



# Anomaly Detection and Alert System for Facilities and Factories

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# Traditional Camera Based Surveillance Systems and their Failures



Average cost of surveillance systems  
**\$5,000 - 25,000**  
per month.

Which can make up  
**More than 5-20%** of  
medium to small company's annual  
expenditure.

## Shortcomings

- **High Cost**
  - In a traditional setting, purchasing business security hardware, installation fees, business monitoring by professionals can be of high expenditures
- **Ineffective Security Surveillance**
  - 2021 March, over 200 Tesla factory and warehouse camera hacked
  - 2022 September, Ford lost 20 Mustang GTs near a surveillance guarded storage facility

# Our System Makes a Difference

- **Kiana, and Cloud-based RTLS**
  - Kiana's Real Time Location Solutions (RTLS) technology tracks the location of objects or people in real time within a building or other contained area.
- **Our Anomaly Detection and Alert System**
  - With Kiana RTLS technology: track the movement of the devices and collect their MacAddress
  - Conduct unsupervised machine learning analysis to recognize collected data
  - Categorize data into different groupings while monitoring their movements
  - Uses SMS portal authentication to extract user phone numbers, and send notifications messages to relevant personnel when abnormal activities are detected
- **What Our System Brings?**
  - Massive pay off in data security and loss reduction
  - Provide the capabilities of venturing on to Industry 4.0 process
  - Requires no additional software or installation costs

# Anomaly Detection and Alert System Performance

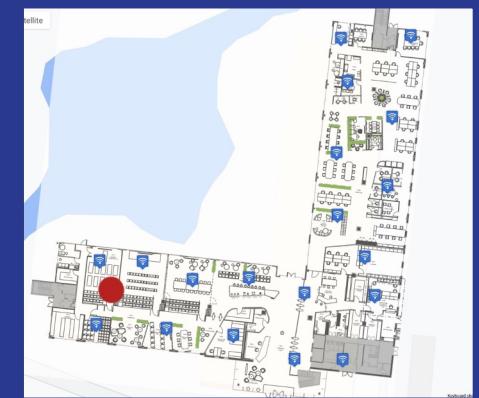
## Traditional Camera System

- Manually check: Time & Energy Consuming
- High maintenance cost
- Not all places monitored
- Not real time



## Anomaly Detection and Alert System

- No additional devices needed
- Low maintenance cost
- All areas covered
- Real time



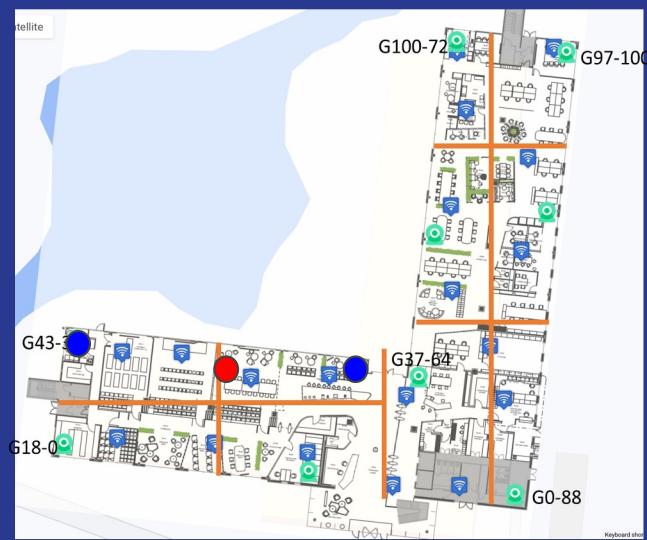
# Anomaly Detection and Alert System Performance

