

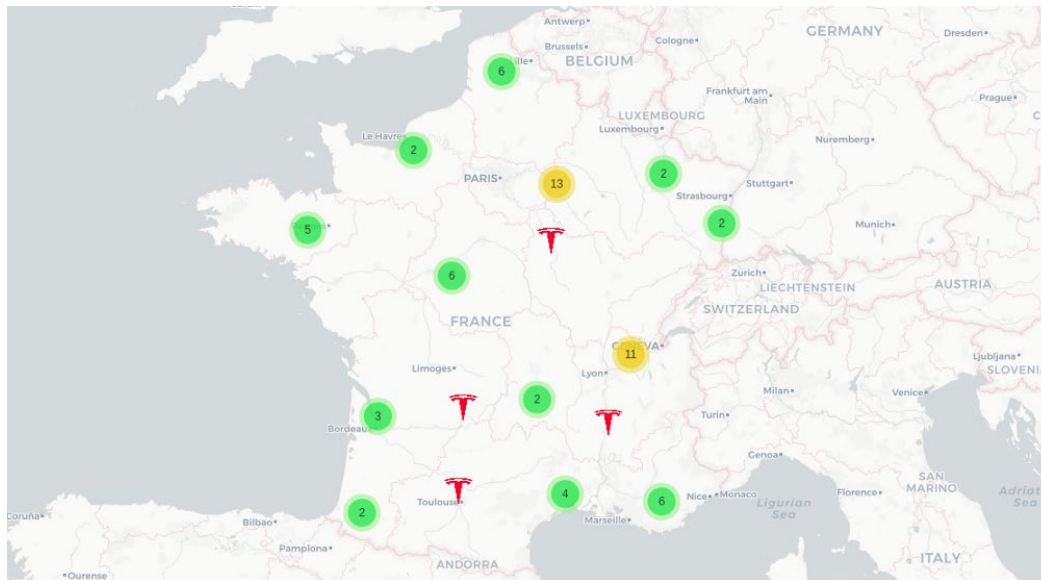
Spatio-Temporal Data Mining & Trajectory Data Mining

- Di Wang, Tomio Miwa and Takayuki Morikawa (2020). Big Trajectory Data Mining: A Survey of Methods, Applications, and Services <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC7472055/>
- Yu Zheng (2015) Trajectory Data Mining: An Overview, ACM Transactions on Intelligent Systems and Technology, 6(3): 29
- Microsoft 2007. <https://www.microsoft.com/en-us/research/project/trajectory-data-mining/>

- Geospatial/ Spatial-temporal data mining
 - Definitions, techniques & use cases
 - Trajectory data mining
-

What is Spatial?

- Relating to or existing in space only
- Take a look at <https://blog.locale.ai/>



What is Spatio-Temporal?

- Relating to *both space and time*
- Spatial (location) and temporal (time) attached
- Changes and movements over time

<https://www.flightradar24.com/51.5,-0.12/6>

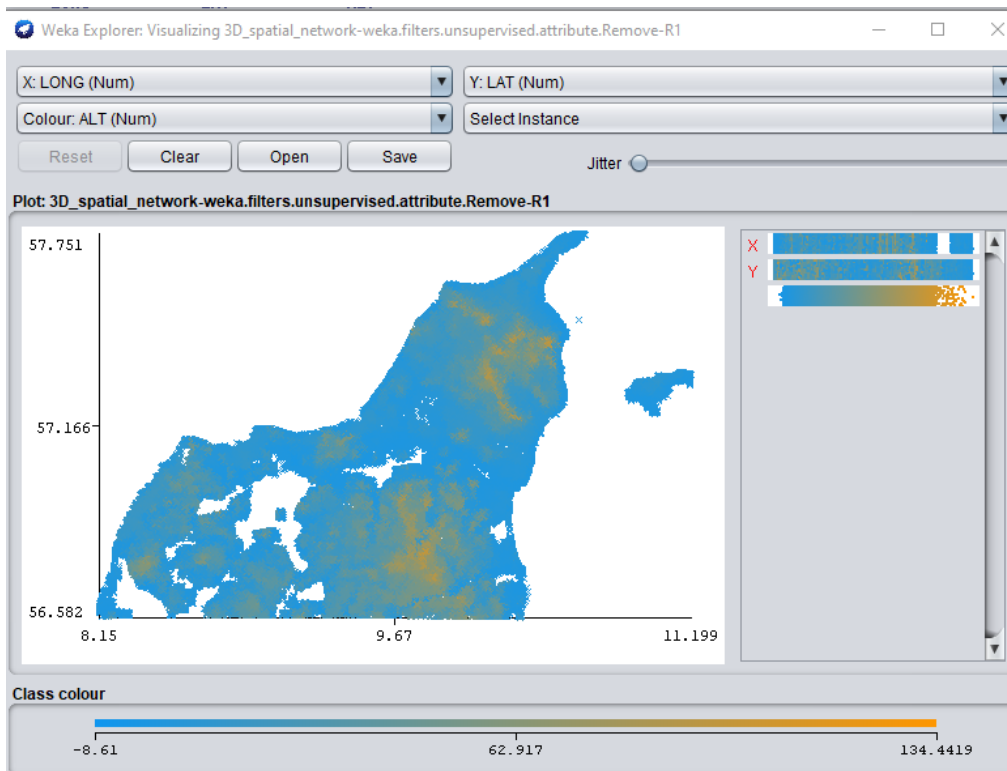


Geo-Spatial data

- An eg. of the data

| VOLCANX020 | NAME | LOCATION | LATITUDE | LONGITUDE |
|------------|-------------|---------------|------------|--------------|
| 509 | Baker | US-Washington | 48.7767982 | -121.8109970 |
| 511 | GlacierPeak | US-Washington | 48.1118011 | -121.1110001 |
| 513 | Rainier | US-Washington | 46.8698006 | -121.7509995 |
| 515 | St.Helens | US-Washington | 46.1997986 | -122.1809998 |

