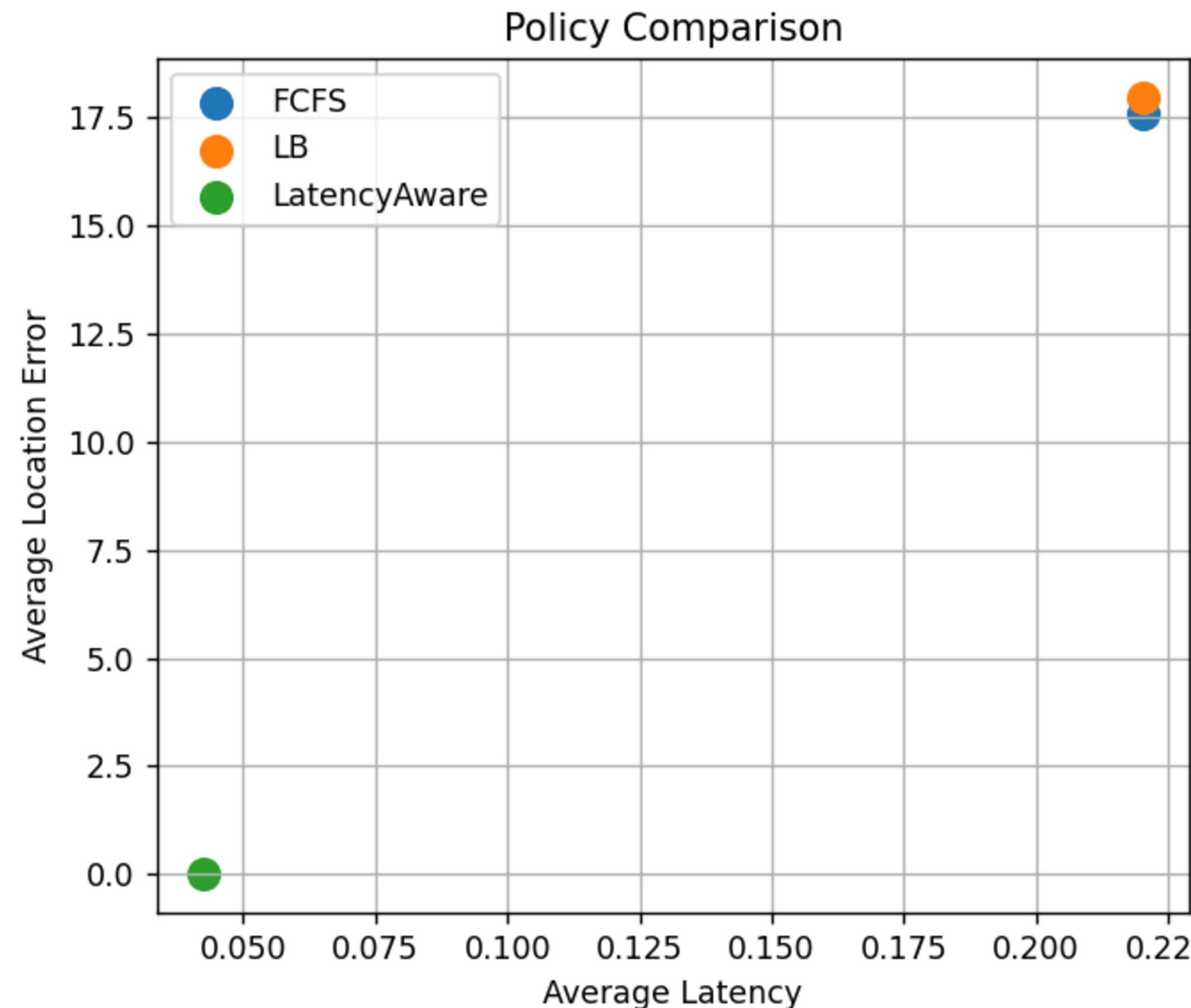
Additional Features

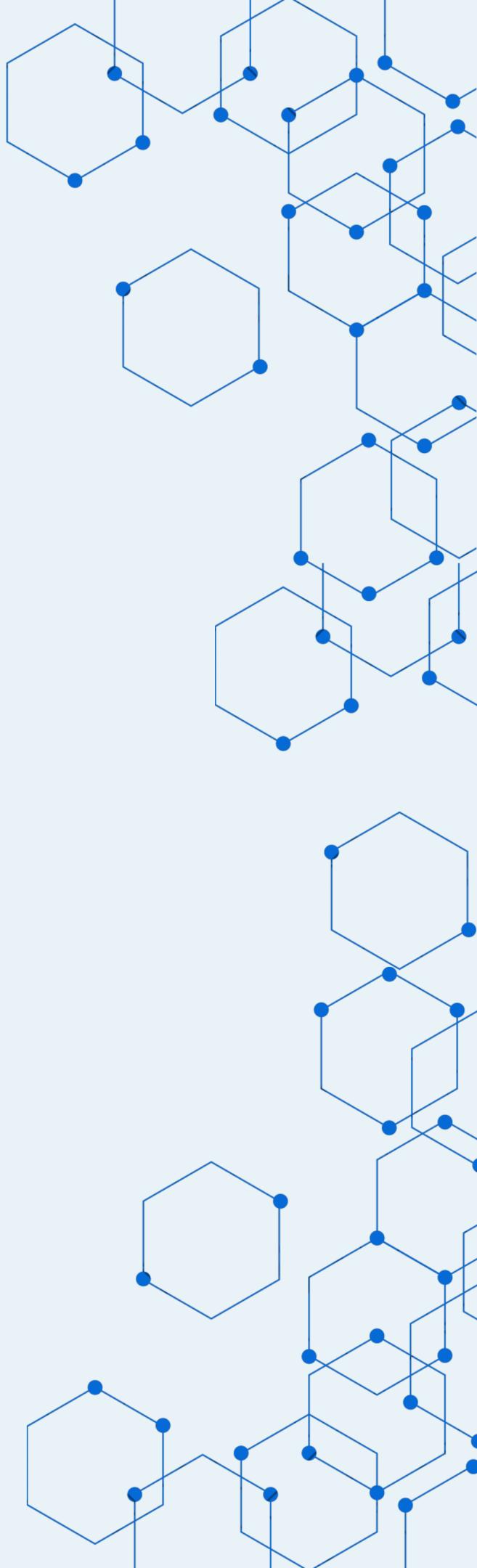
MockBus → Simulates publish/subscribe communication

Integration → Linked with Person 2's processing flow outputs

Figure 1

— □ ×





User Interaction

Implemented :

- A Login page for the workers
- Resource search page which gives instant location