	ClientMacAddr	Alarmtime	Floor	Area	Abnormal_type
0	88:66:a5:33:9f:c2	2019-11-28 12:40:45.961 UTC	3	[64, 85]	human_security
1	a4:c3:f0:a1:fc:2a	2019-12-15 08:27:16.863 UTC	2	[63, 90]	human_security
2	9c:da:3e:69:cb:7a	2020-02-12 12:44:63.961 UTC	2	[64, 86]	human_security
3	90:61:ae:25:4b:2b	2020-06-29 17:54:35.473 UTC	0	[58, 85]	machine_maintenance
4	84:9f:b5:e5:4d:8f	2020-07-08 19:44:63.658 UTC	1	[39, 83]	human_security
5	5c:5f:67:8c:17:7f	2020-08-31 03:35:55.43 UTC	3	[87, 92]	human_security

Output of Detected Anomaly Activities



Alarm Message Example  
 anomaly activities nearby)

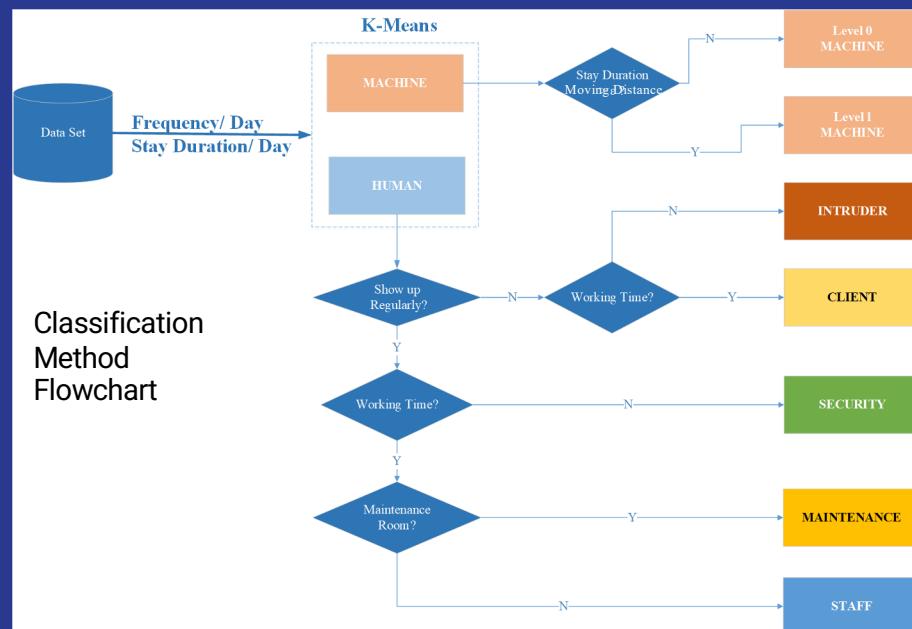


Closest Camera Location (will be activated while

## Machine and Personnel Identification

The system uses unsupervised machine learning models to categorize its recorded devices into two large groups: **Machine** and **Human**.

Each of the two groups have their own subgroups.



