



To: Professor Pisano

From: Beren Donmez, Margherita Piana, Marissa Ruiz, Bennet Taylor, Albert Zhao,

Team: 8

Date: 11/21/24

Subject: Bike Guard First Prototype Test Report

1.0 Equipment and Setup

- **Hardware**

- Raspberry Pi Zero 2 W with 32GB SanDisk SDHC Class 10 card
- Piezo Buzzer
- MPU-6050 Accelerometer
- INIU BI-B61 Portable Charger (22.5W, 10000mAh)
- Raspberry Pi Camera Module V2
- TP-Link Router
- Small breadboard
- Electrical Tape
- Jumper Cables
- USB-C to Micro USB Cable
- Small heat sinks

- **Remote Equipment**

- Laptop

- **Software**

- Raspberry Pi OS 32-bit (Legacy) – Debian Bullseye
- Node.js – for child_process and file system operations
- Python3 – accelerometer data reading
- Flask – to stream camera feed and handle back-end tasks
- Front-end: JavaScript, React, and CSS
- Back-end: Flask, SQL for data storage
- Machine Learning Model: Logistic Regression



- **Setup**

- Hardware components placed in a lockbox mounted to the bike, Raspberry Pi Camera peaks out of the enclosure to capture video data
- Raspberry Pi and all servers are connected to TP-Link Router
- Raspberry Pi is powered by a portable power bank
- Small heat sinks are attached to Raspberry Pi's onboard chip components
- Raspberry Pi connected via SSH using `ssh Team8@Raspberry_pi_IP`
- Accelerometer data collected and saved in CSV files using Node.js and Python scripts (for later use by machine learning model, not currently incorporated)
- Buzzer connected to Raspberry Pi, goes off when Pitch and Roll values reach hardcoded threshold (in future will be handled by machine learning model)
- Pi Camera data is taken using Python and streamed to local host using Python and Flask
- When buzzer goes off, push request is sent to back-end and stored in an SQL database
- Front-end monitors for changes in back-end, uses React to update changes on web interface (`npm start`)
- Front-end embeds camera stream from Raspberry Pi's local host to the web interface (python script to connect front-end and back-end)

2.0 Measurements Taken

2.1 Hardware and Connectivity

- Verified that the Raspberry Pi boots up correctly with the portable power bank by checking that green LED on the Raspberry Pi turns on
- Pinged Raspberry Pi after boot to ensure network connection to router was successful
- Confirmed successful communication between peripherals (accelerometer, buzzer) and Raspberry Pi by running our `accelerometer.py` independently from other processes
- Confirmed camera stream was successful and sent to local host by running `picam3.py` independently of other processes



2.2 Data Recording and Alerts

- Tested accelerometer data logging with Node.js and confirmed that motion is recorded in CSV format
- Verified the bike triggers the buzzer when shaken and sends a message to the front-end
- Observed that the website updates in real-time when the accelerometer detects the shaking
- Observed real-time camera stream on website

2.3 Machine Learning and Classification

- Trained the logistic regression model on pre-collected accelerometer data
- Assessed model performance using the confusion matrix:
 - True Positives: 49
 - False Positives: 3
 - False Negatives: 6
 - True Negatives: 18
- Calculated accuracy as approximately 88.16%

3.0 Conclusions based on test data

- The hardware setup functions as expected: the enclosure does not interfere with the bike's operation, and all components operate reliably.
- The accelerometer successfully detects excessive shaking, triggers the buzzer, and logs data.
- The Raspberry Pi Camera shows a live video feed to the front-end without interruptions.
- The machine learning model achieves good performance, effectively identifying theft scenarios with high accuracy.
- The system meets the measurable criteria for a successful run: real-time alerts, immediate buzzer response, and reliable classification of motion.



4.0 Extra Credit: Our Ideas for the Future

- **Integrate Machine Learning Model to the Product:**

Incorporate the trained logistic regression model directly into the Raspberry Pi system, enabling real-time detection and classification of theft scenarios without external dependencies

- **Incorporation of a Prepaid SIM Card:**

Equip the system with a prepaid SIM card to enable data transmission over cellular networks, ensuring functionality even when Wi-Fi is unavailable

- **GPS Tracking Through the SIM Card:**

Utilize the SIM card for GPS tracking, allowing the owner to monitor the bike's location remotely in case of theft or unauthorized movement.

- **Design and Print our Own Enclosure:**

As it stands, our current enclosure feels too bulky and generalized. In the future we hope to downsize the enclosure and make it blend in with the bike, perhaps disguised as a bike light or other bike component. The smaller we can make it the less noticeable by thieves.

- **Move from web application to mobile application:**

For our demo we showed live alerts and camera stream on a locally hosted website, we hope to move that to a mobile application

- **Switch to replaceable batteries:**

For the sake of the demo, we used a power bank to power our device and make it mobile. In the future, we hope to assess our power budget and move to replaceable batteries

- **Linking user account and device via Bluetooth**

During user setup, user can use Bluetooth to link their phone to their device