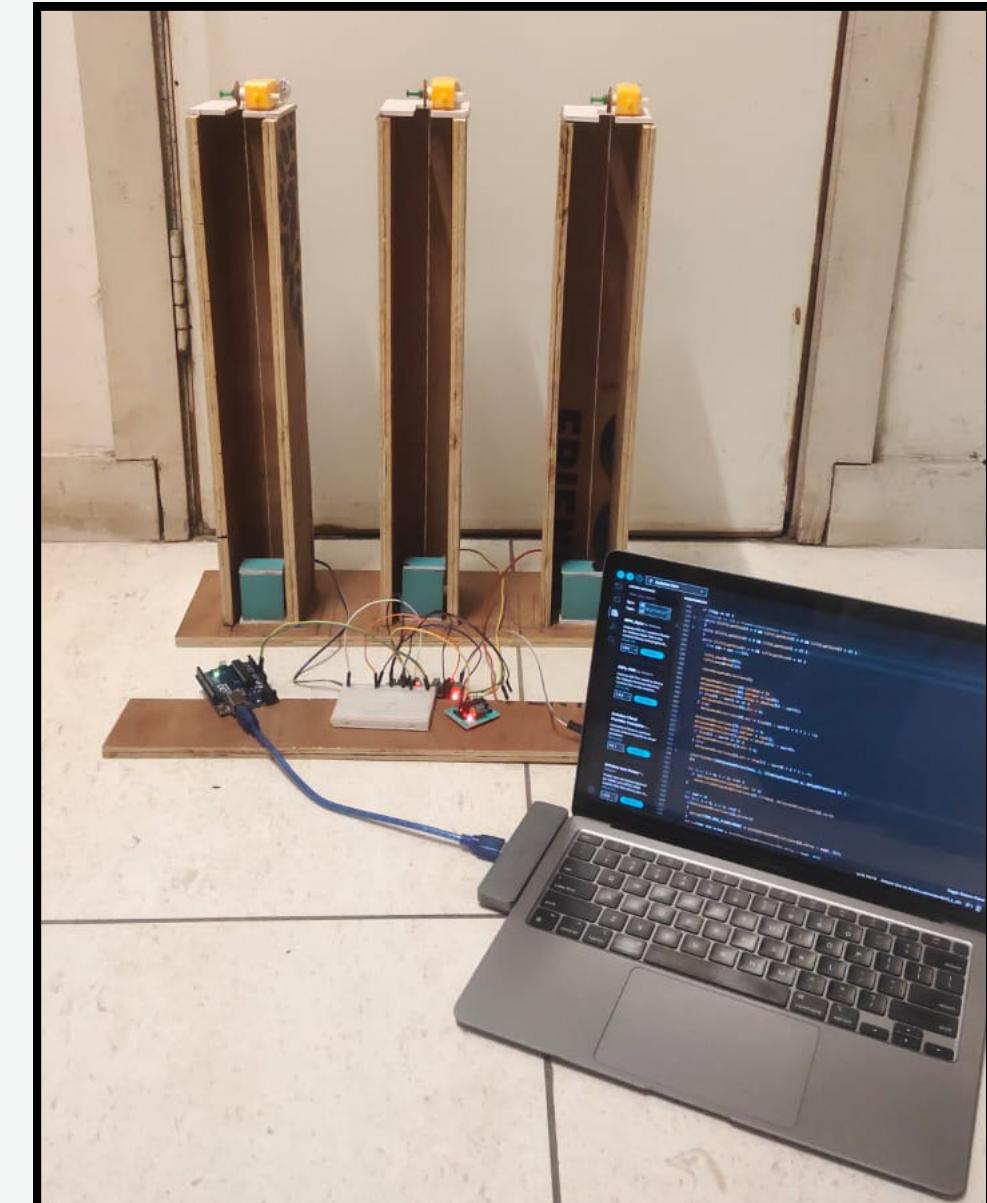


# ELEVATE: Enhancing Elevator Group Management Systems

CPG: 48  
UEC 797: Capstone Project  
CSED Department

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# OVERVIEW

This proposal outlines an innovative approach to elevator systems in a world marked by towering skyscrapers and a fast-paced lifestyle, aiming to transform the urban experience.

The project's primary objective is to balance user experience optimization with environmental responsibility. Prioritizing reduced wait and travel times signifies a commitment to minimizing carbon footprints, thereby contributing to environmental preservation. This dual commitment underscores a dedication to sustainable goals, fostering a more environmentally conscious and responsible future.

## SCOPE AND UTILITY

The implemented algorithms intelligently respond to occupancy levels, user inputs, and traffic patterns, optimizing elevator scheduling and enhancing energy efficiency. The system's scalability and potential for future enhancements promise to usher in new possibilities, making vertical transportation an even more efficient and intelligent aspect of modern building management.

- 
1. Predictive and Proactive Maintenance
  2. Smart Building Integration- control, lighting, and security systems
  3. Energy-saving analytics
  4. User Feedback Mechanism
  5. Accessibility features for visually impaired and handicapped users
  6. Techniques like federation learning to increase security

# OBJECTIVES

- To understand the existing elevator group structure.
- To explore the best machine learning and deep models to optimize elevator group performance.
- To increase the system's efficiency by developing algorithms responding to occupancy levels, user inputs, and traffic patterns routinely used in elevator lobbies.
- To reduce the system's power consumption by incorporating various optimization algorithms.