## Machine Classification Groups

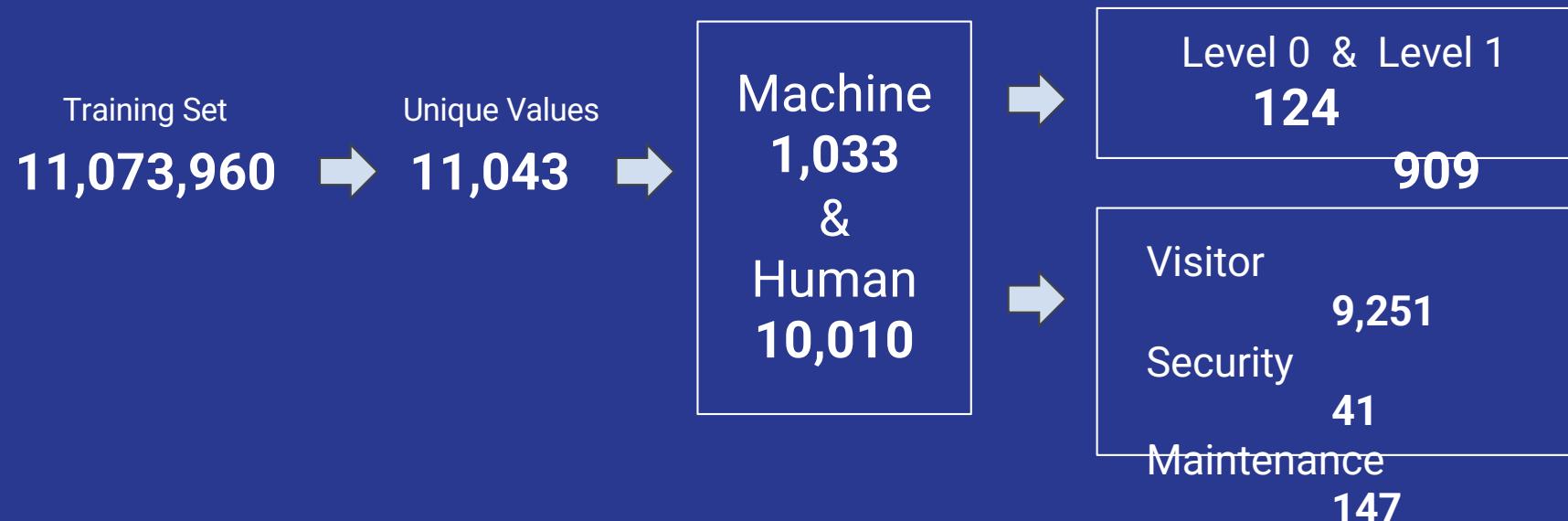
Classification	Description
<b>Level 0 Machine</b>	Machine of great importance, do not usually leave its designated space.
<b>Level 1 Machine</b>	Machine for daily use, can be shifted and moved to other rooms for according to needs.

## Human Classification Groups

Classification	Description
<b>Intruder</b>	Unexpected visitors that enters the building during non-working hours.
<b>Client</b>	Visitors that enters the building during normal working hours.
<b>Security</b>	Security personnel that patrol the secured space at non-working hours and stays in the security room during working hours.
<b>Maintenance</b>	Maintenance team that needs to be called to the scene when technical assistance is needed on Level 1 Machines.
<b>Staff</b>	Working staff at the facility, comes to work and leaves office on schedule.

## Kiana Data Explorations Overviews

- The original pharmacy dataset has been split into training and testing sets, the training dataset consists 70% of the overall data covering 11,073,960 observations of a pharmacy company from 2019 Aug to 2020 Aug.
- Original dataset consists of the observations of devices in the area, with attributes: unique MacAddress, longitude, latitude, locatime, and floor level.

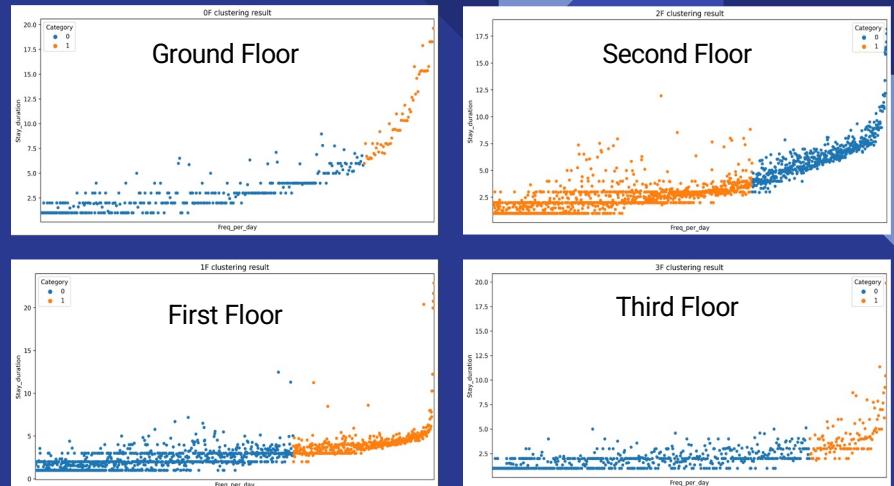


# Machine and Personnel Identification

## Human V.S. Machine

- We conducted a general human/machine splitting using K-means clustering method.
- The model examines
  - **signal frequency per day**
  - **online duration per day**
  - **most commonly appeared location**

