

# OK Google, Tell Me About Myself

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# Last Week's Data Headlines

**DoorDash confirms data breach affected 4.9 million customers, workers and merchants**

**Data Breach Warning For 200 Million Android And iOS Gamers**

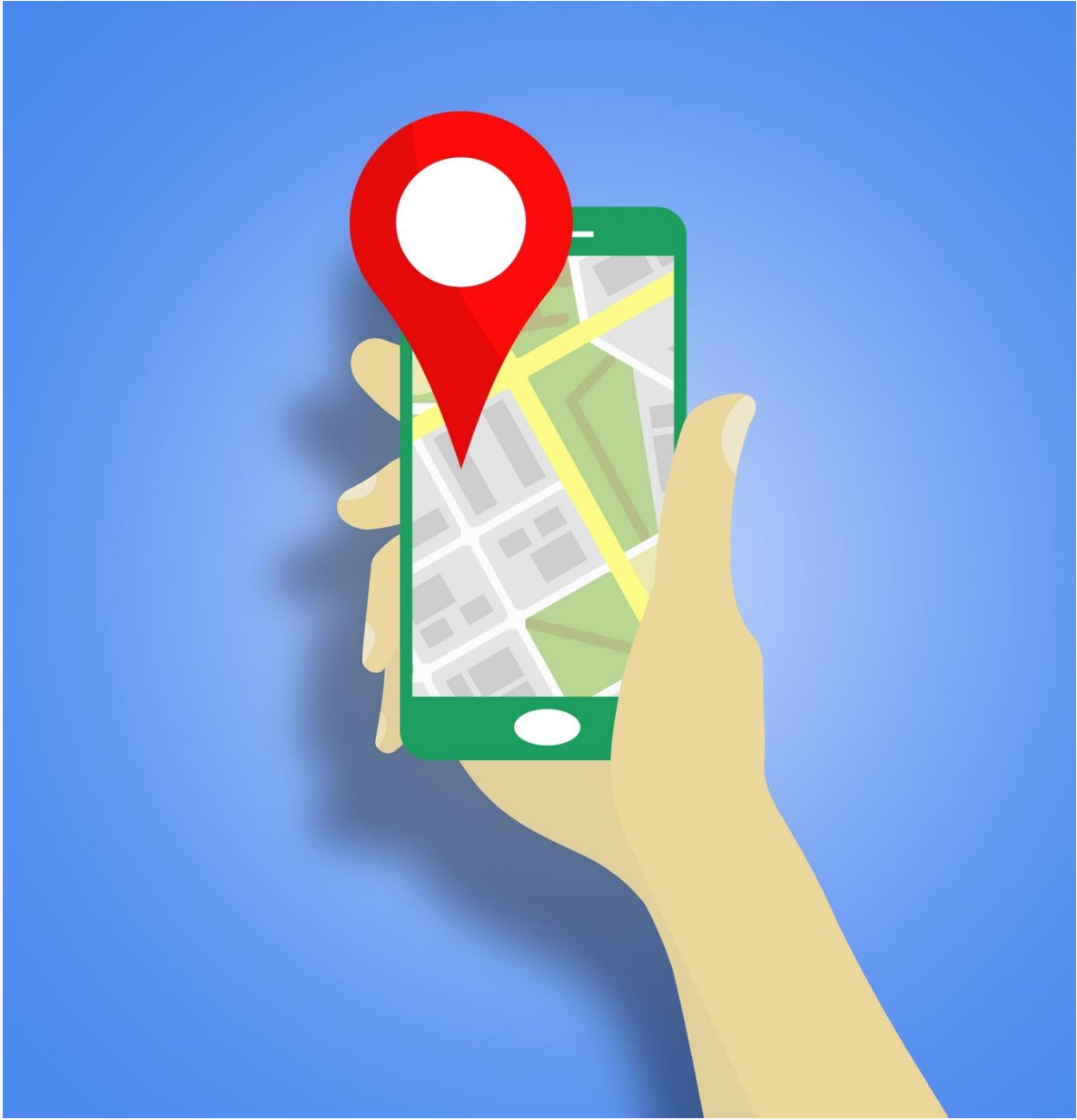
**Kaiser says data breach exposed information on nearly 1,000 Sacramento-area patients**

**Hy-Vee says malware caused payment card data breach**

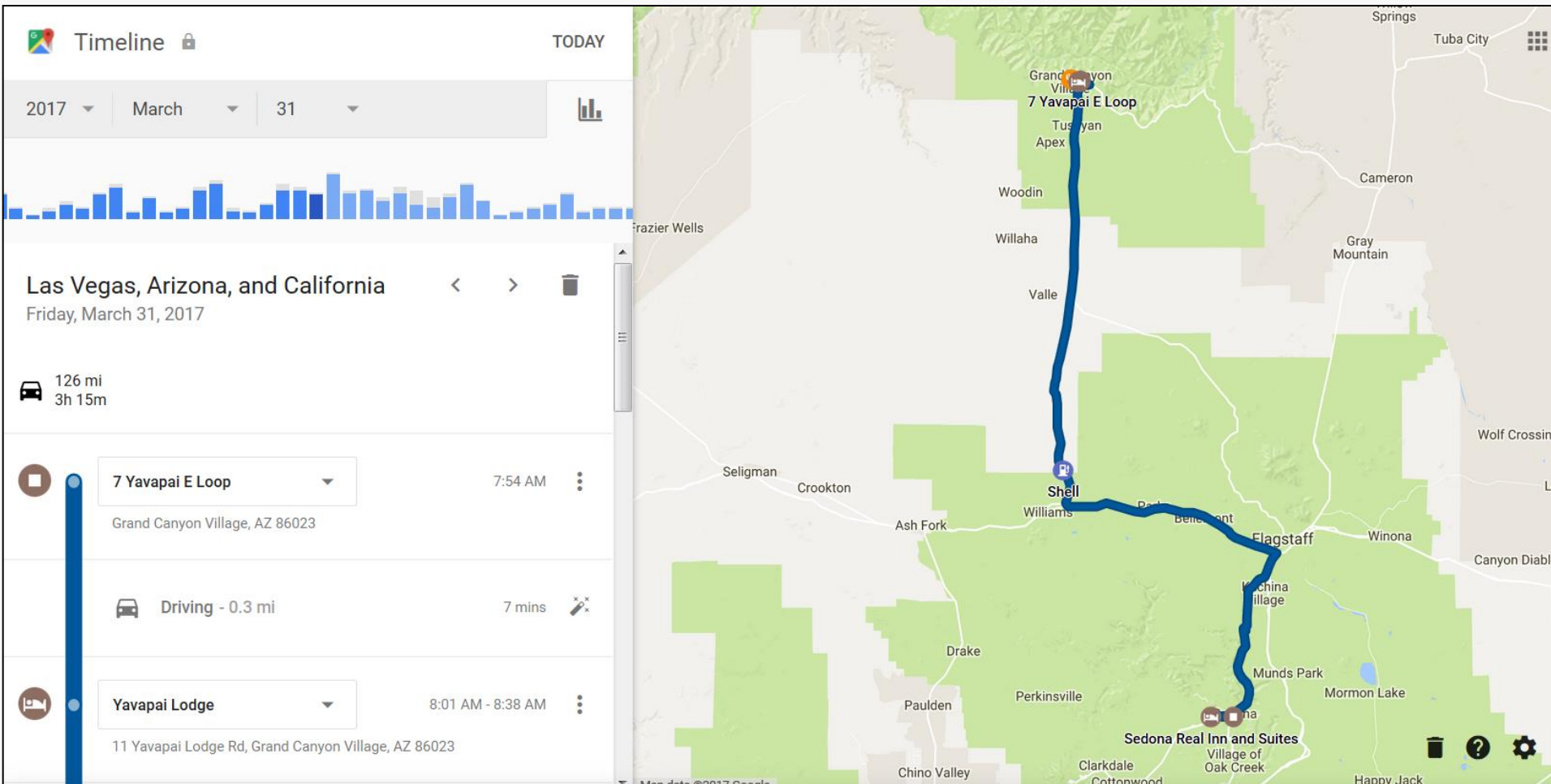
**Zynga data breach exposed 200 million Words with Friends players**



Google™



# Google Timelines





## Your July in review

Your timeline in Google Maps helps you curate the places you've been. Look back on the past month and reminisce about recent trips and past places.

[EXPLORE YOUR TIMELINE](#)



**8 cities visited this month**



**28 places visited this month (5 new)**





## Your activity in timeline



**6 mi (11 km)**  
walked this  
month



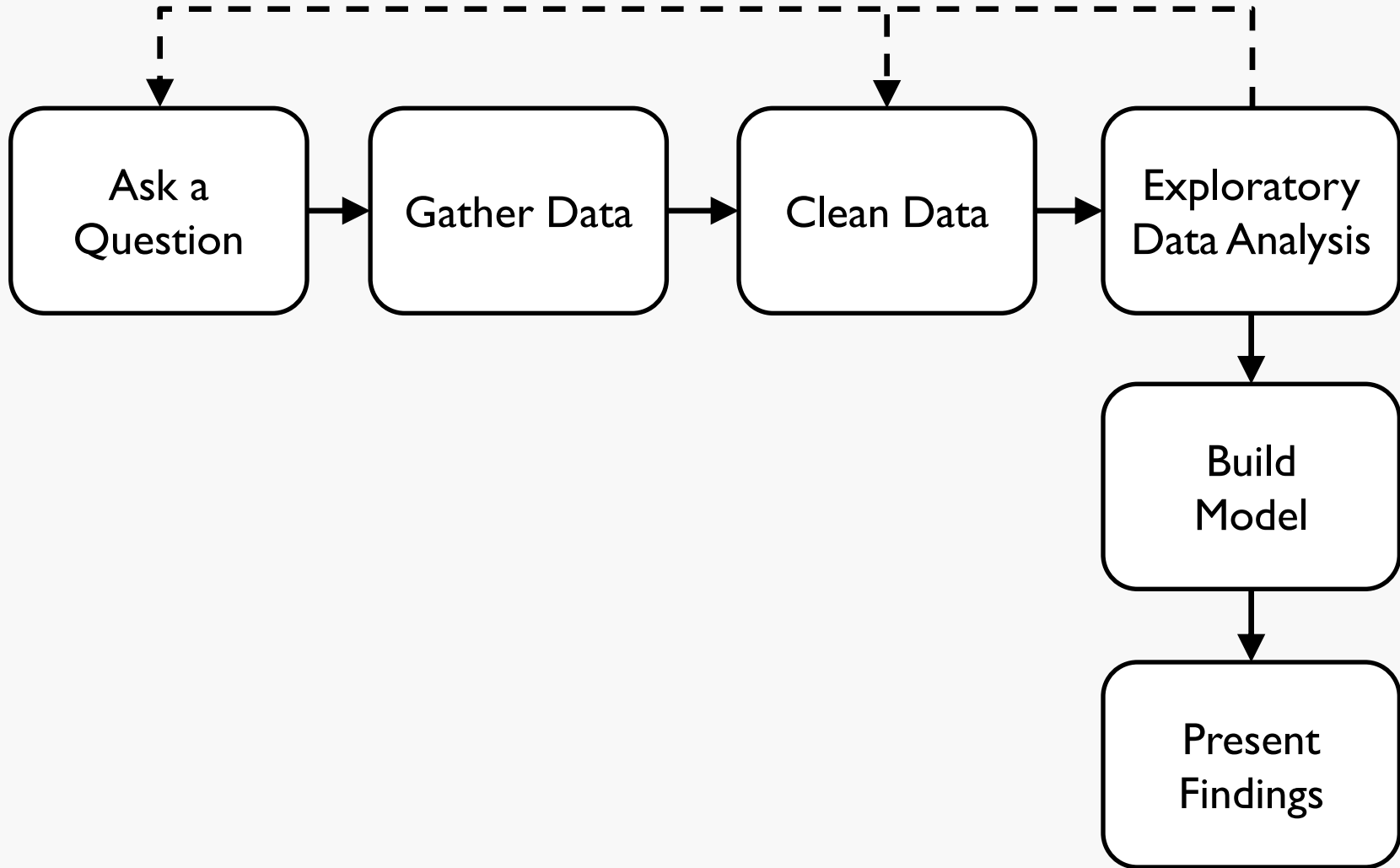
**22 mi (37 km)**  
run this month



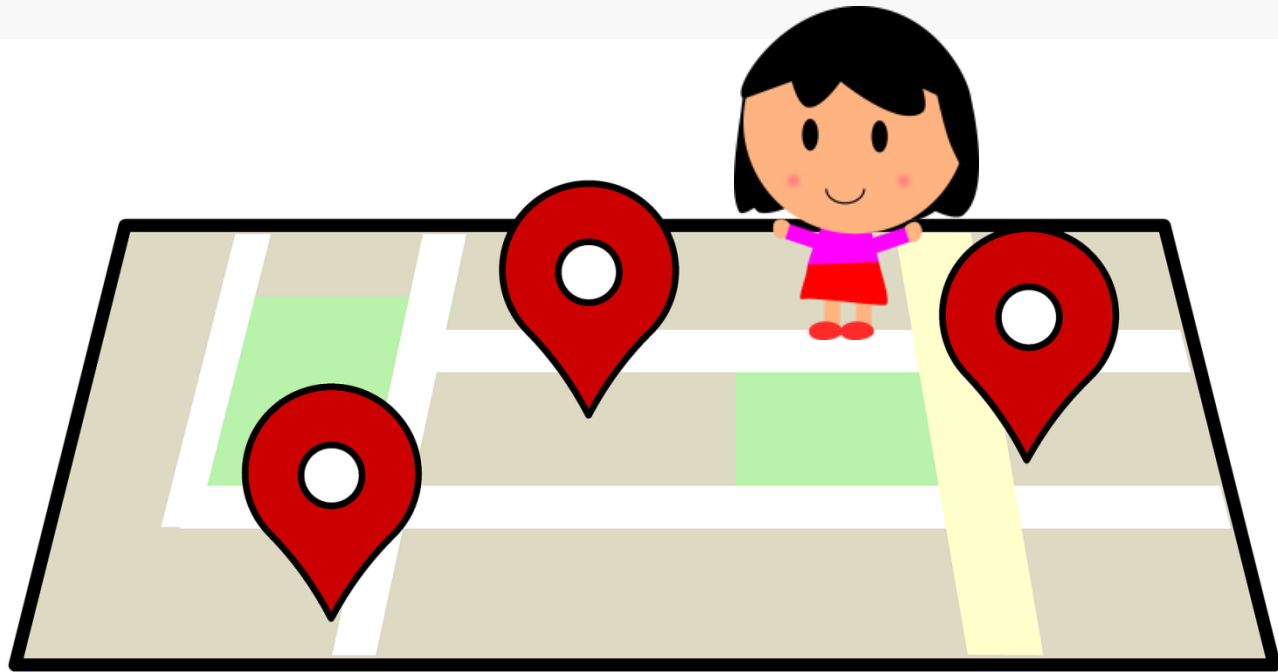
**28 hours** spent  
in a vehicle this  
month



