



Modeling & Management of Mobility Data

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Apr. 2013



"πάντα ρει - Everything changes and nothing remains still" Heraclitus (500 BC)



Outline



- Acquiring trajectories from raw data
 - About mobility data
 - □ The trajectory reconstruction problem
- Location-aware querying
 - From straightforward to advanced mobility queries
- Efficient trajectory indexing and storage in MODs
 - Indexing techniques
 - MOD engines
- The next wave: "semantic" trajectories
- Summary





Acquiring trajectories from raw data

About mobility data

The trajectory reconstruction problem





Raw data: GPS recordings

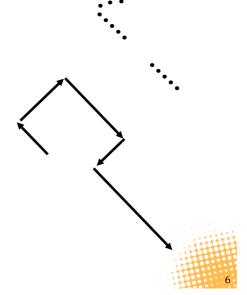
```
objectID, trajectoryID, timestamp, longitude, latitude
201100024,1,2009-01-02 08:54:07,24.609728324369,38.013503319816
201100024,1,2009-01-02 08:54:25,24.6094016577037,38.0127699864845
201100024,1,2009-01-02 08:55:06,24.6086749910399,38.011116653155
201100024,1,2009-01-02 08:55:56,24.6076299910435,38.0092066531597
201100024,1,2009-01-02 08:56:16,24.6071983243782,38.0084733198281
201100034,1,2009-01-02 04:19:26,23.1092366579214,38.5853616531322
201100034,1,2009-01-02 04:19:36,22.927219990932
201100034,1,2009-01-02 04:19:45,23.035993324356
201100034,1,2009-01-02 04:20:05,23.063861657875
...
```

The trajectory reconstruction problem



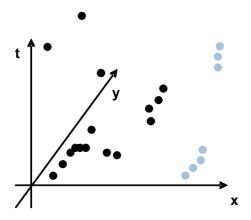
- From raw data, i.e., time-stamped locations
 - Raw data (3D points) arrive either one-by-one or in bulks

- ... to trajectories, i.e., continuous evolutions
 - □ Redundancy is reduced, noise is removed, etc.





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

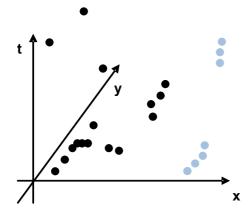


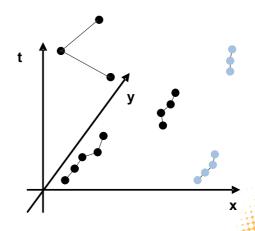
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Reconstructing trajectories



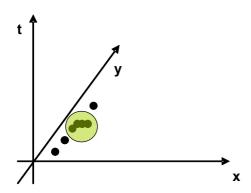
- [Marketos et al. 2008] filters / thresholds that decide whether the new series of data is to
 - be appended to an existing trajectory, or
 - □ initiate a new trajectory, or
 - be considered as noise

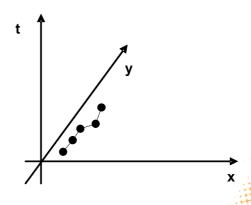






- 1st parameter: tolerance distance
 - □ The tolerance of the transmitted time-stamped positions
 - In other words: the maximum distance between two consecutive time-stamped positions of the same object in order for the object to be considered as stationary

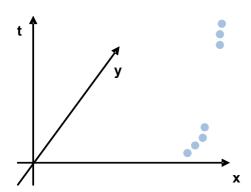


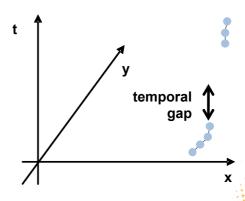


Reconstructing trajectories



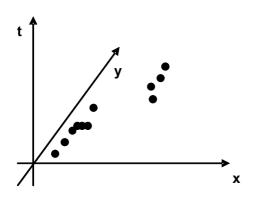
- tolerance distance
- 2nd parameter: temporal gap between trajectories
 - □ The maximum allowed time interval between two consecutive timestamped positions of the same trajectory for a single moving object