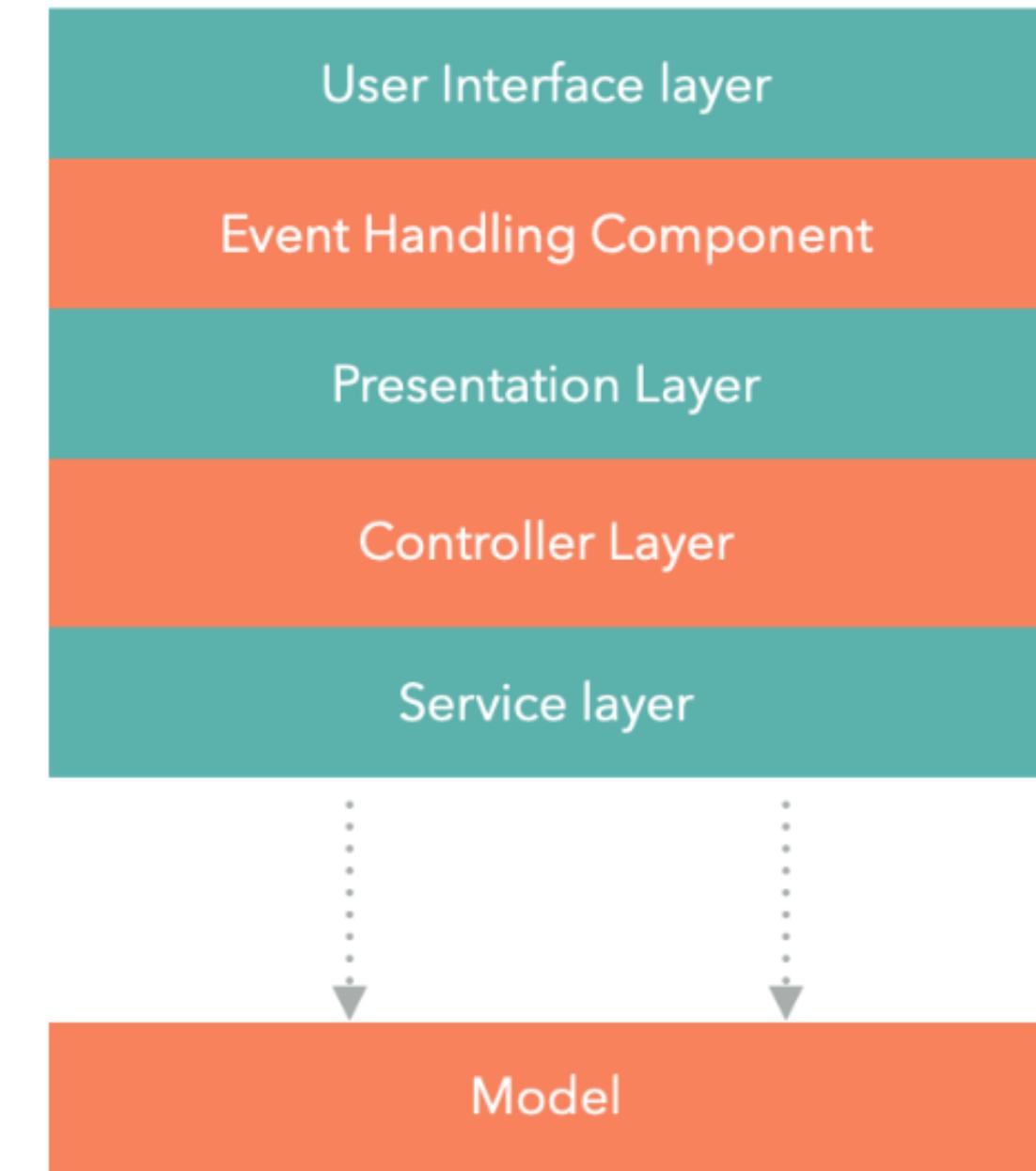
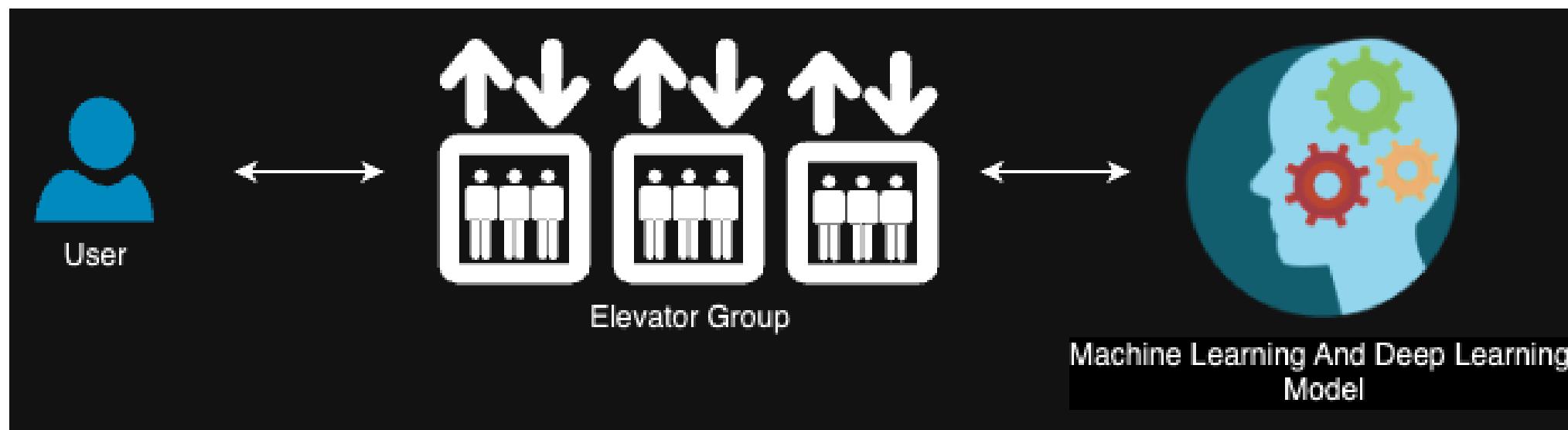
# LITERATURE SURVEY

## Existing Smart Lifts

1. Regenerative Drive Systems (2000s): Captures and reuses energy during descent, enhancing elevator energy efficiency.
2. Mitsubishi Destination Control (2000s): Optimizes elevator routes by grouping passengers with similar destinations for efficient travel.
3. Otis Gen2 Elevator (2000s): Flat-belted steel hoisting system for smoother rides, reduced energy consumption, and space efficiency.
4. Schindler Ahead (2016): IoT-based predictive maintenance system enhancing elevator reliability and minimizing downtime.
5. Thyssenkrupp MULTI (2017): Ropeless elevator using magnetic levitation for vertical and horizontal movement, redefining building design.