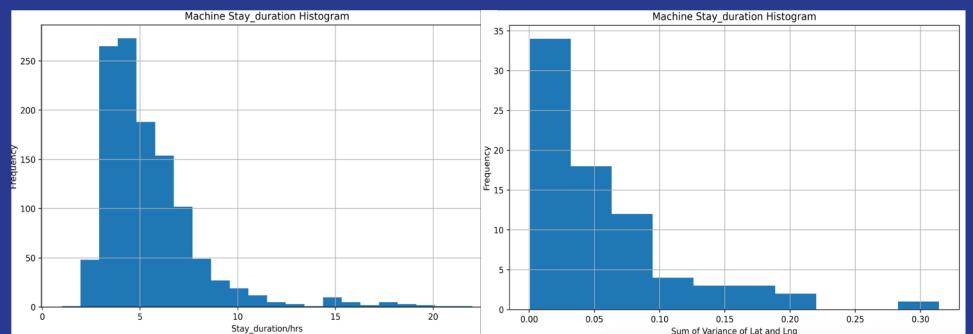
K-means Cluster Analysis on Different Floors



## Level 0 Machine V.S. Level 1 Machine

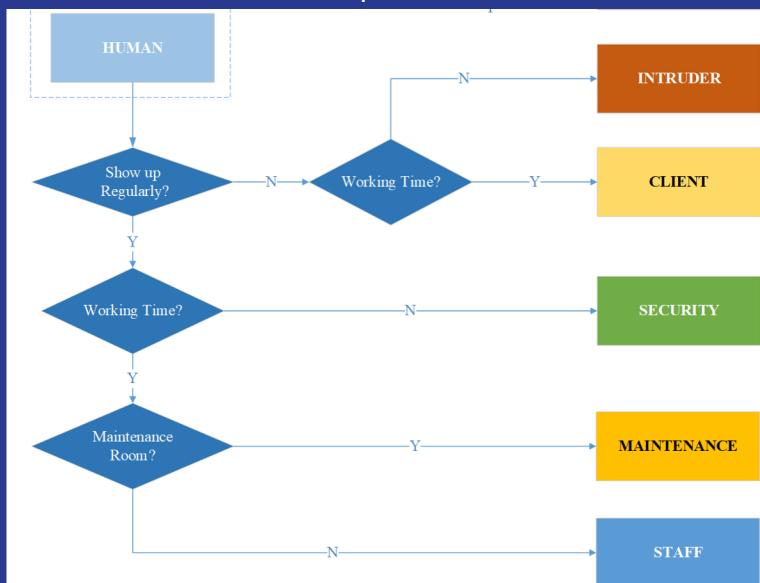
- Further classify machines into high significance machine and regularly used workplace machines.
- The two are separated by
  - **duration of a machine staying in the same spot**
  - **sum of variance of moving distance**

Histogram for Stay Duration and Moving Distance



## Further Personnel Identification

Flowchart for Personnel Separation



## Human V.S. Human

Based on time of appearance, stay duration, and commonly stayed area, the system further divided humans into **security personnel**, **maintenance team**, **working staff**, **client** and **intruder**.