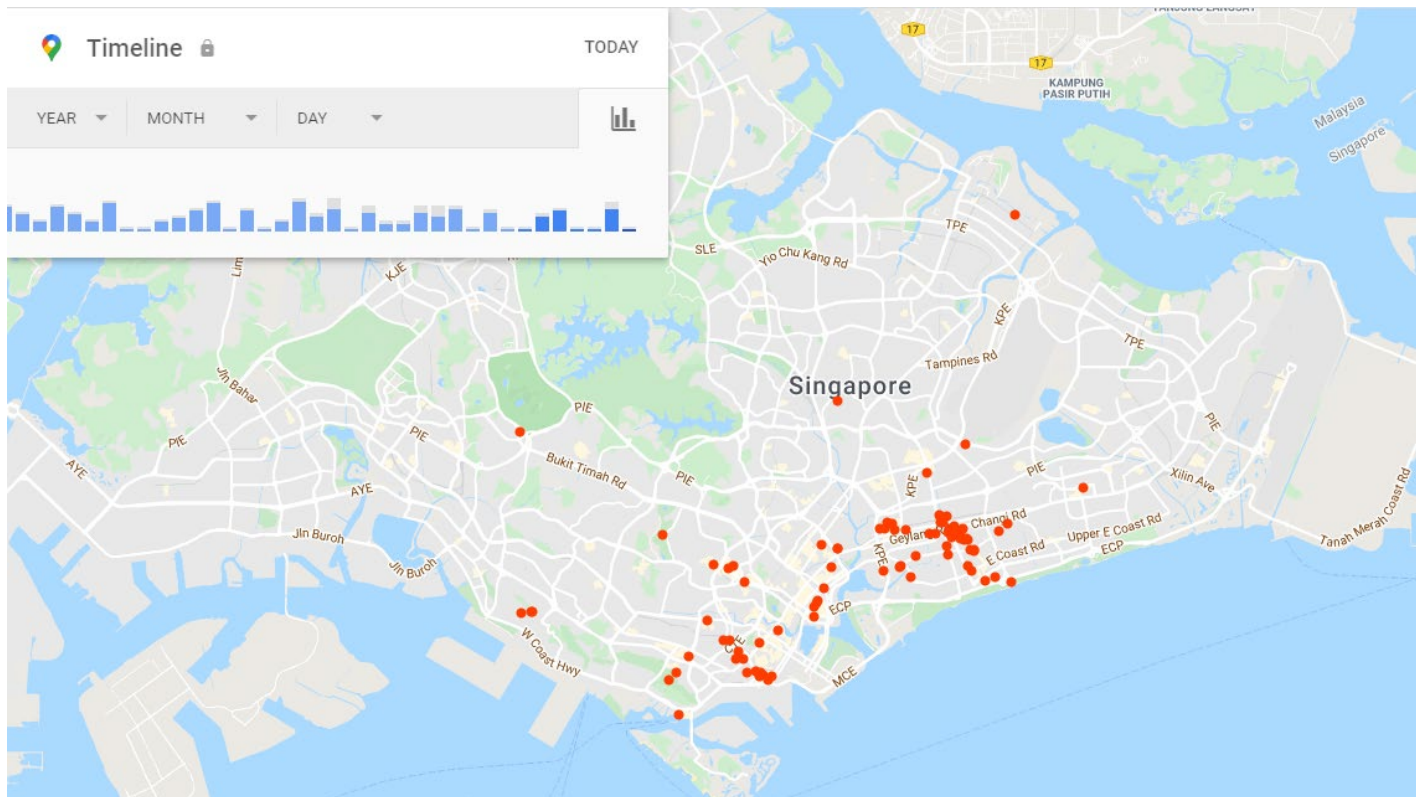
Geospatial data



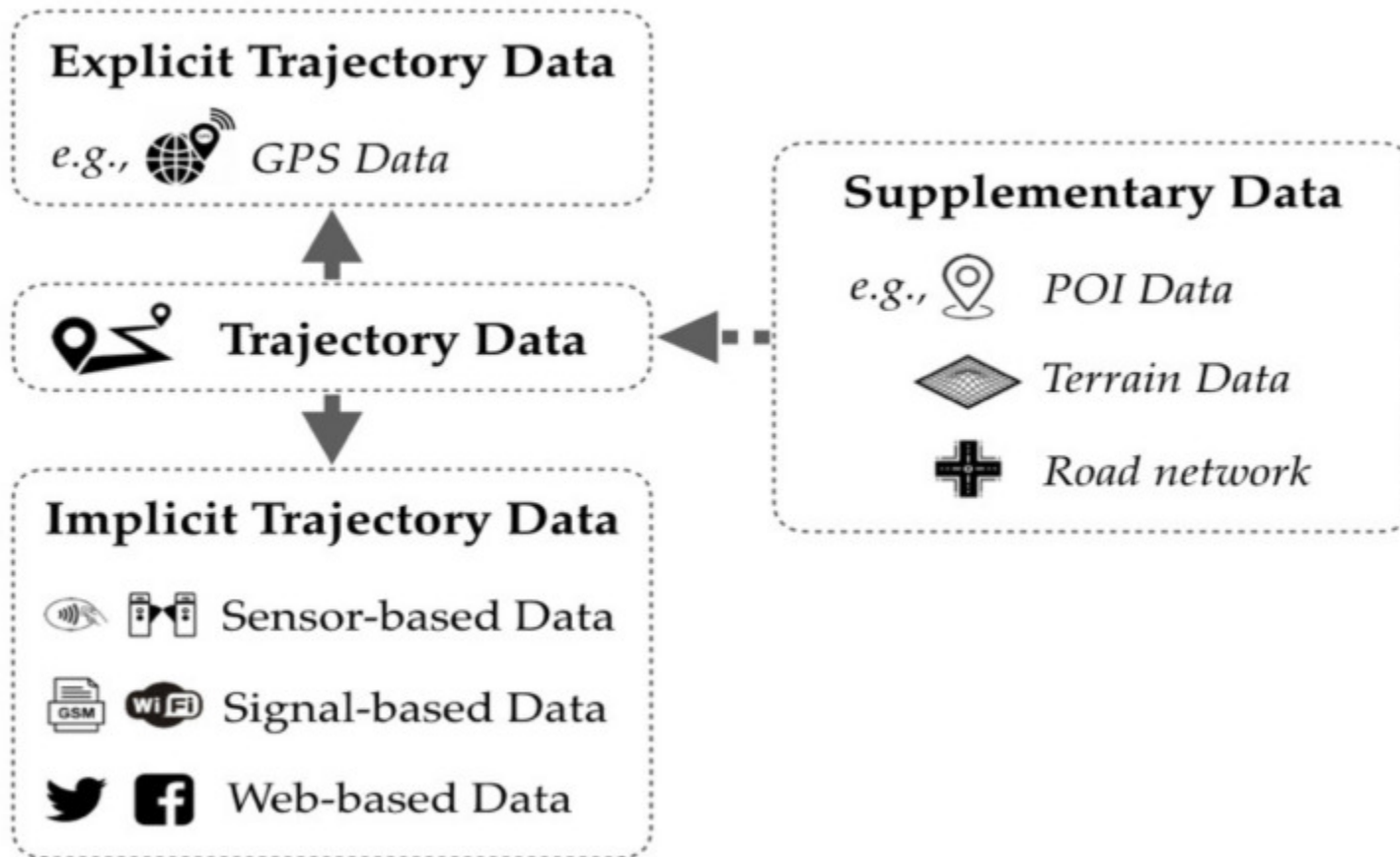
We have looked at it before!
3d_spatial.arff in Lab 4.1

Spatial-temporal Data

- Best example: google maps timeline



Data sources



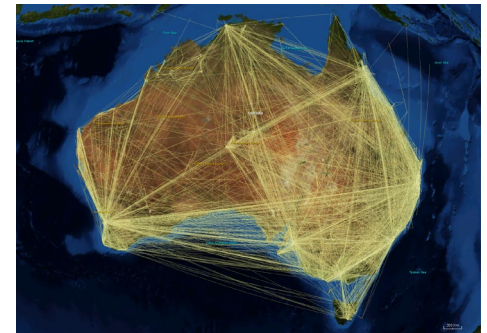
- Space and time are ubiquitous aspects of reality
 - We are living in a space with time dimension...
 - Thus basically all human (things) related data are spatio-temporal in nature
 - Advances in automatic (semi-automatic) data generators (sensors, RFID tags, GPS receivers, mobiles etc) result in MASSIVE spatio-temporal data
 - It is believed that more than 95% of business data are spatial or spatio-temporal
-

- Geospatial -> space only
 - Geospatial temporal -> space and time
 - *Trajectory data mining* => + ***Movement***
 - “a trace generated by a moving object within a certain spatiotemporal context and is generally represented by a series of chronologically ordered points.” (Zhang 2014)
-

Trajectory Examples

- Vehicle trajectories (cars, buses, trucks etc)
- Animal movements (birds, sharks etc)
- People movements (tourists, photo-takers, students etc)
- Mouse click movements (HCI, software design etc)

Can you name other eggs?



- Animal movements
 - Cows frequent visits to shades, but rare visits to grazing areas => indication of sickness?
 - Bees periodic visits to hive from flowers => useful for beekeeping
 - Human movements
 - Frequent visits to fast food restaurants but rare visits to gyms/parks/beaches => indication of health risk
 - Frequent visits to Indian/Korean/Japanese restaurants => Asian?
-

Moving Objects



Questions to ask: Where, when, why, what?

Raw GPS Trajectory

```
dog_walk.txt
1, -16.898353, 145.694882, 2016-11-06T18:13:07, MOVING
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1, -16.898271, 145.694818, 2016-11-06T18:13:10, MOVING
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1, -16.898232, 145.69477999999998, 2016-11-06T18:13:13, MOVING
1, -16.898222, 145.694772, 2016-11-06T18:13:14, MOVING
1, -16.898207, 145.694757, 2016-11-06T18:13:15, MOVING
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1, -16.898584, 145.694296, 2016-11-06T18:14:05, MOVING
1, -16.898592, 145.694288, 2016-11-06T18:14:06, MOVING
1, -16.898606, 145.694274, 2016-11-06T18:14:07, MOVING
```

How the data looks like?