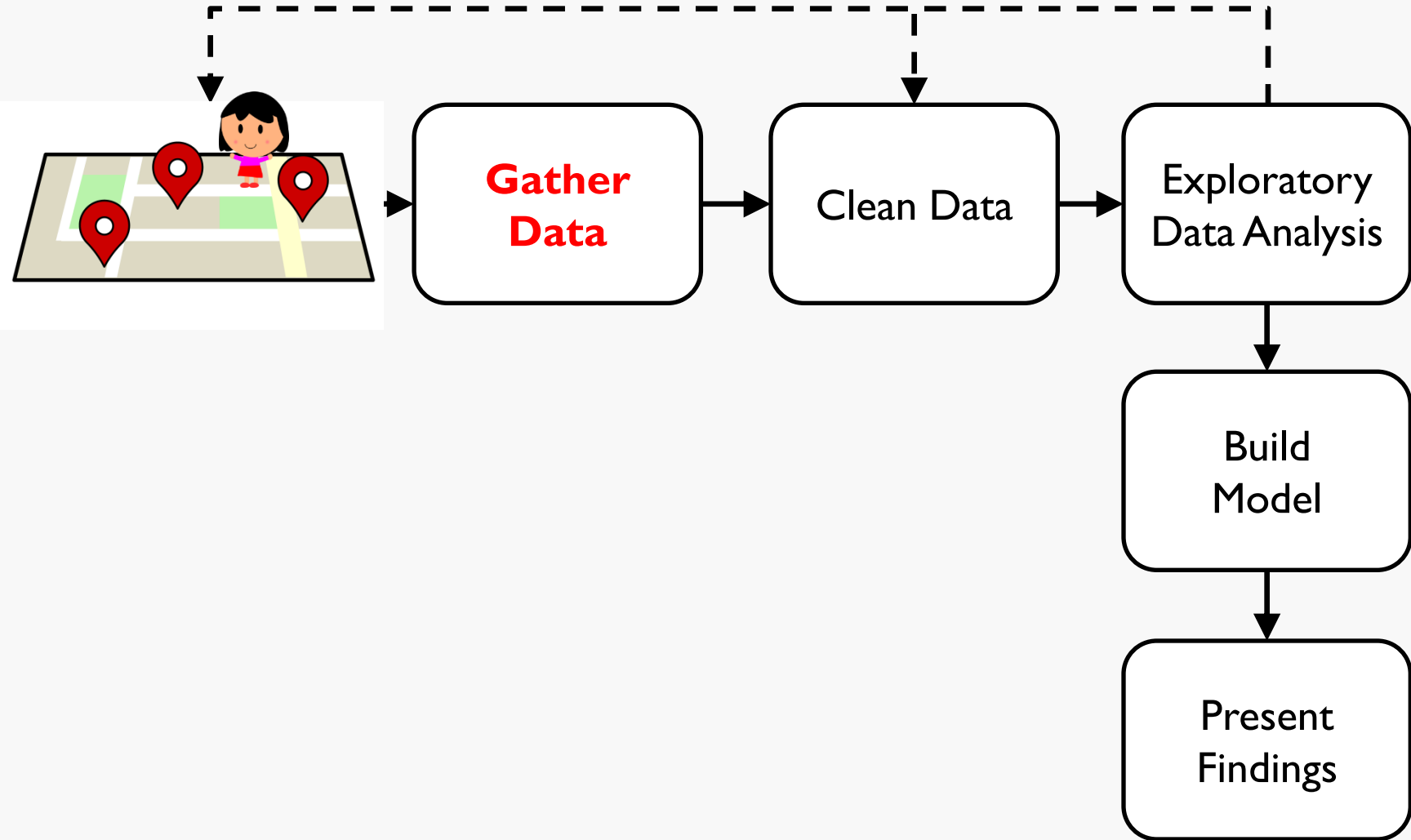
# Data Science Process



Can I create a model of my life from  
stored location data?



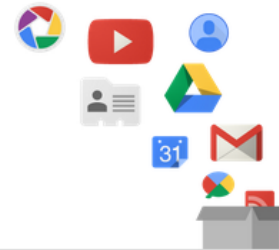
# Data Science Process



Your account, your data.  
Export a copy.

Create an archive with your data from Google products.

Manage archives



Location History

JSON format



Mail

All mail



Maps (your places)



My Maps



Searches



Tasks



Voice



YouTube

All data types  
OPML (RSS) format



# KML format example

```
<when>2017-03-30T22:16:05Z</when>  
<gx:coord>-112.1206089 36.0538447 2110</gx:coord>  
  
<when>2017-03-30T22:15:32Z</when>  
<gx:coord>-112.1206895 36.0541252 2108</gx:coord>  
  
<when>2017-03-30T22:14:41Z</when>  
<gx:coord>-112.1161455 36.0566548 2117</gx:coord>  
  
<when>2017-03-30T22:13:41Z</when>  
<gx:coord>-112.1110006 36.0585582 2123</gx:coord>
```

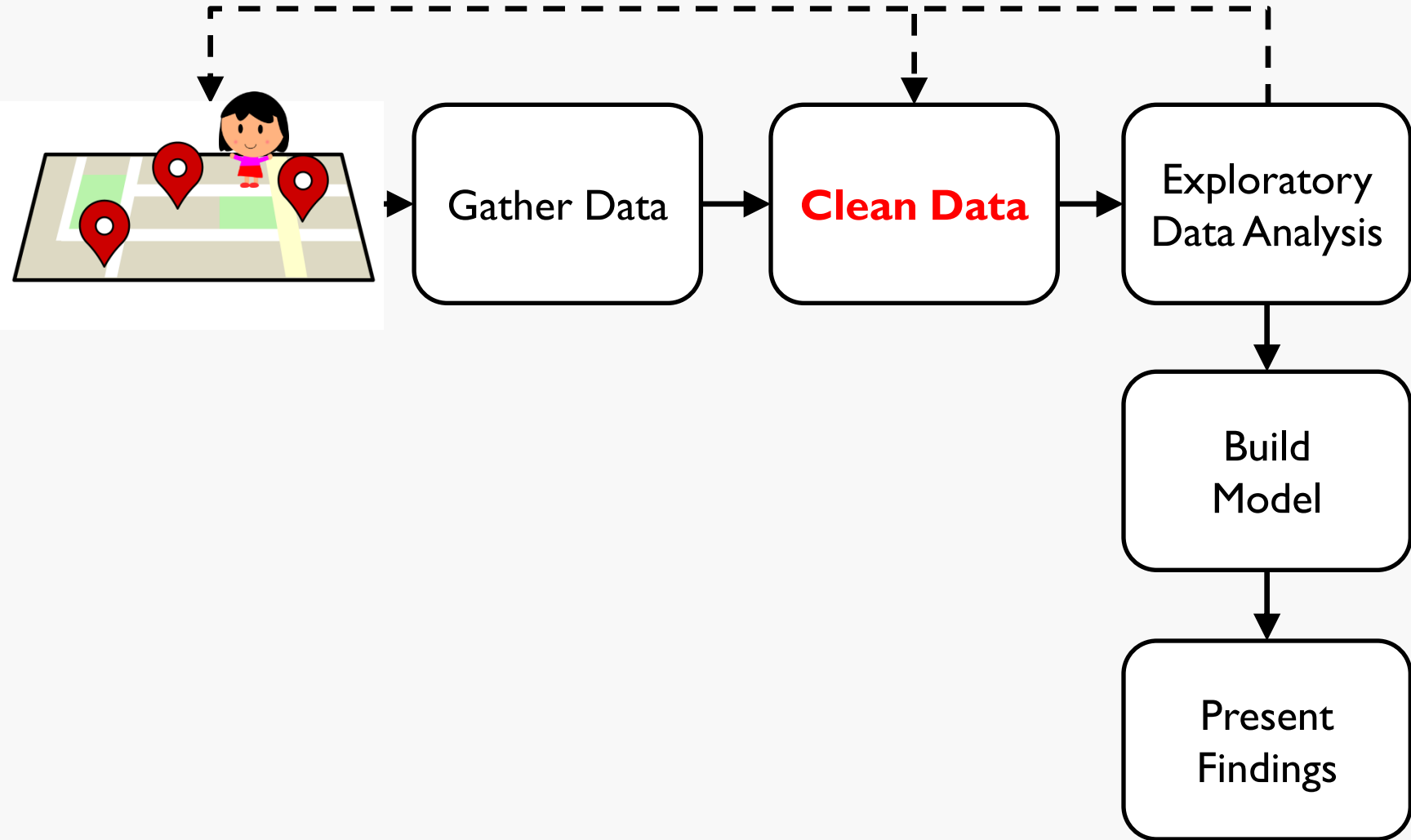
# JSON format example

```
"timestampMs" : "1490998907806",  
"latitudeE7" : 348600316,  
"longitudeE7" : -1118161027,  
"accuracy" : 21,  
"activity" : [ {  
  "timestampMs" : "1490998831576",  
  "activity" : [ {  
    "type" : "STILL",  
    "confidence" : 75  
  }, {  
    "type" : "ON_FOOT",  
    "confidence" : 10  
  }, {  
    "type" : "IN_VEHICLE",  
    "confidence" : 5  
  }, {  
    "type" : "ON_BICYCLE",  
    "confidence" : 5  
  }, {  
    "type" : "UNKNOWN",  
    "confidence" : 5  
  }, {  
    "type" : "WALKING",  
    "confidence" : 5  
  }  
]
```

