

# Using Location-based Data in Regional Parks Visitors Research

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# Agenda

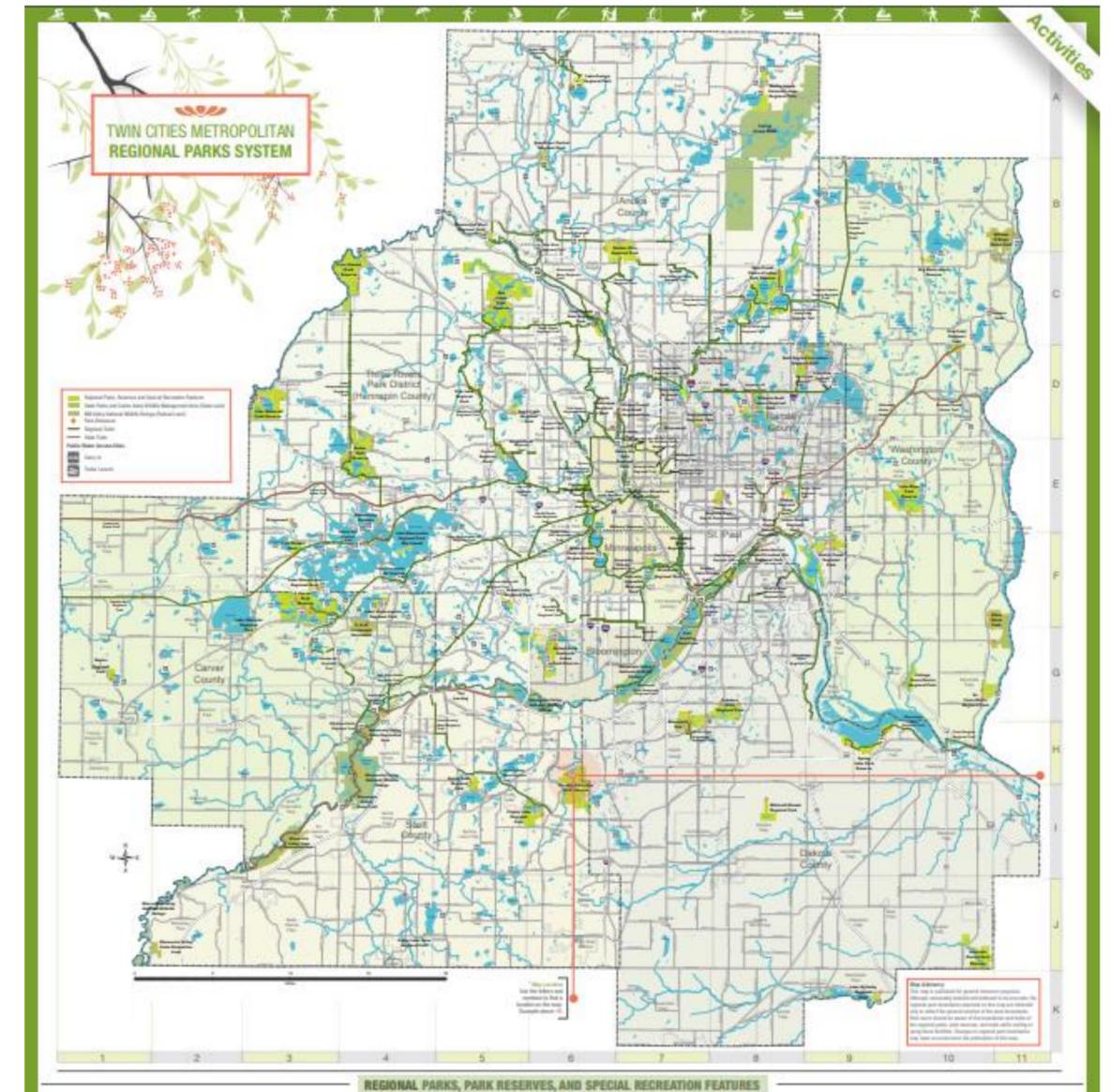
- Introduction and background
  - Regional Parks
- Location-based services
  - StreetLight data characteristics and limitations
- Case study: Which parks are most popular?
- Case study: Measuring activity within Como Regional Park

**Location-based data challenges the way we research regional parks.**

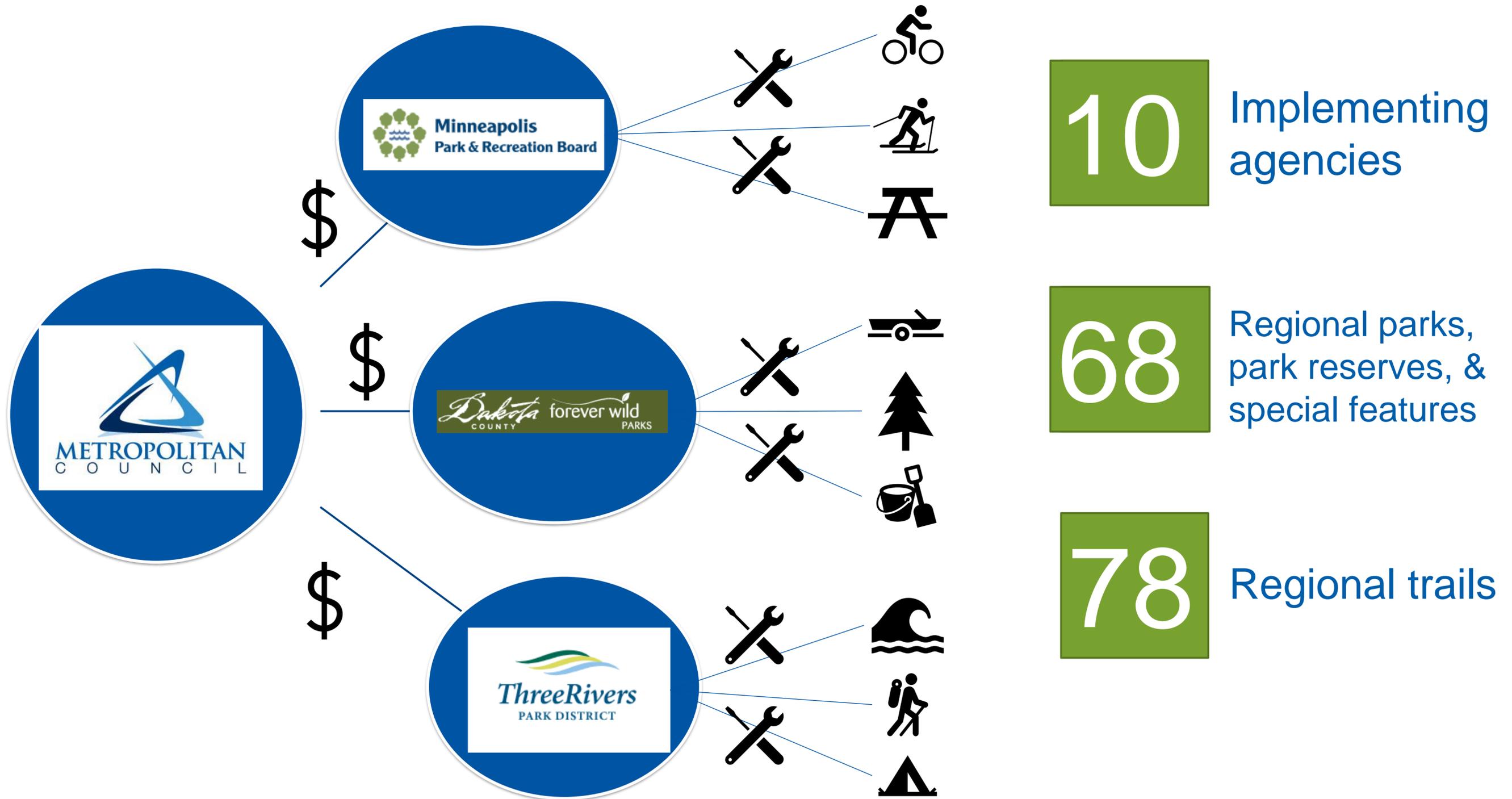
# Why regional parks?

# Why regional parks?

- The Metropolitan Council is required by law to estimate how many people visit the regional parks system annually
- Regional parks are owned and managed by implementing agencies
- There are 10 regional parks implementing agencies



Map: [Metropolitan Council](#)



# Most Popular Parks

Minneapolis Chain of Lakes



Image: [Minneapolis Park and Recreation Board](#)

Como

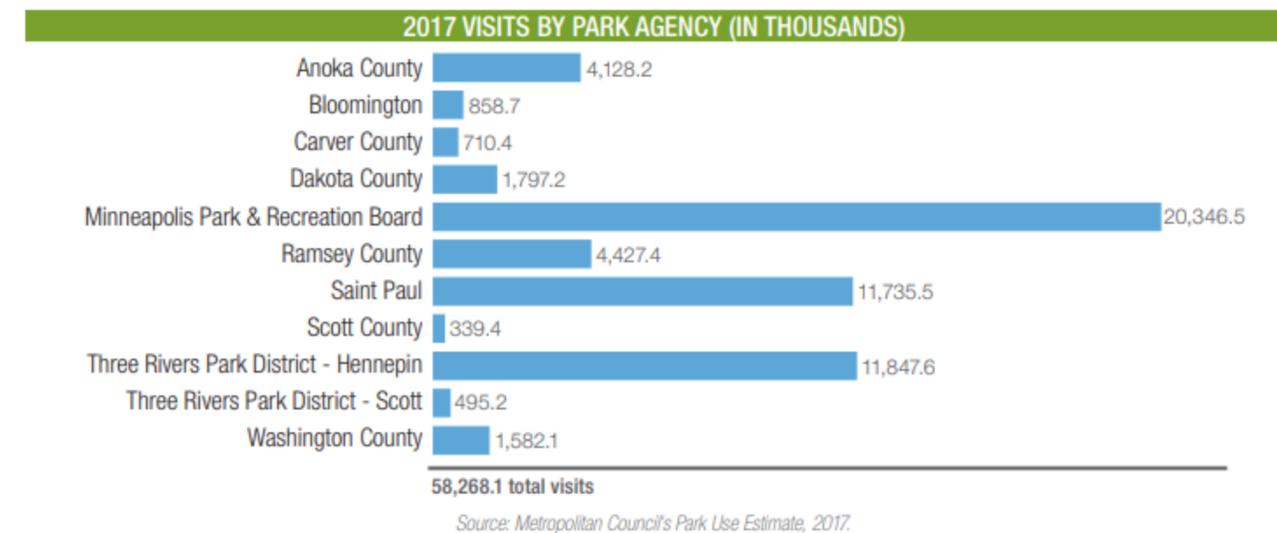


Image: [CityPages](#), 2019

How do you know?