# ARCHITECTURE USED

The project considers user interaction and requests and then processes the requests, allocating the best possible elevator based on the ML models for serving the request.



# TECHNIQUES AND TOOLS USED

## Tools

- Arduino UNO
- DC Motors
- L-293D

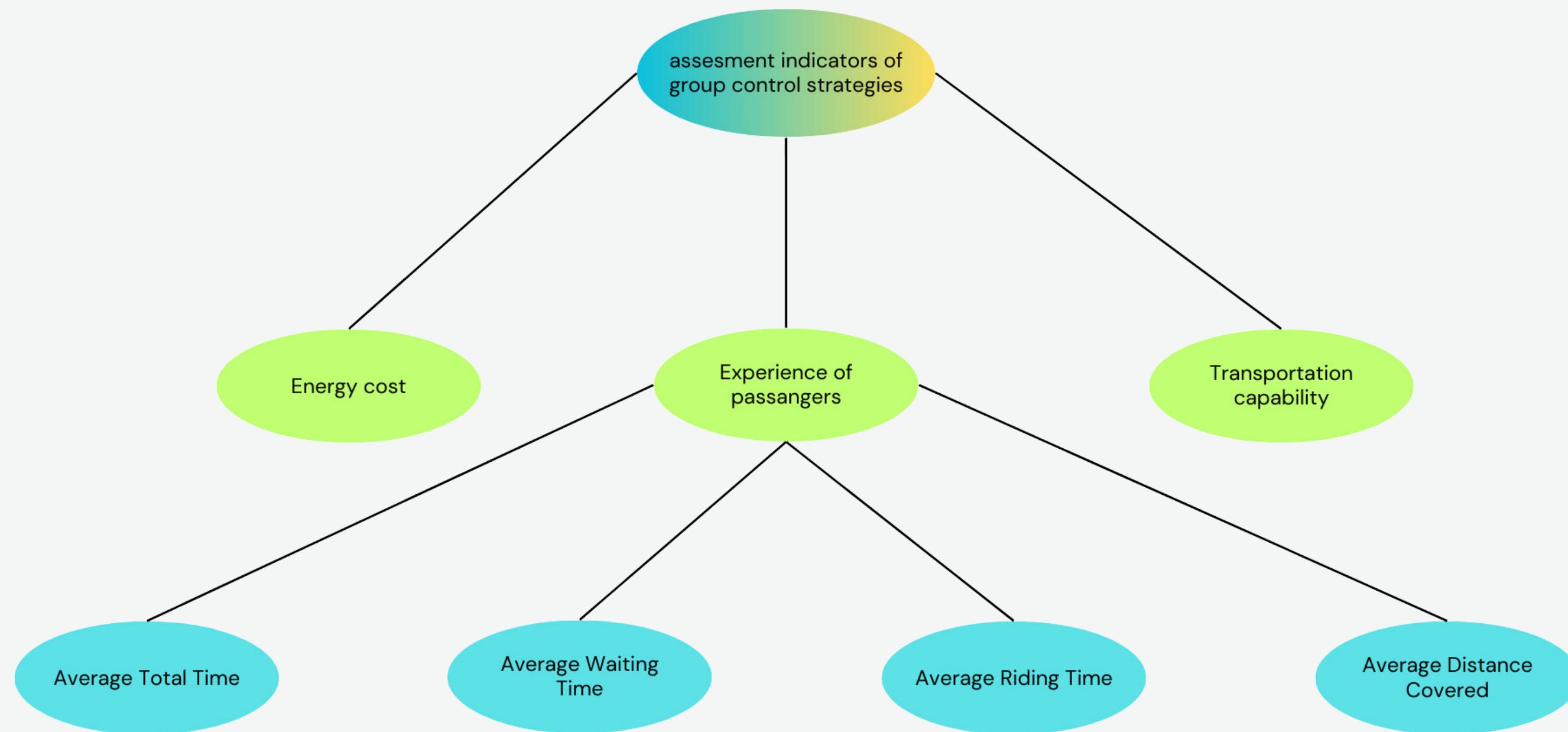
## Algorithms

- Zoning
- First Come, First Serve
- ACO algorithm
- Predictive Dispatching

## Software

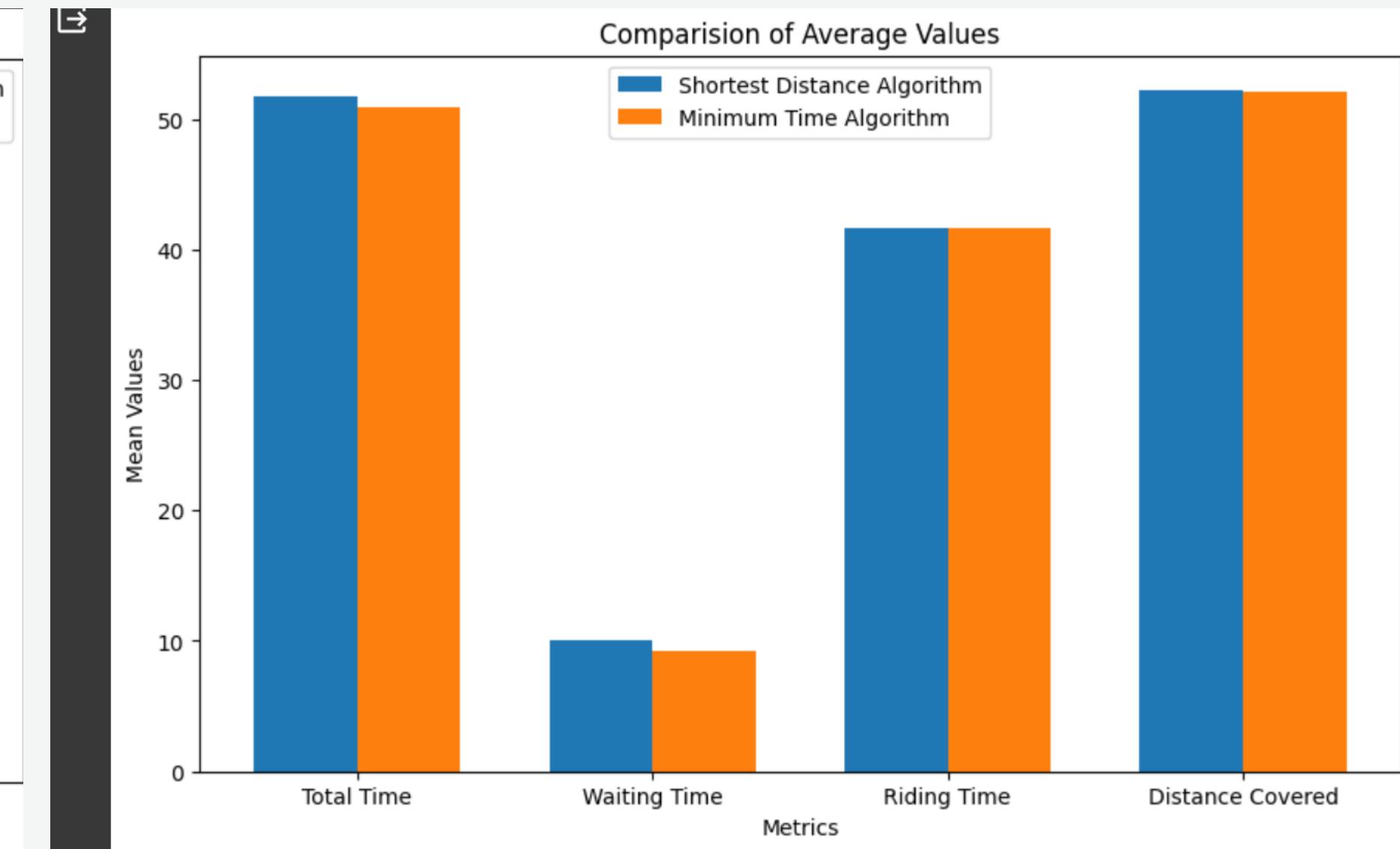
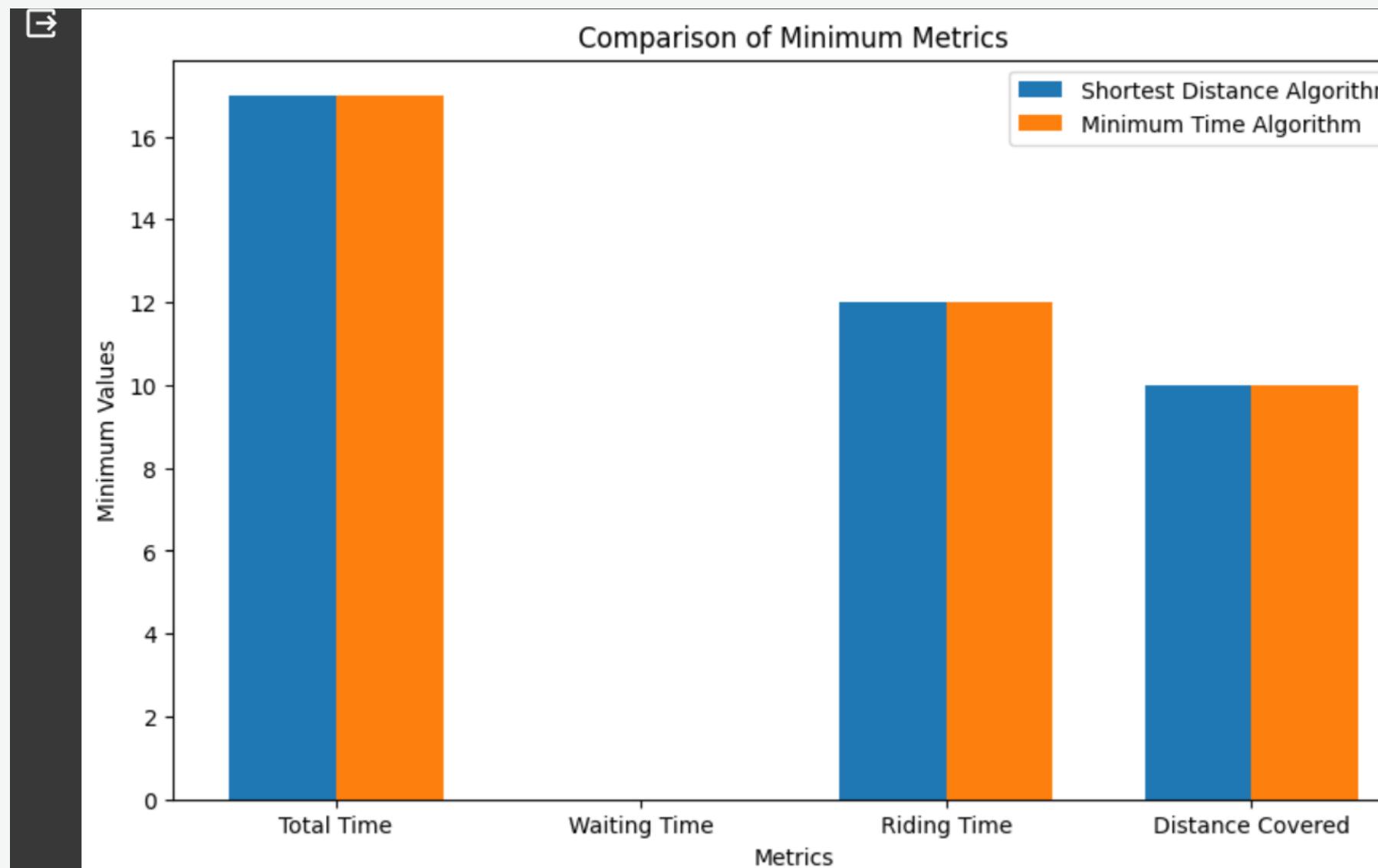
- Arduino IDE
- Google Colab
- Python