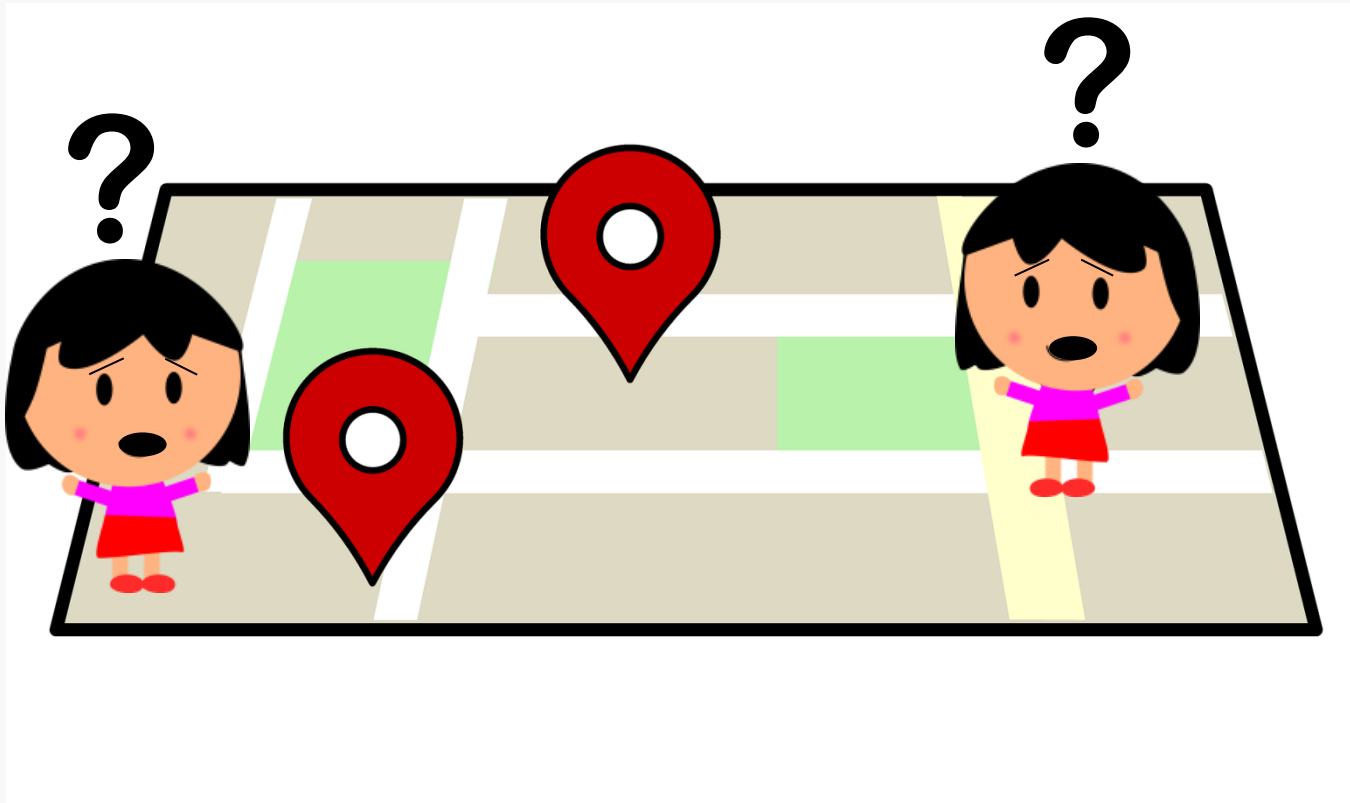
Not always  
available



# Data Science Process

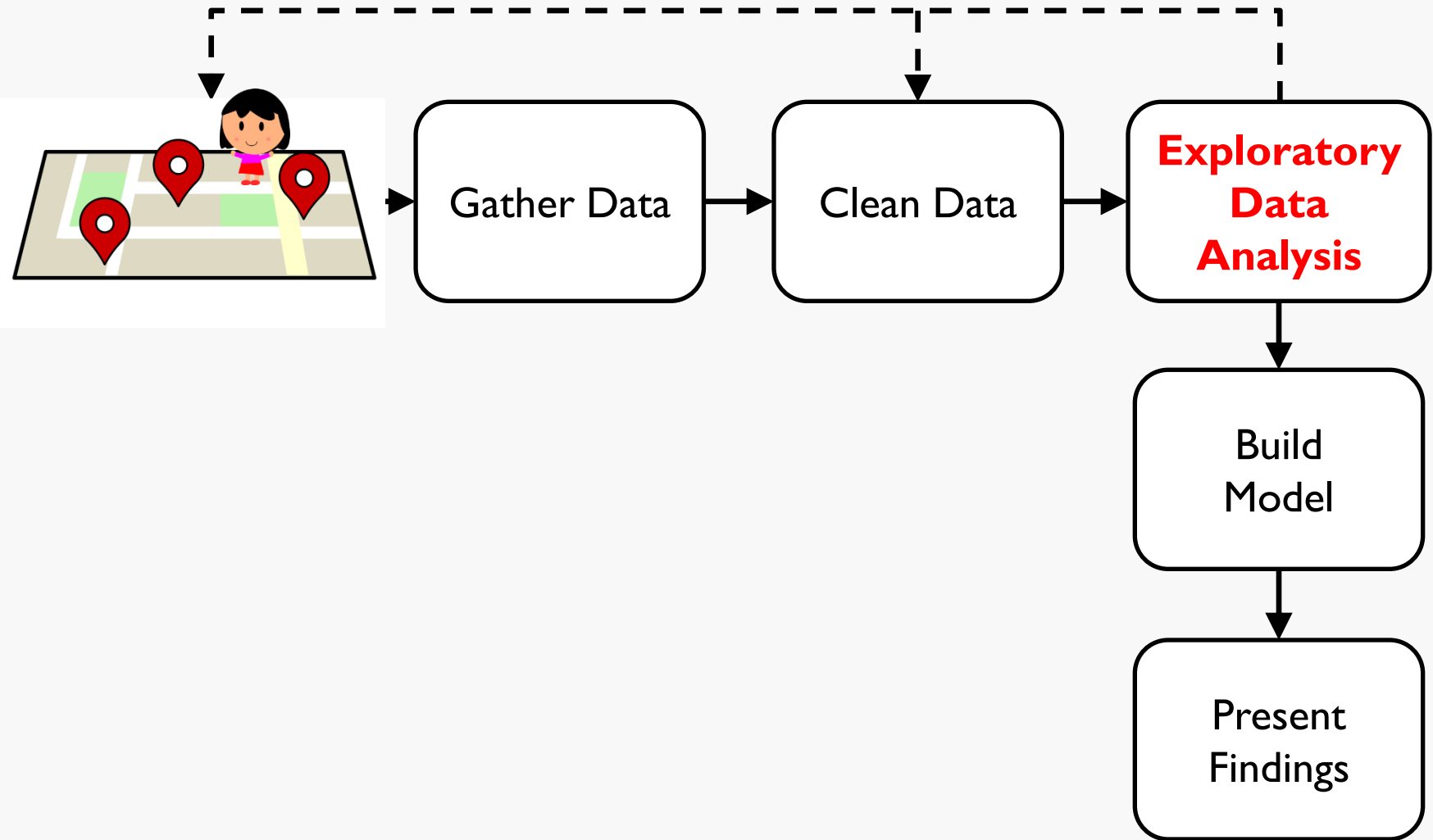


# Traveling at the speed of light





# Data Science Process



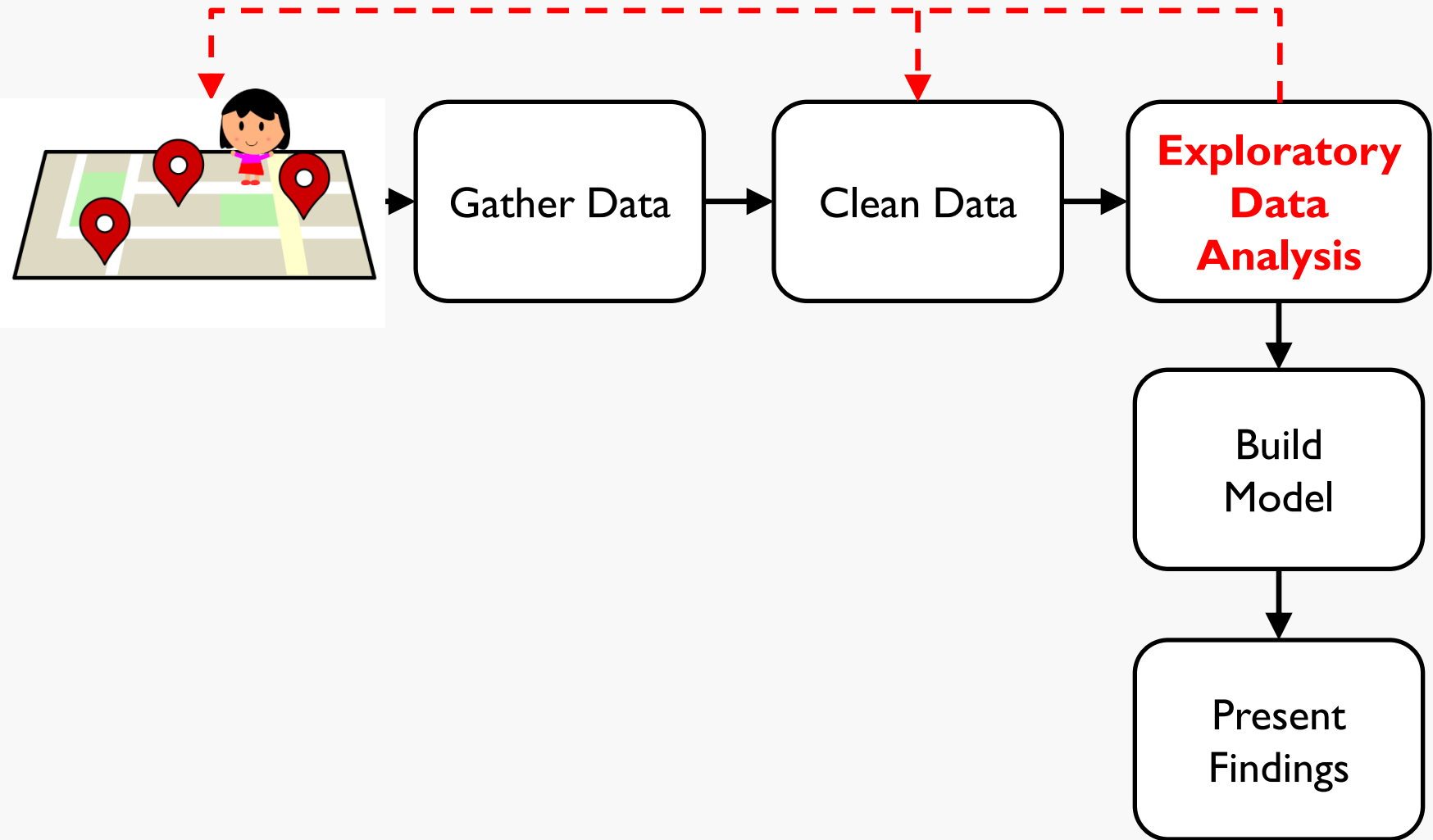
# What is EDA?

- Define characteristics
  - Trends
  - Biases
  - Variability
  - Breadth



- Test Assumptions
- Visualize

# Data Science Process



# What's in the data?

```
<when>2017-03-30T22:16:05Z</when>
```

```
<gx:coord>-112.1206089 36.0538447 2110</gx:coord>
```

```
<when>2017-03-30T22:15:32Z</when>
```

```
<gx:coord>-112.1206895 36.0541252 2108</gx:coord>
```

```
<when>2017-03-30T22:14:41Z</when>
```

```
<gx:coord>-112.1161455 36.0566548 2117</gx:coord>
```

```
<when>2017-03-30T22:13:41Z</when>
```

```
<gx:coord>-112.1110006 36.0585582 2123</gx:coord>
```

