

An open-source system for monitoring activity in a built environment combining edge and fog computing

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Summary

At Executive Park 6, we developed a low-cost indoor monitoring system preserving the privacy using Raspberry Pi 4, Respeaker USB microphone array, Raspberry Pi Camera and Google Coral USB TPU for people suffering from cognitive impairment diseases like Dementia, Alzheimer's disease etc. with their privacy preserved. With custom real-time algorithms, we detect the position of participants and try to assess their engagement with each other through poses. To collate all the data and effective monitoring, we developed EP6 Dashboard which helps combining all the data from Audio, Camera and Bluetooth pipelines together and presents the analysis of this data using Computer Vision, Signal Processing and Machine Learning Techniques. EP6 Dashboard is one stop solution for monitoring the sensor network, analyzing and visualizing the multimodal data collected.

Statement of need

As detection techniques have improved over the years, Alzheimer's and other different forms of dementia have come to receive an increasing amount of attention. In addition to being the sixth most common cause of mortality in the US, Alzheimer's disease is also thought to be the possible cause of dementia in more than 6 million people. While the average person experiences a natural reduction in cognitive function with aging, some people experience a far more rapid decline, which frequently progresses to Alzheimer's disease or another form of dementia. These people are classified as having mild cognitive impairment, making them the study's population.

EP6 Dashboard is a unified portal developed using python packages and React framework to monitor the indoor activities through audio, visual and spatial tracking. It monitors following activities: 1. Audio 2. Visual 3. Indoor Temperature and Humidity 4. Bluetooth

System Architecture

To implement the aforementioned system, we have followed a scalable three tier architecture using Flask as application server hosted with nginx as load balancer and reverse proxy. Frontend is designed with React for and served through nginx as webserver. We are using influxdb as database for storing the time series data generated by the edge devices. Redis is used as key-value store to interact with background python processes, whose output is consumed on the dashboard. MySQL database is used for storing the authentication and authorization of users.

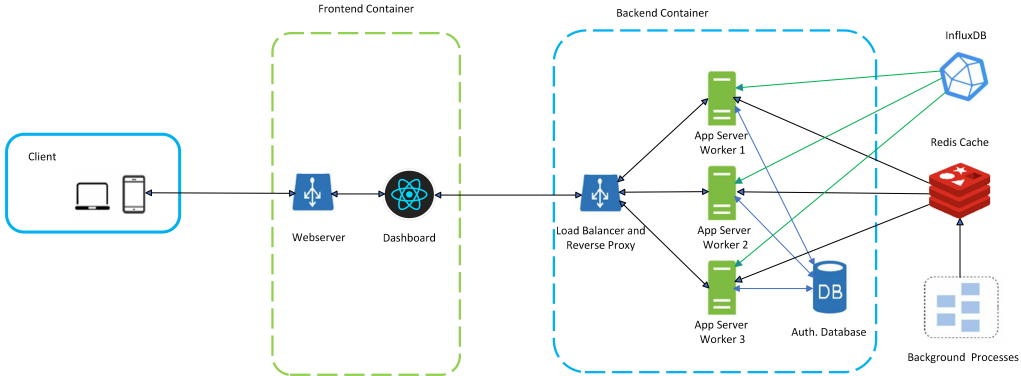


Figure 1: Architecture Diagram of EP6 Dashboard

39 **Monitoring the Sensor Network**

40 For a distributed sensor network application like this, it becomes very important to ensure that
41 all of the sensors are working faithfully and recording the data. To ensure this, we developed a
42 robust mechanism to check the health of each of the Raspberry Pi and sensors mounted to it.
43 The results from this upstream system is sourced to the dashboard.

44 Section A) in fig. II represents position of each Raspberry Pi on EP6 schematic with region
45 monitored by them. If clicked on a particular region, it shows the status of sensors connected
46 to that Raspberry Pi as shown in section C) of fig. II. Lastly, table shown in section B) of
47 fig. II represents the list of all Raspberry Pi in EP6 with their status and an option to reboot
48 them remotely.

