PetSTAR

B7: Rebecca Manley, Brandon Wei

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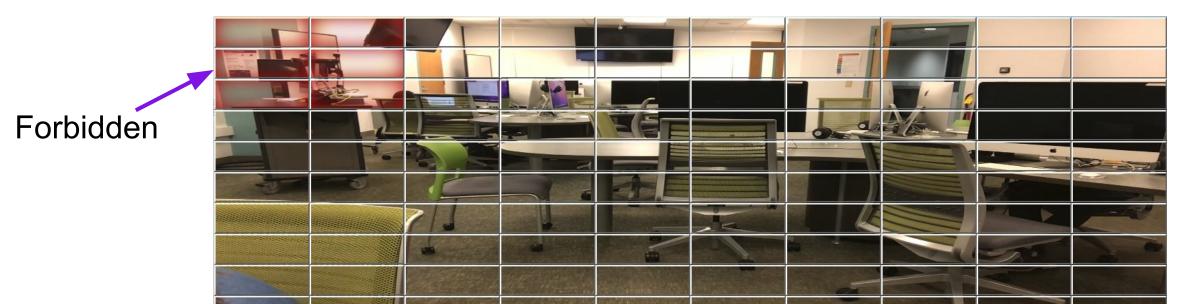
Product Pitch

We set out to create a pet monitoring system with features beyond currently available camera systems: keeping track of pet activity and notifying the user if a pet tries to enter somewhere it shouldn't. To this end, we needed our system to track pet movements accurately - estimate each animal's position within 1 foot - and respond to any events quickly - forbidden location detection within 1 second, and a notification delivered in less than 10 seconds.

Our final product utilizes a Raspberry Pi and web application to meet these requirements. We have achieved an average accuracy within **3-6 inches**, and an average speed of **0.6 seconds** to detect and **1.1 seconds** to report.