# Parameters to Work on

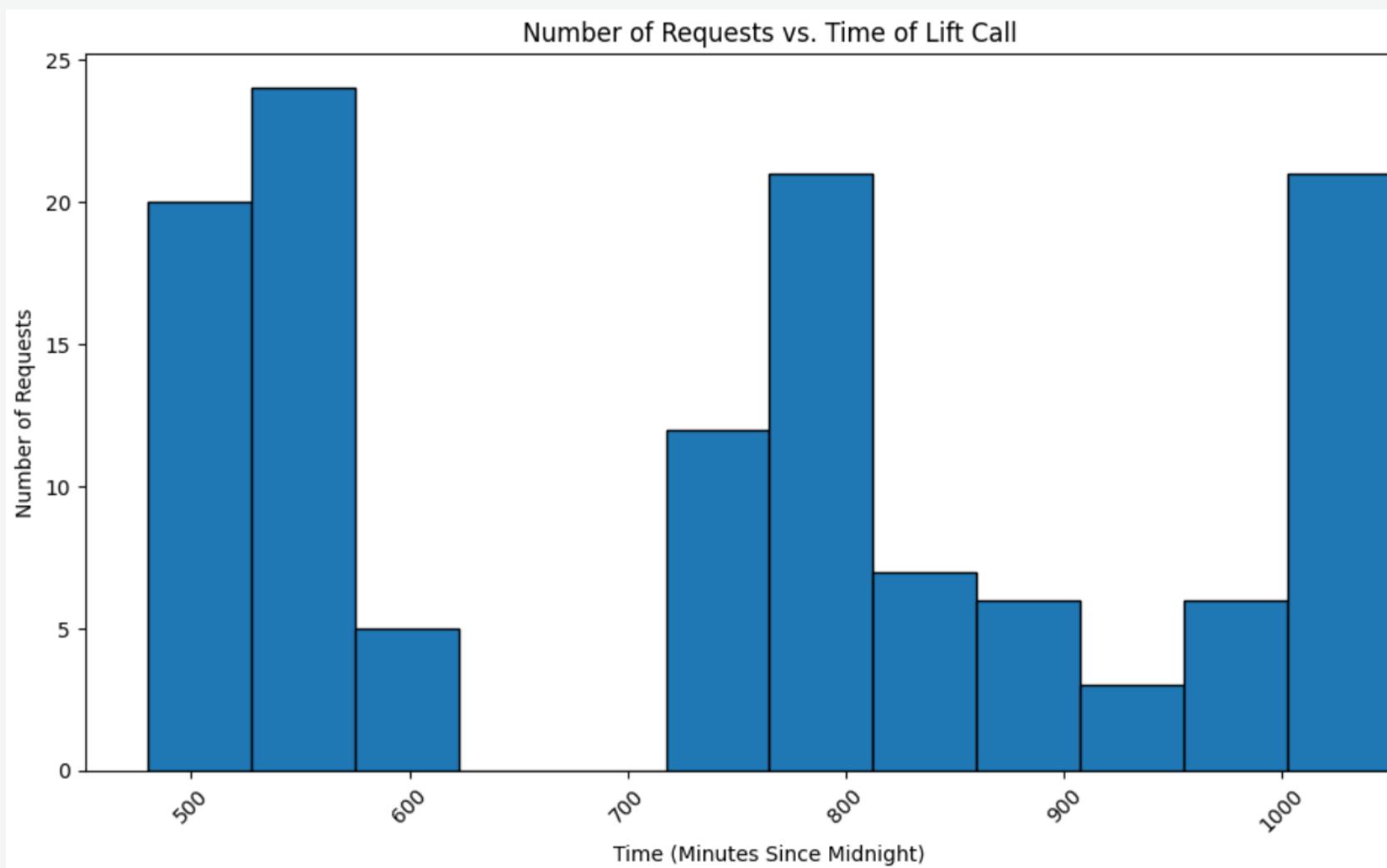


## SNAPSHOTS OF THE PROJECT

# Comparision of Metrics



# Peak Traffic Patterns



**Assessment result of Mornign Peak mode:**

---

Algorithms	Average Total Time	Average Waiting Time	Average Riding Time	Average distance covered	Total Stops
0 Shortest Distance	48.500000	20.833333	27.666667	46.666667	15
1 Minimum Time	40.166667	12.500000	27.666667	38.333333	16

**Assessment result of Morning Normal mode:**

---

Algorithms	Average Total Time	Average Waiting Time	Average Riding Time	Average distance covered	Total Stops
0 Shortest Distance	48.500000	20.833333	27.666667	46.666667	15
1 Minimum Time	40.166667	12.500000	27.666667	38.333333	16

**Assessment result of Lunch mode:**

---

Algorithms	Average Total Time	Average Waiting Time	Average Riding Time	Average distance covered	Total Stops
0 Shortest Distance	49.314286	9.142857	40.171429	49.428571	95
1 Minimum Time	51.200000	11.028571	40.171429	51.714286	105