# How we usually research visitors

- Use estimates
  - Representative at the agency level
- Visitor Studies
  - In-person sampled surveys
  - Representative at the agency level
- Significant limitations
  - Every park has its unique features and geography
  - Self-selection survey bias
  - Funding restrictions



Image, chart: [Metropolitan Council](#)

# Preliminary 2018 Estimated Visits

Minneapolis Chain of Lakes

7.3 million

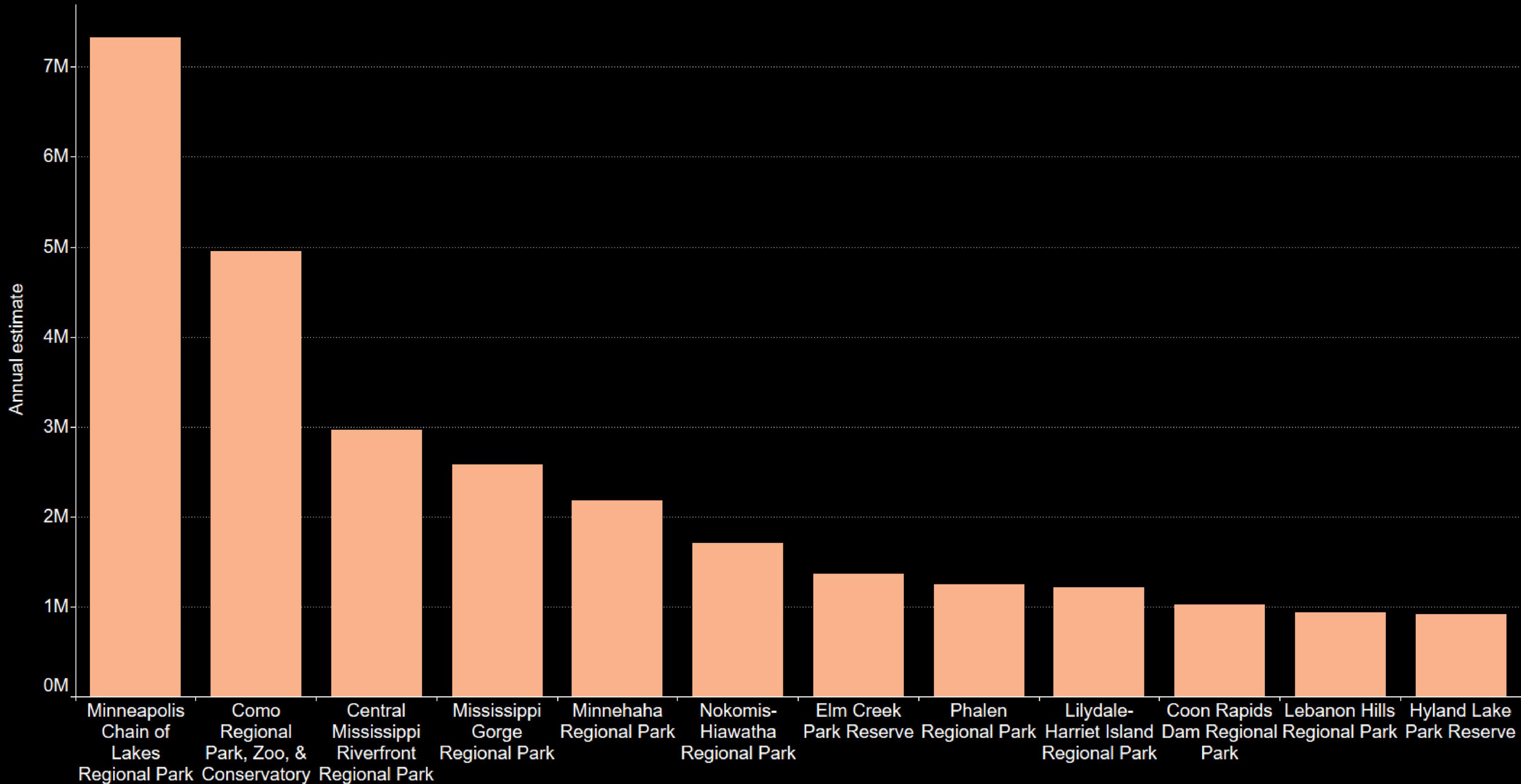


Como

4.9 million



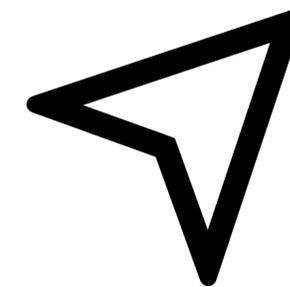
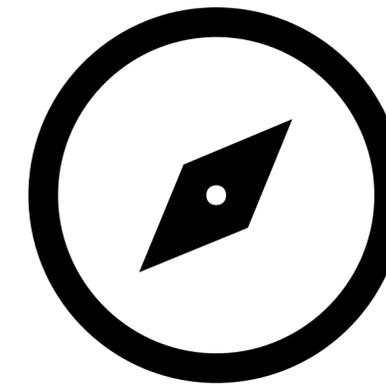
# 2018 Annual Use Estimate



# Why location-based services?

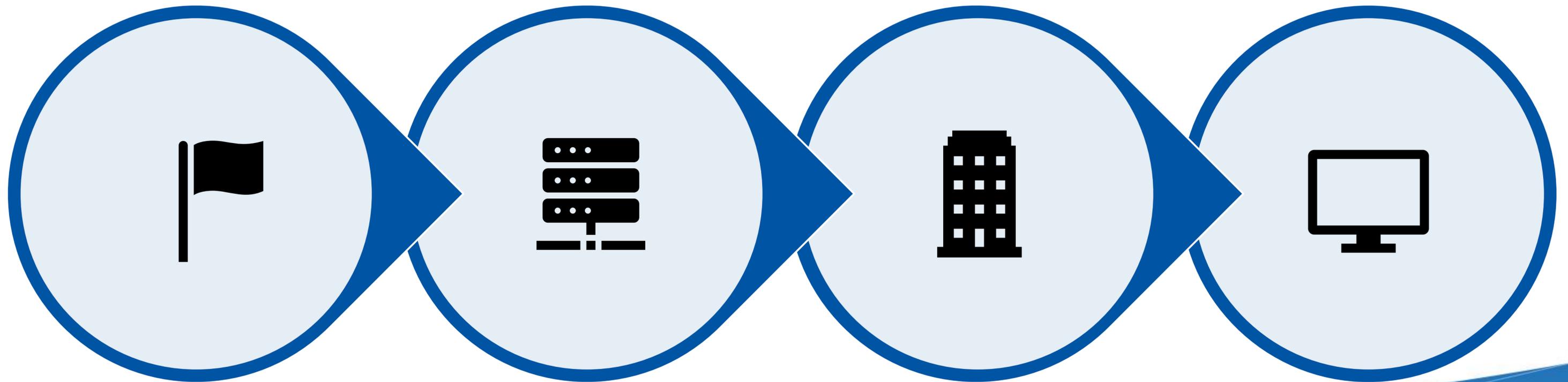
# Take out your phone!

- Go to your settings
  - Or try swiping down from the top of your screen, and look for an icon like these
- If your location is turned on, then your data is being collected by *someone*
- Apps and services use your location



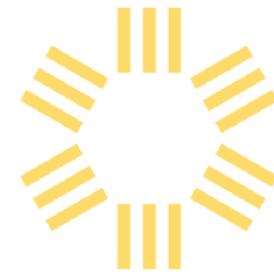
# So what is LBS data?

