



# Mobility Data Management

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*“πάντα ρει - Everything changes and nothing remains still”*  
Heraclitus (500 BC)

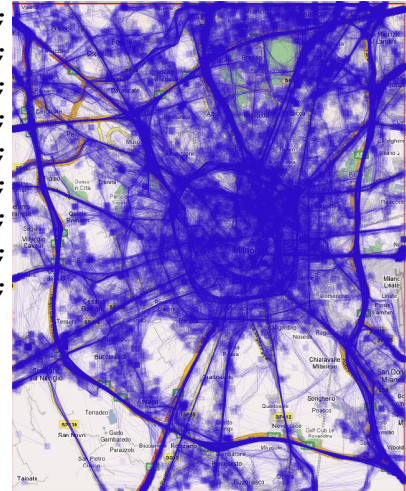


# Our starting point: GPS data feeds

## ■ Raw data: GPS recordings

N; Time; Lat; Lon; Height; Course; Speed; PDOP; State; NSat

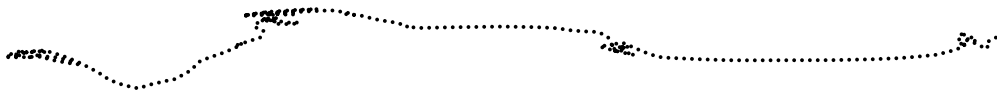
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...
8;22/03/07 08:51:52;50.777132;7.205580; 67.6;345.4;21.817;3.8;1808;4
9;22/03/07 08:51:56;50.777352;7.205435; 68.4;35.6;14.223;3.8;1808;4
10;22/03/07 08:51:59;50.777415;7.205543; 68.3;
11;22/03/07 08:52:03;50.777317;7.205877; 68.8;
12;22/03/07 08:52:06;50.777185;7.206202; 68.1;
13;22/03/07 08:52:09;50.777057;7.206522; 67.9;
14;22/03/07 08:52:12;50.776925;7.206858; 66.9;
15;22/03/07 08:52:15;50.776813;7.207263; 67.0;
16;22/03/07 08:52:18;50.776780;7.207745; 68.8;
17;22/03/07 08:52:21;50.776803;7.208262; 71.1;
18;22/03/07 08:52:24;50.776832;7.208682; 68.6;
...
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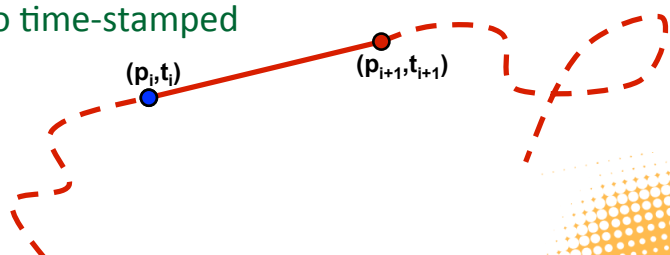
## What is a (GPS-based) trajectory?

- A **trajectory** is a model for a motion path of a moving object (animal, car, human, ...)
  - (due to discretization) a sequence of sampled time-stamped locations  $(p_i, t_i)$  where  $p_i$  is a 2D point  $(x_i, y_i)$  and  $t_i$  is the recording timestamp of  $p_i$



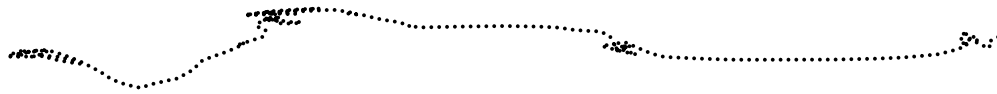
- A common representation in MOD is a **3D polyline** in the plane where vertices correspond to time-stamped locations  $(p_i, t_i)$

- and **linear interpolation** is assumed between  $(p_i, t_i)$  and  $(p_{i+1}, t_{i+1})$

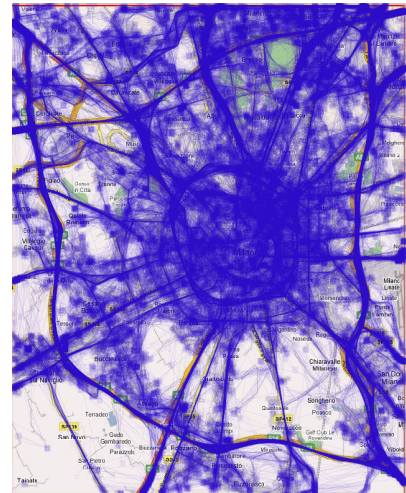


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# Key questions that arise



- How to **reconstruct a trajectory** from raw logs?
- How to **store and query trajectories** in a DBMS?
  - What is a “trajectory” data type?
    - simply a sequence of (x, y, t) tuples?



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## Acquiring trajectories from raw data

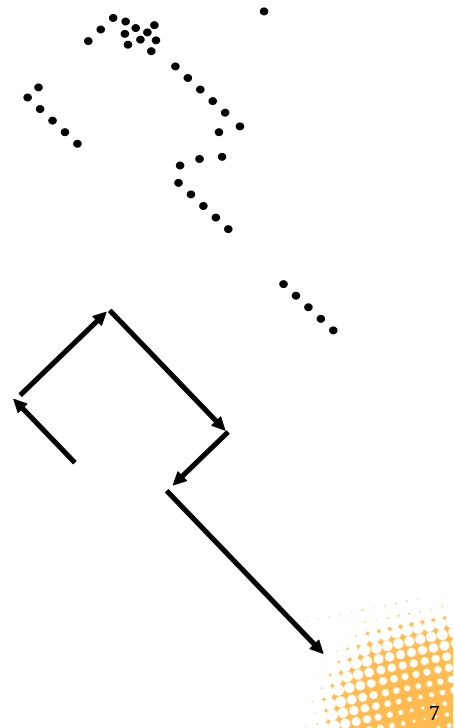
About mobility data

The trajectory reconstruction problem

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# The trajectory reconstruction problem