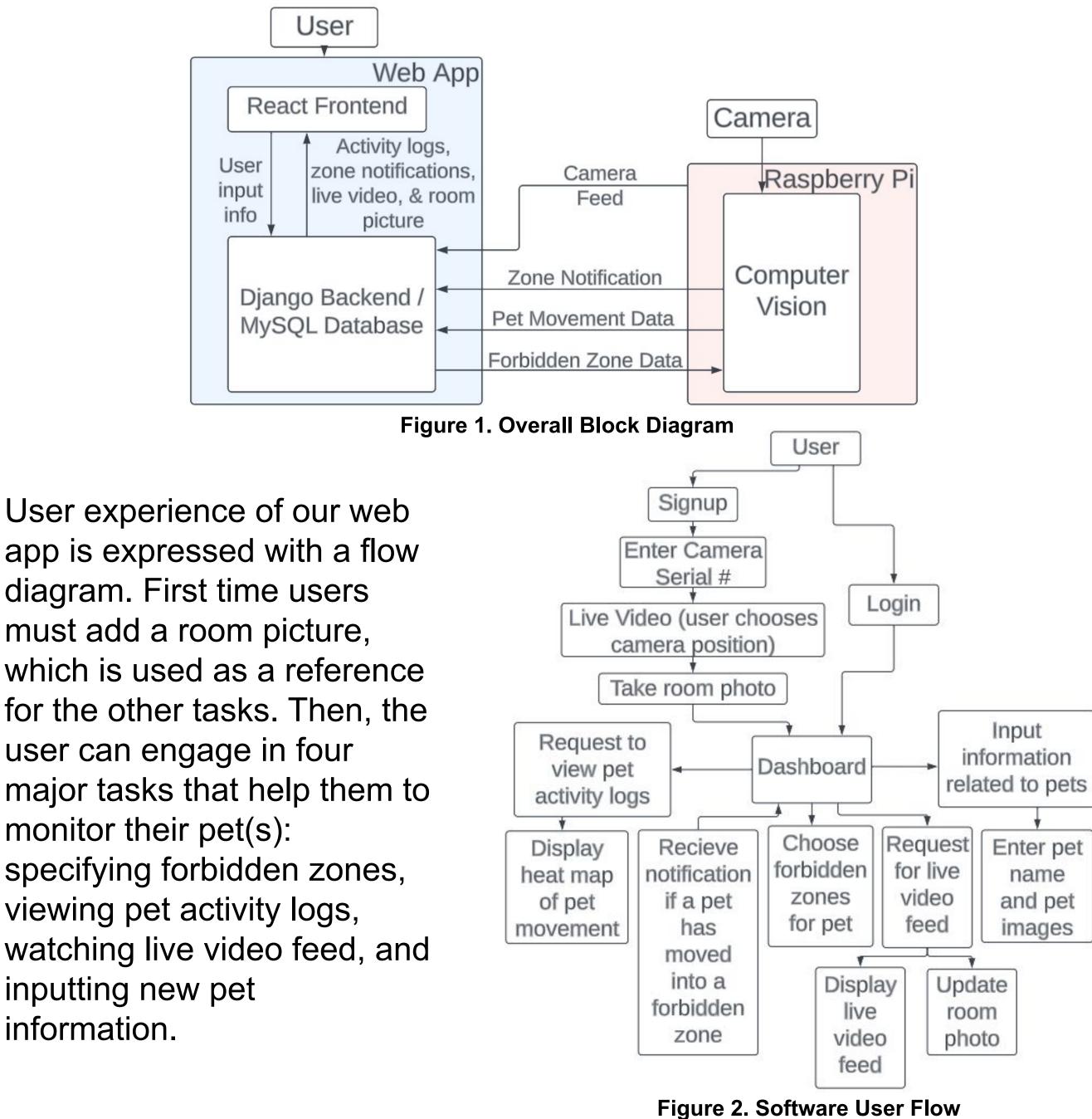
System Description

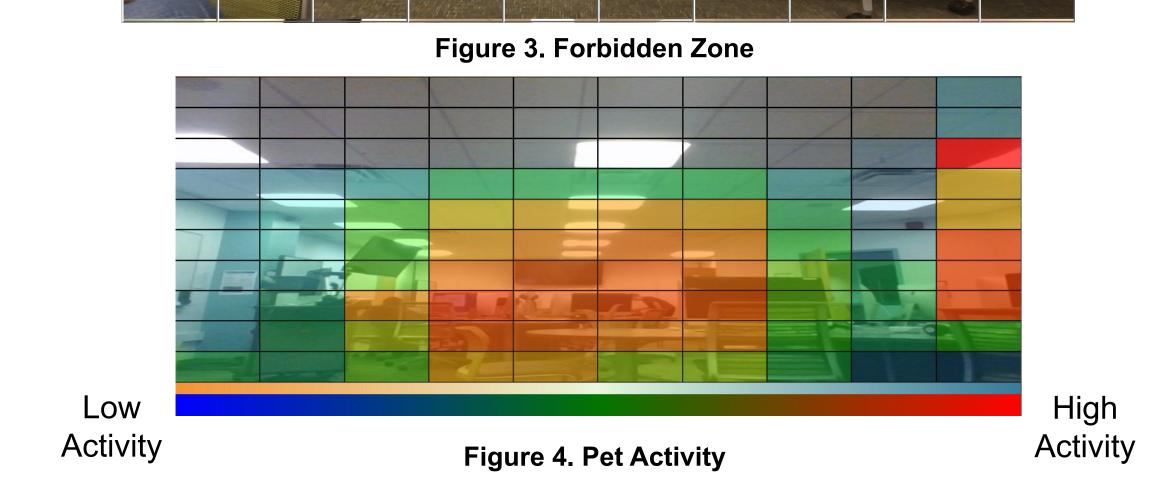
The web application uses React for the frontend and Django for the backend. Communication between these parts is done using Axios to handle requests and Django REST Framework to format data for the backend. Two unique features of the web application are specifying forbidden zones where pets should not go and viewing pet activity via a heat map.



System Architecture

Two major subsystems: the web app and Raspberry Pi. These systems communicate relevant pet data back and forth via requests. The web application consists of a frontend (user interaction) and backend (data storage). The Raspberry Pi hosts the computer vision tasks to process video feed and report to the web app.





The hardware for our project consists of a Raspberry Pi Model 4B connected to an RPi camera module 3. The RPi runs computer vision algorithms implemented with OpenCV. It detects new movement via pixel differences and tracks identified movements using OpenCV built-in tracking algorithms.