- LBS data is information about a device's location in space and time



# StreetLight Data

- Subscription through partnership with MNDOT
- Provides platform for running individual “projects” on trips
- Gathered from apps, such as shopping, dating, weather, productivity
- Accurate to 20 meters (65 feet)



STREETLIGHT  
DATA

*“StreetLight Data is the first company to make using real-world transportation data easy, efficient, and affordable.”*

Image, quote: [“About Us” page](#)

# What is a trip?

- A trip must be greater than 500m and longer than 3 minutes
- Trips can be
  - Trip Start
  - Trip End
  - Trip Pass-Through
  - Trip All

# StreetLight data characteristics

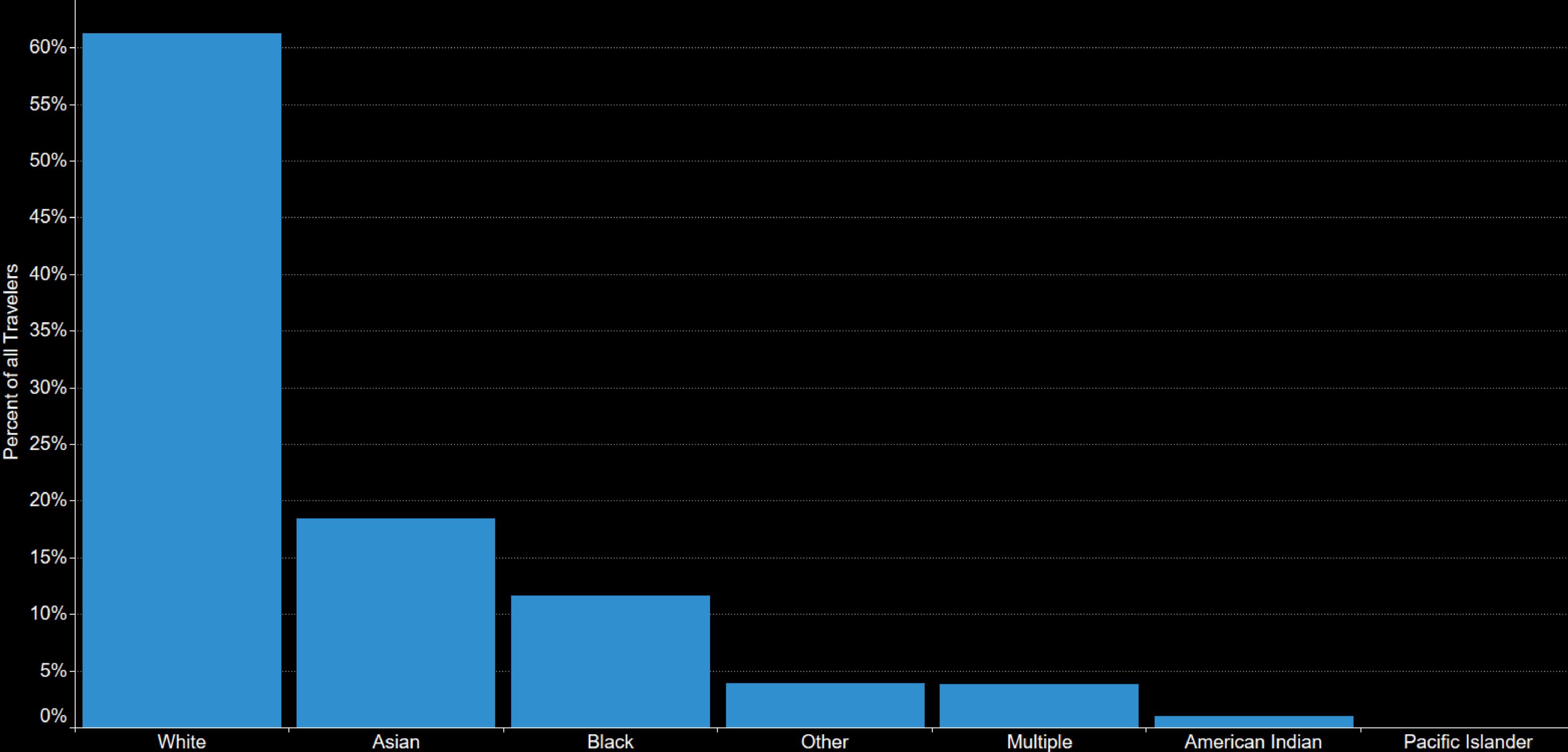
- **StreetLight Traffic Index** a normalized measure of the relative traffic in a given area
- Specific time periods, down to the day (3 months recommended)
  - Day type (i.e. weekday, weekend)
  - Day part (i.e. retail hours, morning, afternoon)
- Large sample size
- Bike and pedestrian (limited)

# StreetLight data characteristics (cont.)

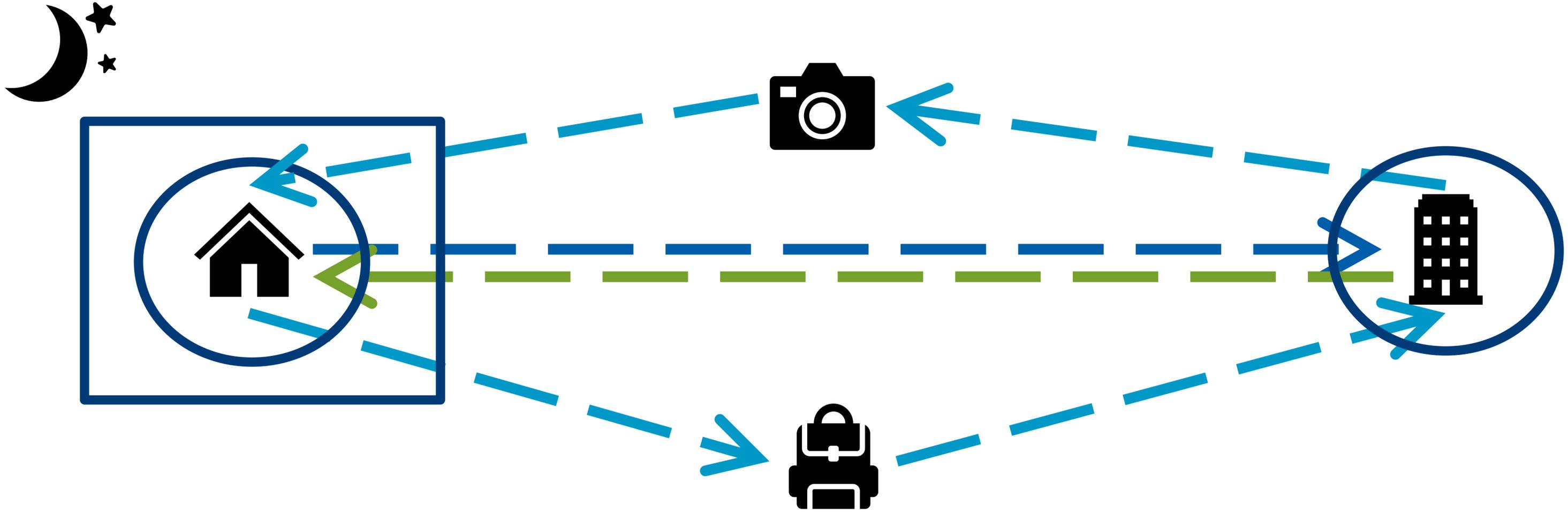
- Inferred traveler demographics
  - Race
  - Household income
  - Ethnicity
  - Family status (children, no children)
  - Education

# Phalen - Traveler Race

## Retail Hours 2018



# A weekday

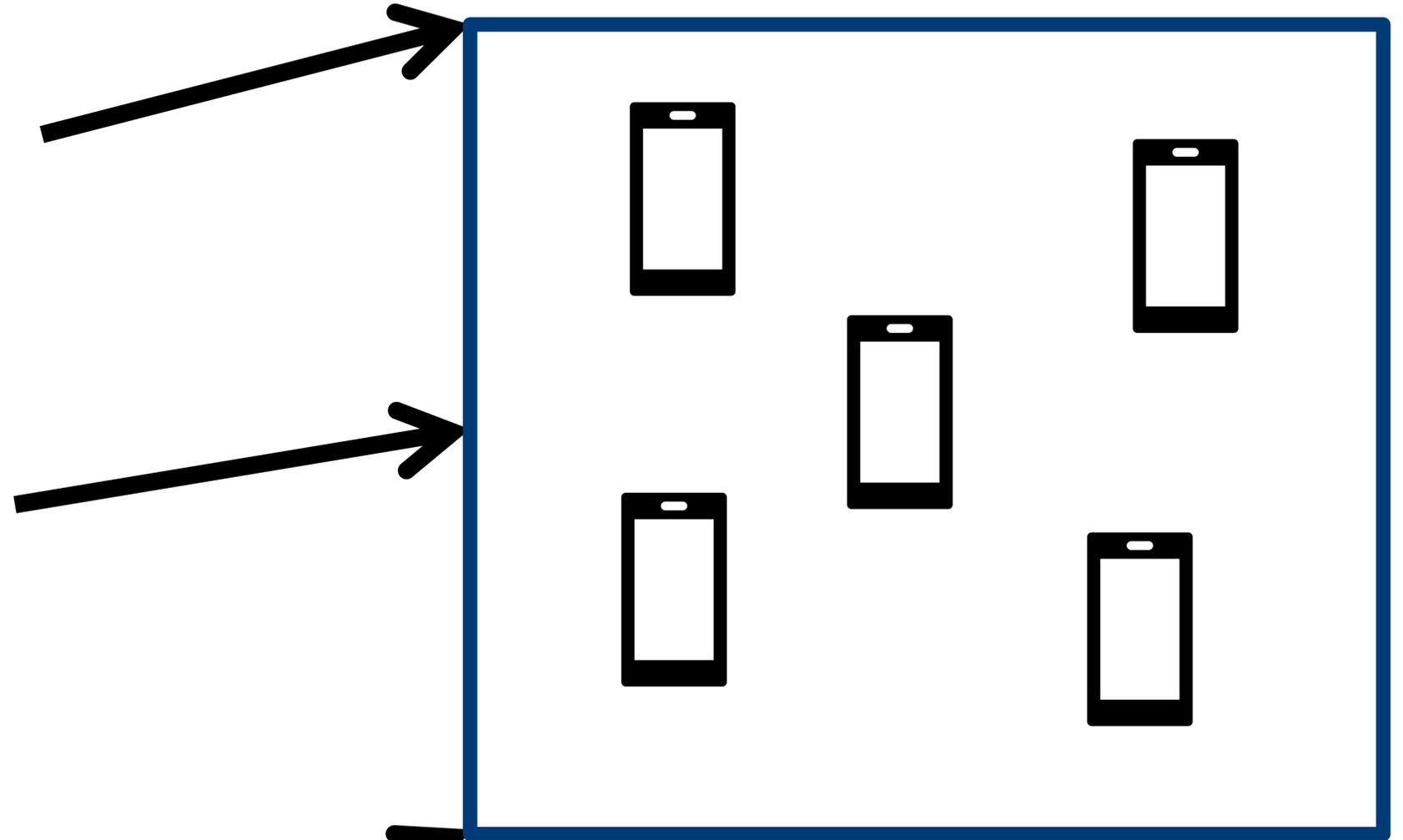


StreetLight uses other data sources to determine the demographics for a given area