# Location

<when>2017-03-30T22:16:05Z</when>

<gx:coord>-112.1206089 36.0538447 2110</gx:coord>

<when>2017-03-30T22:15:32Z</when>

<gx:coord>-112.1206895 36.0541252 2108</gx:coord>

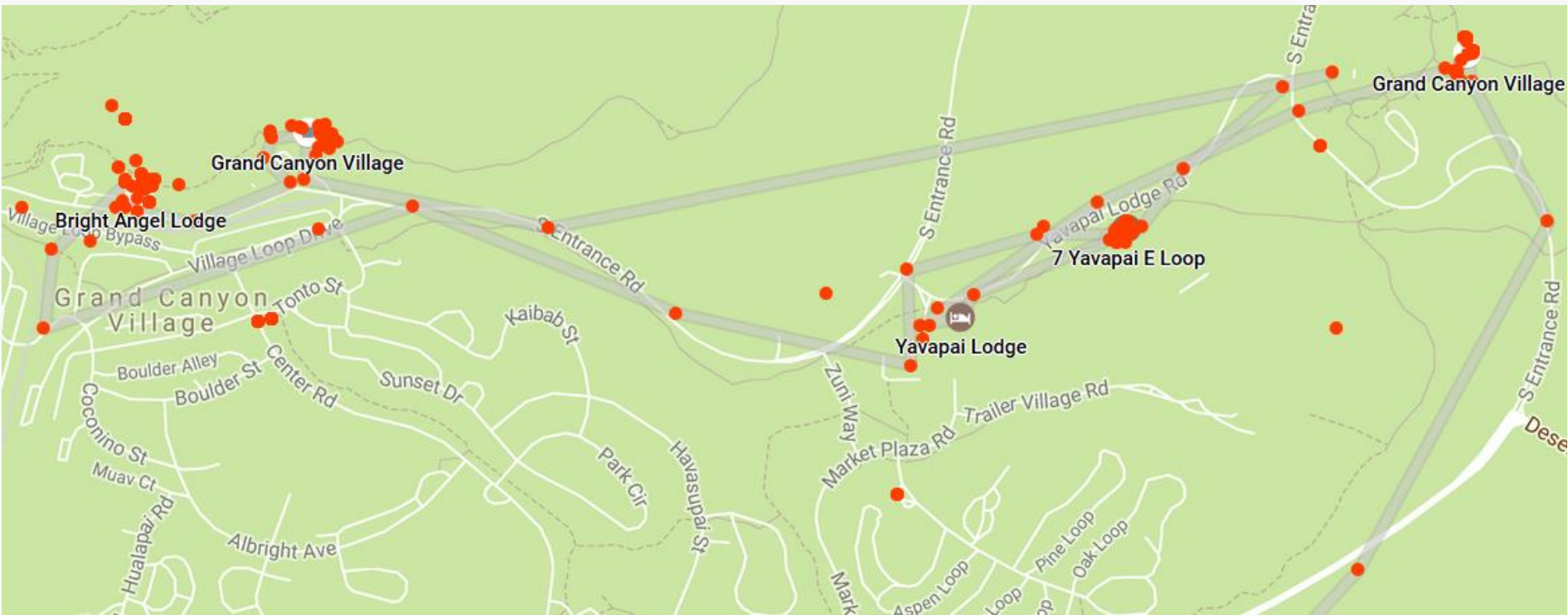
<when>2017-03-30T22:14:41Z</when>

<gx:coord>-112.1161455 36.0566548 2117</gx:coord>

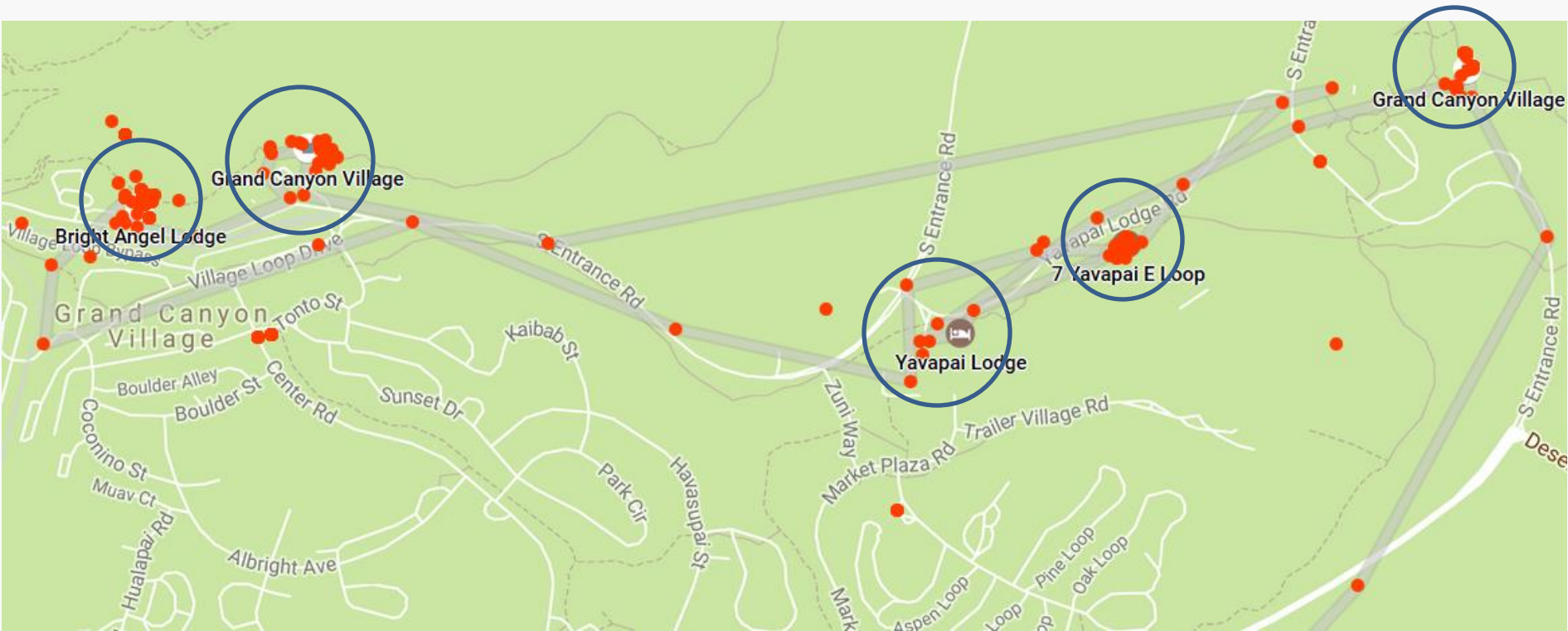
<when>2017-03-30T22:13:41Z</when>

<gx:coord>-112.1110006 36.0585582 2123</gx:coord>

# Grand Canyon

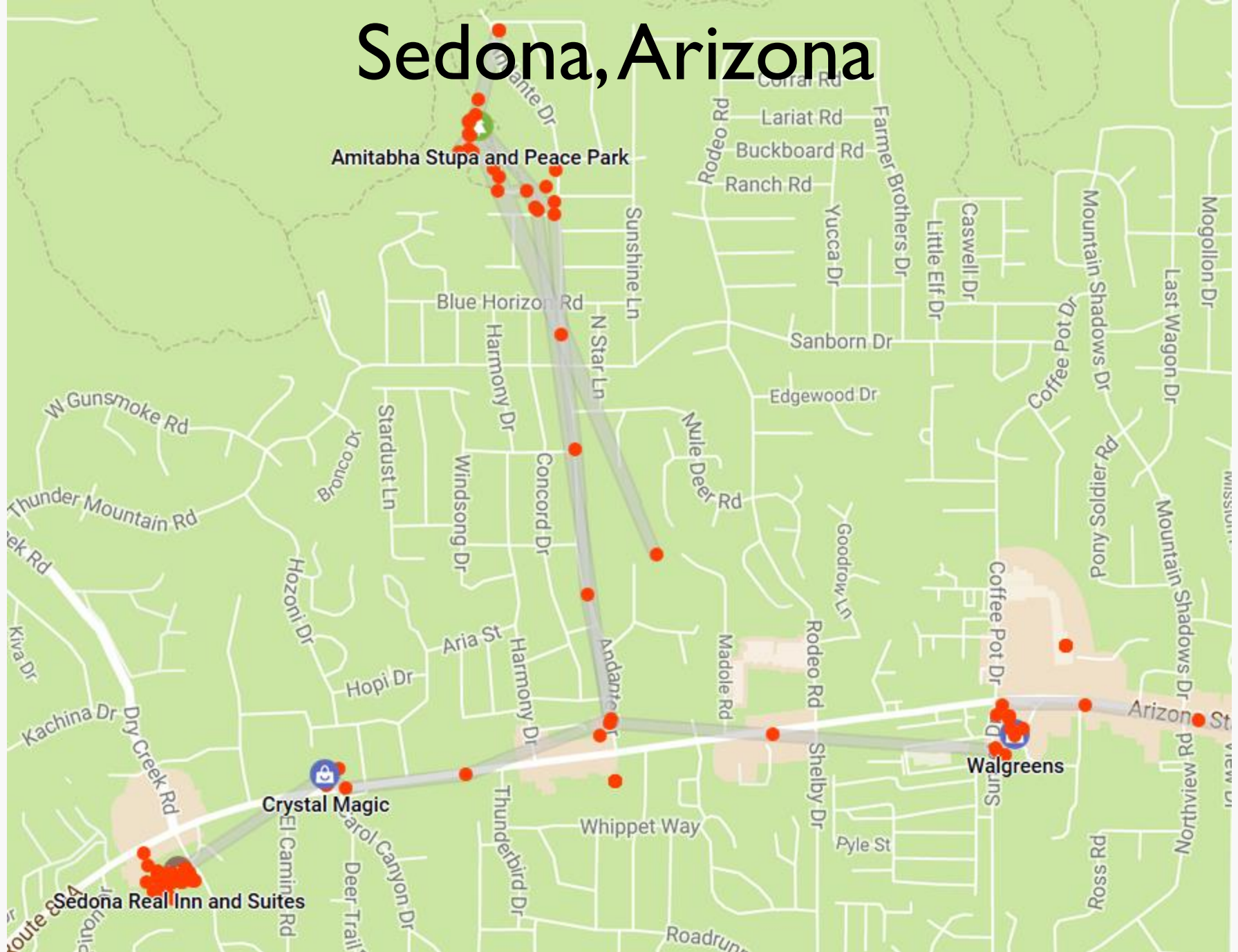


# Grand Canyon



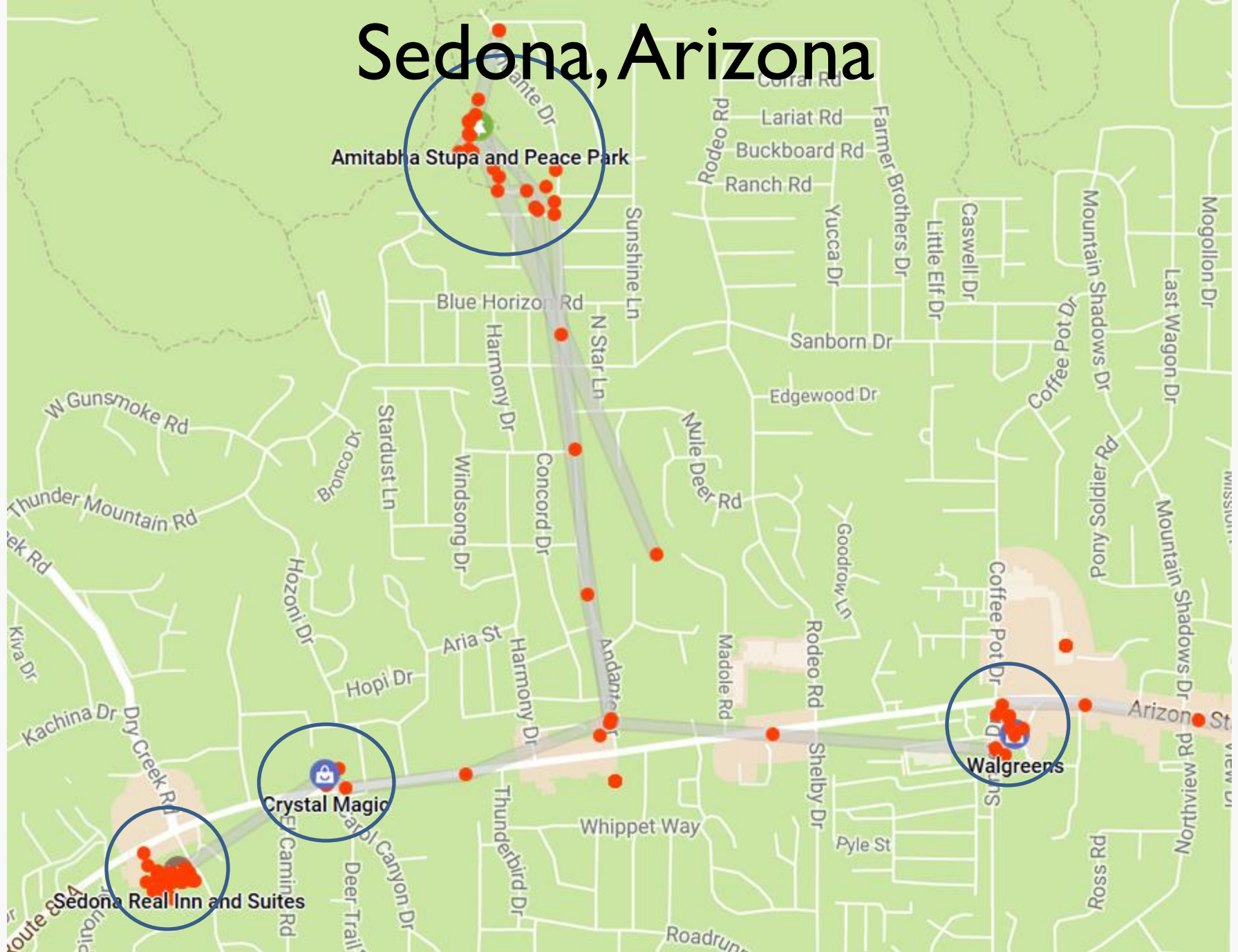


# Sedona, Arizona

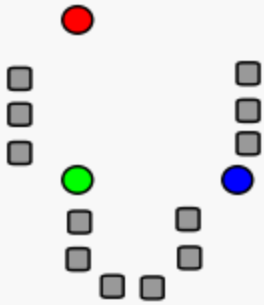




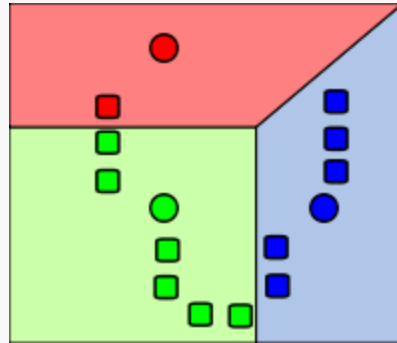
# Sedona, Arizona



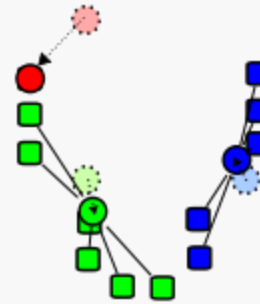
# K-Means



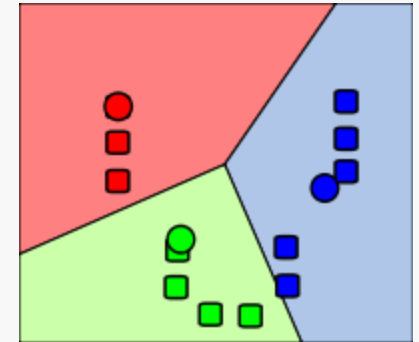
Randomly pick  
 $K = 3$  points  
(initial  
centroids)



Assign each  
point to its  
closest  
centroid



Using points in  
clusters,  
calculate new  
centroids



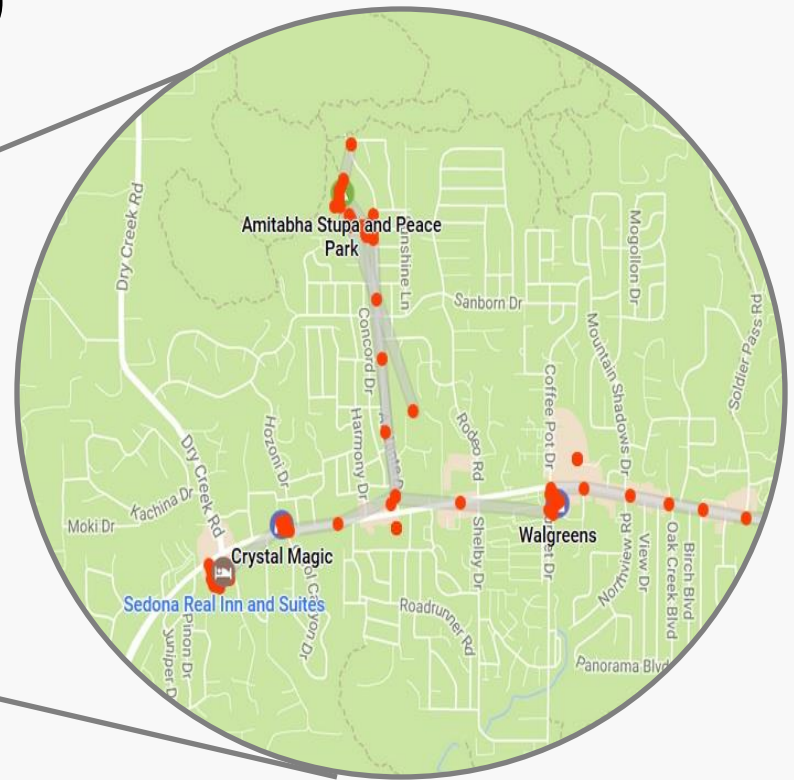
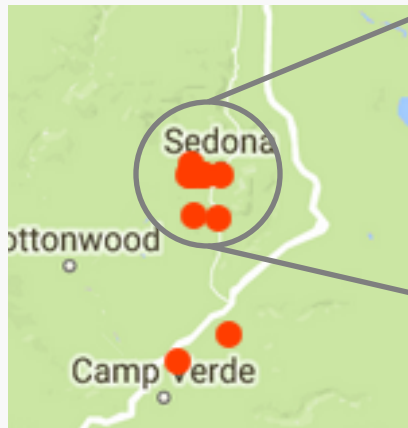
Assign each  
point to its  
closest centroid

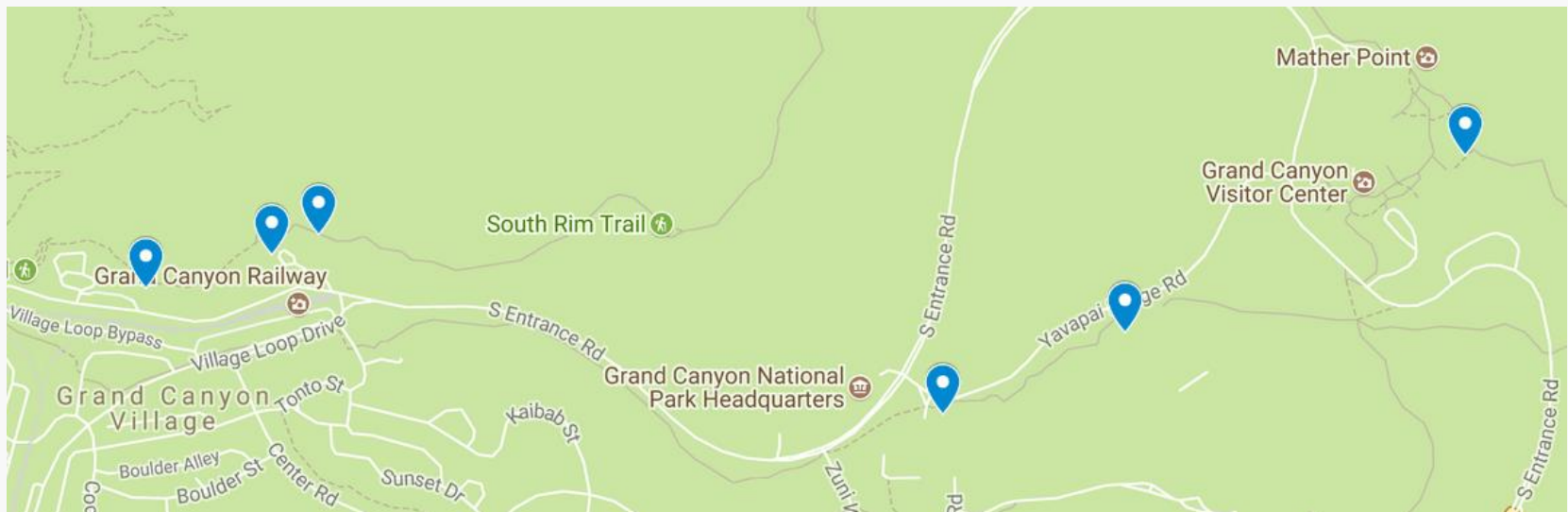
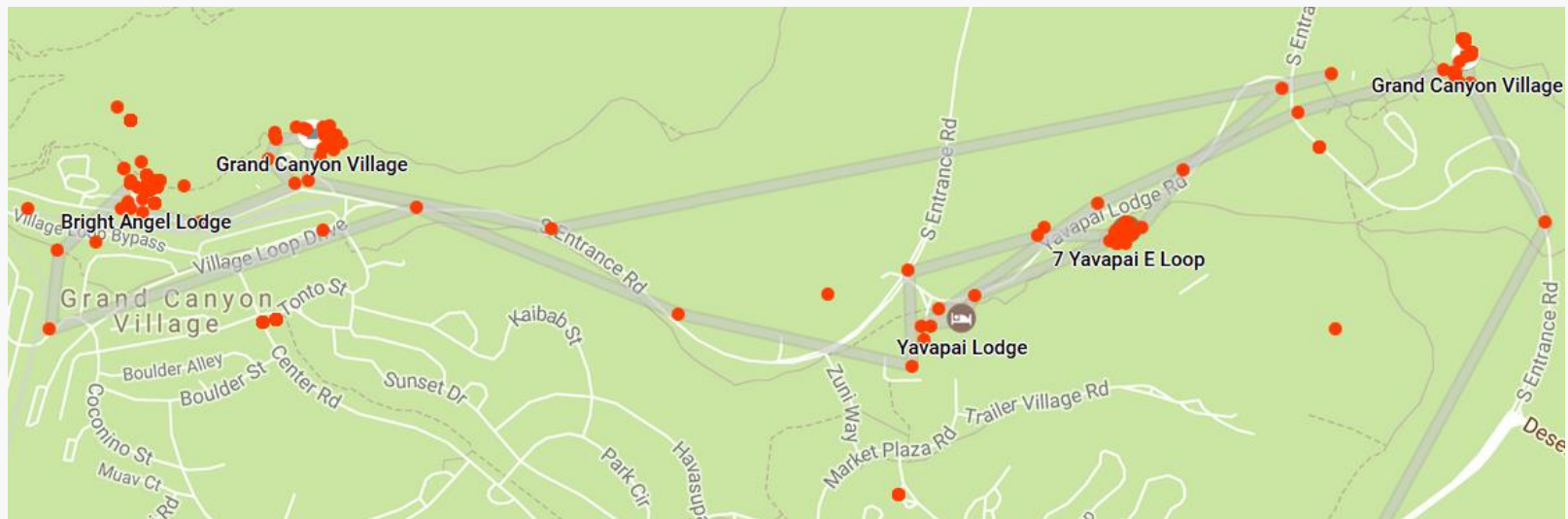
[K Means Example Step 1.svg](#),  
[K Means Example Step 2.svg](#),  
[K Means Example Step 3.svg](#),  
[K Means Example Step 4.svg](#)