Overview of Trajectory Data Mining

Filtering noise*

Map-matching*

Interpolation

Stay point detection*

Preprocessing

Periodic Patterns*

Regions-of-Interest

Sequential Patterns

Trajectory Clustering*

Trajectory Patterns

Trajectory Classifier*

Pattern Mining



Spatio-temporal Trajectories

- What?
 - The process of fitting raw trajectory recordings onto an underlying map structure before data mining
 - How?
 - Very different from ‘structured data’
 - The idea is how do you combine location (map) with time data?
 - Noisy with GPS etc.
-

Issues with GPS Trajectories

- Spatial uncertainties
- Errors and noisy
- Irregular
- Could be too densely recorded or too coarsely recorded

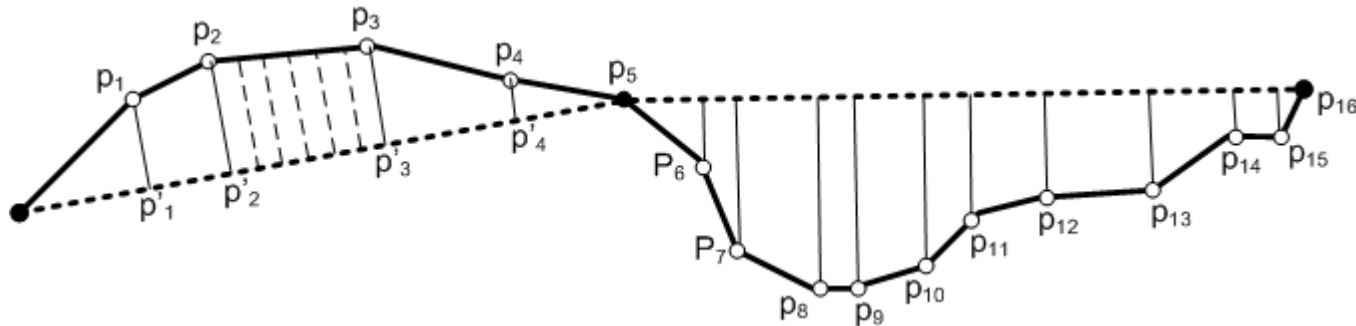
→ Preprocessing

Trajectory Simplification

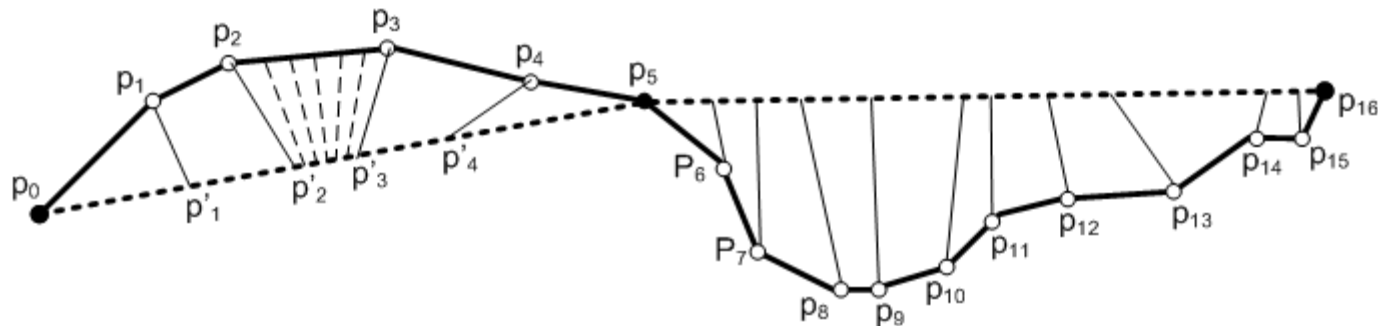
- Aim
 - Reduce the complexity of an input trajectory
 - Sensors capture as much movement details as possible by oversampling but still want to preserving the motion of the tracked entity
 - Performance metrics
 - Reduce processing time
 - Reduce Error measure
 - What error measure?
 - Criteria include perpendicular Euclidean distance and time synchronized Euclidean distance
-