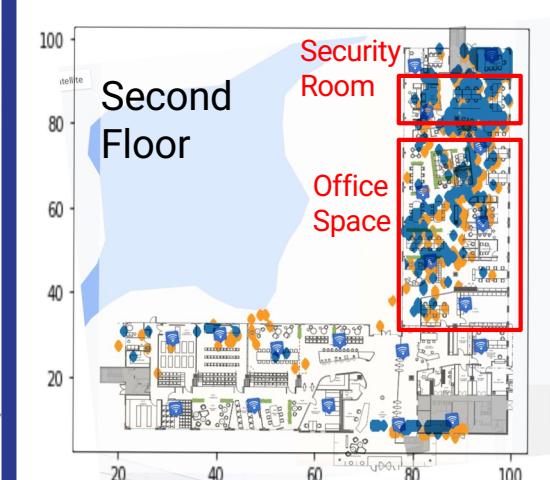
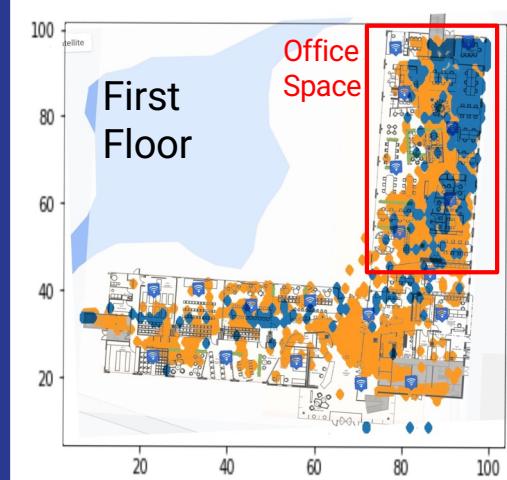
- **Time of Presence**
  - Security, maintenance and staff usually have a fixed schedule of presence
  - Client and intruder visits at random
- **Overall Stay Duration**
  - Maintenance and staff usually have a fixed working hours
  - Security personnel stays in the facility longer
- **Commonly Stayed Area**
  - Security, maintenance and staff have their designated working area
  - Client and intruder do not

## Further Personnel Identification

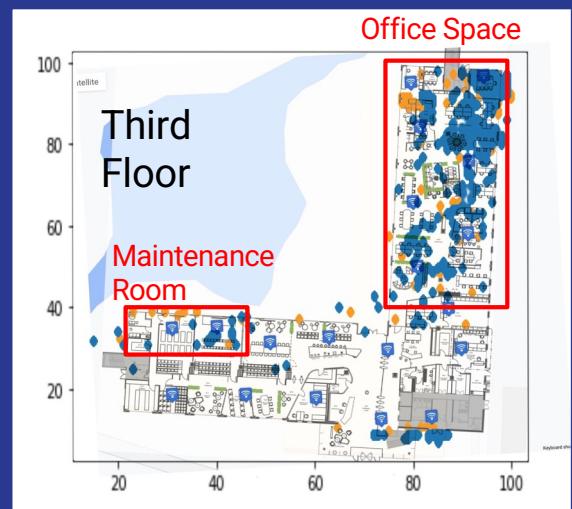
- 1) Gridding System that divide floor map into  $100 \times 100$  grids.



2) Assigned Floor Map for First Floor and Second Floor



- A gridding system is created to simplify computation.
- We used grids to assign different rooms for convenience of personnel recognition and dividing security levels.
- The rooms are appointed with our conjectures. Rooms can be easily modified by changing the assigned grids.



## Abnormal Activity Detection: Levels of Floors and Humans



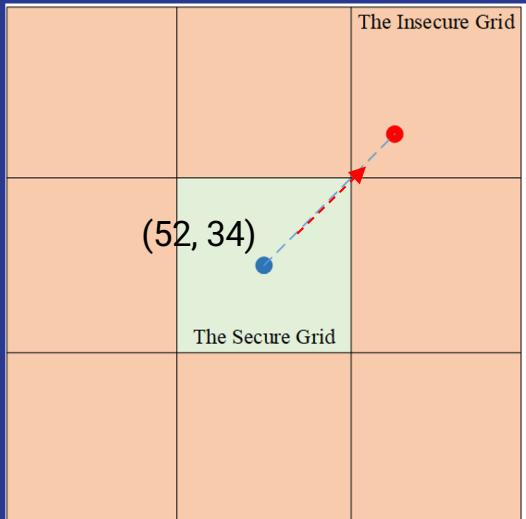
Classification	Security Level	Regional Level	Areas allowed to enter
Security	3	3	Area 0, Area 1, Area 2, Area 3
Maintenance and High level Staff	2	2	Area 0, Area 1, Area 2
Low level Staff	1	1	Area 0, Area 1
Visitor	0	0	Area 0