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# Recursive K-Means

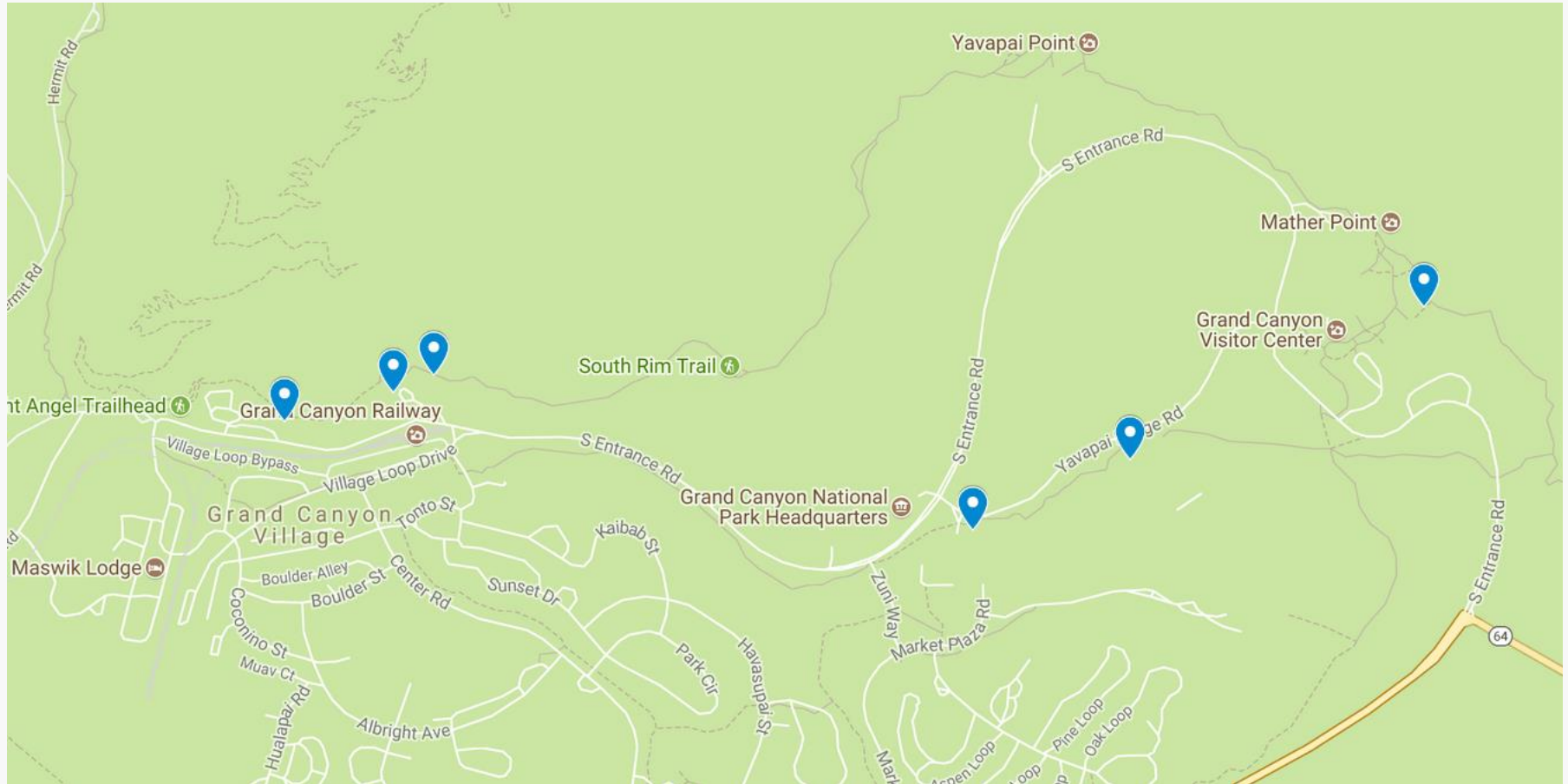
- Max radius: 0.1 miles
- Min points in cluster: 250







# Now what?



# Time

```
<when>2017-03-30T22:16:05Z</when>
```

```
<gx:coord>-112.1206089 36.0538447 2110</gx:coord>
```

```
<when>2017-03-30T22:15:32Z</when>
```

```
<gx:coord>-112.1206895 36.0541252 2108</gx:coord>
```

```
<when>2017-03-30T22:14:41Z</when>
```

```
<gx:coord>-112.1161455 36.0566548 2117</gx:coord>
```

```
<when>2017-03-30T22:13:41Z</when>
```

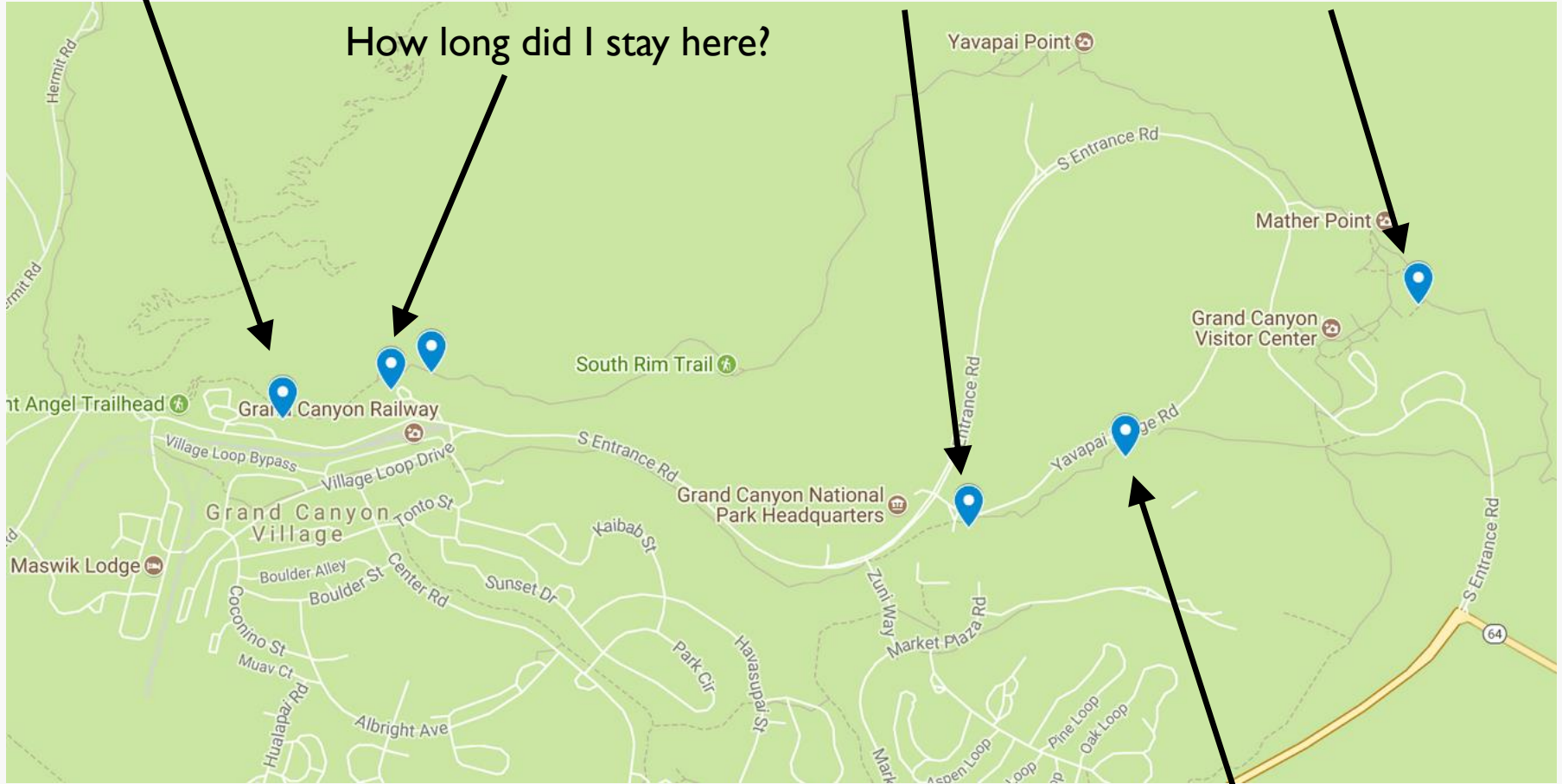
```
<gx:coord>-112.1110006 36.0585582 2123</gx:coord>
```

How often do I visit this location?

Am I usually here on weekends, weekdays, or both?

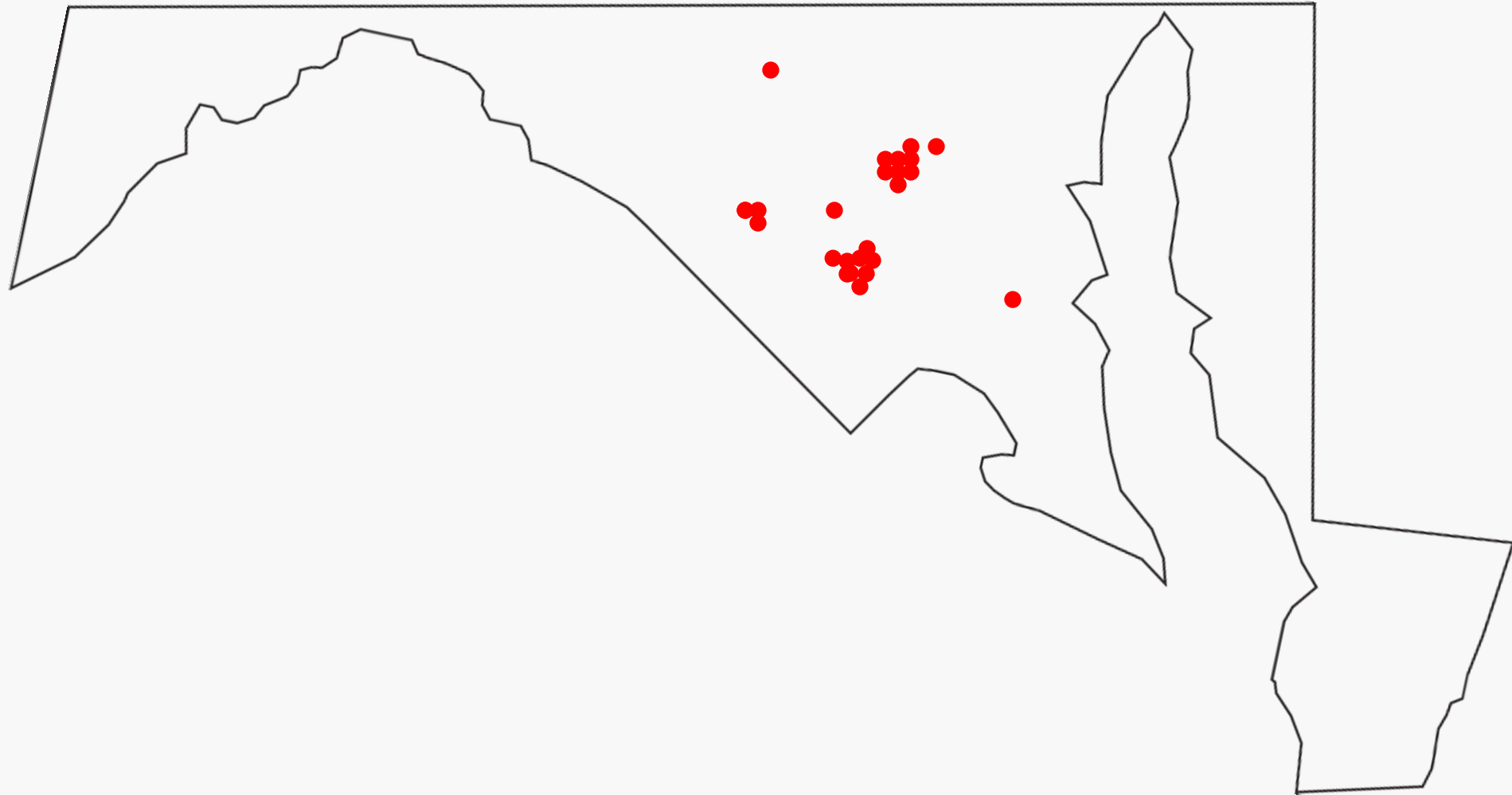
When was I last here?

How long did I stay here?



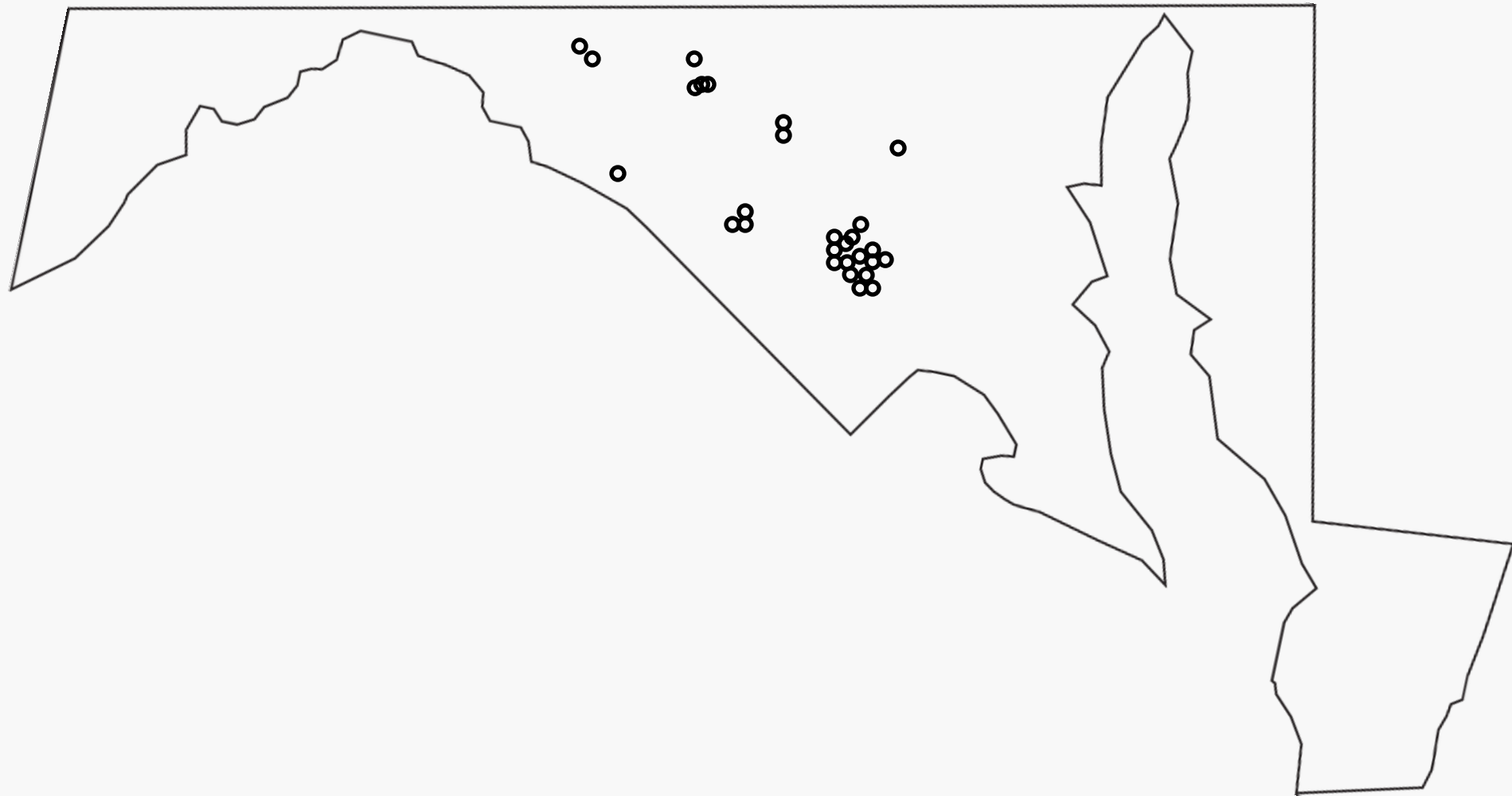
What times of the day can I be found here?