Illustration of Error Measures

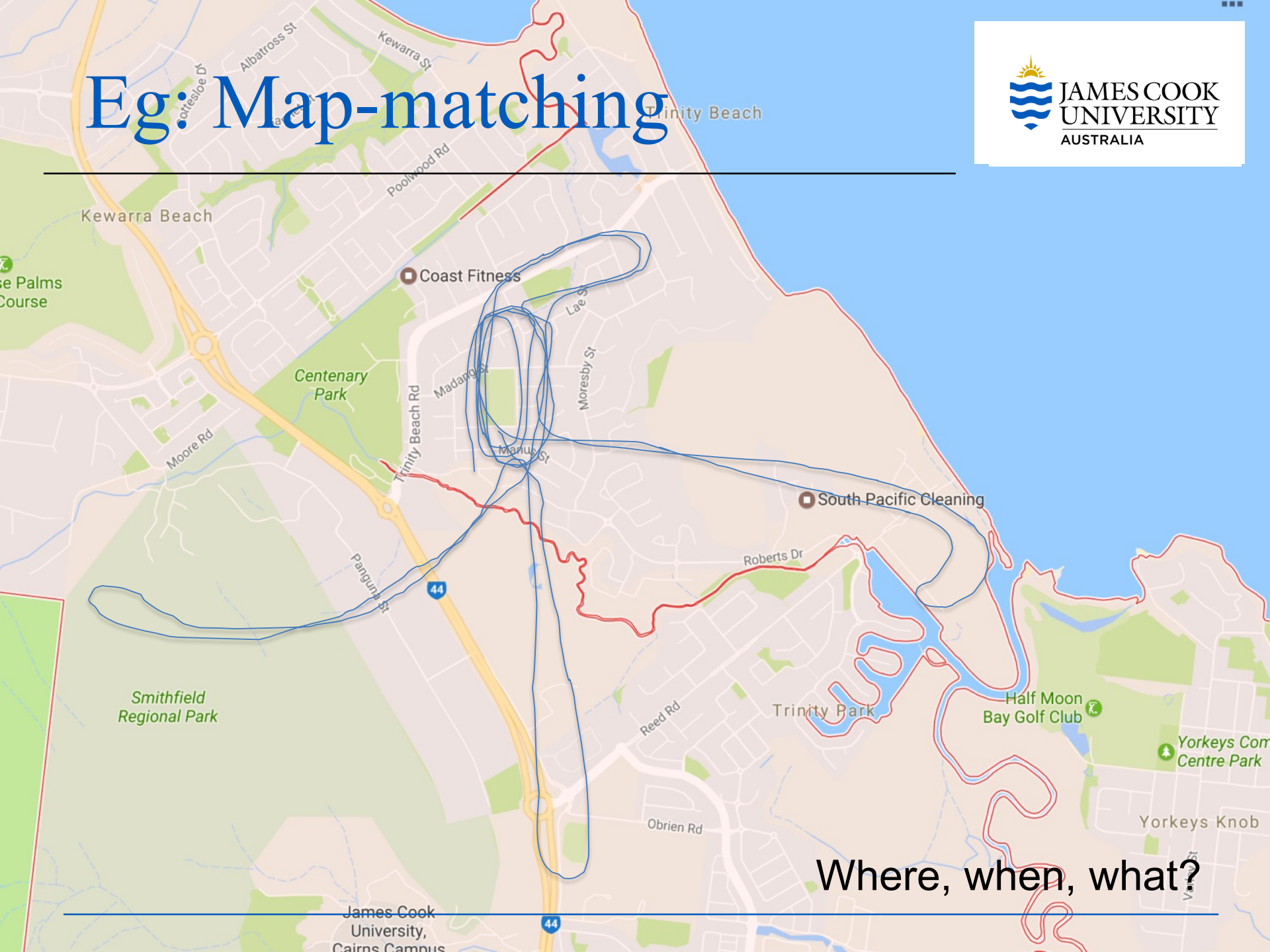
- Perpendicular Euclidean Distance



- Time Synchronized Euclidean Distance

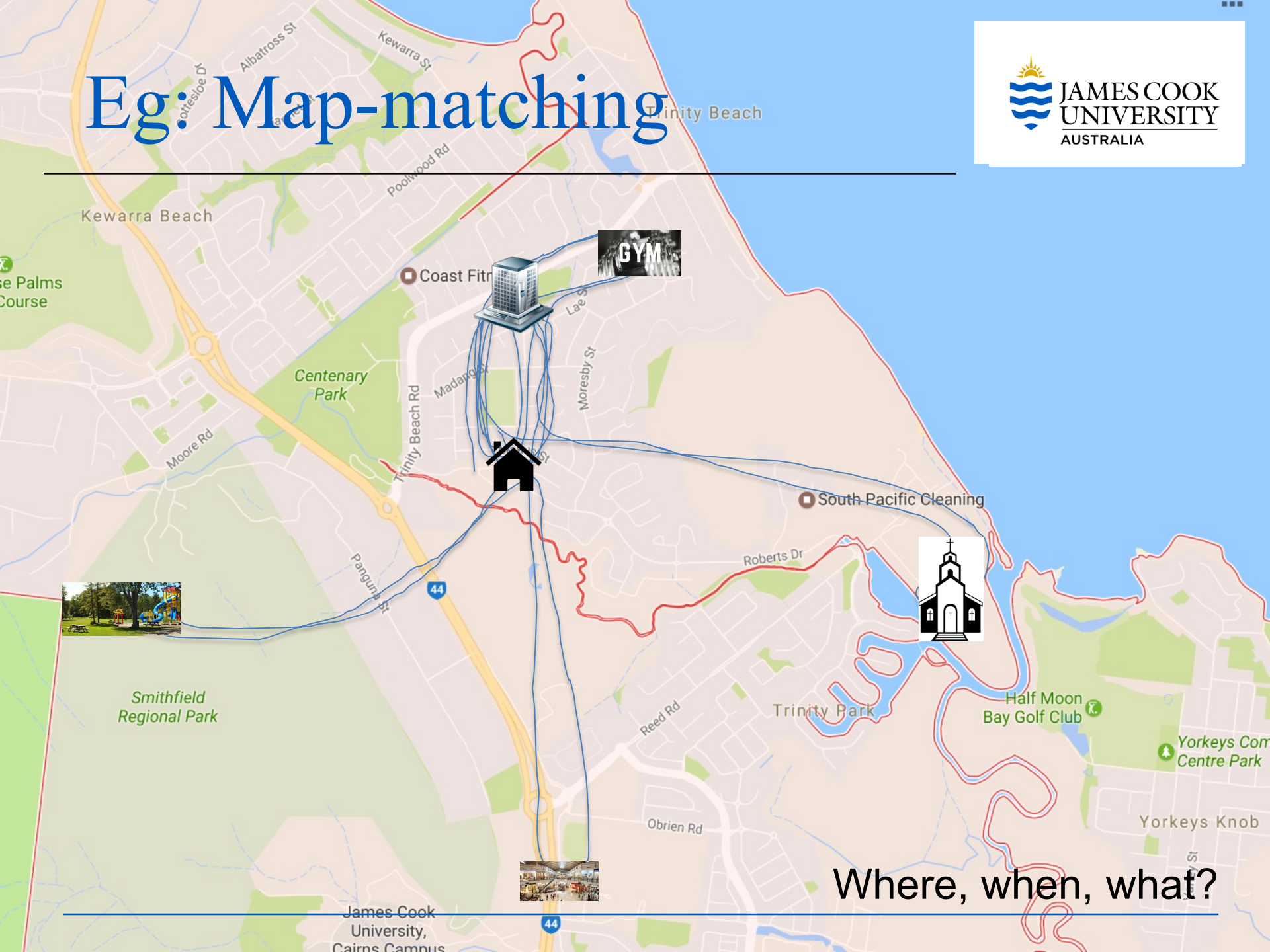


Eg: Map-matching



Where, when, what?

Eg: Map-matching



Where, when, what?

Eg: Map-matching

Monday, Wednesday 12pm

Monday, Wednesday 1pm

9am

5pm

Sunday 11am

Sunday 12pm

Saturday 4pm

Saturday 5pm

Thursday 8pm

Thursday 6pm

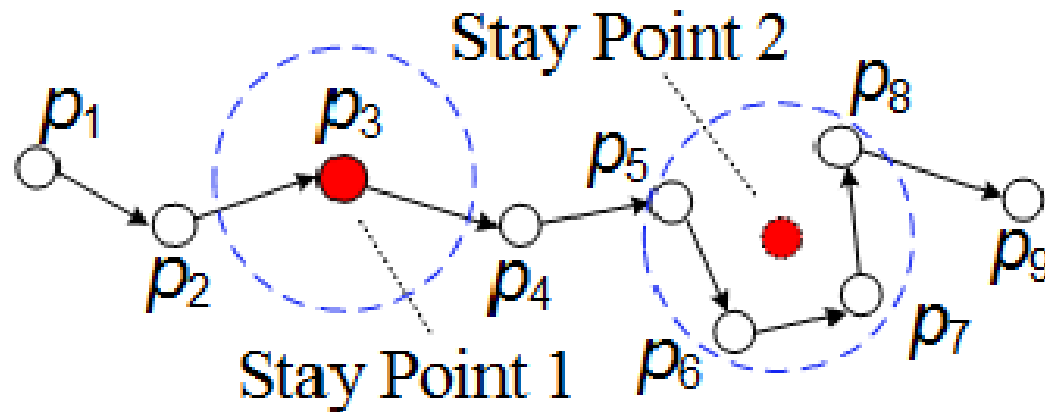
Where, when, what?



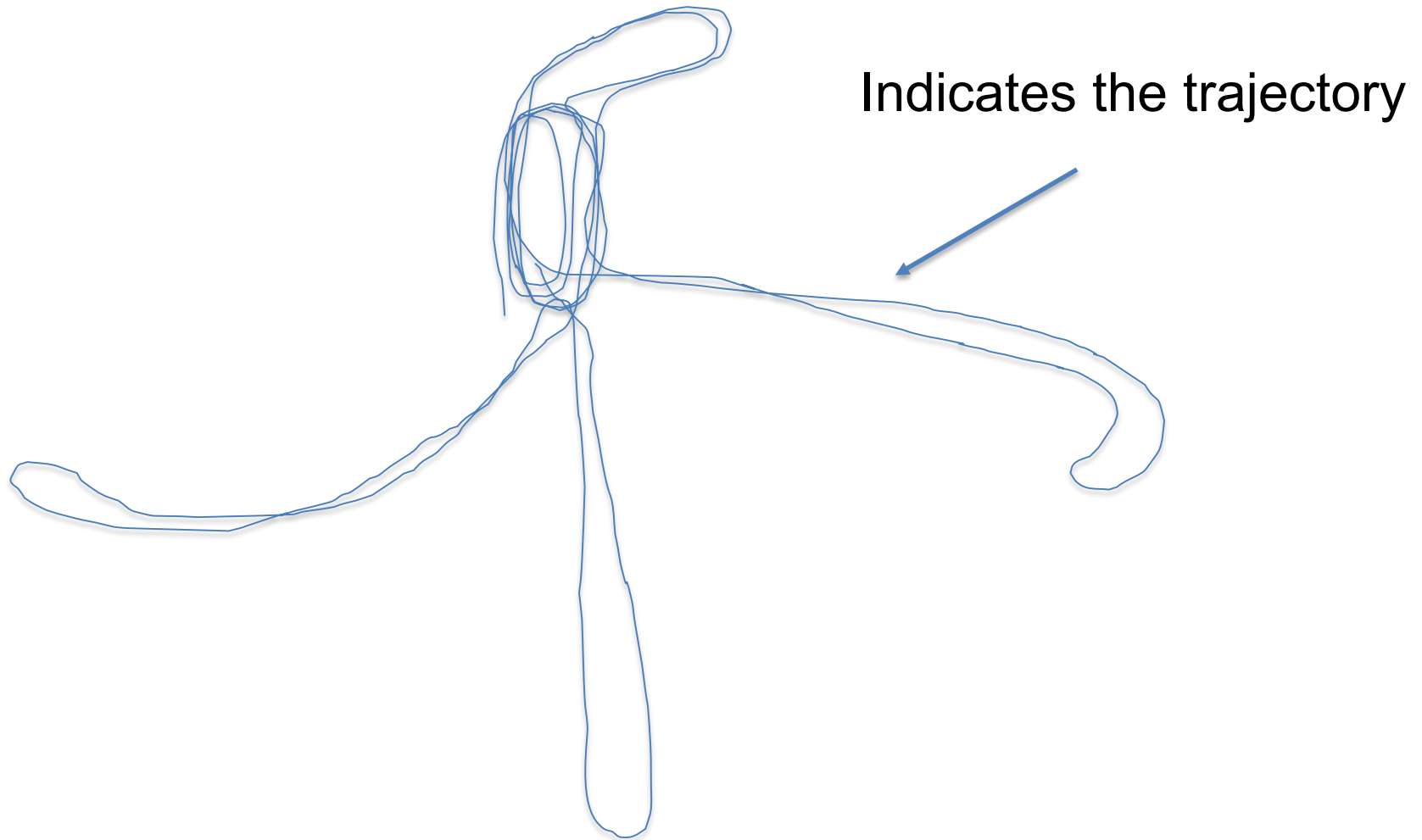
Stay Point Detection

- The identification of a location a moving object has stayed for a while within a certain distance threshold
- These stay points can indicate interesting insights for eg. at a restaurant/ shopping mall.
- Uses clustering technique studied earlier eg DBSCAN