Classification Groups and their Assigned Security Levels

- 4 Levels: 0, 1, 2, 3
- Assigned levels based on cumulative population density: the higher the density, the higher the security level

Replicable : )

## Abnormal Activity Detection: Levels of Machines and Humans



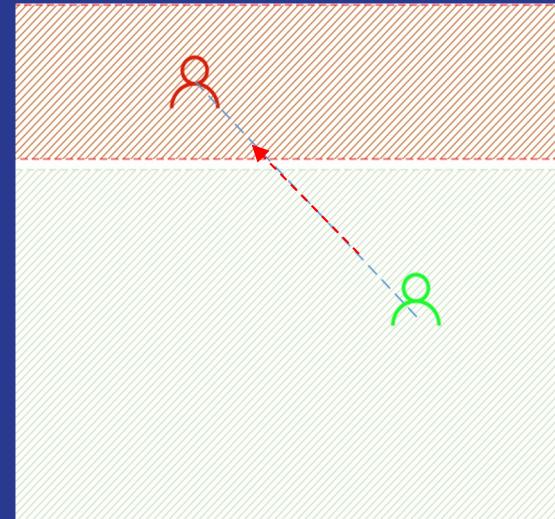
Security area for machines (10x10 surrounding grids)

### Level 1 Machines

- Leave secure grid  $\rightarrow$  Level 3 alarm (Security)
- Signal disappears  $\rightarrow$  Level 3 alarm (Security)

### Level 0 Machines

- Leave company building  $\rightarrow$  Level 3 alarm (Security)
- Signal disappears  $\rightarrow$  Level 1 alarm (Maintenance)



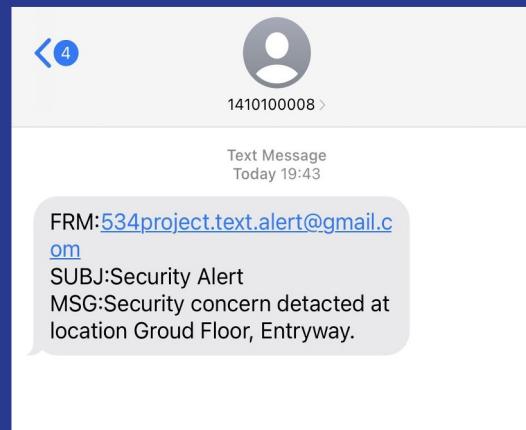
Security area for humans

- Enter unpermitted area  $\rightarrow$  level 3 alarm (Security)

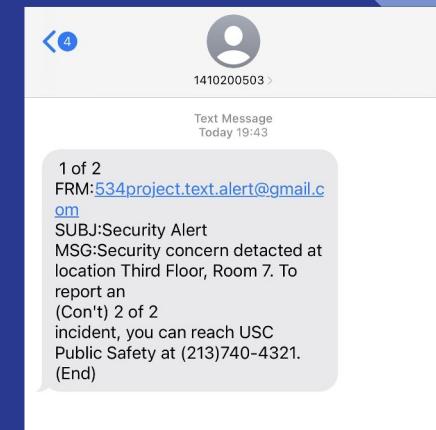
## Abnormal Activity Detection: Anomaly Alert