# Weekday Day Points

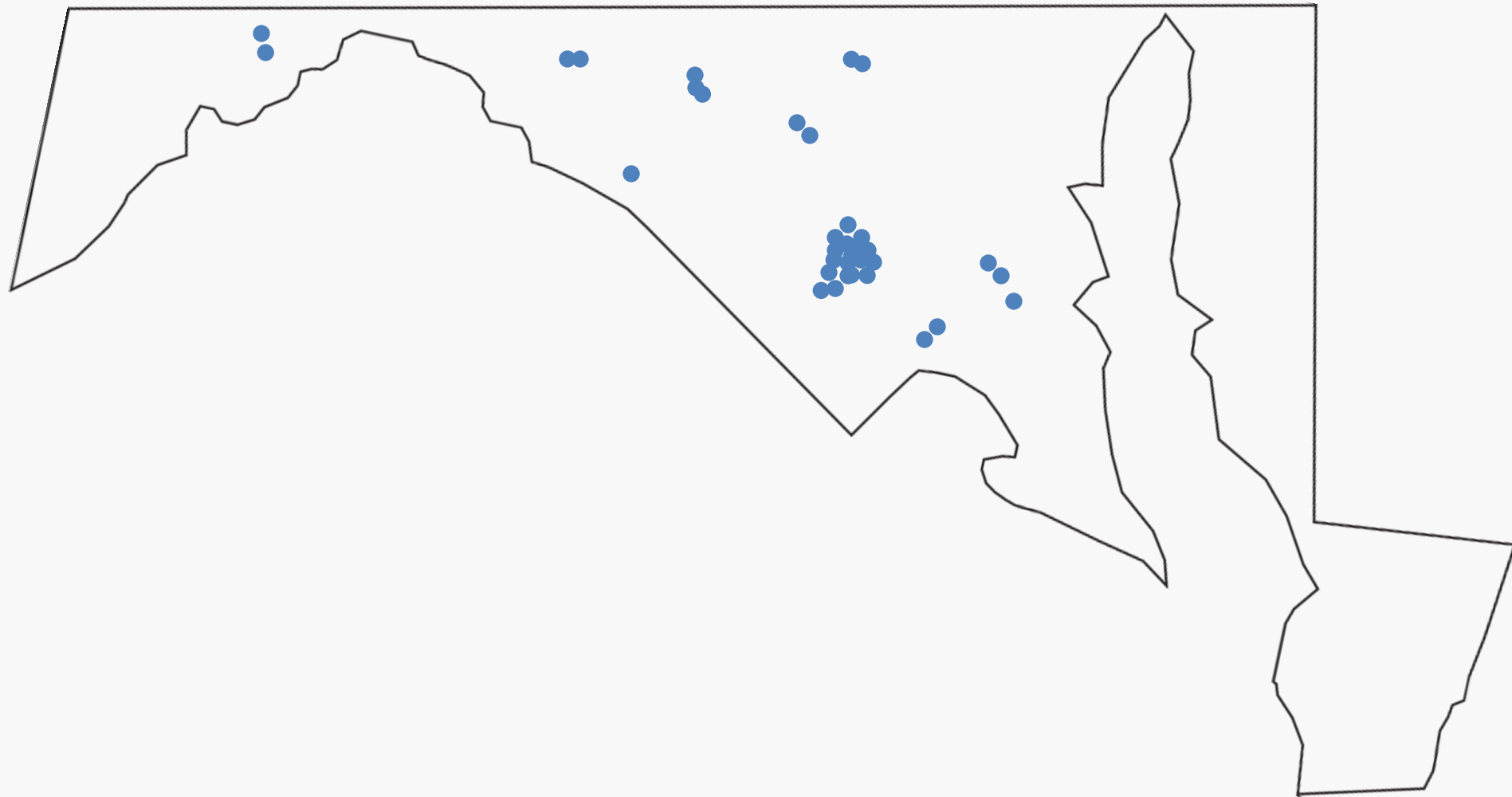




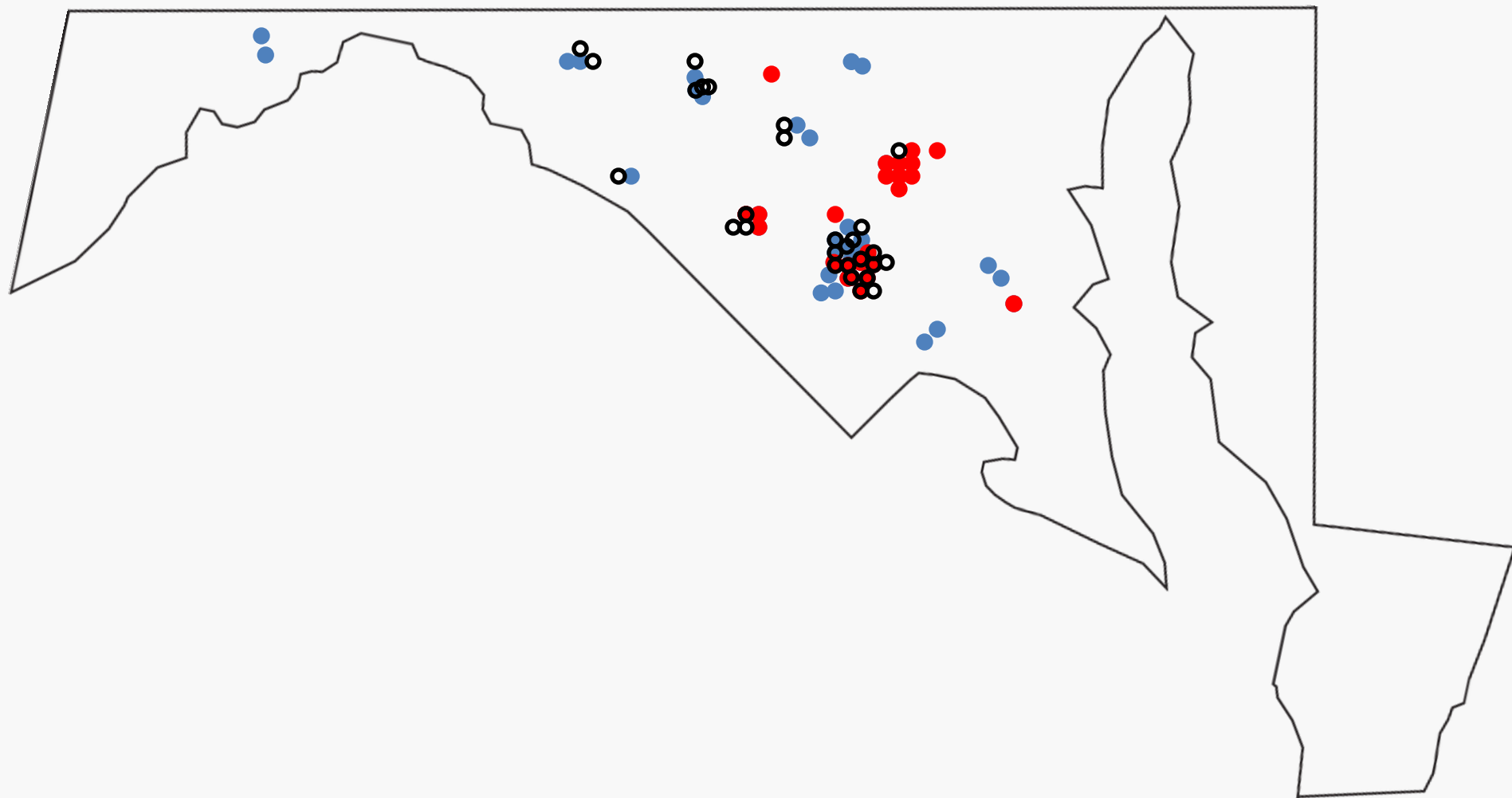
# Weekday Evening Points



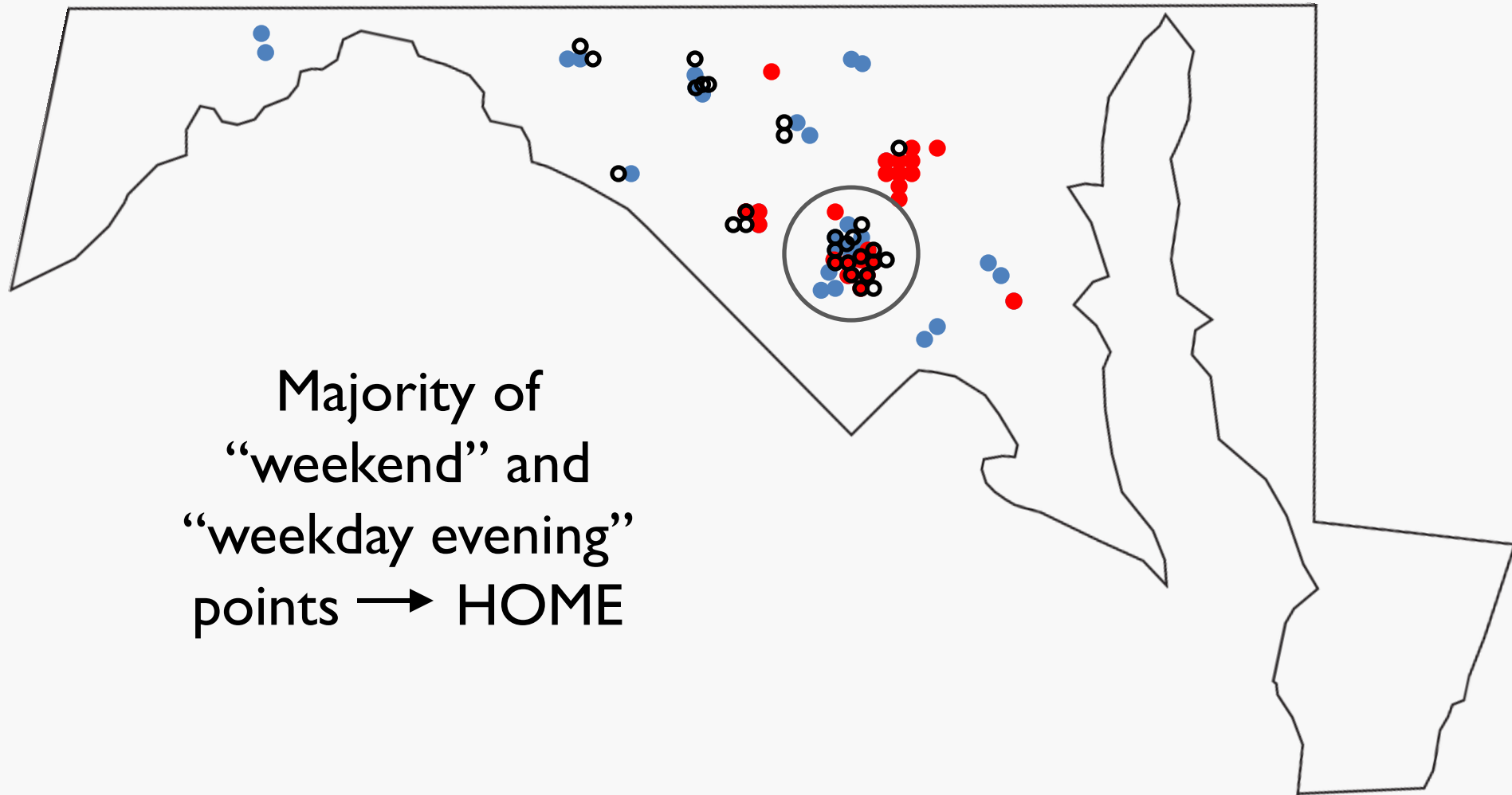
# Weekend Points



# All Points



# All Points



# If I Know Where “Home” Is, I Can Calculate...



- At a specific time / day of week, probability of being:
  - home
  - away from home

# If I Know Where “Home” Is, I Can Calculate...



- Average local travel radius
  - Use points when I was home at **both** the beginning/end of the day
- Max local distance I've driven

# If I Know Where “Home” Is, I Can Calculate...

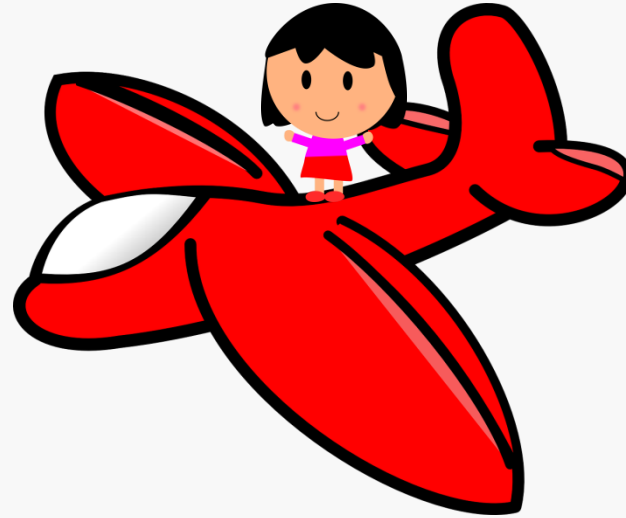


- Distance from home for any recorded data point
  - Assume traveling if:
    - Not home that day
    - Home only at beginning/end of day