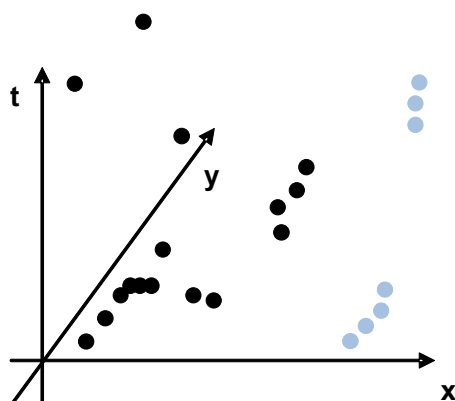
- From raw data, i.e., time-stamped locations
  - Raw data (3D points) arrive either one-by-one or in bulks
- ... to trajectory data, i.e., continuous evolutions
  - Linear interpolation is assumed between samples,
  - Redundancy is reduced, noise is removed,
  - etc.



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## Reconstructing trajectories

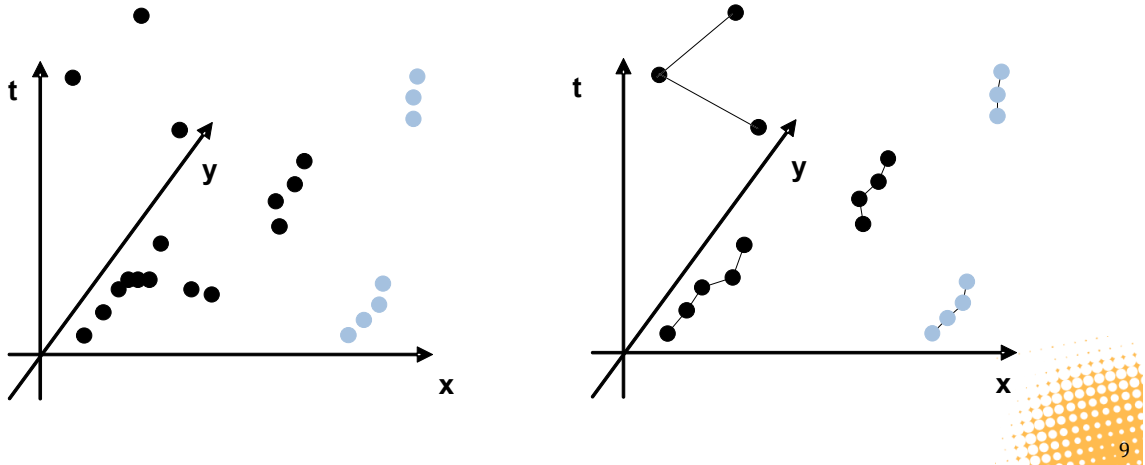
- Collected raw data represent time-stamped geo-locations
- Raw data (3D points) arrive either one-by-one or in bulks
- **Any idea?**



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# Reconstructing trajectories

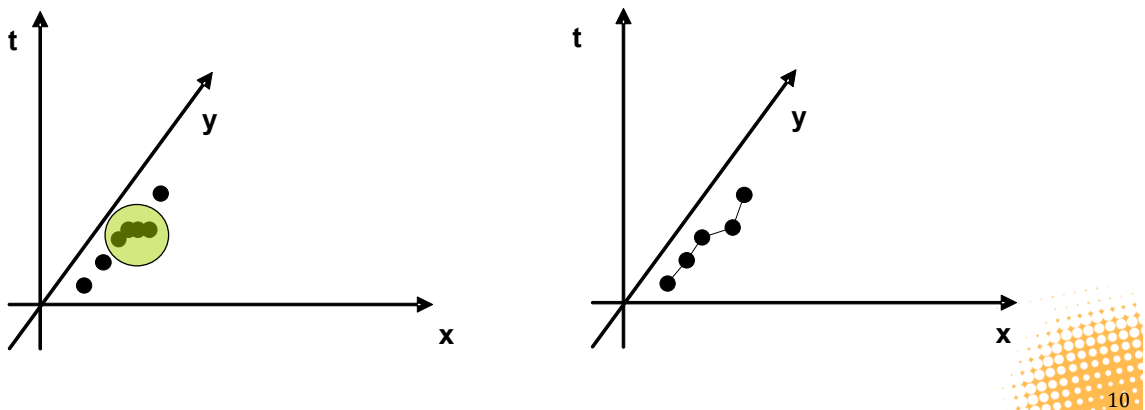
- [Marketos et al. 2008] proposes filters / thresholds that decide whether the new series of data is to
  - be appended to an existing trajectory, or
  - initiate a new trajectory, or
  - be considered as noise



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# Reconstructing trajectories

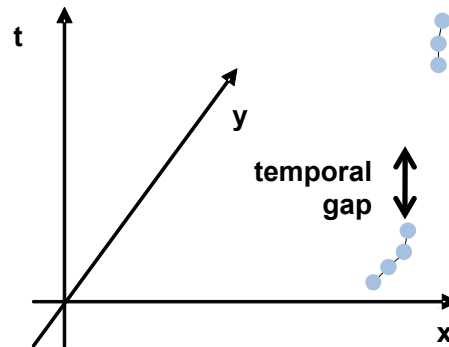
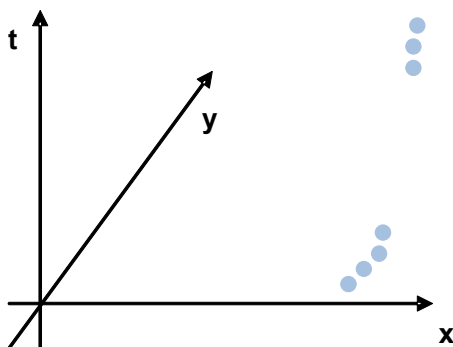
- 1<sup>st</sup> parameter: tolerance distance
  - The tolerance of the transmitted time-stamped positions
  - In other words: the **maximum distance between two consecutive time-stamped positions** of the same object in order for the object to be considered as **stationary**



