

# WiFi Sensing Based Occupancy Monitoring

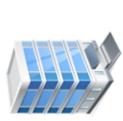
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### Background

- Utilize WiFi signals to detect the physical environment
- Measure the variations in the waves as they interact
- Variations in Channel State Information (CSI) describe how the signal propagate throughout space
- Machine learning uses CSI data to make predictions
- Doesn't require intrusive sensors or cameras

## Why It Matters

- Emergency services can determine where to go
- Identify bottlenecks within buildings
- Building management can make informed decisions about space usage
- Leverage existing WiFi infrastructure







# **Experiment Equipment**

- Used to collect CSI a. ESP32 d1 mini

data from ESP32 b. Raspberry Pi Used to extract

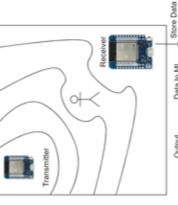
 Used to train c. Keras API

ML models





## High-Level Design



- transmitter outputs WiFi signal ESP32 configured as a
- Wave is disrupted by occupants
- receiver receives WiFi signal ESP32 configured as a

CSI data stored on Raspberry

- Data file passed to ML model
- ML predicts occupancy

Data to ML

Process Data & Make Prediction

Trial	Data Collected On	Accuracy
1	3-Nov-23	33%
2	3-Dec-23	%09
3	13-Feb-24	71%
4	21-Feb-24	95%

Validation accuracy with different data sets

5.19%	29.14%	93.24%
17.69%	45.33%	6.76%
77.12%	25.52%	0.00%
) Person	octual Data 2 People	A 9lqo99 £

Model Predicted 1 Person 2 People

Confusion matrix for 1-, 2-, and 3-person

occupancy (validation accuracy: 71.58%) 8.60% 0.82% 90.41% 4.12% 95.06% 0.99% 3 People Actual Data

1 Person 3 People

91.13%

8.87%

0.00%

Confusion matrix for 1-, 3-, and 5-person occupancy (validation accuracy: 92.2%) Model Predicted