Decennial census

Land use

American Community Survey



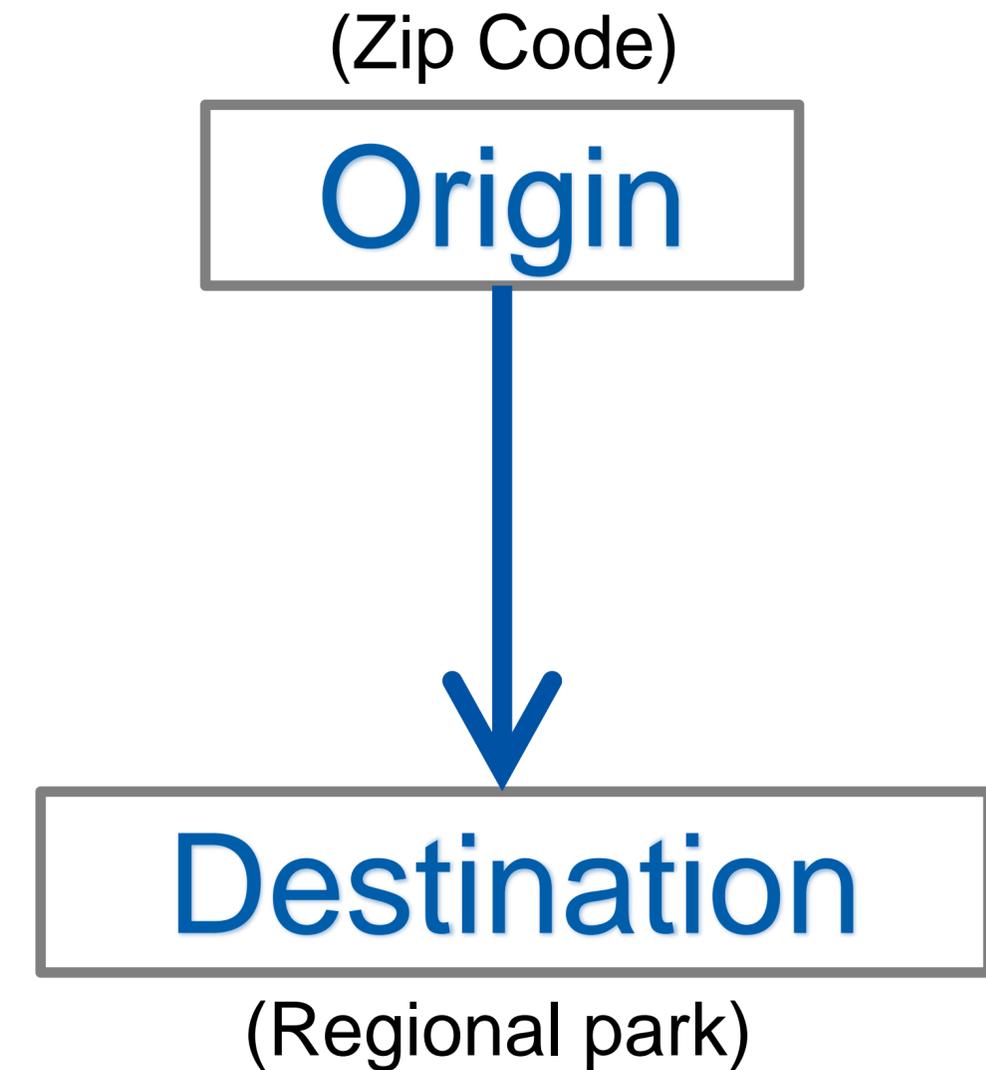
# StreetLight data limitations

- Disparities in cell phone ownership
- Bias in what kinds of apps groups of people use
- Spatial accuracy
- StreetLight Traffic Index “black box”

**Case study:**  
**Which parks are most popular?**

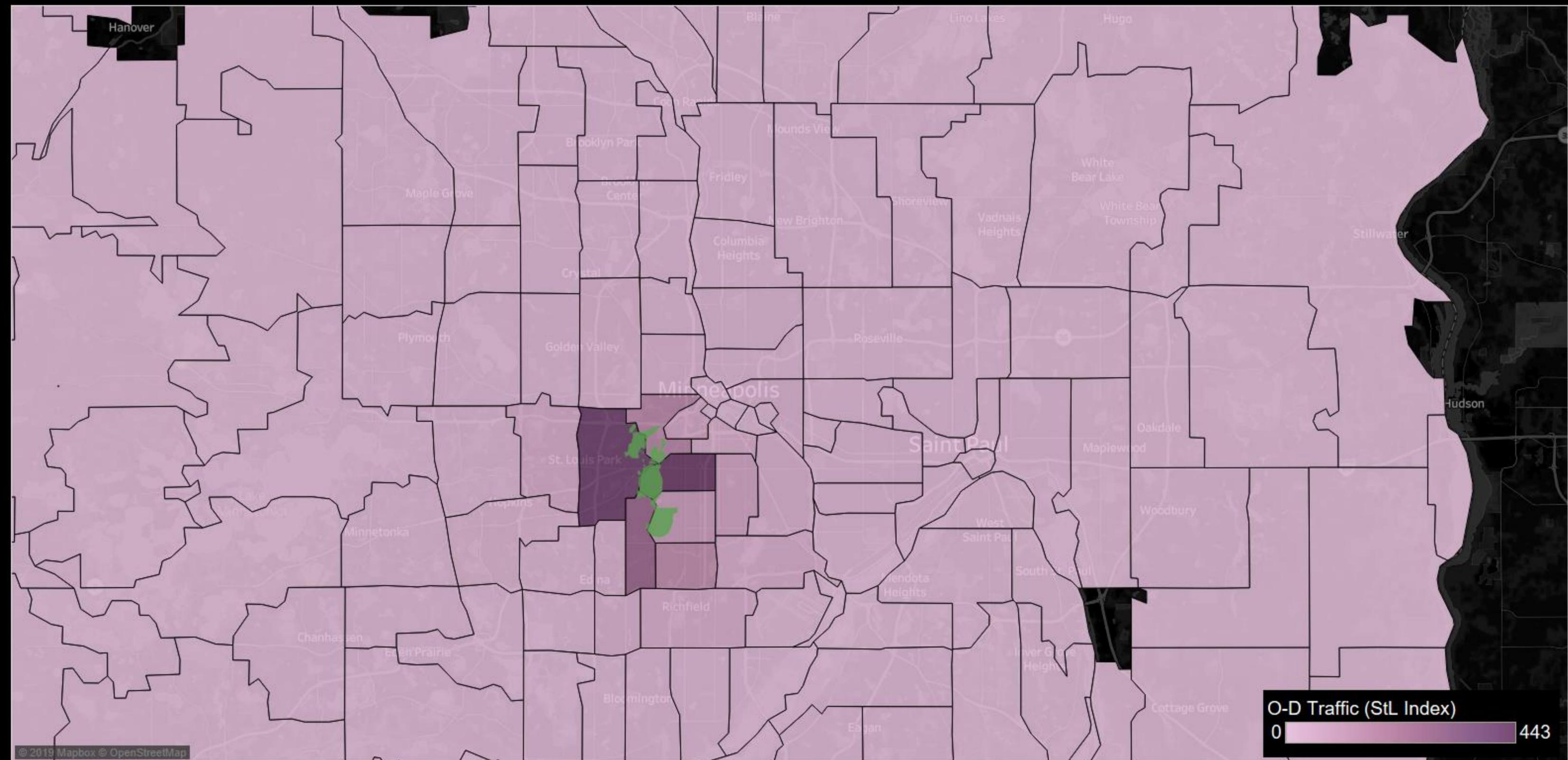
# Origin-Destination Analysis

- Measures the relative traffic to each regional park from each zip code
- Includes Trips that end in the park
- Filter data to only include Average Day (M-Su), retail hours (6am-10pm), all 2018
- 84,000 devices
- 207,000 trips