**Assessment result of Evening mode:**

---

Algorithms	Average Total Time	Average Waiting Time	Average Riding Time	Average distance covered	Total Stops
0 Shortest Distance	49.285714	8.428571	40.857143	50.000000	54
1 Minimum Time	46.523810	5.666667	40.857143	48.095238	54

**Assessment result of After Hours mode:**

---

Algorithms	Average Total Time	Average Waiting Time	Average Riding Time	Average distance covered	Total Stops
0 Shortest Distance	0.0	0.0	0.0	0.0	0
1 Minimum Time	0.0	0.0	0.0	0.0	0

# Zone Control

Zoning Method Criteria	Zone 1	Zone 2	Zone 3
Numerical range of requested floors	Floors 1 to 3	Floors 4 to 6	Floors above 6
Average waiting time for requested floor	Avg. waiting time <= 5 sec	5 sec < Avg. waiting time <= 10 sec	Avg. waiting time > 10 sec
Number of occupants on a floor	0-2 occupants	2-5 occupants	More than 5 occupants
Number of stops made by elevators	0-2 stops	2-4 stops	More than 4 stops
Joint criteria of waiting time and stops	Avg. waiting time <= 5 sec and Avg. stops <= 2	5 sec < Avg. waiting time <= 10 sec and 2 < Avg. stops <= 4	Avg. waiting time > 10 sec or Avg. stops > 4

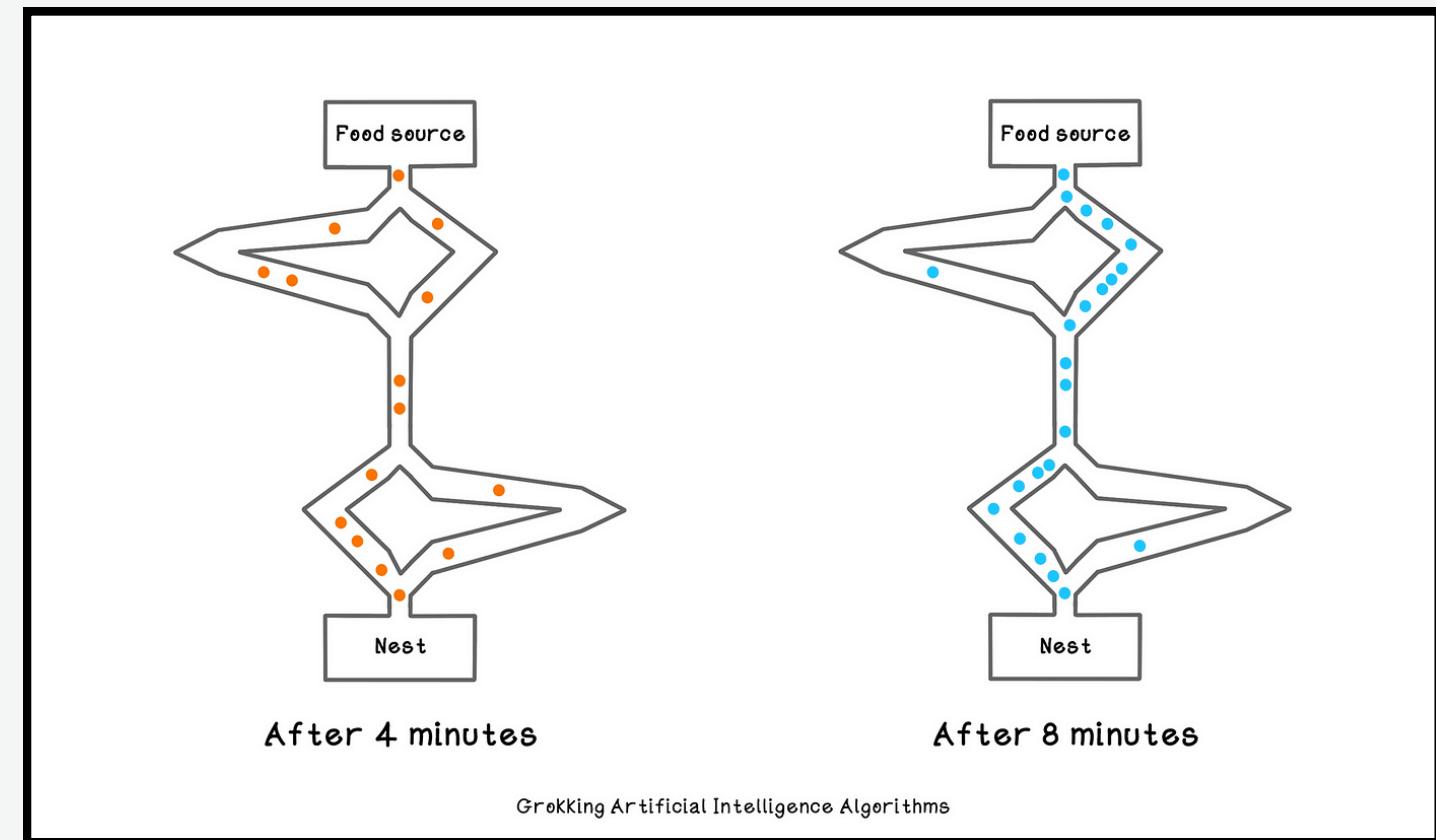
# Predictive Dispatching

```
[156] rfff = df['Requested From Floor'].value_counts().sort_index()
      rtff = df['Requested To Floor'].value_counts().sort_index()
```

```
→ Elevator 1 is moving to floor 3
  Sent elevator 1 to floor 3 with 17 requests.
Elevator 1 is moving to floor 8
  Sent elevator 1 to floor 8 with 17 requests.
Elevator 1 is moving to floor 6
  Sent elevator 1 to floor 6 with 17 requests.
Elevator 1 is moving to floor 1
  Sent elevator 1 to floor 1 with 14 requests.
Elevator 1 is moving to floor 2
  Sent elevator 1 to floor 2 with 14 requests.
Elevator 1 is moving to floor 4
  Sent elevator 1 to floor 4 with 13 requests.
Elevator 1 is moving to floor 7
  Sent elevator 1 to floor 7 with 11 requests.
Elevator 1 is moving to floor 9
  Sent elevator 1 to floor 9 with 11 requests.
Elevator 1 is moving to floor 5
  Sent elevator 1 to floor 5 with 11 requests.
<ipython-input-156-8243772315c1>:6: FutureWarning: iteritems is deprecated and will be removed in a future version. Use .items instead.
    for floor, count in floor_requests.iteritems():
```