# Significant Location Types Identified



**HOME**



**AWAY  
FROM  
HOME**

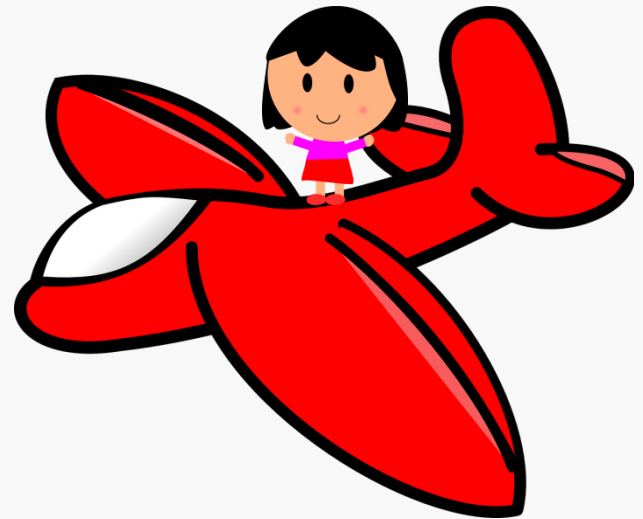


**LOCAL**



# Away From Home Locations

- Hotels (Vacation, Conferences)
- Tourist Venues



# Local Locations

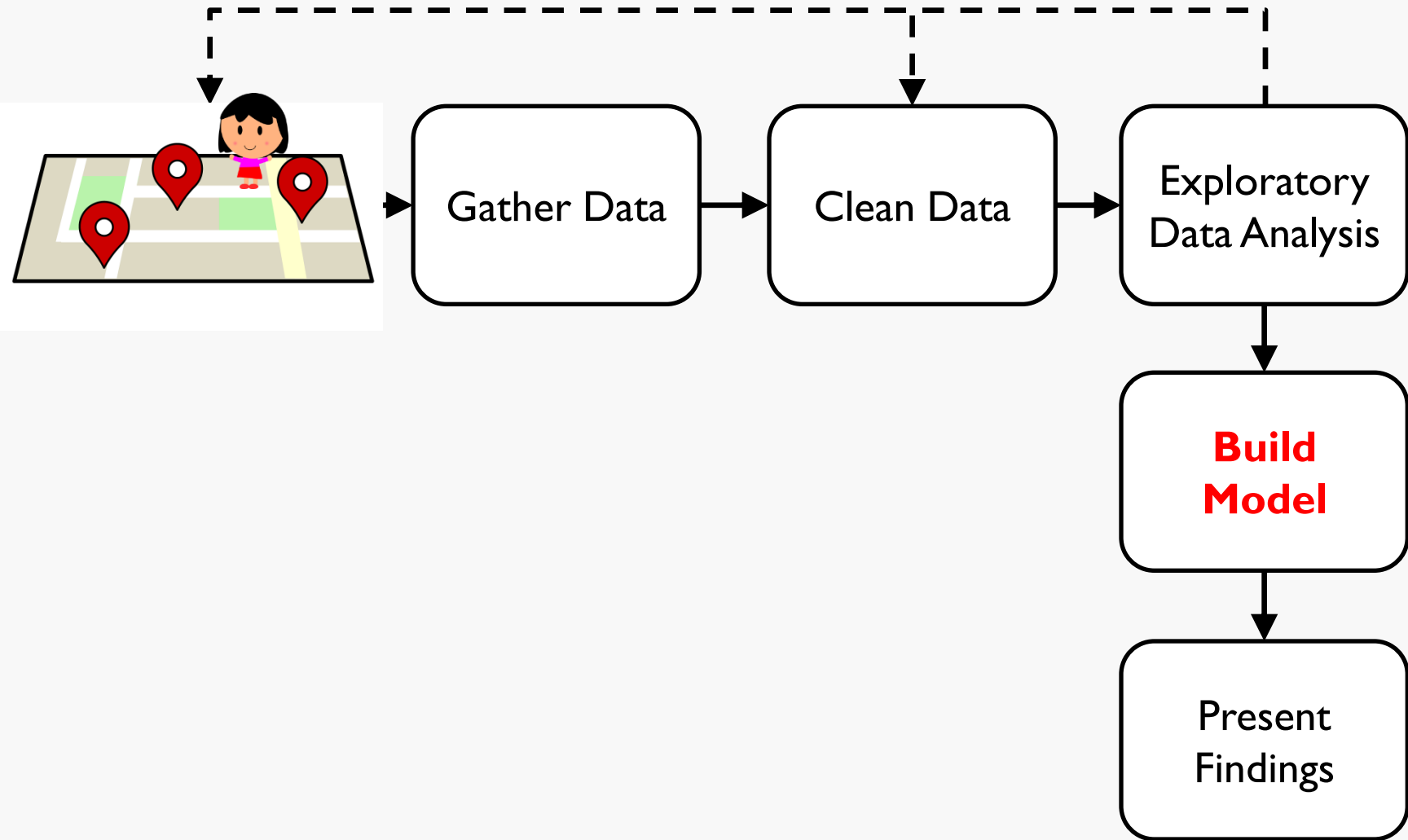
- Favorite Breakfast / Lunch / Dinner Spots
- Grocery Store
- Running Trails



# Specific Local Locations

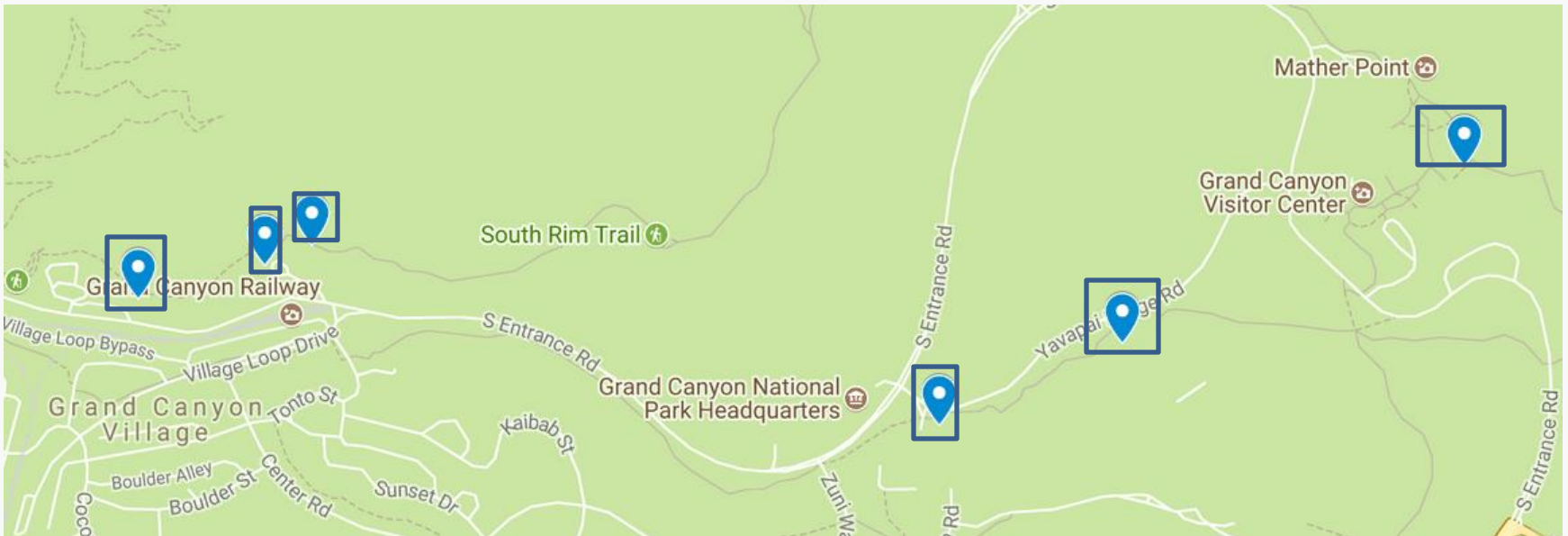
- *Work*: If I'm there on weekdays  $\geq 5$  hours
- *Weekend*: If I'm only there on Sat / Sun
  - Concert Venue
- *Same Day*: If I'm only there on a specific day of the week
  - Farmer's Market
  - Trivia Night

# Data Science Process



# Significant Location Details

- Lat / Long Boundaries
- Location Type
  - *AwayFromHome*
  - *Local (home, work, weekend, sameDay)*



# Data Point Details

## Original

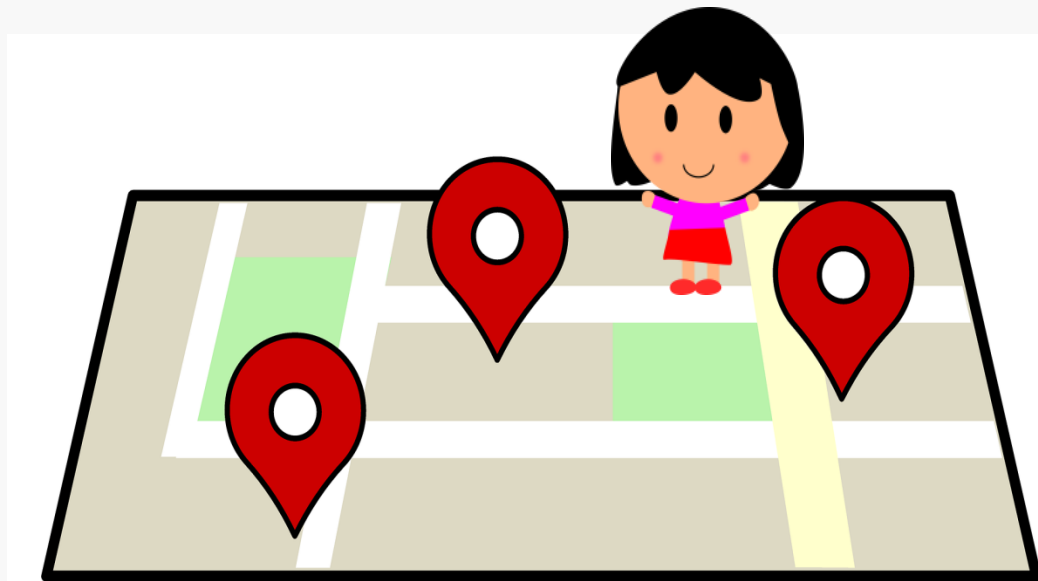
```
<when>2017-03-30T22:16:05Z</when>  
<gx:coord>-112.1206089 36.0538447 2110</gx:coord>
```

## New

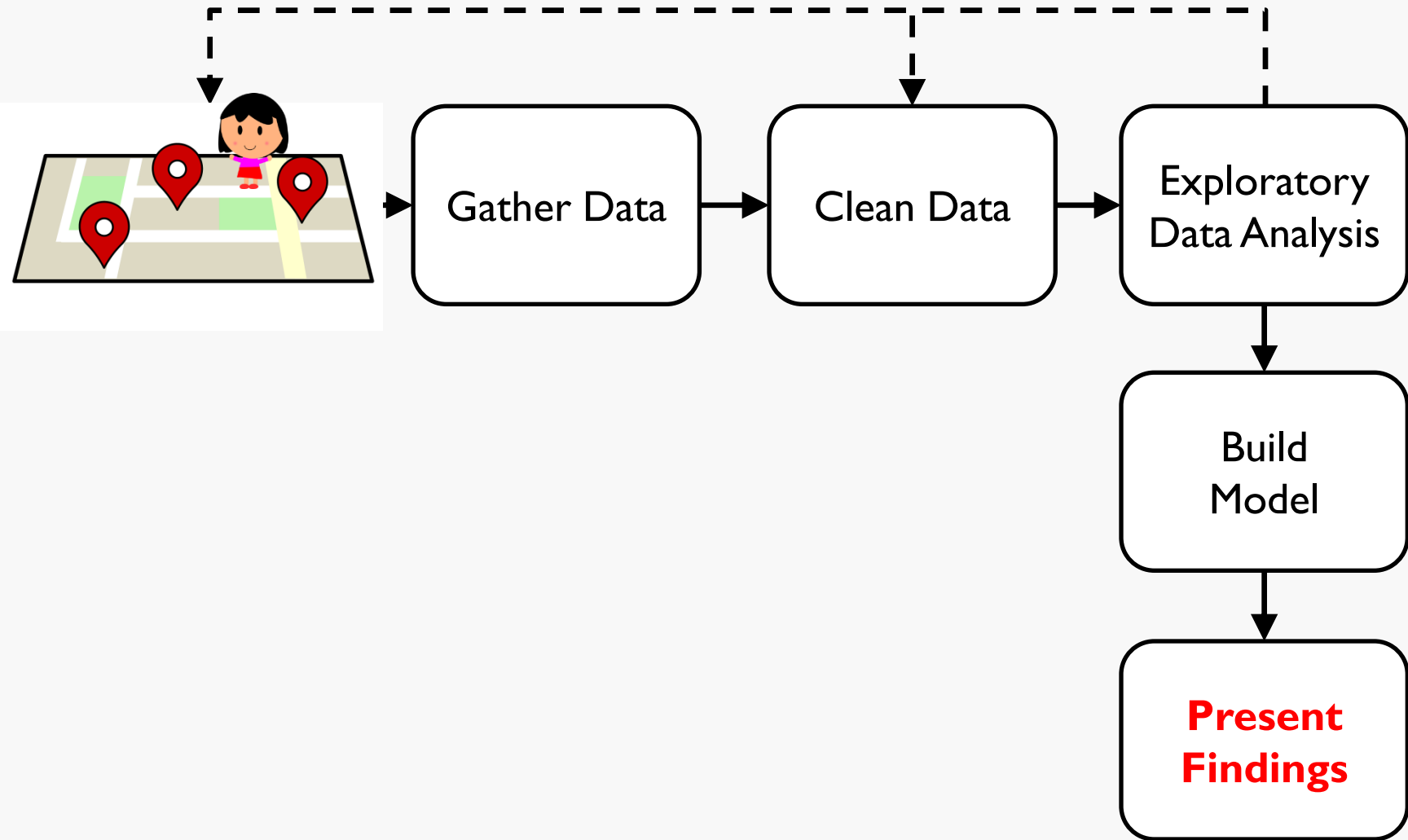
```
<distanceFromHome>1949.46</distanceFromHome>  
<locationLabel>cluster3</locationLabel>  
<description>awayFromHome</description>
```

# Model of Me

- Dates away from home
- Local travel radius
- Likelihood of being at a location by day / time
- Significant locations



# Data Science Process





**DEMO**

# Questions I can ask the data

- Where was I on August 9, 2017 at 2:18PM ?
- Predict where I will be on Monday at 8:45AM.
- Predict when I am likely to be away on Saturday.
- Predict whether I'll be home on Sunday at 10PM.

# Expanded Questions

- How many days was I out of town in July?
- When was I at work on a weekend?
- How many times did I visit the grocery store last month?
- How long does it usually take to drive to work?
- When was I last at the Grand Canyon?