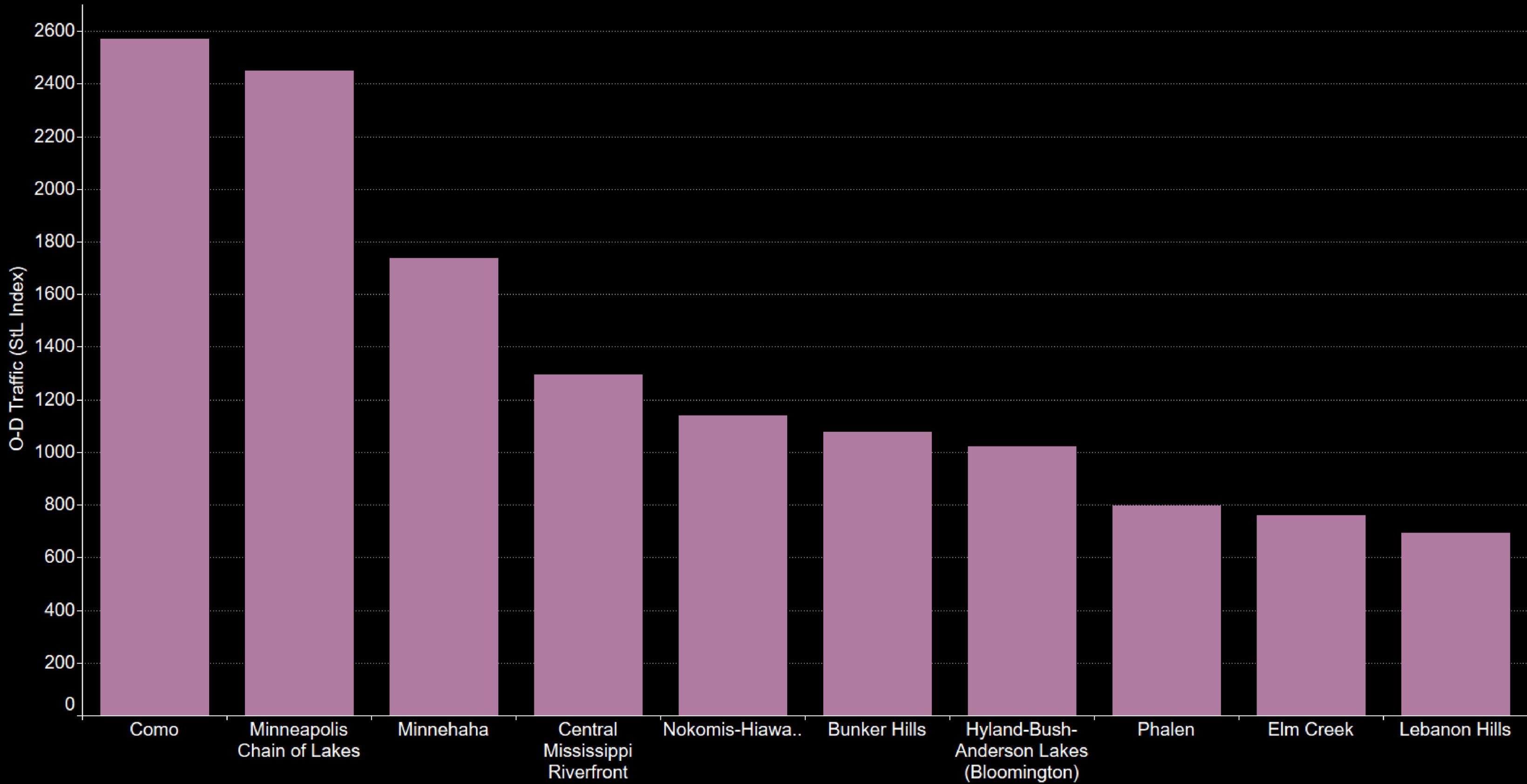
# Minneapolis Chain of Lakes

## OD Traffic by Zip Code



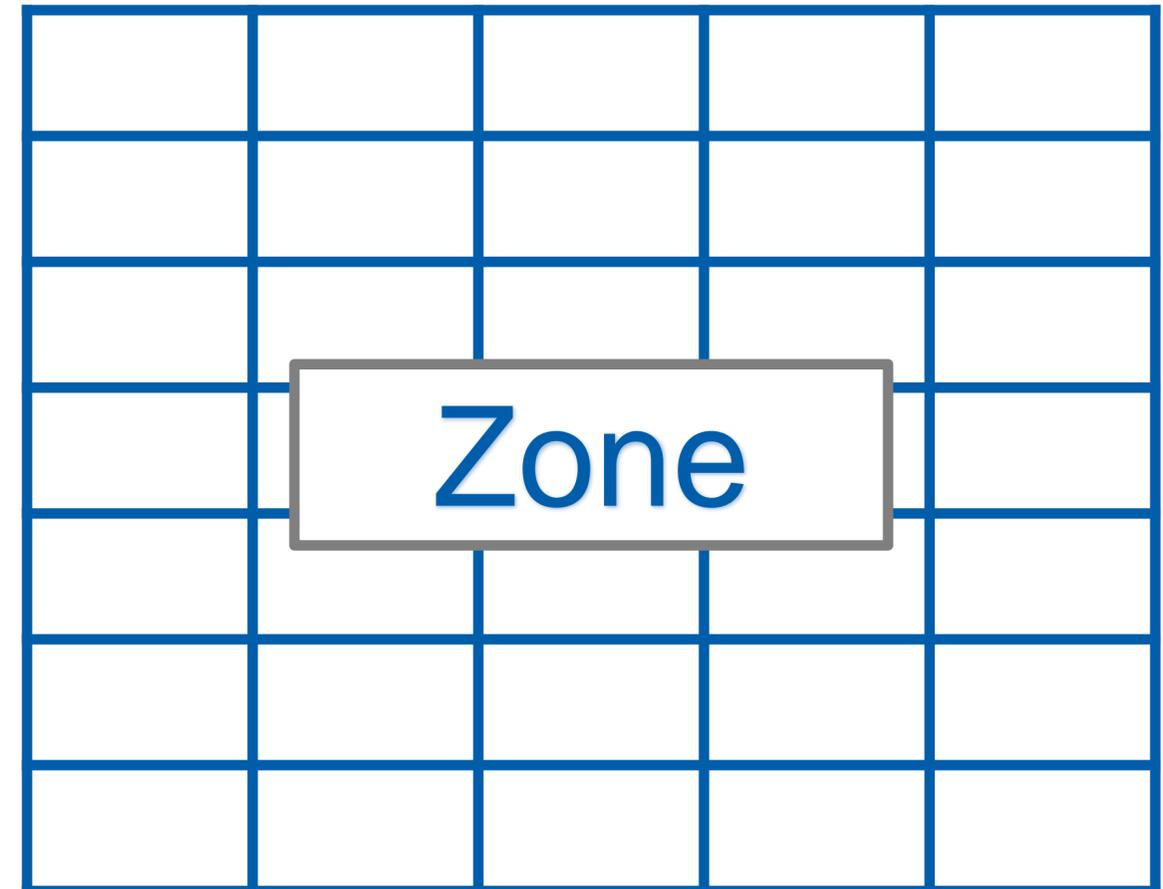


# Total OD Traffic Index by Park



# Zone Activity Analysis with Home Zip Codes

- Measures the relative traffic to each regional park from each home zip code
- Includes All Trips (Start, End, Pass-through)
- Filter data to only include Average Day (M-Su), retail hours (6am-10pm), all 2018
- 485,000 devices
- 673,000 trips