- Manual “Item found button”
- Resource location Map
- Integrated with backend

Tools Used :

FLASK,HTML,CSS,MATPLOTLIB,PYTHON

Search Material

Search

Location: AP0

View Map

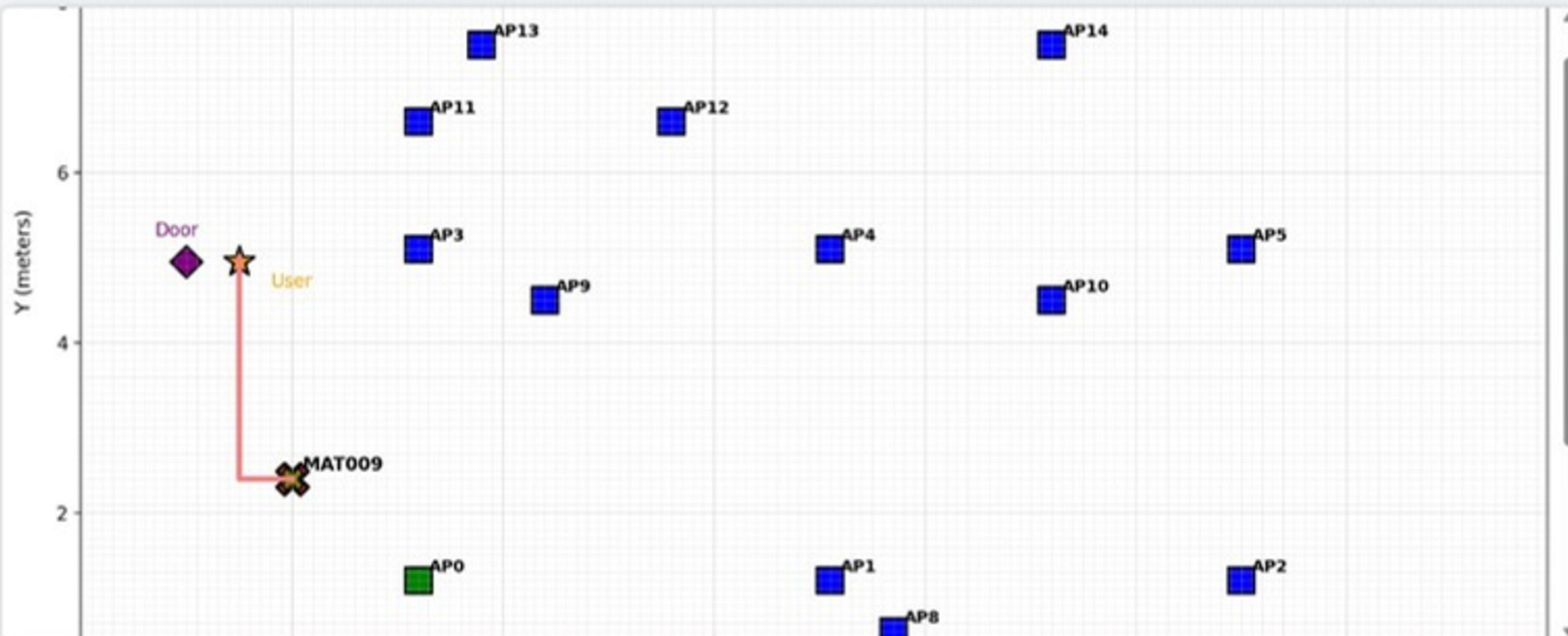
Worker Login

Username

Password

Login

Allocated Resource:



Back to Search

Mark as Picked