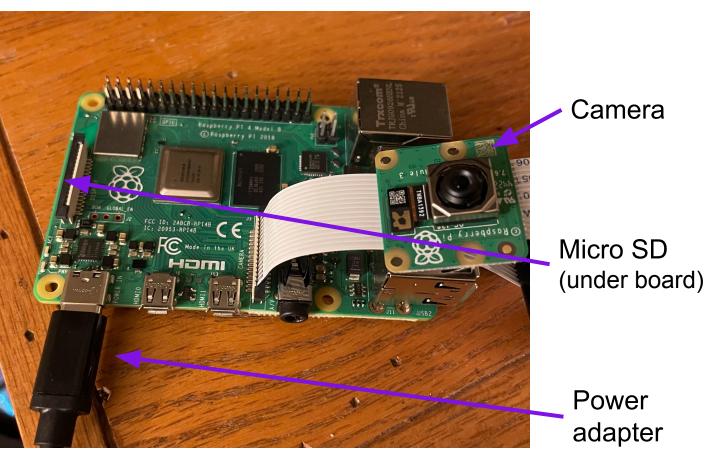


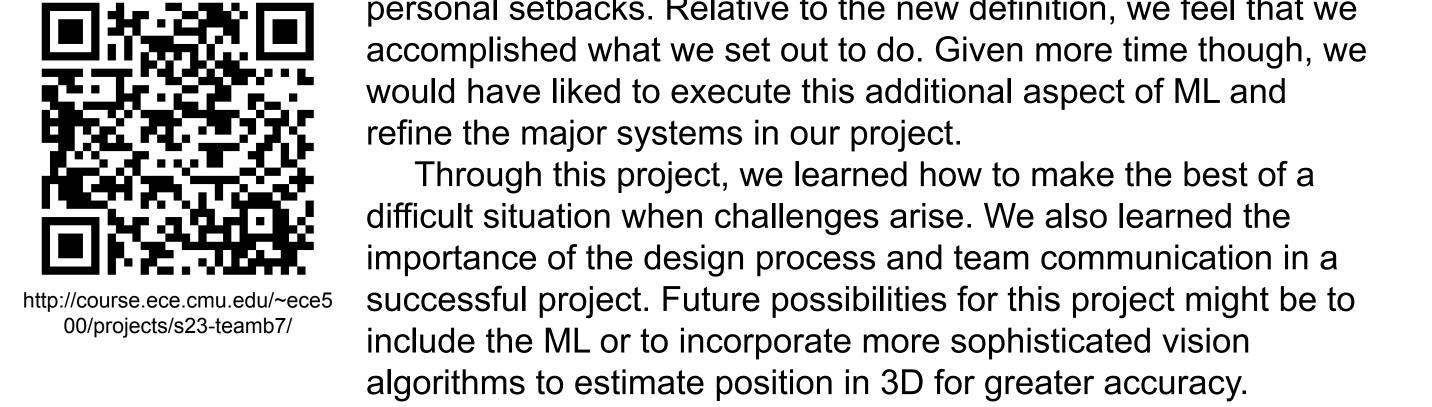
Figure 5. Raspberry Pi Configuration

System Evaluation

The approach to any speed-related tests was to use a slow motion camera to more precisely calculate the time between two visual events. Accuracy was measured by comparing human chosen position to computer chosen position.

diagram. First time users must add a room picture, which is used as a reference for the other tasks. Then, the user can engage in four major tasks that help them to monitor their pet(s): specifying forbidden zones, viewing pet activity logs, watching live video feed, and inputting new pet information.

Conclusions & Additional Information



Our project initially aimed to incorporate an aspect of machine learning - to differentiate between multiple animals in the home but needed to be rescoped in the middle due to unavoidable personal setbacks. Relative to the new definition, we feel that we

Test	Method	Metric	Goal	Result*
Zone detection	Slow-Mo	Pet enters zone in real life ->	< 1	~0.625
speed	Video	CV detection	second	seconds
Zone notification	Slow-Mo	Pet enters zone in real life ->	< 10	~1.125
speed	Video	Web app notification	seconds	seconds
Accuracy of	Human vs.	Distance between centers of	Within 1	Generally
tracking	Computer	human-chosen vs. computer- chosen bounding boxes	foot	3-6 inches
Detection speed of	Slow-Mo	Animal enters frame in real life	< 5	~0.75
new animals	Video	-> bounding box appears	seconds	seconds

The biggest tradeoff we faced for speed was the frequency of requests between the CV, backend, and frontend. Faster request rates means info on the web app is more up to date (including notifications), but the extra computational power/time uses more web server resources and lowers our frame rate.

*Numbers listed above were computed running the CV code on a laptop.