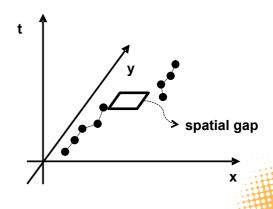






- tolerance distance, temporal gap
- 3rd parameter: spatial gap between trajectories
 - □ The maximum allowed distance in 2D plane between two consecutive time-stamped positions of the same trajectory

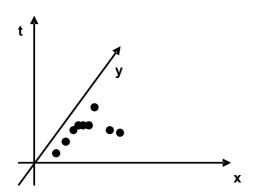


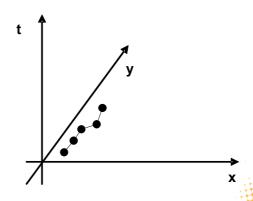


Reconstructing trajectories



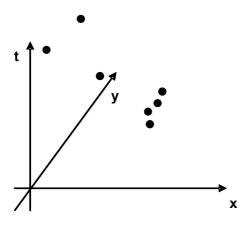
- tolerance distance, temporal gap, spatial gap
- 4th parameter: maximum speed
 - Decides whether a reported time-stamped location is noise, hence to be discarded from the output trajectory

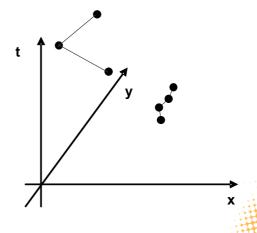






- tolerance distance, temporal gap, spatial gap, maximum speed
- 5th parameter: maximum noise duration
 - □ The maximum duration of a noisy part of a trajectory. If 'noise' continues longer than $noise_{max}$, most probably it is not noise but, instead, the start of a new trajectory!







Location-aware querying

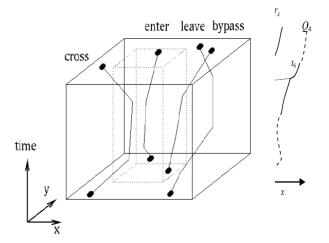
From straightforward to advanced mobility queries



What kind of queries?



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

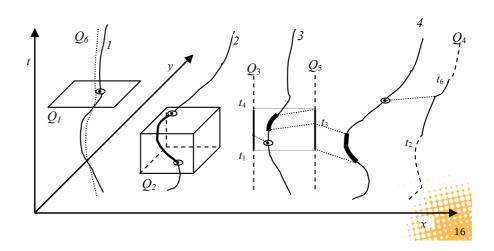




Coordinate-based queries



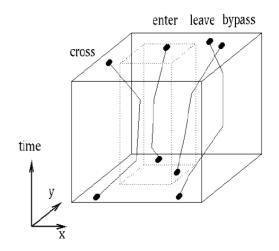
- Spatial (range or NN) search
 - "Find all trajectories that were inside area A at time instant t (or time interval I)" or
 - "Find the trajectory that was closest to point B at time instant t (or time interval I)"



Trajectory-based queries



- Topological queries: Search for trajectories that
 - entered (crossed, left, bypassed, etc.) a reference area, or
 - intersected (moved in parallel, etc.) a reference trajectory
- Directional queries: Search for trajectories that were located
 - west of (south of, etc.) a reference area, or
 - left of (right of, in front of, etc.) a reference trajectory





Taxonomy of location-aware queries



- Dimensions:
 - □ **Type**: range, NN, reverse NN, closest-point, ...
 - Time: past, present, future
 - Duration: snapshot, continuous
 - Query (reference) object: stationary, moving
 - Data objects: stationary, moving
- Queries: several combinations of the above dimensions