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# Reconstructing trajectories

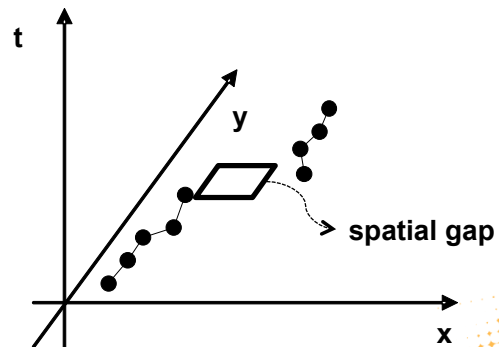
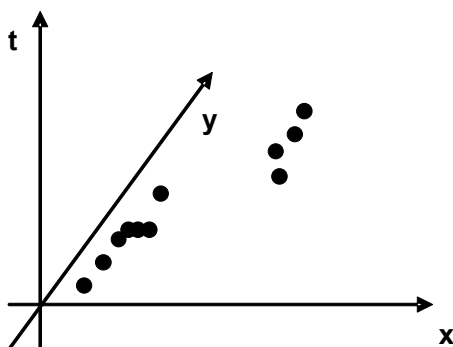
- tolerance distance
- 2<sup>nd</sup> parameter: temporal gap between trajectories
  - The **maximum allowed time interval** between two consecutive time-stamped positions of the same trajectory for a single moving object



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# Reconstructing trajectories

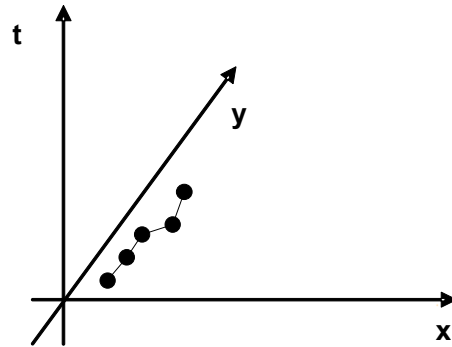
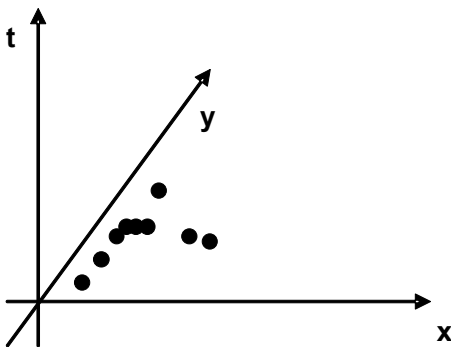
- tolerance distance, temporal gap
- 3<sup>rd</sup> parameter: spatial gap between trajectories
  - The **maximum allowed distance** in 2D plane between two consecutive time-stamped positions of the same trajectory



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# Reconstructing trajectories

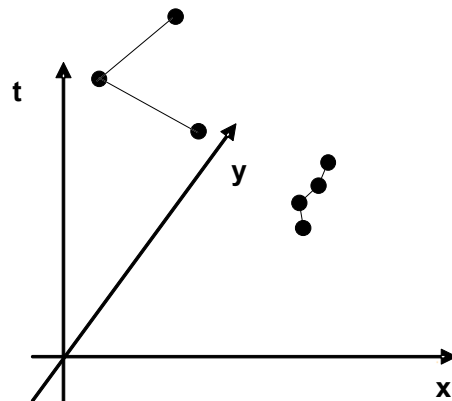
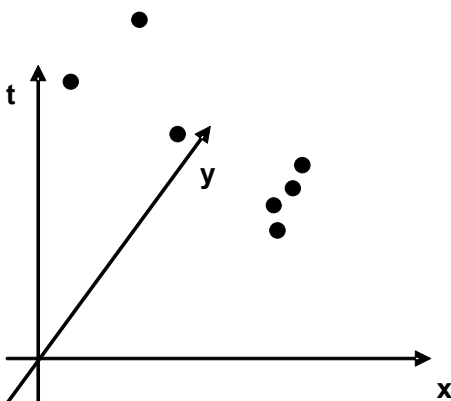
- tolerance distance, temporal gap, spatial gap
- 4<sup>th</sup> parameter: maximum speed
  - Decides whether a reported time-stamped location is **noise**, hence to be discarded from the output trajectory



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# Reconstructing trajectories

- tolerance distance, temporal gap, spatial gap, maximum speed
- 5<sup>th</sup> parameter: maximum noise duration
  - The **maximum duration of a noisy part** of a trajectory. If 'noise' continues longer than  $noise_{max}$ , most probably it is not noise but, instead, the starting positions of a new trajectory!



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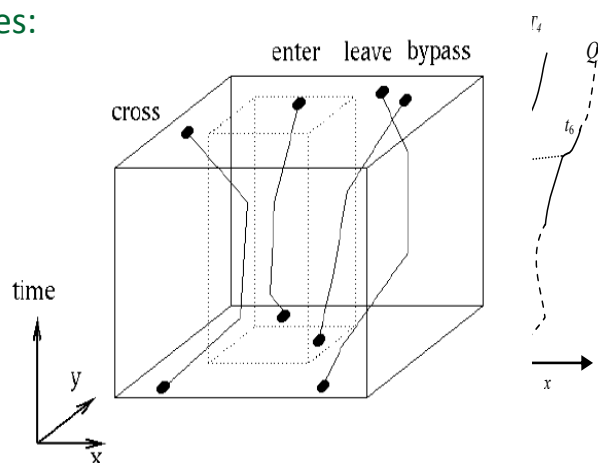
# Location-aware querying

From primitive to advanced MOD queries

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## What kind of queries?