Level 1 alert messages



Level 2 alert messages



Level 3 alert messages

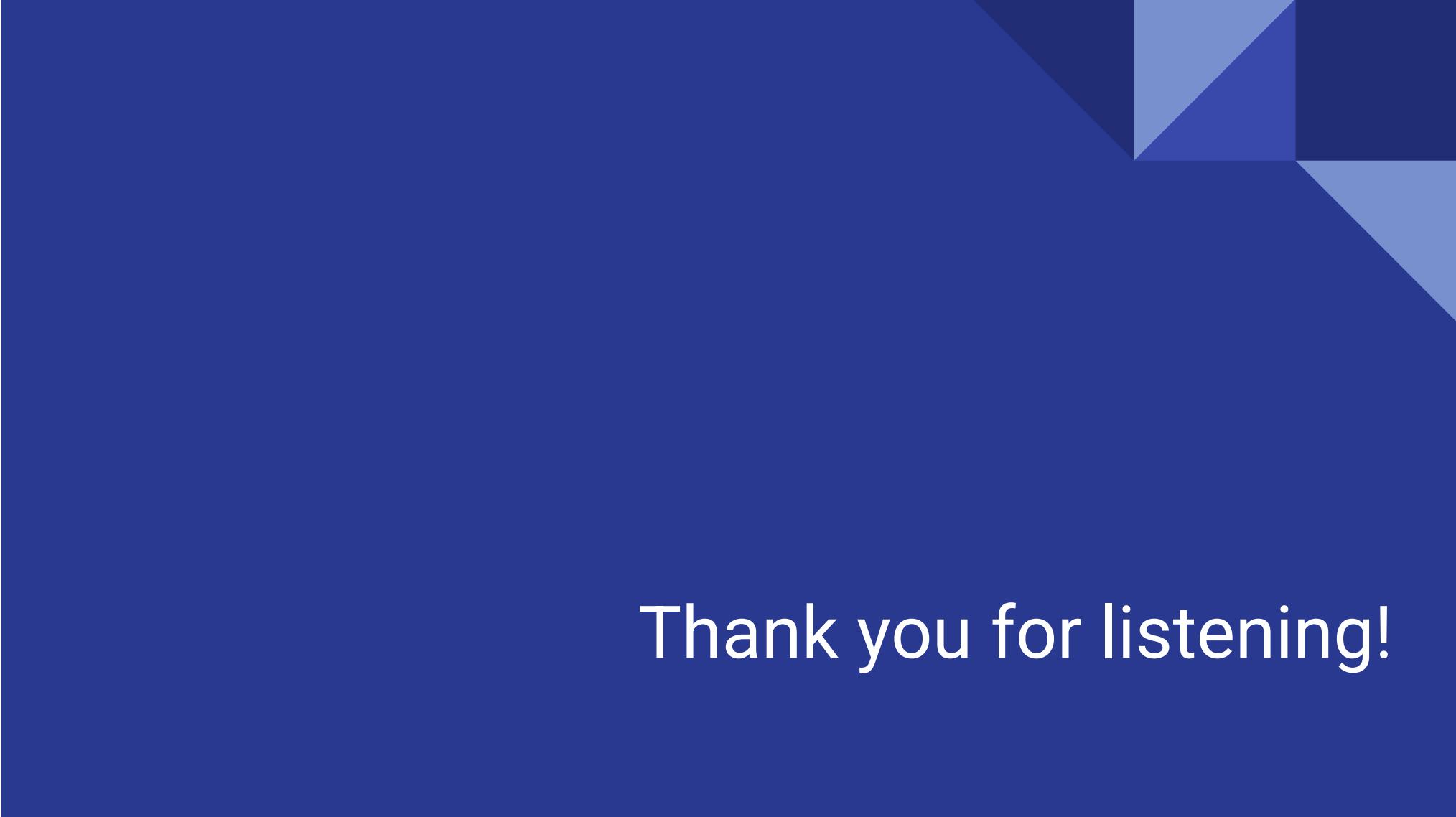
- Connect to WIFI → obtain phone number through SMS Portal authentication
- Security & maintenance are responsible for different areas
- Notify type & location of incidents
- System connects to security cameras
- Our testing model sent alert messages through the built-in python package smtplib, for communication on an enterprise scale, there are other platforms, e.g., Twilio, to provide APIs programmable communication tools