Stay point detection



Stop/Move Representation

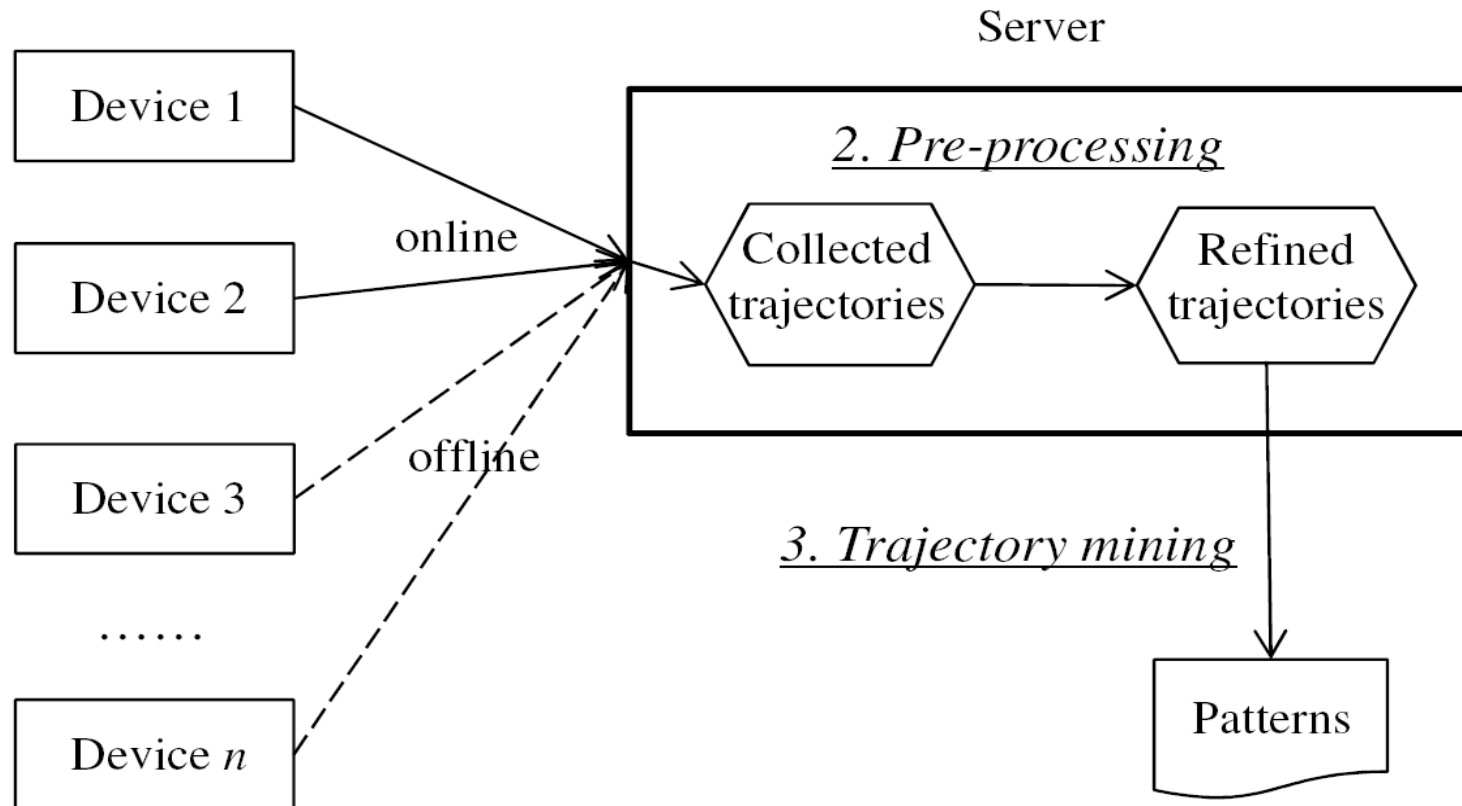


Stop/Move Representation



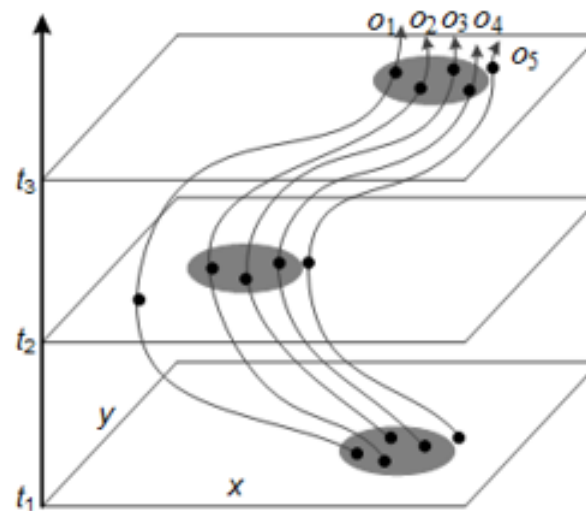
Trajectory Data Mining

1. Data collection

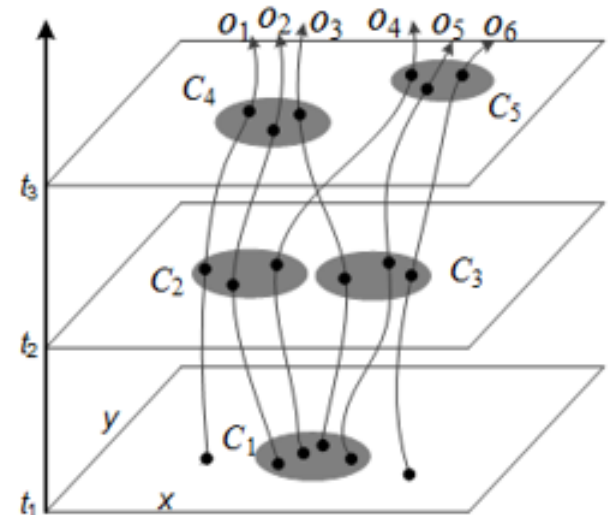


Trajectory Data Mining

- Categories of patterns:
 - moving together patterns,
 - trajectory clustering,
 - periodic patterns and
 - frequent sequential patterns