# Minneapolis Chain of Lakes

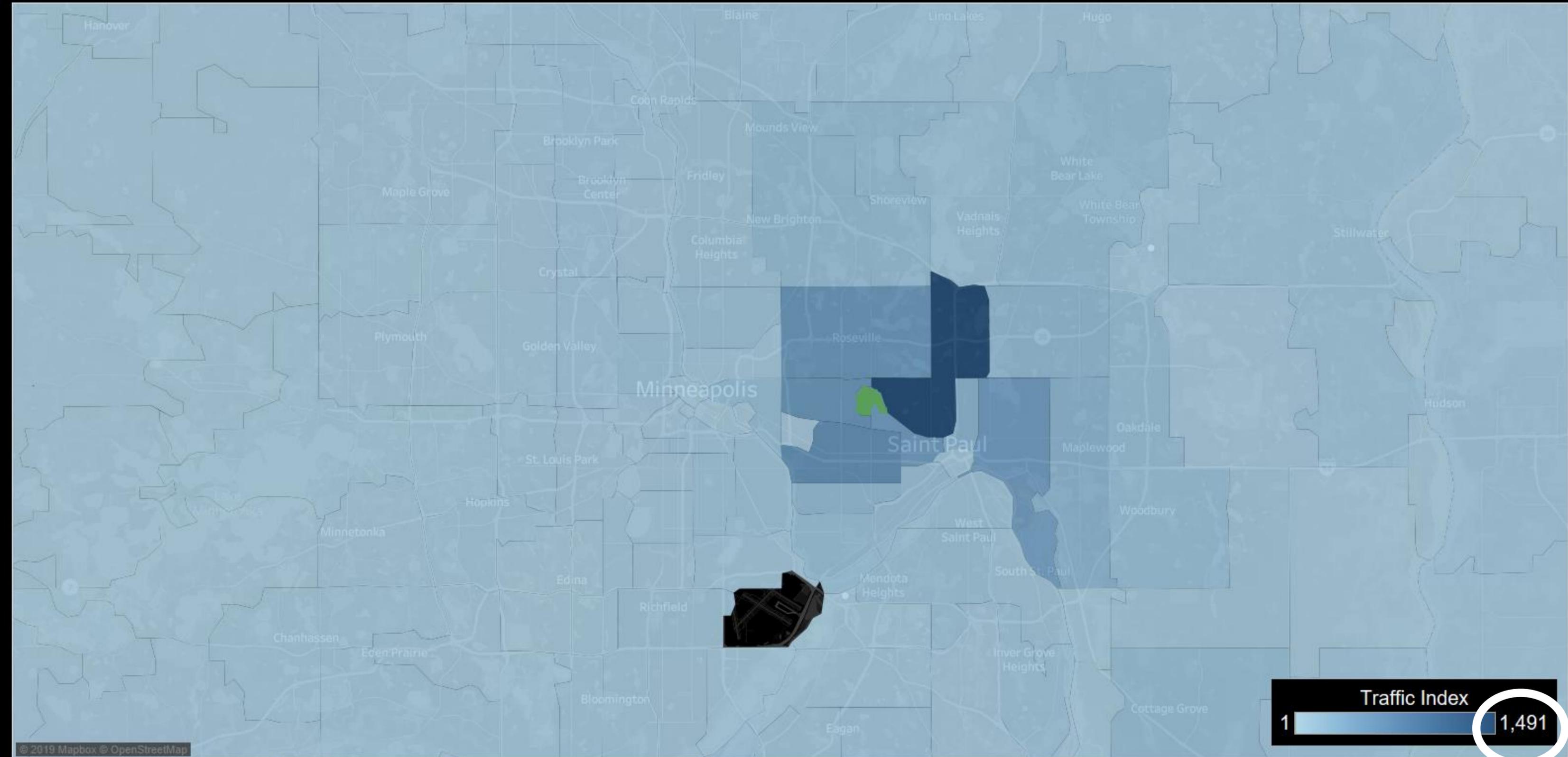
## Visitor Home Zip Codes



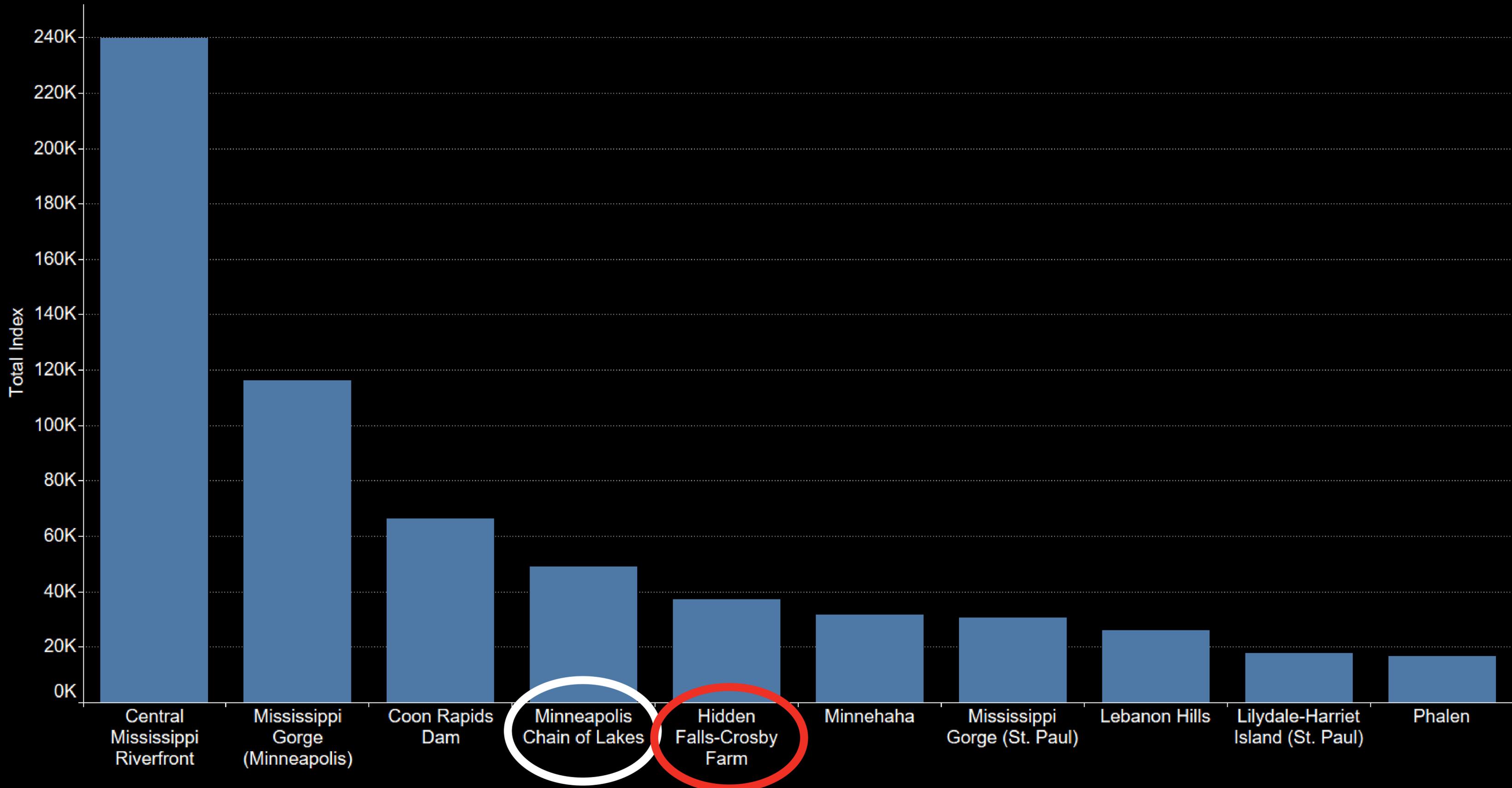


# Como

## Traffic by Home Zip Code



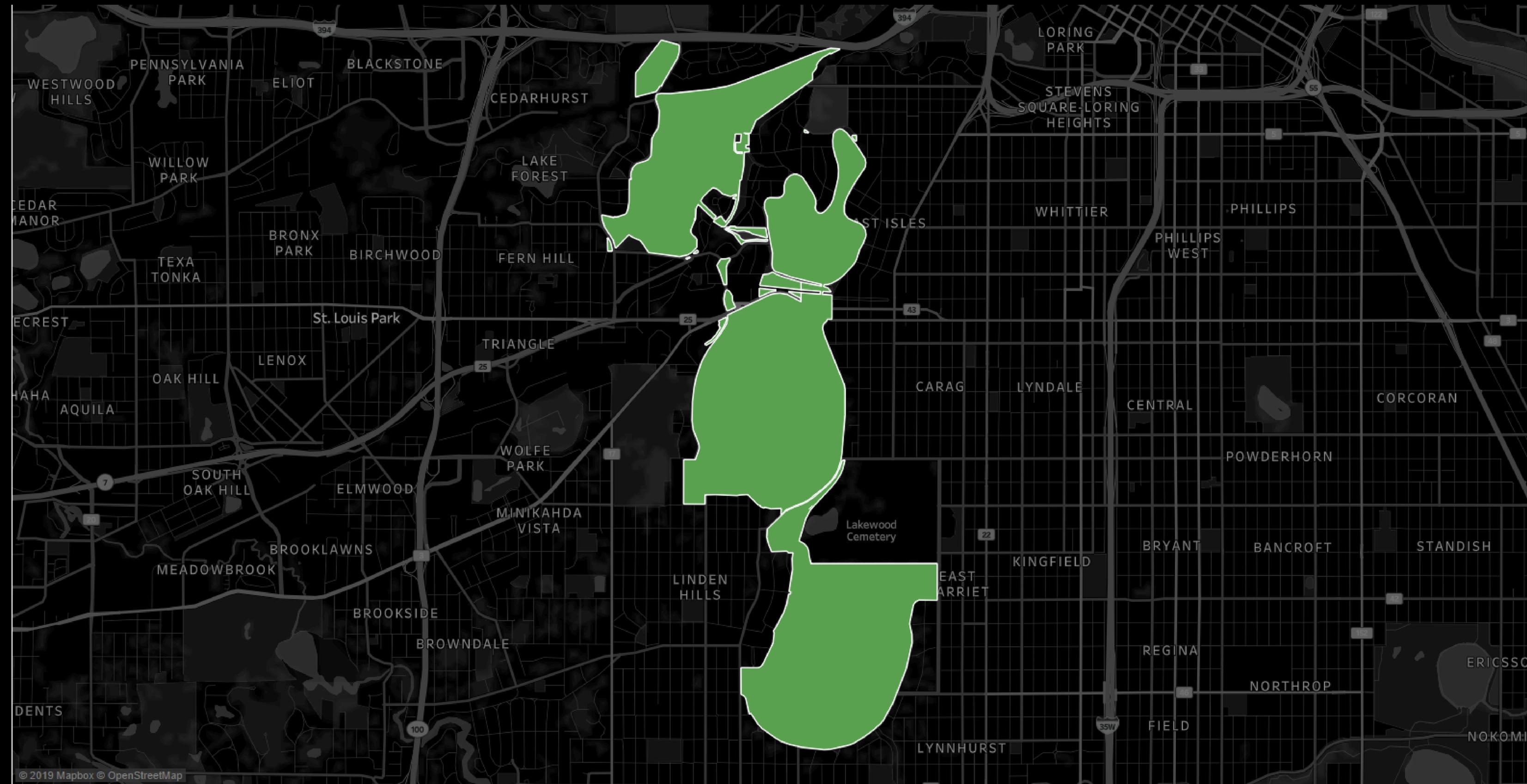
# Total Traffic Index by Park



# Hidden Falls-Crosby Farm



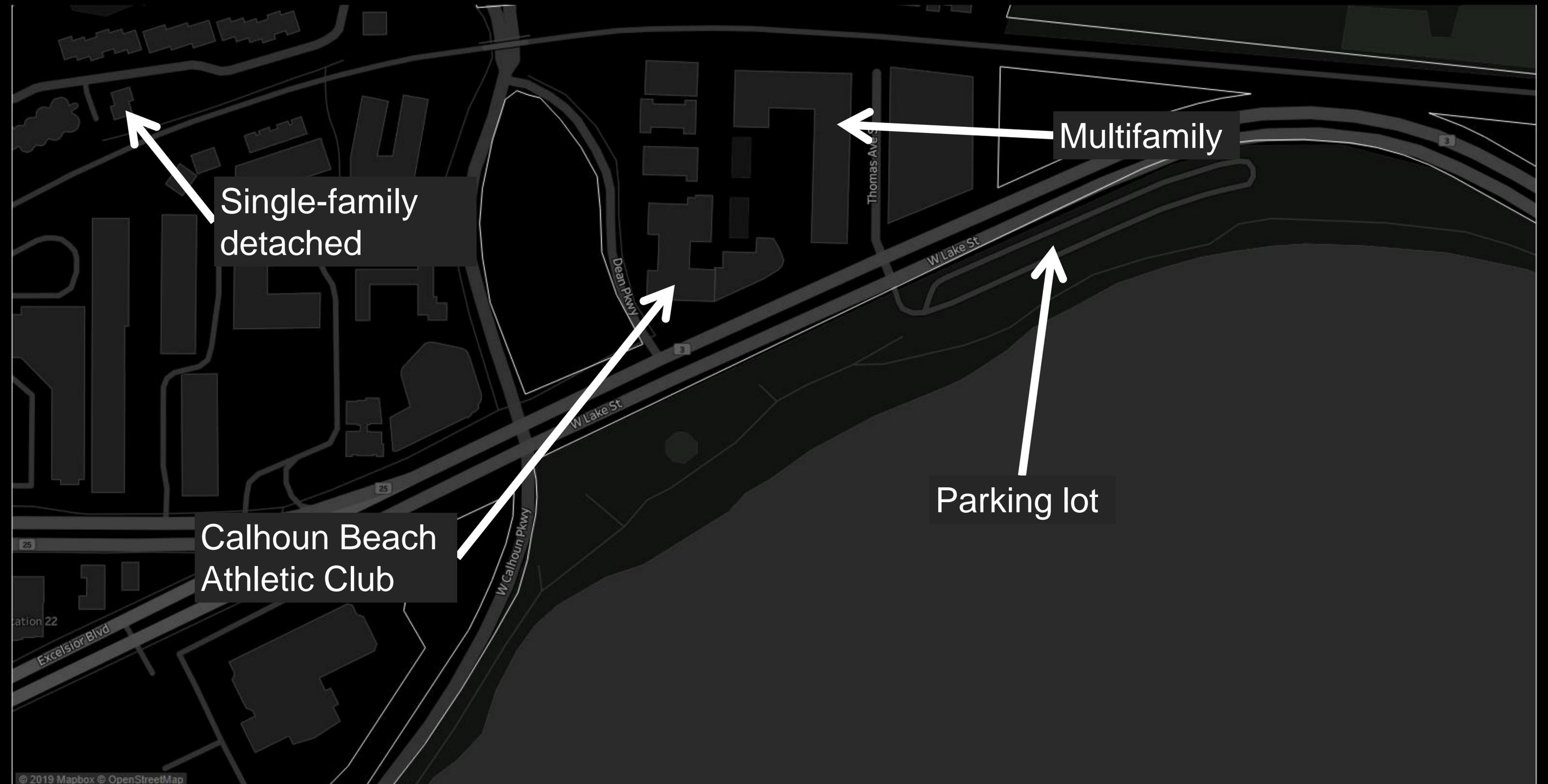
