

# Facility Location Optimization Report

## Introduction

This project tackles a facility location optimization problem, aiming to minimize costs and travel times for locating a central facility in Atlanta, GA. The analysis leverages geospatial techniques and Google Maps APIs to compare two approaches: Haversine distance (straight-line) and driving distance (actual travel).

## Objectives

1. Identify the optimal location for a facility based on cost minimization.
2. Compare results using Haversine and driving distance methods.
3. Visualize the optimal location and routes on an interactive map.

## Methodology

1. **Data Collection:**
  - Locations in Atlanta, GA with associated weights and fixed costs.
  - Geographic coordinates retrieved using the Google Maps API.
2. **Haversine Distance Optimization:**
  - Calculated "as-the-crow-flies" distances between points.
  - Used Scipy's `minimize` function to determine the optimal location.
3. **Driving Distance Optimization:**
  - Leveraged Google Maps Directions API to compute actual travel distances.
  - Used Scipy's `brute` function for integer-based optimization.
4. **Visualization:**
  - Mapped the optimal locations and routes using Folium and Google Maps.

## Results

### Haversine Distance Method

- **Optimal Location:** 618 Cresthill Ave NE, Atlanta, GA 30306, USA.
- **Coordinates:** [33.7837, -84.3674].
- **Minimum Cost:** \$192,150.77.

### Driving Distance Method

- **Optimal Location:** 1470 Ashwood Way, Lawrenceville, GA 30043, USA.
- **Coordinates:** [34.0, -84.0].
- **Minimum Cost:** \$2,954,751.00.

## Comparative Analysis

- The driving distance method accounts for real-world travel constraints, leading to higher costs compared to the Haversine approach.
- Both methods provide actionable insights for facility placement but differ in precision based on real-world constraints.

## Discussion

- **Advantages:**
  - Haversine distance offers quick estimates for facility placement.
  - Driving distance provides realistic results suitable for practical implementation.
- **Challenges:**
  - Haversine distance ignores road networks and travel conditions.
  - Driving distance depends on API accuracy and incurs computational overhead.

## Recommendations

1. Use Haversine distance for preliminary analysis to identify potential locations.
2. Apply driving distance optimization for final decision-making in practical scenarios.
3. Regularly update location data and parameters to reflect changes in road networks and demand patterns.

## Conclusion

This project demonstrates the value of geospatial optimization in solving facility location problems. By combining mathematical models with real-world data, businesses can make informed decisions to minimize costs and improve operational efficiency.

## References

- Google Maps API for geocoding and driving distances.
- Python libraries: Pandas, Scipy, Folium.