# Ant Colony Optimisation

```
AntColonyOptimization:  
def __init__(self, elevator_system, requests, iterations=10, evaporation_rate=0.5, alpha=1.0, beta=2.0):  
    self.elevator_system = elevator_system  
    self.requests = requests  
    self.iterations = iterations  
    self.evaporation_rate = evaporation_rate  
    self.alpha = alpha  
    self.beta = beta  
  
def run(self):  
    for _ in range(self.iterations):  
        for request in self.requests:  
            floor, requesting_elevator = request  
            self.elevator_system[requesting_elevator].request_floor(floor)  
  
        for elevator in self.elevator_system.values():  
            elevator.operate()  
  
        # Update pheromones based on the efficiency of each elevator  
        total_waiting_time = sum(elevator.current_floor for elevator in self.elevator_system.values())  
        pheromones = np.ones((len(self.elevator_system), len(self.elevator_system)))  
        for elevator_id, elevator in self.elevator_system.items():  
            efficiency = 1 / (elevator.current_floor + 1)  
            pheromones[:, elevator_id - 1] *= (1 - self.evaporation_rate)  
            pheromones[:, elevator_id - 1] += (efficiency + 1) * self.alpha  
  
        # Select the elevator with the highest pheromone level for each request  
        for request in self.requests:  
            floor, requesting_elevator = request  
            probabilities = pheromones[:, requesting_elevator - 1] # Adjust indexing  
            selected_elevator = np.argmax(probabilities)  
            self.elevator_system[selected_elevator + 1].request_floor(floor) # Adjust indexing
```



# BalanceTrade Off



Assessment result based on tradeoff:

	Algorithms	Average Total Time	Average Waiting Time	Average Riding Time	Average distance covered
0	Shortest Distance	51.968	10.320	41.648	52.24
1	Minimum Time	50.872	9.224	41.648	51.76
2	General Tradeoff	51.968	10.320	41.648	51.60
3	Tradeoff based passsanger flow	52.280	10.632	41.648	53.20