- The nature of trajectory data provides us with the ability to query them with a variety of operators.
- Primitive queries on trajectories:
  - Coordinate-based
    - Range, NN
  - Trajectory-based
    - Topological, Directional



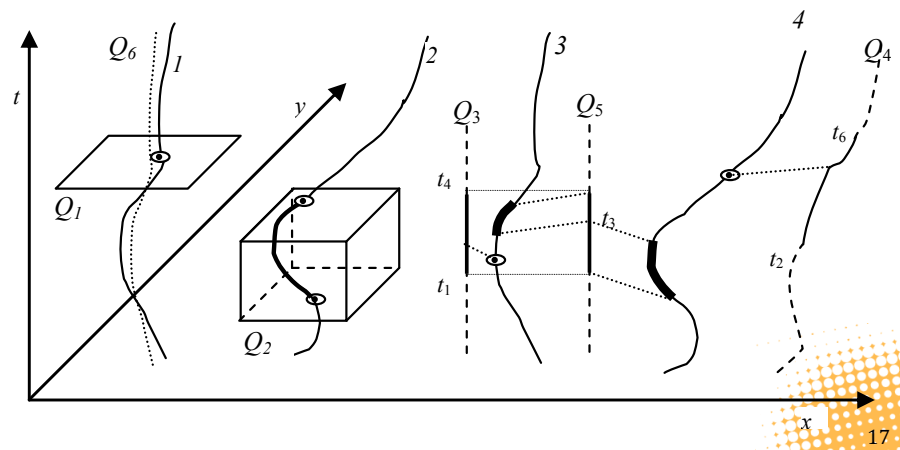
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# Coordinate-based queries

## ■ Spatial (range or NN) search

- “Find all trajectories that were inside area A at time instant  $t$  (or time interval  $I$ )” or
- “Find the trajectory that was closest to point B at time instant  $t$  (or time interval  $I$ )”

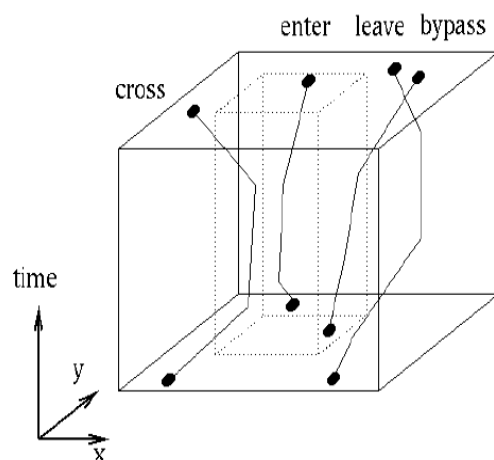


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# Trajectory-based queries

## ■ Topological / directional search

- “Find all trajectories that entered (crossed, left, bypassed, etc.) or were located west (south, etc.) of an area” or
- “Find all trajectories that crossed (met, etc.) or were located left of (right of, in front of, etc.) a query trajectory TQ



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# Taxonomy of location-aware queries



- Dimensions [Mokbel & Aref, 2007]:
  - **Type**: range, NN, reverse NN, closest-point, ...
  - **Time**: past, present, future
  - **Duration**: snapshot, continuous
  - **Query (reference) object**: stationary, moving
  - **Data objects**: stationary, moving
- Queries: every (?) possible combination of the above dimensions

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## Examples

slides from [Mokbel & Aref, 2007]



*Where are my nearest McDonalds for the next hour?*

- **Type**: Nearest-Neighbor query
- **Time**: Future
- **Duration**: Continuous
- **Query**: Moving
- **Object**: Stationary



*Send E-coupons to all cars that I am their nearest gas station*

- **Type**: Reverse NN query
- **Time**: Present
- **Duration**: Snapshot
- **Query**: Stationary
- **Object**: Moving

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# Examples

slides from [Mokbel & Aref, 2007]



*Continuously report the number of cars in the freeway*

- Type: Range query
- Time: Present
- Duration: Continuous



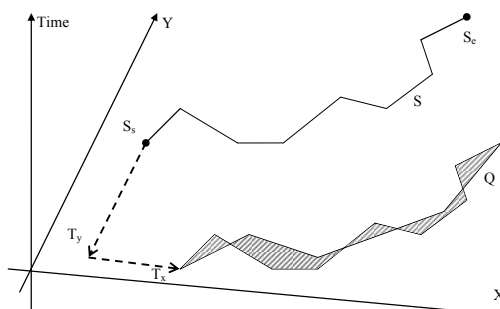
*What was the closest dist. between Taxi A & me yesterday?*

- Type: Closest-point query
- Time: Past
- Duration: Snapshot
- Query: Moving
- Object: Moving

## Advanced trajectory-based queries

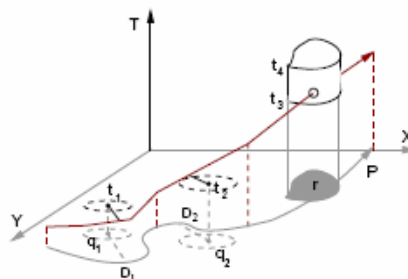
### ■ Trajectory similarity queries

- [Frentzos et al. 2007]  
“Given a query trajectory  $Q$ , find the  $k$ - most similar trajectories to  $Q$  (perhaps, constrained is space and/or time)”



### ■ Spatio-temporal pattern queries

- [Hadjieleftheriou et al. 2005]  
e.g. “Find objects that crossed through region  $A$  at time  $t_1$ , came as close as possible to point  $B$  at a later time  $t_2$  and then stopped inside circle  $C$ ”



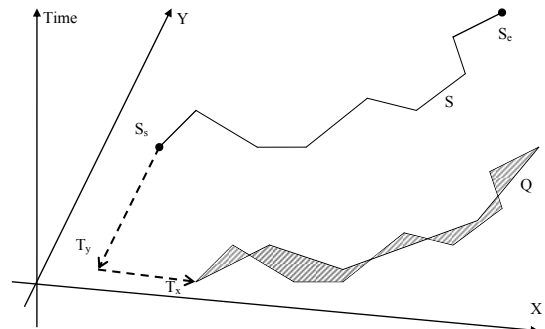
# Trajectory Similarity Queries

## ■ Key question:

- How do we measure **distance** or **(dis-)similarity** between two trajectories?