# **DISCUSSION**

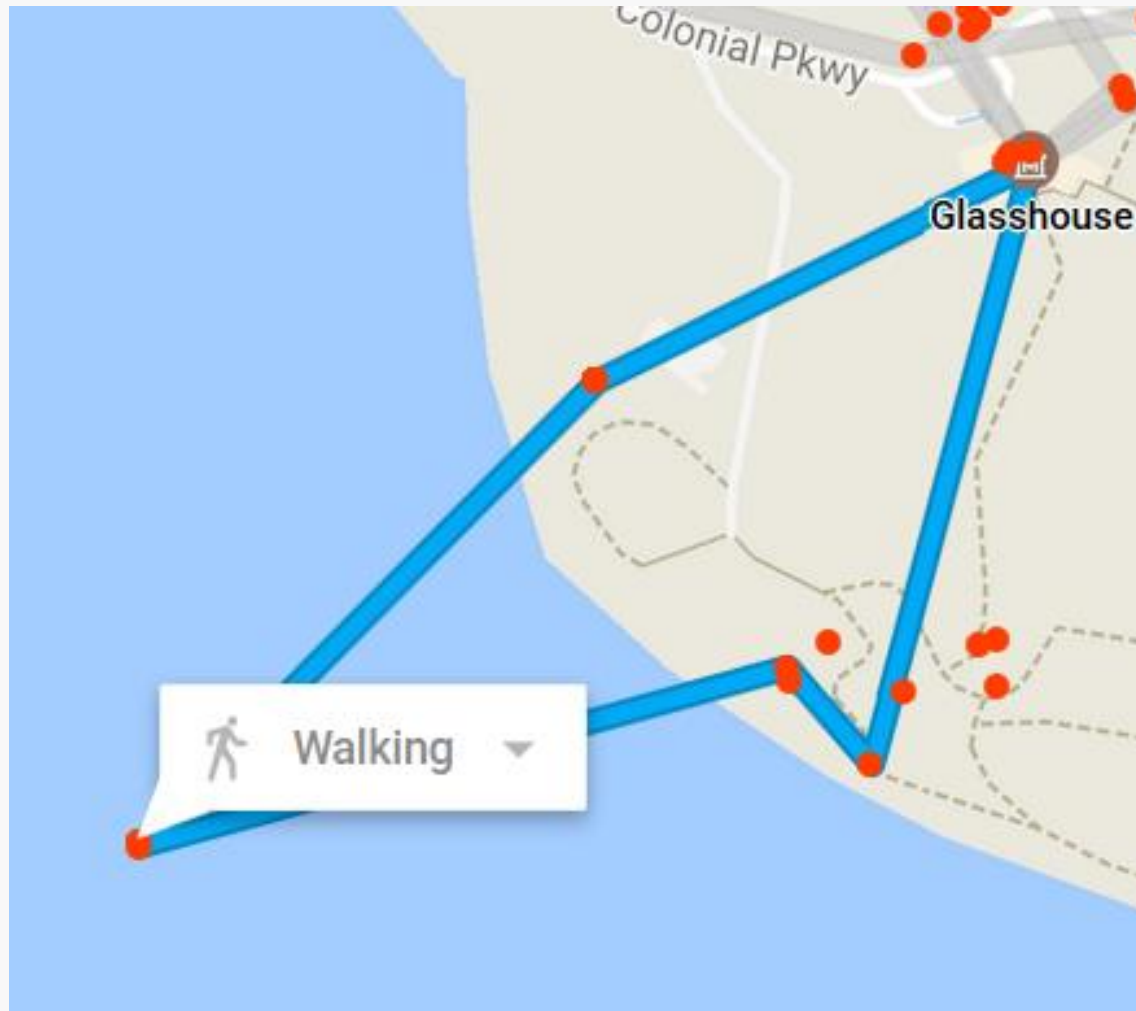
# Assumptions

- Regular schedule
- “Normal” work habits
- Home
  - More often than anywhere else
  - More often on weekday evenings & weekends

# When This Doesn't Work

- Irregular schedule / lots of travel
- Not enough points
- Bad technology
  - signal
  - hardware

# Bad Technology



# Cautions

- Analysis is a general pattern of behavior
- Locations may be inaccurate (Google itself asks for corrections)
- A long traffic light can be a “location”

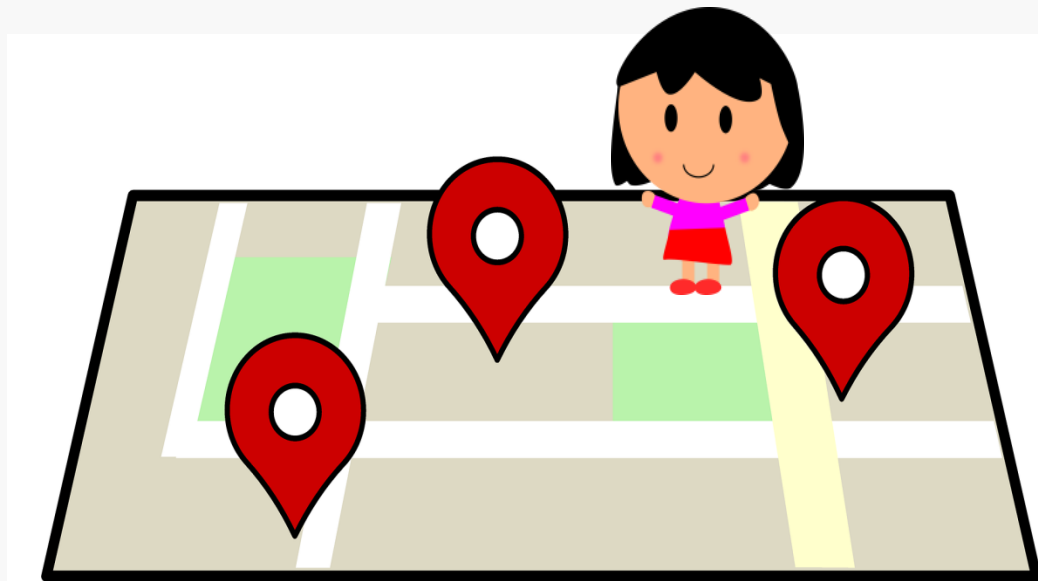


Google already asks for  
your Home and Work  
addresses...

...which means that they  
already know your  
significant locations!

# Would you share this info?

- Dates away from home
- Local travel radius
- Likelihood of being at a location by day / time
- Significant locations



# Who could have it

- Products and apps
- Companies that access data
- Companies that buy / share data

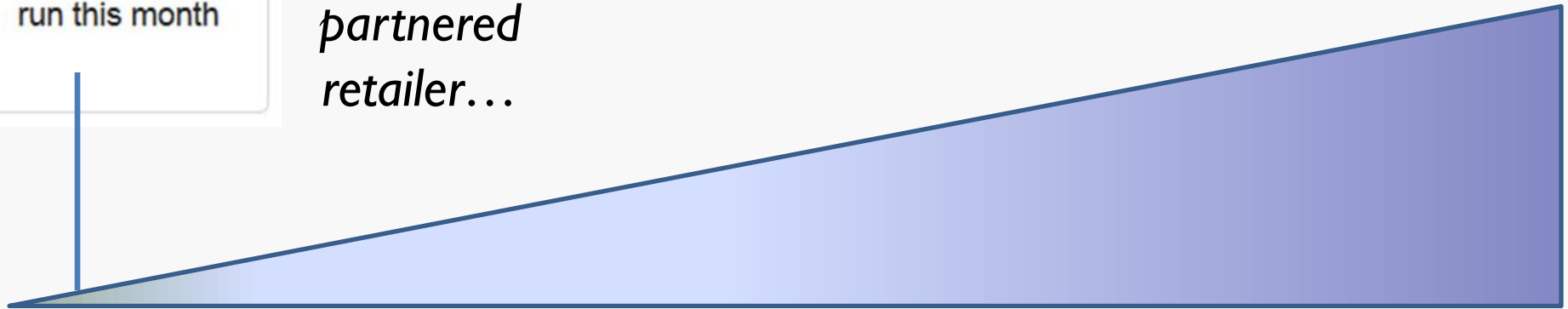


# Implications



22 mi (37 km)  
run this month

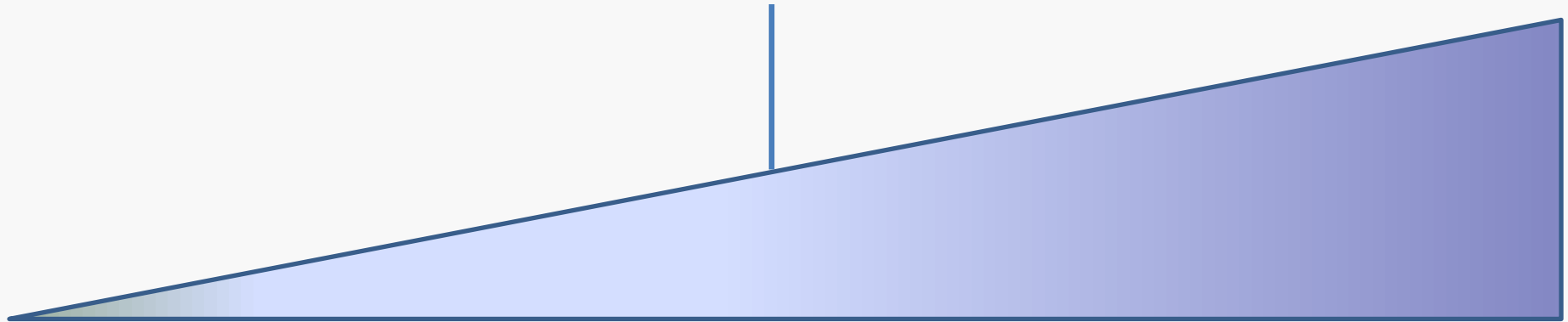
*Time to buy  
new shoes!  
Get \$20 off  
at a  
partnered  
retailer...*



Benign

# Implications

*Your insurance  
claim was  
denied due to...*

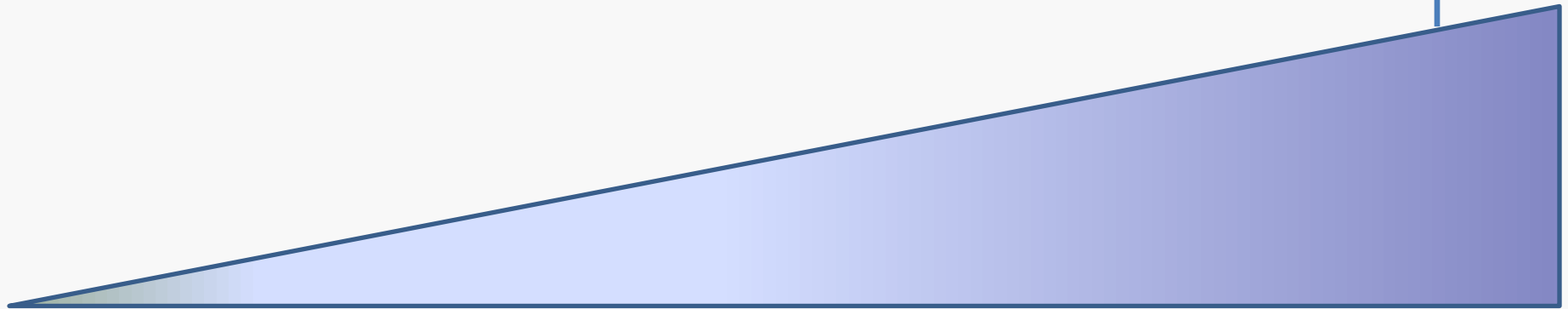


Benign

Worrisome

# Implications

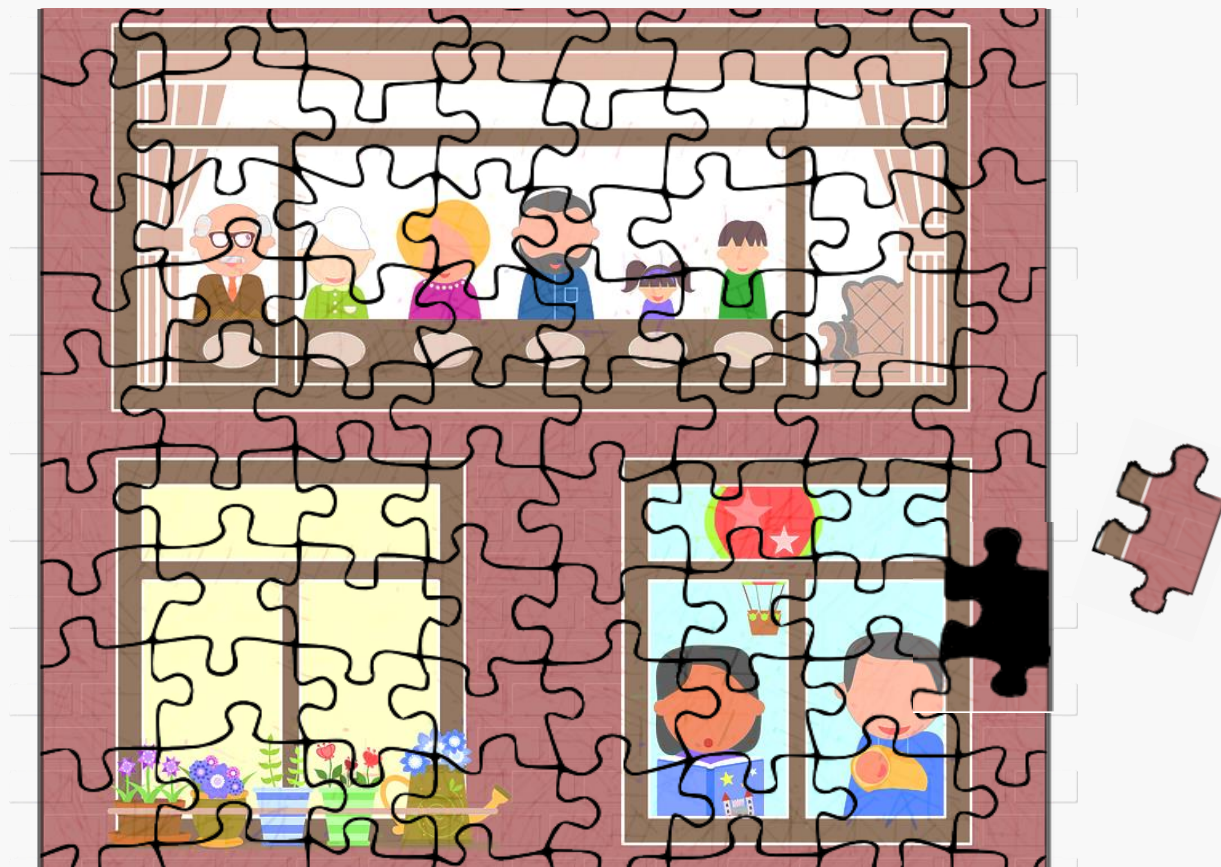
Companies  
share data



Benign

Worrisome

!@#\$%^!&





**Your Data, Your Choice**

# Further Information

- Code (Jupyter Notebook)

<https://github.com/laconicllama>

- Contact

[laconicllama@hotmail.com](mailto:laconicllama@hotmail.com)

# Questions?