## Abnormal Activity Detection: Third Party Connections



### Cooperating Security Cameras

- Each security camera is responsible for specific grids
- System connects to security cameras
- Send video, type and location of the incidents as needed

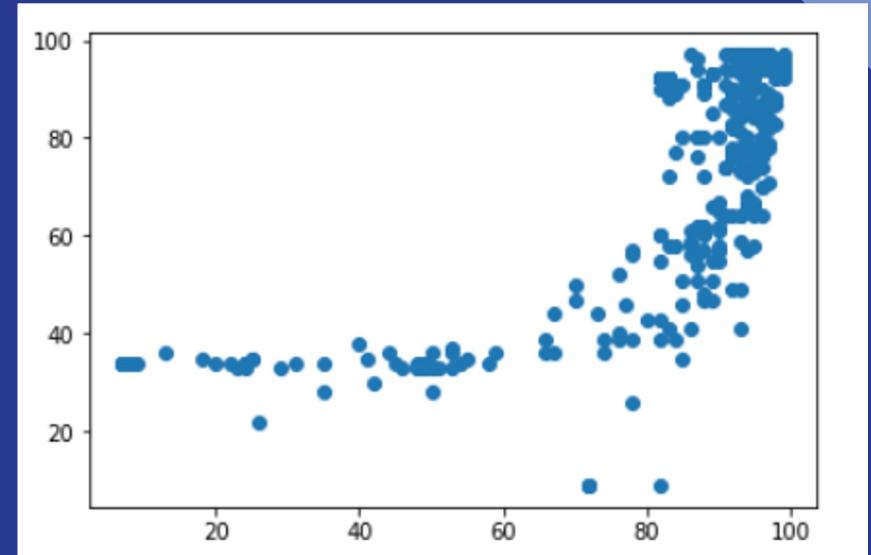


Thank you for listening!

## Appendix

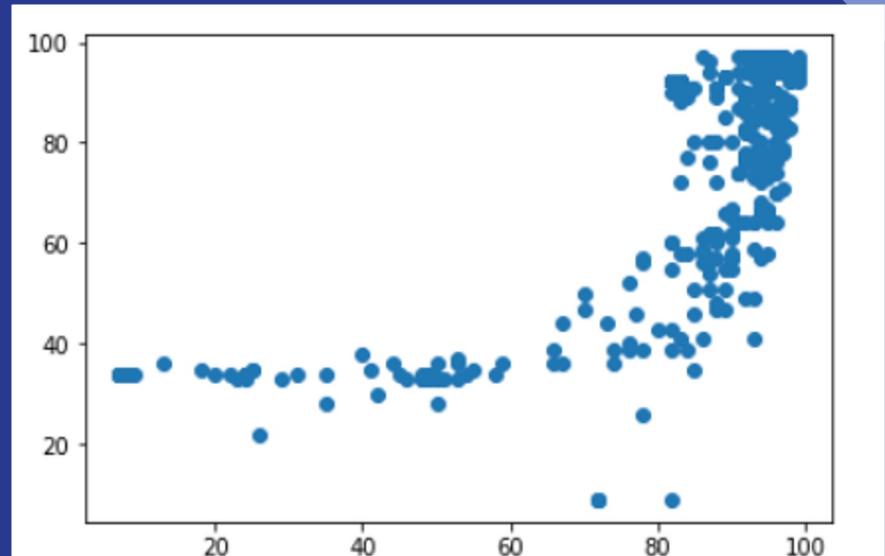
The Cost of Cyber Attacks on Businesses. <https://www.solvereone.com/cost-of-cyber-attacks-on-businesses/>

## Appendix: Levels of Floors and Distributions of Level 0 Machines



Distribution of level 0 machines on floor 1

## Appendix: Levels of Floors and Distributions of Level 0 Machines



Distribution of level 0 machines on floor 2

## Appendix: Levels of Floors and Distributions of Level 0 Machines