## ■ Similarity variations:

- [Pelekis et al. 2007, 2011]  
Similarity in space and/or time,  
or wrt. derived information (e.g. speed or direction)



## ■ Similarity queries have been studied extensively in time-series literature

- But, things are different here! Both where and when are important

# Trajectory Similarity Queries (cont.)

## ■ Different points of view: **Moving clusters, Flock queries**

- What is a flock?
  - a large enough subset of objects moving along paths close to each other for a certain time
- In the flock, identify **leaders** and **followers**

## ■ Solutions:

- [Benkert et al. 2008],  
[Gudmundsson & van Kreveld, 2006]



# Efficiently trajectory indexing and storage in MODs

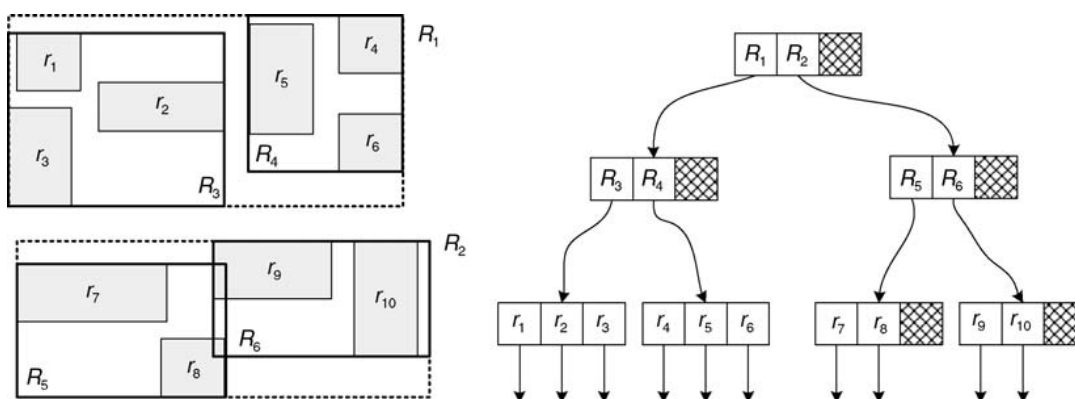
Indexing techniques

MOD engines

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## R-trees for spatial data

- For d-dimensional point or region data
- Is it portable to mobility data?
  - In other words, is space + time simply a 3D space?

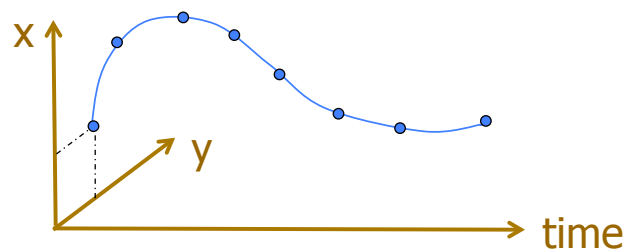


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# What ?? for mobility data

## ■ Challenges:

- Both space and time are equally important!
- But! time is not simply a 3<sup>rd</sup> dimension
  - e.g. it is monotonously increasing
- How does a trajectory approximation look?
  - Approximation is necessary for indexing

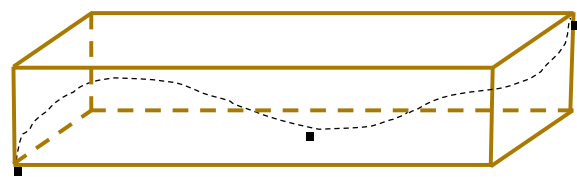


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## Two approaches: native vs. parametric space

## ■ Typically approximate using MBRs; then index these MBRs

- we can use R-trees etc. ☺
- trajectories are lines, thus MBRs add extensive empty space ☹

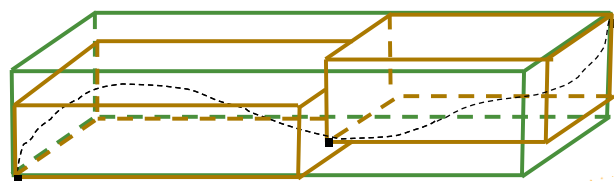


## ■ How many MBRs per trajectory?

- One MBR per trajectory (too much empty space...) or one MBR per segment (too many MBRs...)