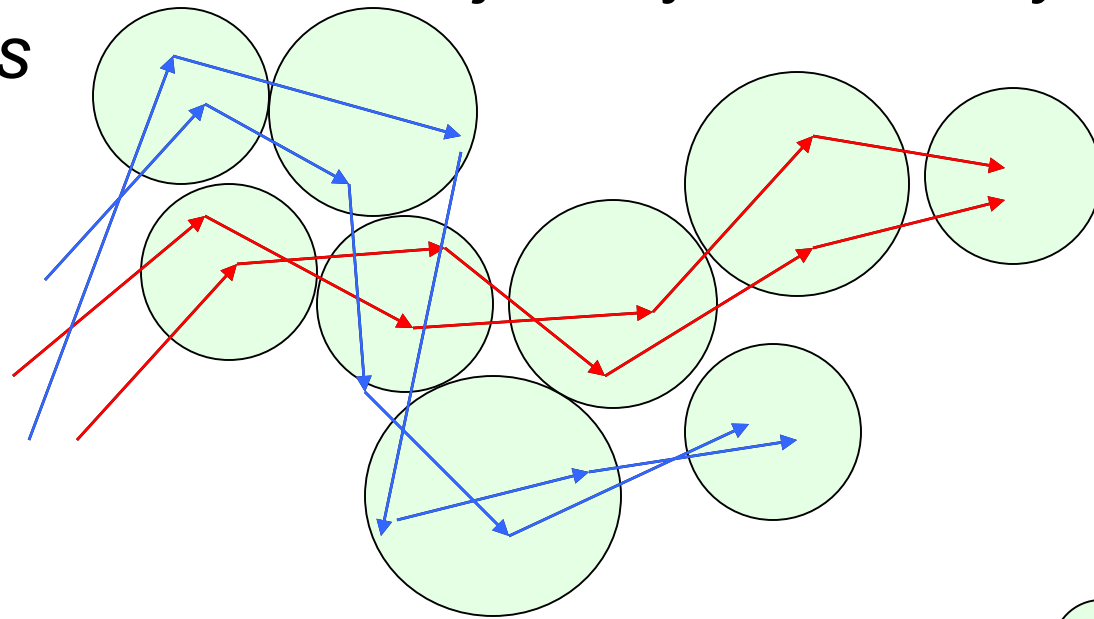
A) Flock, convoy and swarm




B) Gathering

Trajectory Clustering

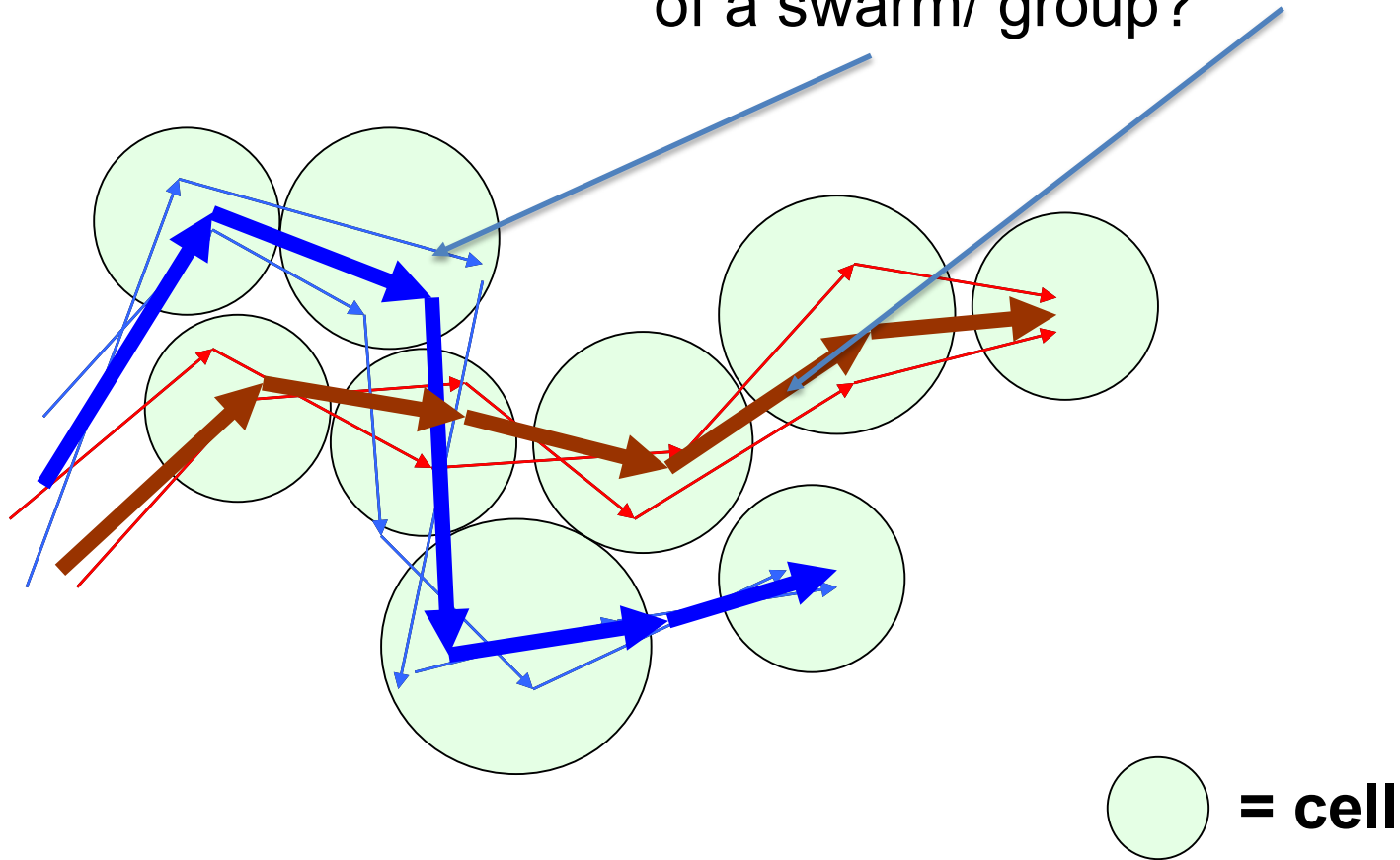
- Group similar trajectories geometric proximity in spatial/spatiotemporal space.
- Find a representative trajectory from *many trajectories*



 = cell

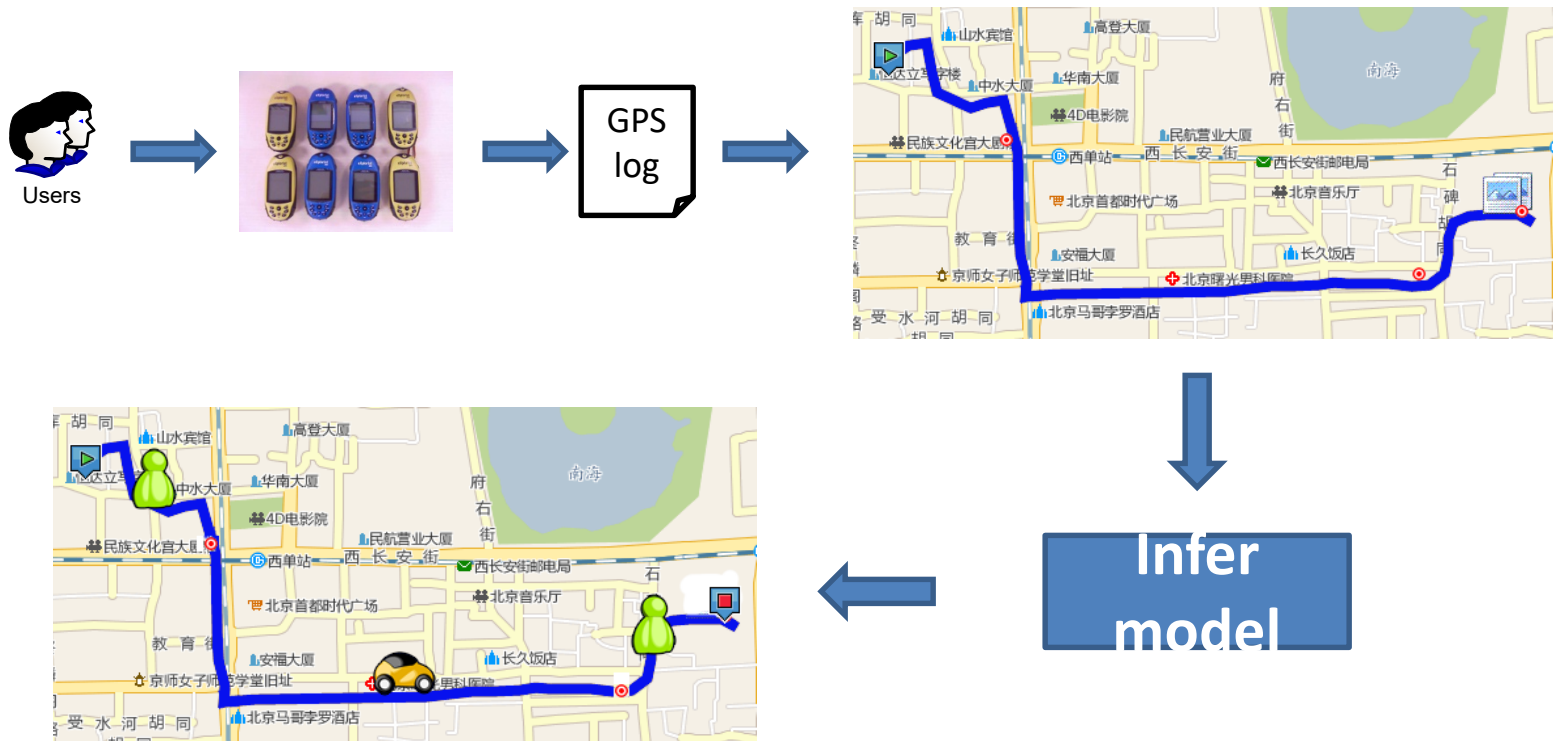
Trajectory Clustering

Representative trajectory
of a swarm/ group?