# Minneapolis Chain of Lakes



# Minneapolis Chain of Lakes



# Bde Maka Ska North Beach



Single-family detached

Multifamily

Parking lot

Calhoun Beach Athletic Club

# What did we learn?

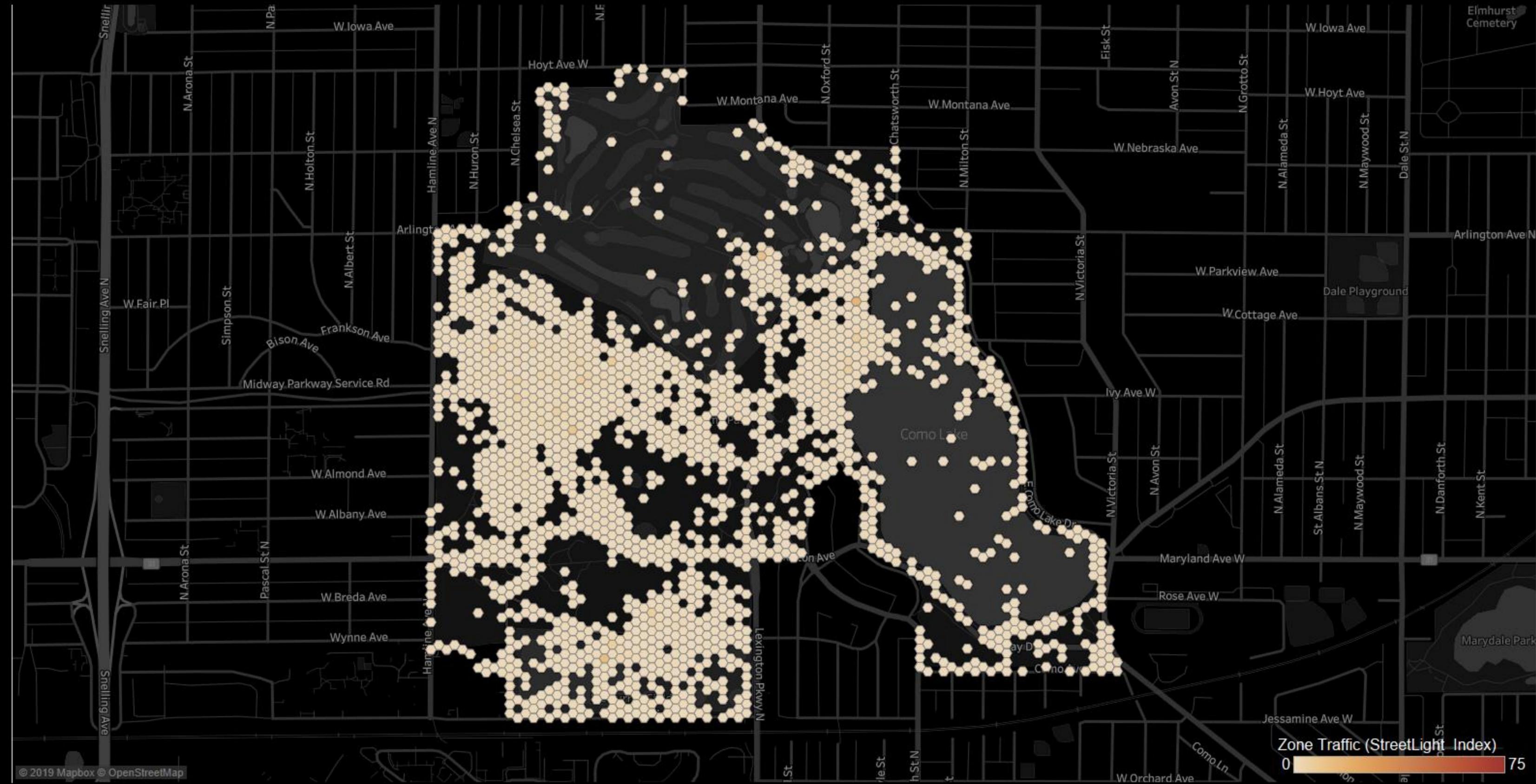
- There isn't a consistent way to define who counts as a visitor across all regional parks
- Each park requires detailed review