## ■ Can we do anything better?

- Smart “partitioning” for MBRs [Hadjieleftheriou et al. 2002]



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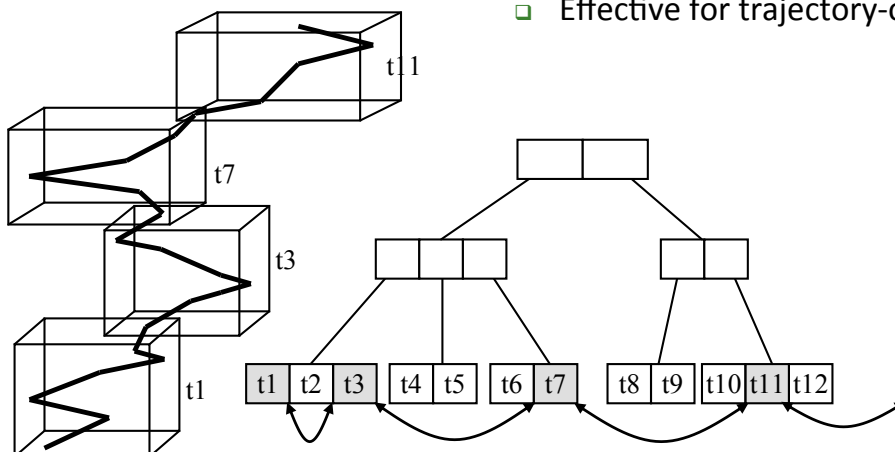
- Indexing the past (= trajectories)
  - unconstrained movement : the trajectory-bundle tree (TB-tree) [Pfoser et al. 2000]
  - network-constrained movement: the fixed-network-restricted tree (FNR-tree) [Frentzos, 2003]
- Indexing the present (and anticipated future)
  - Data partitioning: TPR-tree [Saltenis et al. 2000], TPR\*-tree [Tao et al. 2003]
  - Space partitioning: B<sup>x</sup>-tree [Jensen et al. 2004], ST<sup>2</sup>B-tree [Chen et al. 2008]
- (Hybrid solution for) Indexing the past & present
  - R<sup>PPF</sup>-tree [Pelanis et al. 2006]
- We focus on historical tracks of moving objects (trajectories) → concentrate on the 1<sup>st</sup> group

## TB-tree

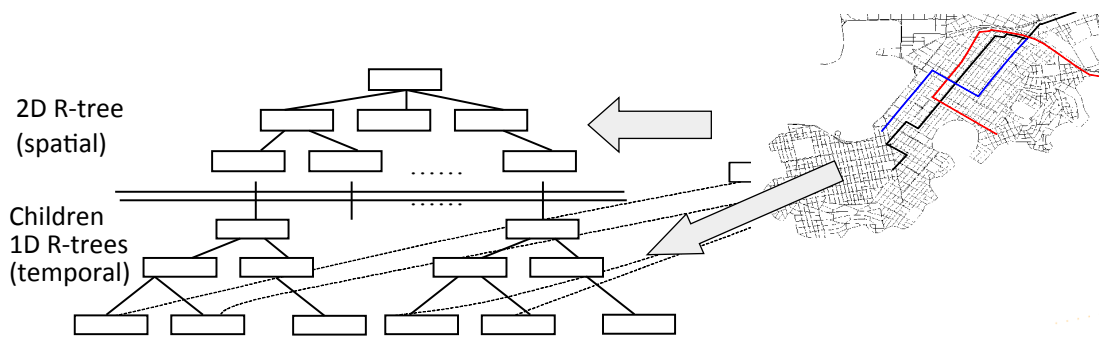
- [Pfoser et al. 2000] Maintains the 'trajectory' concept

- Each node consists of segments of a single trajectory
- nodes corresponding to the same trajectory are linked together in a chain

- Effective for trajectory-oriented queries



- (Frentzos, 2003) a forest of 1D (temporal) R-trees on top of a 2D (spatial) R-tree
  - There is an additional “Parent” 1D R-tree which indexes the temporal intervals of the 1D R-trees leaf nodes



## Moving Objects Database Systems

- From traditional DBMS to **Moving Object Database (MOD) engines**
  - Data types, indices, query processing & optimization strategies for trajectories
- Spatial and temporal dimensions are considered as first-class citizens.
- State-of-the-art prototype MOD engines
  - **SECONDO** (Güting et. al.) IDEAS'00, ICDE'05, MDM'06
  - **HERMES** (Pelekis et. al.) EDBT'06, SIGMOD'08



<http://dna.fernuni-hagen.de/Secondo.html/index.html>

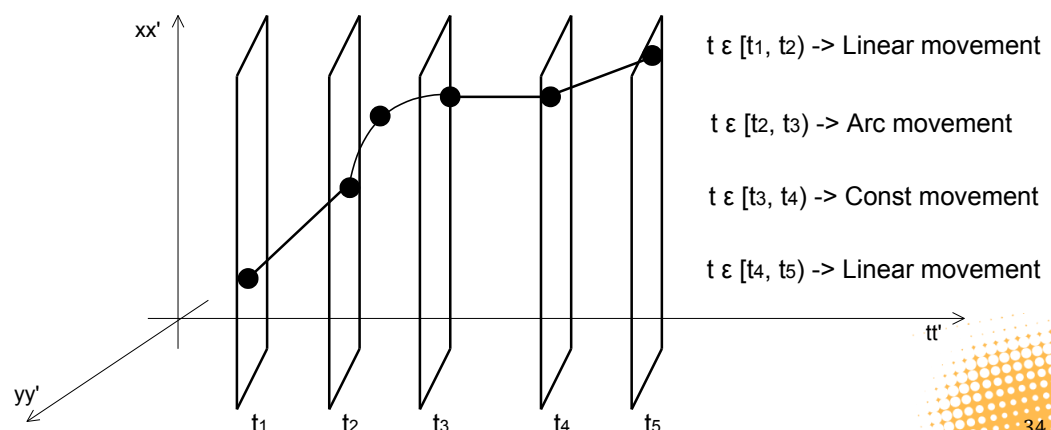
- A generic DBMS framework that can be filled with implementation of various data models (R, OR, XML) and data types (spatial data, moving objects)
  - Built on top of Berkeley DB.
- A MOD is a set of SECONDO objects of the form  $(name, type, value)$ , where  $type$  is one of the implemented algebras
- About 20 implemented algebras
  - standard algebra, relational algebra, R-Tree algebra, spatial algebra, etc.
- Query optimizer includes optimization of conjunctive queries, selectivity estimation, and implementation of an SQL-like query language