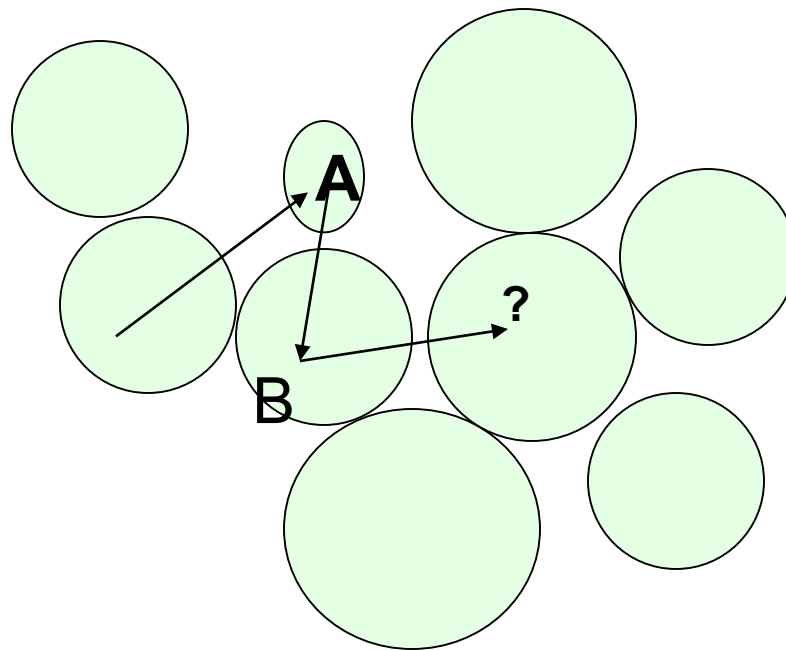
Trajectory Classification

- With supervised learning, classify trajectories into activities like hiking/ dining or different modes (walking/ driving)



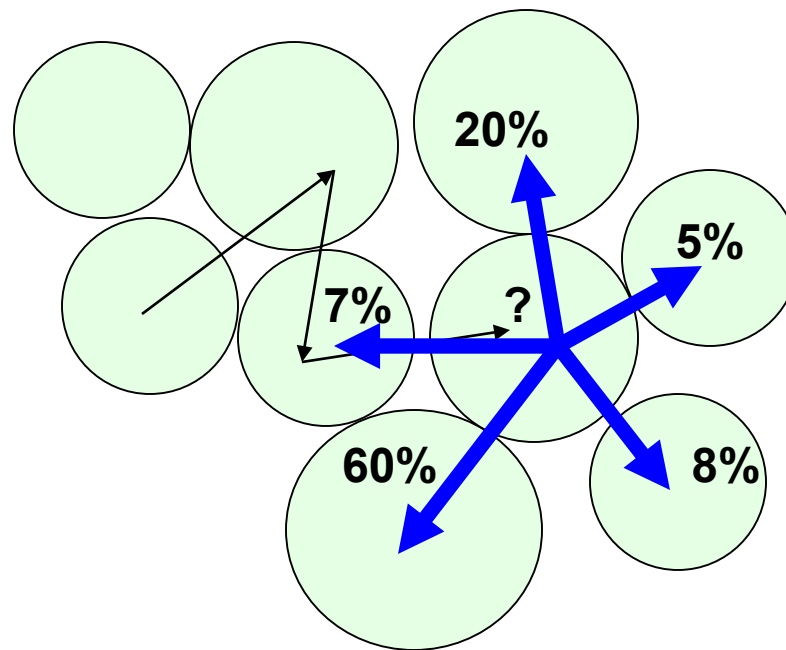
Trajectory Classification

- Predict next move.
 - If it is driving activity, where is next place of interest after A / B?



Trajectory Classification

- Obtain next destination with probability.
After drinks and eating, next?



Spatio-Temporal Periodic Pattern Mining

- Periodic patterns are trajectories periodically executed by a moving object. For eg. regular movement patterns from office staff, which are rather similar each working day.
 - There are 2 main approaches:
 - Fixed Period Approach
 - Reference Spot Approach
-

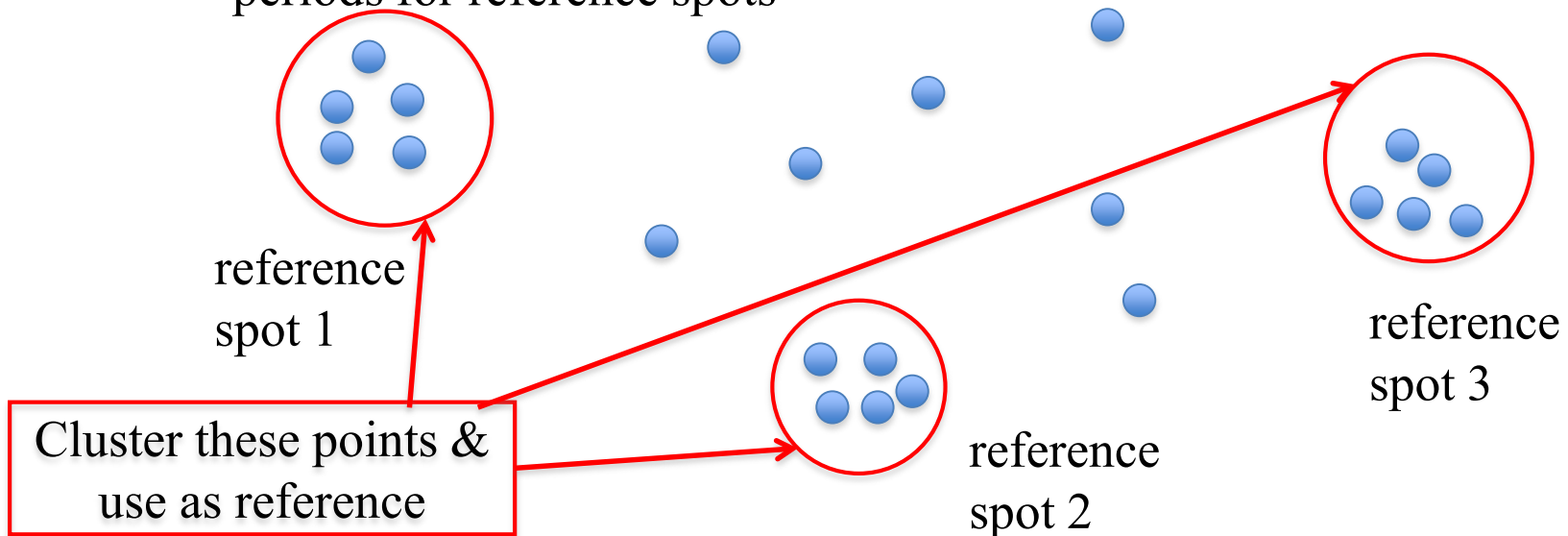
Spatio-Temporal PPM

■ Fixed (Time) Period Approach

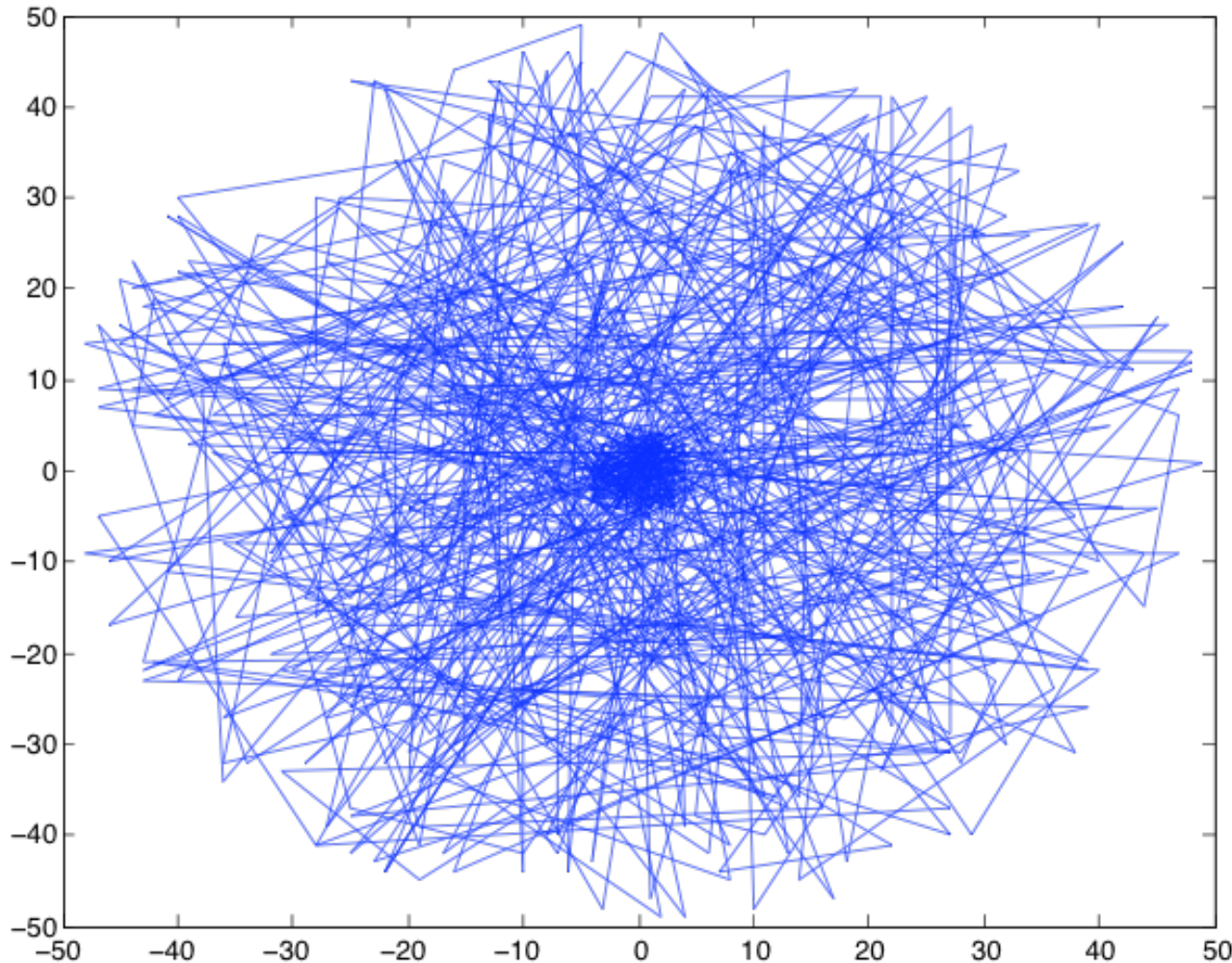
- To segment the long trajectory into a set of smaller (shorter) sub-trajectories based on a given fixed time period