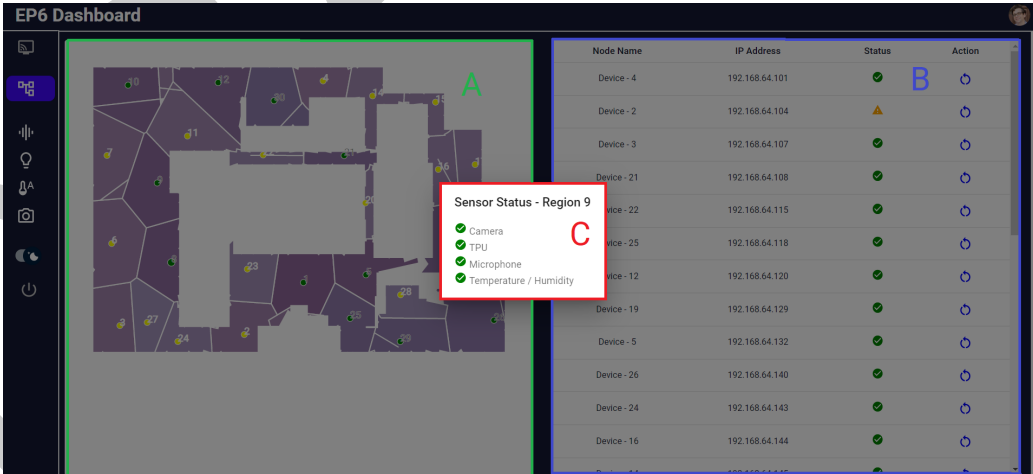


Figure 2: Sensor Network Monitoring (A) Schematic of EP6 defining the positions of Raspberry Pi with region monitored by them (B) Status of each Raspberry Pi with option to reboot them remotely (C) Status of Sensors connected to Raspberry Pi in Region 9

49 **Audio Pipeline Analysis**

50 As part of audio pipeline, we collect the audio features of the participants through respeaker
51 USB microphone arrays placed on ceiling. We don't store the audio itself but its computed
52 feature to preserve participant privacy. Through these features, we perform speaker diarization

53 followed by tagging the respective participants groups. Through these, we are measuring the
54 degree of engagement in each group.

55 Occupancy analysis plays an important role in identification of how physical location of the
56 EP6 correlates with the interaction among the participants. fig. III represents the image of
57 Audio section on EP6 dashboard. Section A) shows the physical location of microphone arrays
58 in the EP6 lab. To have an hour dependent view of occupancy, we also plot the heatmap of
59 occupancy based on the audio signals as shown in section C) which can be controlled through
60 slider shown in section B) of fig. III. One can vary the hours and see the occupancy change
61 with respect to hours.

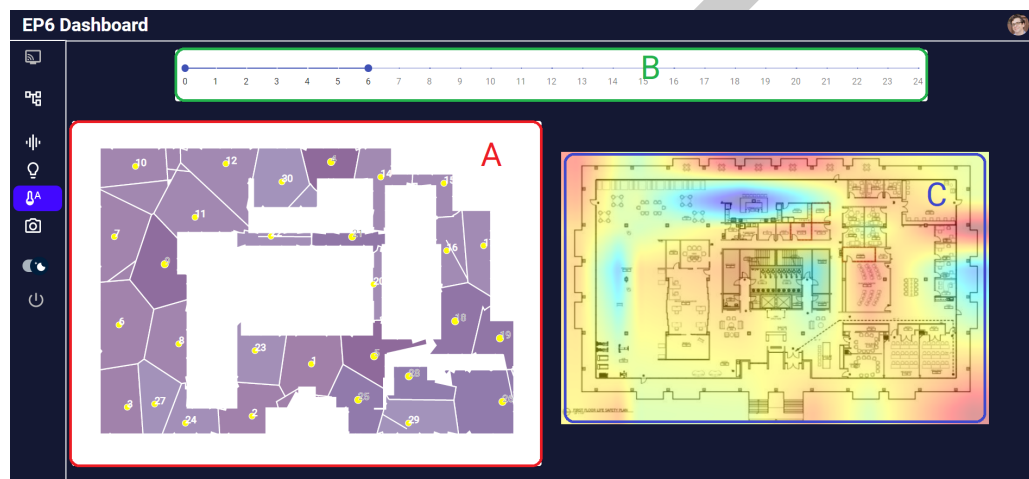


Figure 3: Audio Pipeline (A) EP6 Schematic with position of microphone arrays (B) Slider to analyze the activity between two specified hours (C) Heatmap depicting the activity level in EP6

62 Visual Pipeline Analysis

63 Through the means of vision pipeline, we are trying to record and analyze the features in
64 movement, action, and interaction of participants/patients through computer vision techniques,
65 which are later displayed in the vision segment of the dashboard. With movement, we are
66 trying to analyze the path taken on the EP6 floor in a certain time frame over multiple
67 iteration for every individual human subject. With action, we try to analyze posture of every
68 single individual (sitting, standing, leaning to talk). With interactions, we aim to register
69 f-formations for further research purposes that occurs during participants interactions. All
70 the above-mentioned processing is done through the collection of Image feed from all the
71 individual raspberry Pi's at every second. However, it is unsaid truth that Pi's don't always
72 gather data at every instance of time due to processing fluctuations. With the said state of
73 image collection, an algorithm manages to stitch images from all the cameras, of the given
74 instance in time, into one singular image that shows the overview of the EP6 floor and activity
75 going on.

76 Dashboard displays the processed data in 4 different ways, Heatmap, Occupancy Map, and
77 Camera Location. Heatmap, as the name suggests, displays the occupancy in terms of heat
78 signature to visualize the population distribution throughout the EP6 floor. On contrary,
79 Occupancy Map gives more specific location of all the individuals, at any given time, in the
80 EP6. Lastly, Camera Location provides the location of the places from where the image
81 originates.

82 **Bluetooth Pipeline Analysis**

83 **Mathematics**

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