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<http://infolab.cs.unipi.gr/hermes/>

- 2 implementations: Oracle Spatial vs. PostgreSQL
- Supports several data types: moving point, moving line, moving polygon, etc.
- 3DR-tree and TB-tree indexing



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# Summary



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## Summary on Mobility Data Management

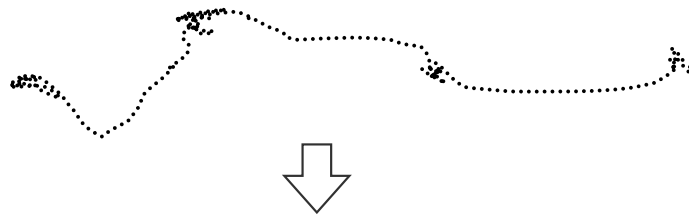
- From (stationary) spatial to moving object databases
- Current research agenda:
  - **Modeling** issues
    - From 'raw GPS' to semantically-annotated trajectories
  - **Benchmarking** issues
    - Generating realistic synthetic trajectory data
    - Querying based on emerging **location-based social networking** (LBSN) apps
  - **Advanced MOD** implementations
    - centralized vs. distributed vs. noSQL architectures



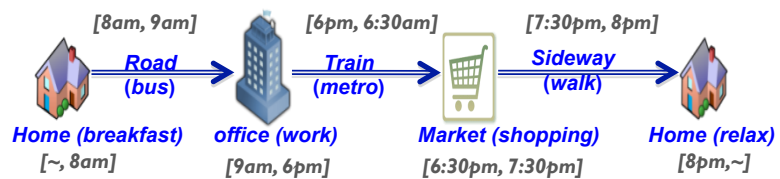
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# From “raw” to “semantic” trajectories

raw mobility data  
sequence (x,y,t) points  
e.g., GPS feeds



meaningful mobility tuples  
<place, time<sub>in</sub>, time<sub>out</sub>, tags>

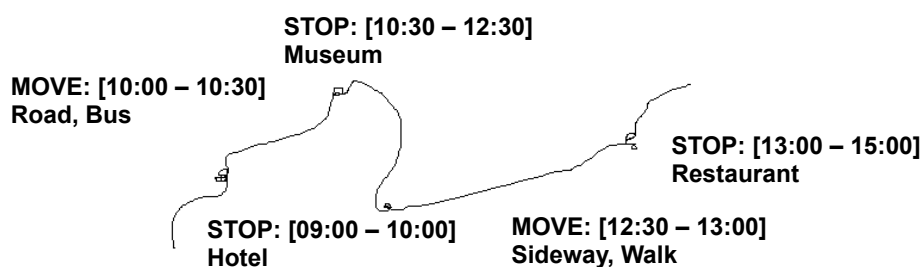


- Semantic Trajectory:  $T = \{e_{first}, \dots, e_{last}\}$
- Episode:  $e_i = (STOP \mid MOVE, t_{from}, t_{to}, place, tag)$

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# From “raw” to “semantic” trajectories

- **Stops** are places (points, regions) where the object stays “static”
- **Moves** are the parts of the object’s trajectory in between two Stops, i.e. where the object is “moving”
- **Tags** are meta-data associated with Stops and Moves
  - information about (at least...) **when?** **where?** (also...) **how?** **what?** **why?**

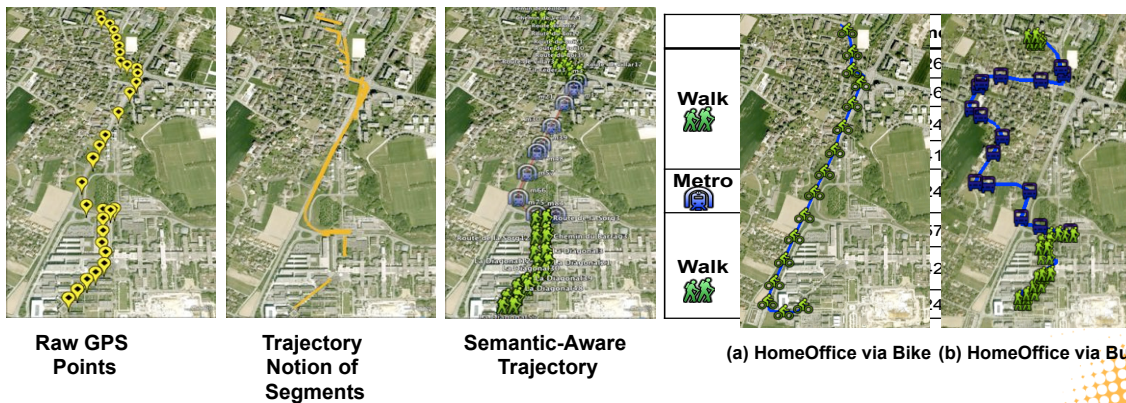


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# Why semantic trajectories?