■ Reference Spot Approach

- Find reference spots using clustering algorithms and then find associated periods for reference spots

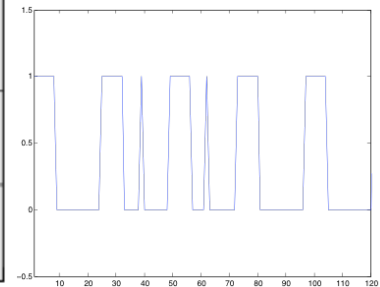
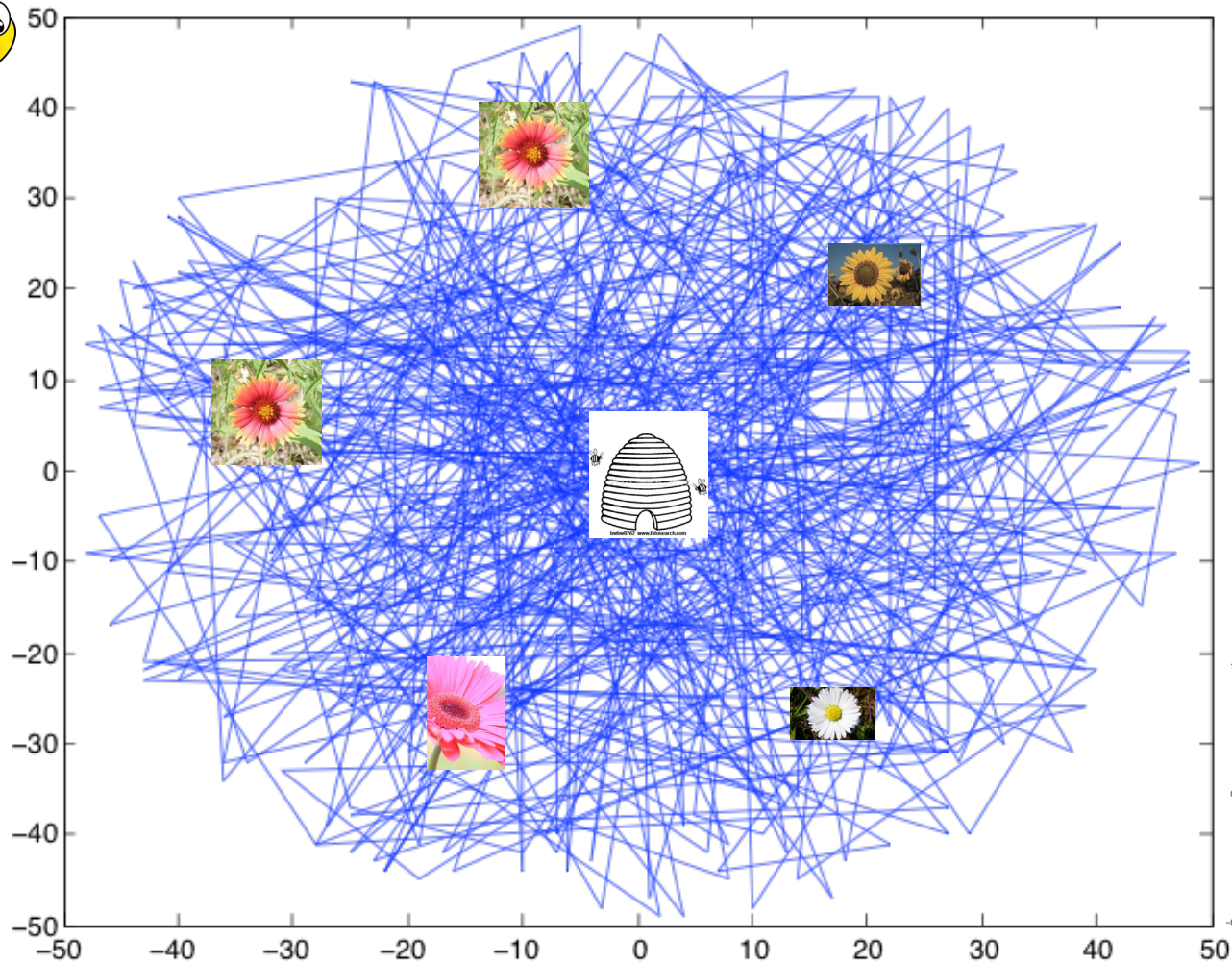


Periodic Pattern Mining



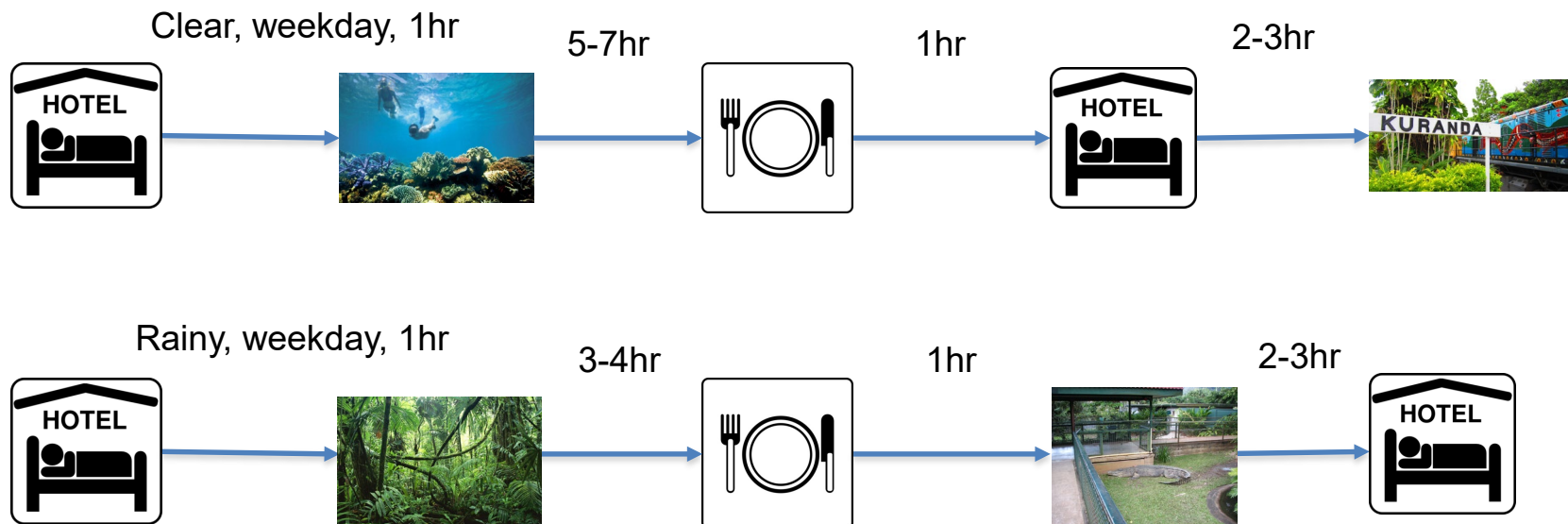
Not easy! Eg.
movements
of a bee (or
bees)

Periodic Pattern Mining



Trajectory Pattern Mining

- TPM considers spatio-temporal information
- In addition, add on aspatial *semantic information* to produce richer patterns



- Incorporate semantics – semantic trajectory data mining by incorporating aspatial information
 - Techniques largely the same:
 - classification is still in its infancy
 - Association mining (more used)
 - Lots of pre-processing with uncertainties and noise handling
-