# Case study: Measuring activity within Como

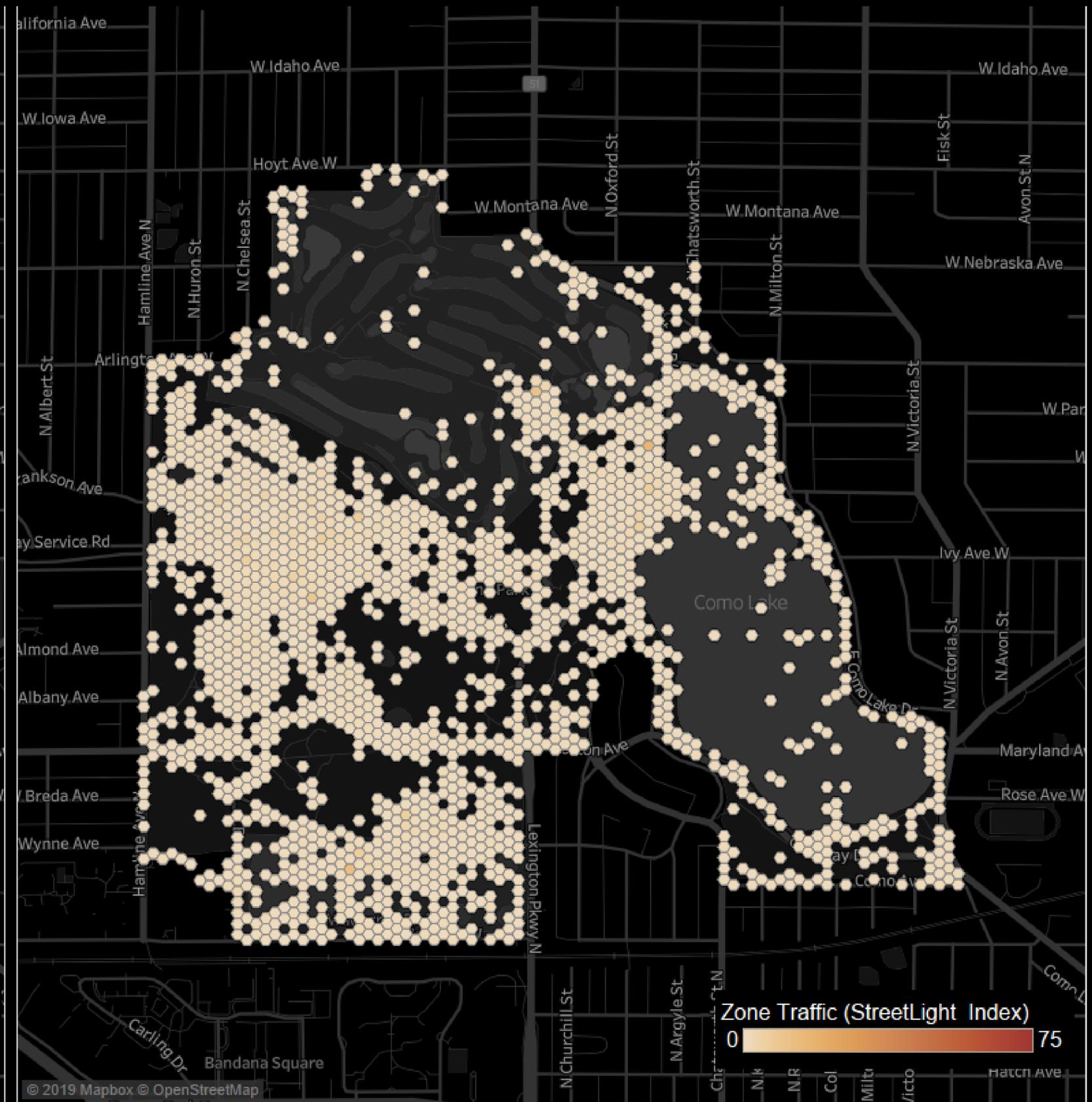
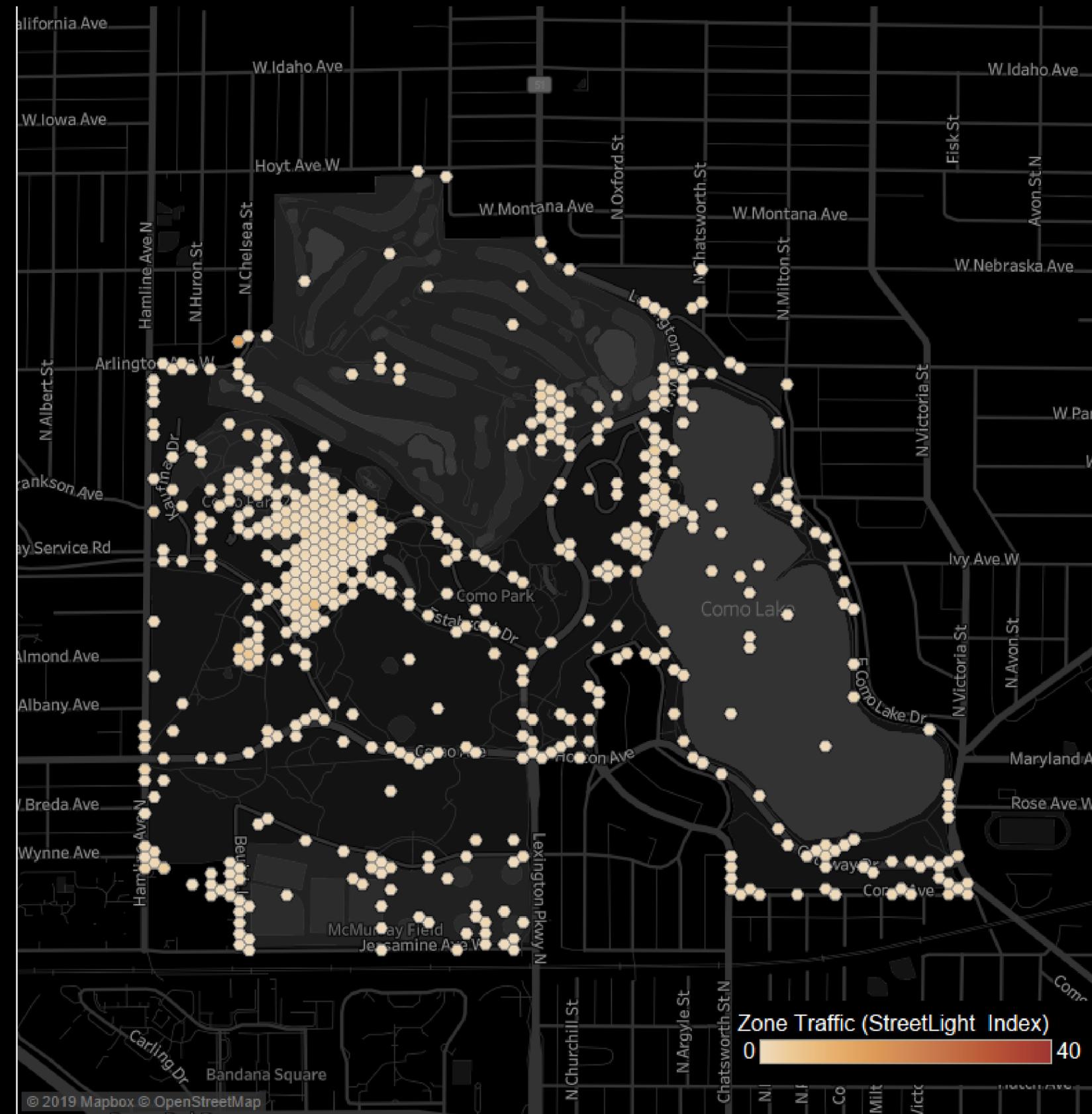


# Summer 2018



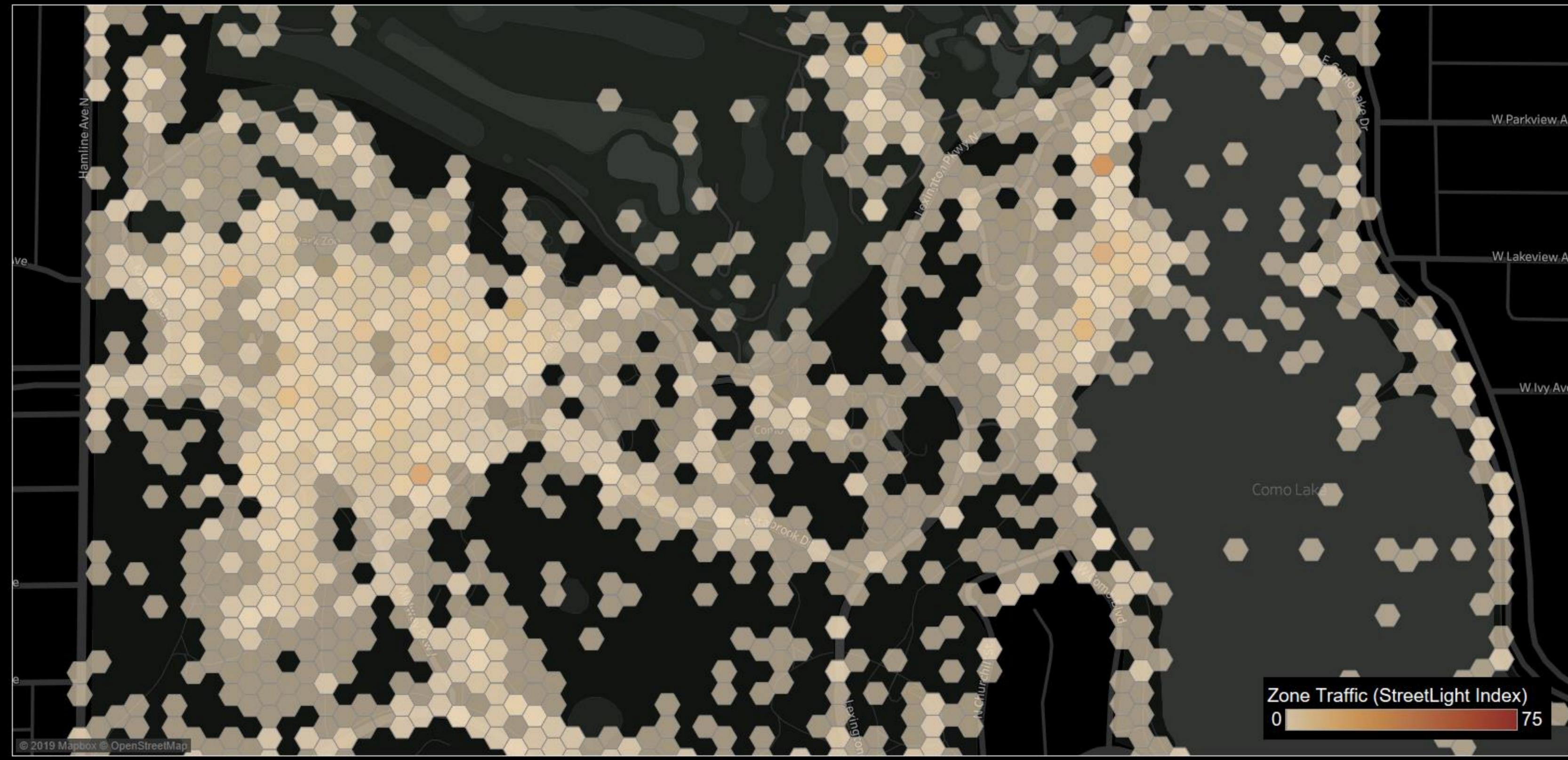
# Winter 2018

# Summer 2018



# Summer 2018

## May 2018 - August 2018

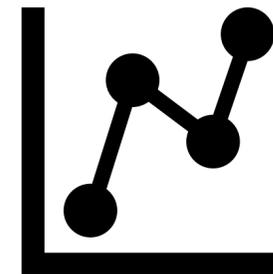
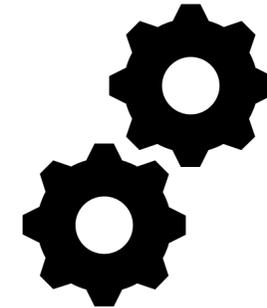


# Implications

- Implement stronger sampling plans
- Target resources to different areas over the year
- Avoid placing survey staff where there are few people

# What now?

- Location-based data challenges the way we think about regional parks visitors.
  - Examine disparities in park use
  - Differentiate passthrough and non-passthrough visitors
  - Locate busy areas within parks
  - Allow low-cost park-level analysis
  - Open the door for advanced research



# Thank you!

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