- Detection of homogenous fractions of movement,
  - Trajectory is reconstructed **as a sequence of episodes (stops/moves)**
    - E.g., home, shopping, move by bus, biking, ...
- Semantic data abstraction & compression (efficiency/effectiveness)

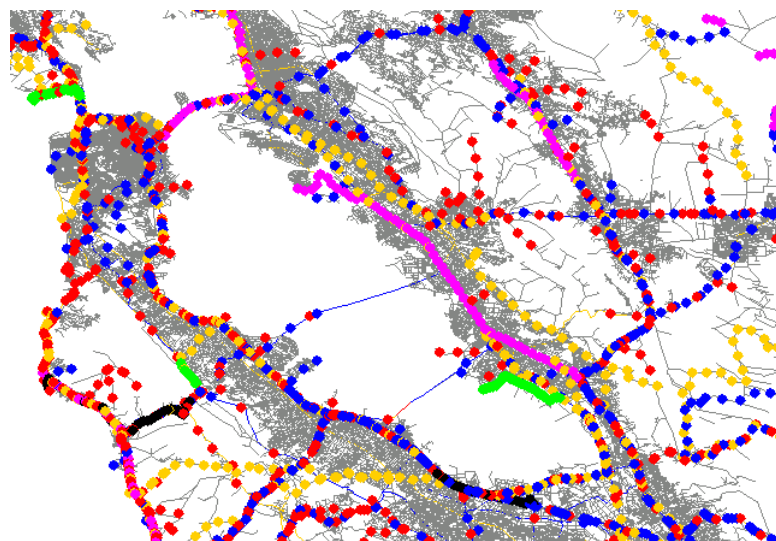
## Home-office trajectory examples



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# Synthetic trajectory data generators

- Objectives:
  - flexibility,
  - realism,
  - scalability,
  - ...
- State-of-the-art
  - Brinkhoff
  - BerlinMOD (on top of SECONDO)
- Challenge:
  - Generate-by-example



source: [www.fh-ooe.de/institute/iapg/personen/brinkhoff/generator](http://www.fh-ooe.de/institute/iapg/personen/brinkhoff/generator)

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## Reading list



# Mobility data modeling & MOD engines



## (1)

- de Almeida, V.T. et al. (2006) Querying Moving Objects in SECONDO. Proceedings of MDM.
- Behr, T. and Güting, R.H. (2005) Fuzzy Spatial Objects: An Algebra Implementation in SECONDO. Proceedings of ICDE.
- Brinkhoff, T. (2002) A Framework for Generating Network-based Moving Objects. Geoinformatica, 6(2): 153-180.
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- Dieker, S. and Güting, R.H. (2000) Plug and Play with Query Algebras: SECONDO – A Generic DBMS Development Environment. Proceedings of IDEAS.
- Düntgen, C. et al. (2009) BerlinMOD: a Benchmark for Moving Object Databases. The VLDB Journal, 18: 1335-1368.
- Güting, R.H. et al. (2000) A Foundation for Representing and Querying Moving Objects. ACM Transactions on Database Systems, 25(1):1-42.
- Güting, R.H. et al. (2006) Modeling and querying moving objects in networks. VLDB Journal, 15(2): 165-190.



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# Mobility data modeling & MOD engines



## (2)

- Marketos, G. et al. (2008) Building real-world trajectory warehouses. Proceedings of MobiDE.
- Pelekis, N. et al. (2006) Hermes - A Framework for Location-Based Data Management. Proceedings of EDBT.
- Pelekis, N. et al. (2008) HERMES: aggregative LBS via a trajectory DB engine. Proceedings of ACM SIGMOD.
- Theodoridis, Y. (2003) Ten Benchmark Database Queries for Location-based Services. The Computer Journal, 46(6): 713-725
- Theodoridis, Y. and M. Nascimento (2000) Generating Spatiotemporal Datasets on the WWW. SIGMOD Record, 29(3): 39-43.



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# MOD query processing



## (1)

- Benetis, R. et al. (2002) Nearest Neighbor and Reverse Nearest Neighbor Queries for Moving Objects. Proceedings of IDEAS.
- Benkert, M. et al. (2008) Reporting Flock Patterns. Computational Geometry, 41: 111-125.
- Frentzos, E. et al. (2005) Nearest Neighbor Search on Moving Object Trajectories. Proceedings of SSTD.
- Frentzos, E. et al. (2007) Index-based Most Similar Trajectory Search. Proceedings of ICDE.
- Gedik, B., and Liu, L. (2004) MobiEyes: Distributed Processing of Continuously Moving Queries on Moving Objects in a Mobile System. Proceedings of EDBT.
- Gudmundsson, J. and M. van Kreveld (2006) Computing longest duration flocks in trajectory data. Proceedings of ACM-GIS.
- Jensen, C.S. et al. (2003) Nearest Neighbor Queries in Road Networks. Proceedings of ACM-GIS.
- Li, F. et al. (2005) On Trip Planning Queries in Spatial Databases. Proceedings of SSTD.



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# MOD query processing



## (2)

- Mokbel, M.F. and W.G. Aref (2007) Location-aware Query Processing and Optimization: A Tutorial. Proceedings of MDM.
- Papadias, D. et al. (2003) Query Processing in Spatial Network Databases. Proceedings of VLDB.
- Pelekis, N. et al. (2007) Similarity Search in Trajectory Databases. Proceedings of TIME.
- Porkaew, K. et al. (2001) Querying Mobile Objects in Spatio-Temporal Databases. Proceedings of SSTD.
- Shekhar, S. and Yoo, J. S. (2003) Processing In-Route Nearest Neighbor Queries: A Comparison of Alternative Approaches. Proceedings of GIS.
- Sankaranarayanan, J. et al. (2005) Efficient Query Processing on Spatial Networks. Proceedings of ACM-GIS.
- Tao, Y. et al. (2002) Continuous Nearest Neighbor Search. Proceedings of VLDB.
- Xia, T. and Zhang, D. (2006) Continuous Reverse Nearest Neighbor Monitoring. Proceedings of ICDE.



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## (1)

- Cai, Y. and Ng, R.T. (2004) Indexing Spatio-Temporal Trajectories with Chebyshev Polynomials. Proceedings of ACM SIGMOD.
- Chen, S. et al. (2008) ST<sup>2</sup>B-tree: A Self-Tunable Spatio-Temporal B+-tree Index for Moving Objects. Proceedings of ACM SIGMOD.
- Frentzos, E. (2003) Indexing Objects Moving on Fixed Networks. Proceedings of SSTD.
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