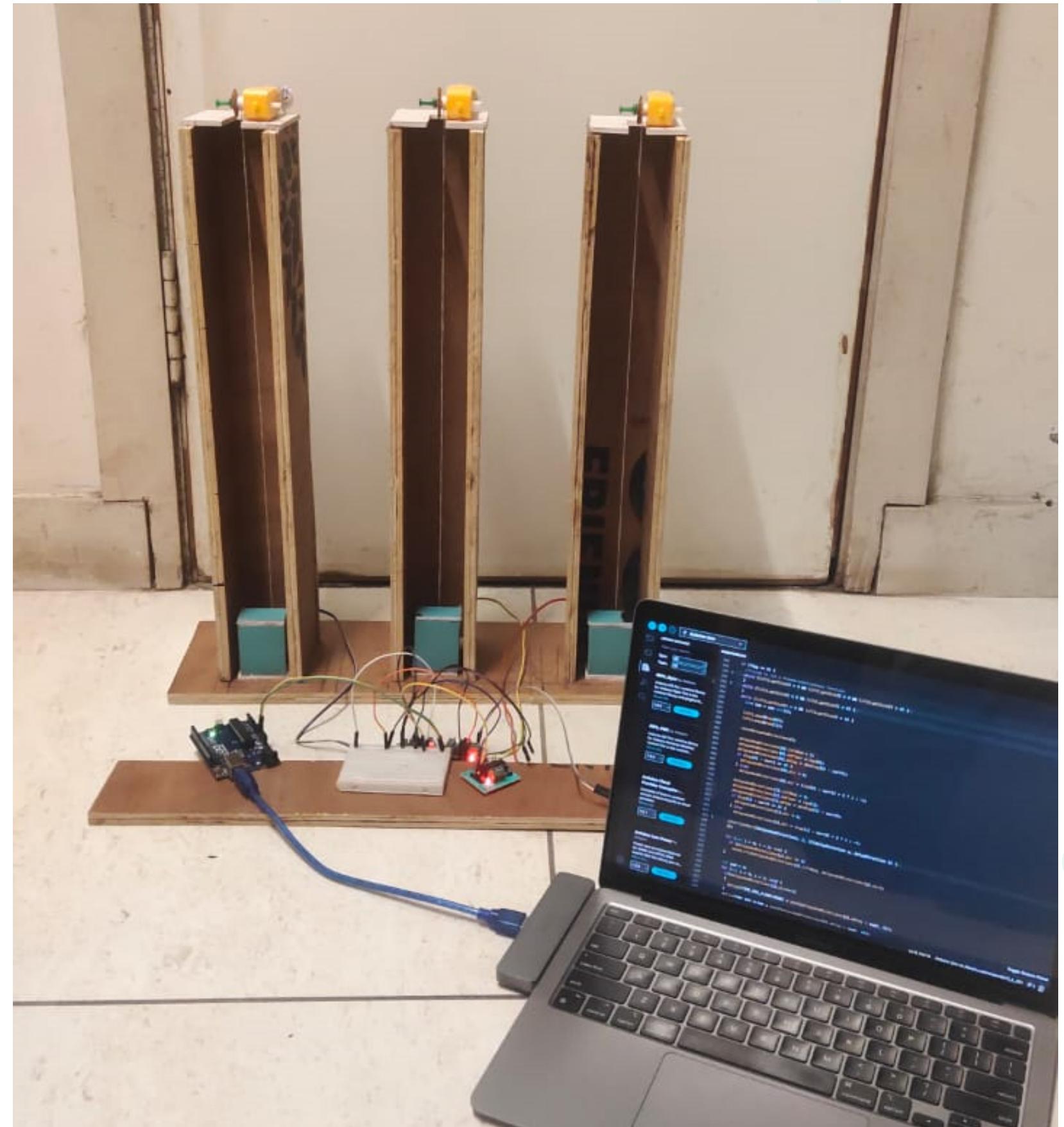
Passenger  
satisfaction

Shortest distance optimisation  
vs  
Minimum wait time optimisation

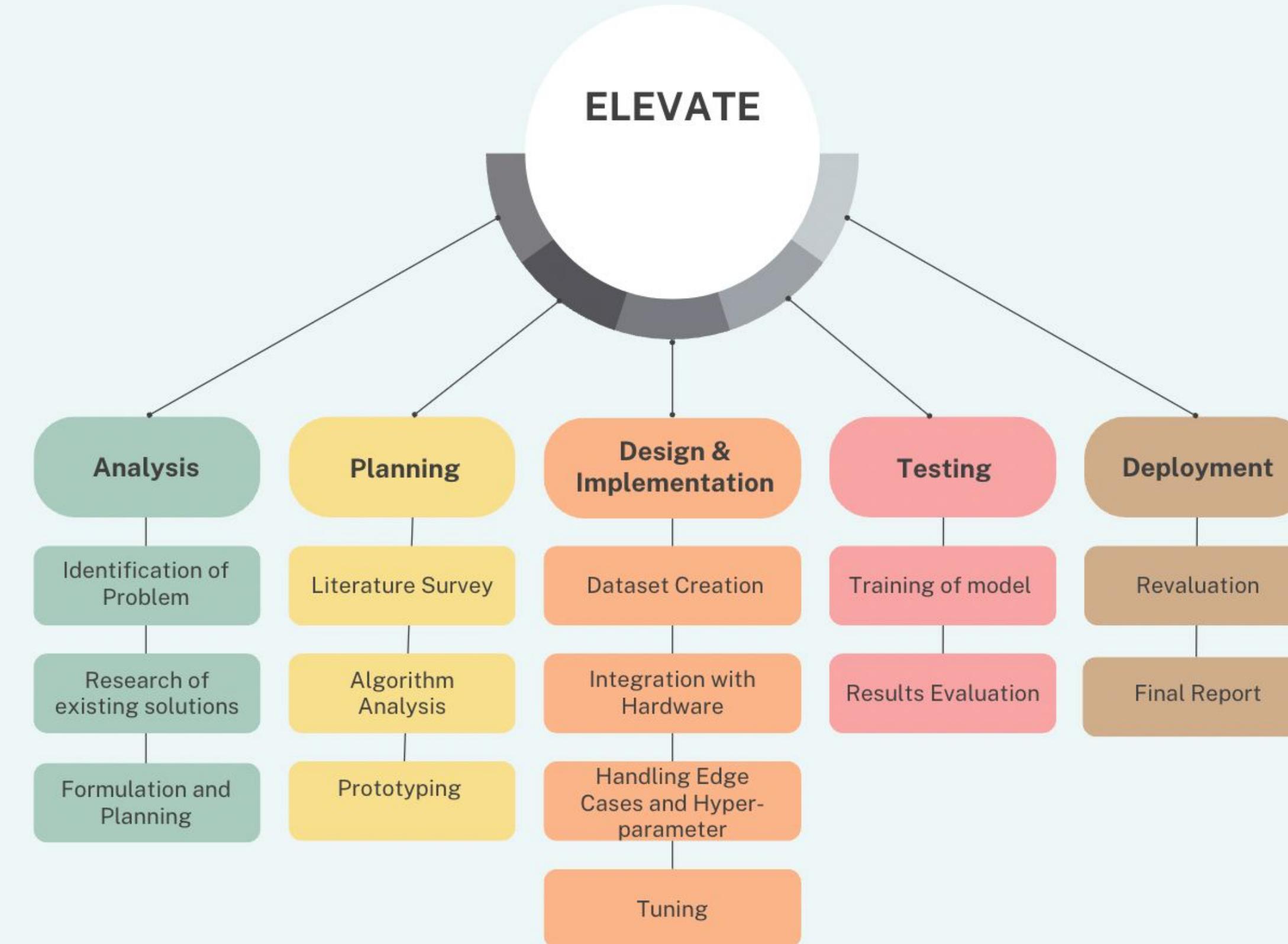
Energy  
efficiency

# HARDWARE

The hardware simulation uses Arduino and 3 DC motors to replicate the Lift Lobby. L239D ICs have been used to facilitate the movement of engines in both directions; efforts have been made to make the system near perfect.



# METHODOLOGY



# PROFESSIONAL AND TECHNICAL LEARNING

All the team members learned the importance of punctuality in attending the group meetings to discuss the upcoming responsibilities. The team was also regularly at the evaluation destination for panel and mentor evaluation.

Apart from professional discipline, the team learned new tools like working with IOT-related boards like Arduino and associated hardware; the group also enjoyed implementing what we learned in our lecture classes, like the H-bridge circuit.

We also enjoyed exploring domains like deep learning and industry-standard optimization technique.

# INDIVIDUAL ROLES

Names	Software Implementation	Research	Documentation	Hardware implementation
Jahnvi Gangwar (Leader)	✓	✓	✓	✓
Veer Daksh Agarwal	✓	✓	✓	✓
Shamayla Jindal	✓	✓	✓	✓
Chahat Joneja	✓	✓	✓	✓
Mukul Singhal	✓	✓	✓	✓

# **THANK YOU!!**