Querying the Past

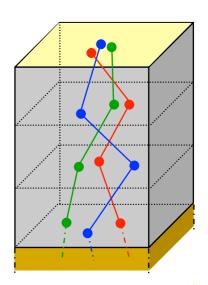


Examples:

- Querying along the Temporal Dim.: What was the location of a certain object from 7:00 AM to 10:00 AM yesterday?
- Querying Along the Spatial Dim.: Find all objects that were in a certain area at 7:00 AM yesterday
- Querying Along the Spatio-temporal Dim.:
 Find all objects that were close to each other from 7:00 AM to 8:00 AM yesterday



- Large number of historical trajectories
- Persistent read-only data





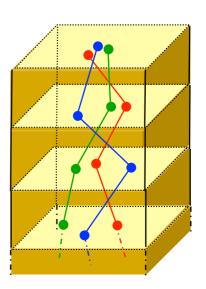
Querying the Present



- Time is always NOW
- Example Queries:
 - □ Find the number of objects in a certain area
 - What is the current location of a certain object?

Features:

- Continuously changing data
- Real-time query support is required
- Index structures should be update-tolerant



Present data is always accessed through continuous queries



Querying the Future

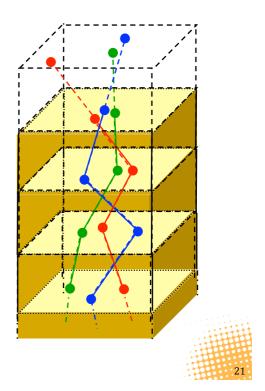


Examples:

- What will my nearest restaurant be after 30 minutes?
- Does my path conflict with any other cars for the next hour?

Features:

- Predict the movement through a velocity vector
- Prediction could be valid for only a limited time horizon in the future



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







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



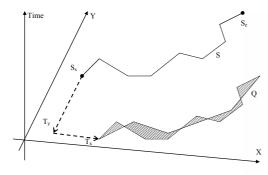
... as well as more advanced queries



Trajectory similarity queries

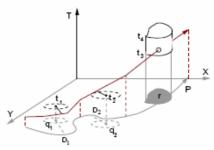
□ [Frentzos et al. 2007]

"Given a query trajectory Q,
find the k- most similar
trajectories to Q (perhaps,
constrained is space and/or
time)"



Motion 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 Cduring interval (t_3, t_4) "





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 w.r.t. derived information (e.g. speed and 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]







Efficient trajectory indexing and storage in MODs

Indexing techniques

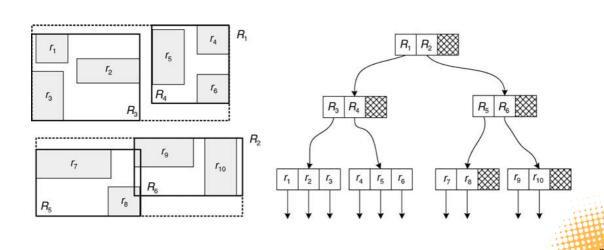
MOD engines



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?



What ?? for mobility data



- Challenges:
 - □ Both space and time are equally important!
 - □ But! time is not simply a 3rd dimension
 - think of its monotonicity
 - Also, density in the 'spatio-temporal' space could be quite different from place to place
- Diversity in space, in time, in space-time ...



Diversity in space

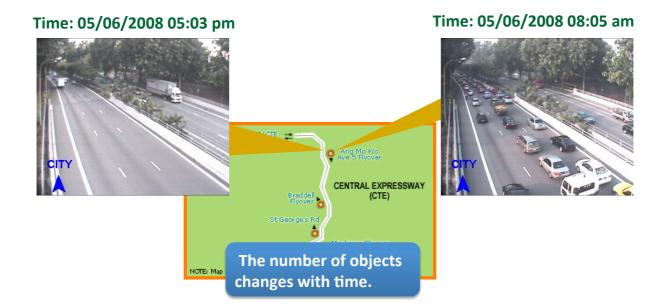






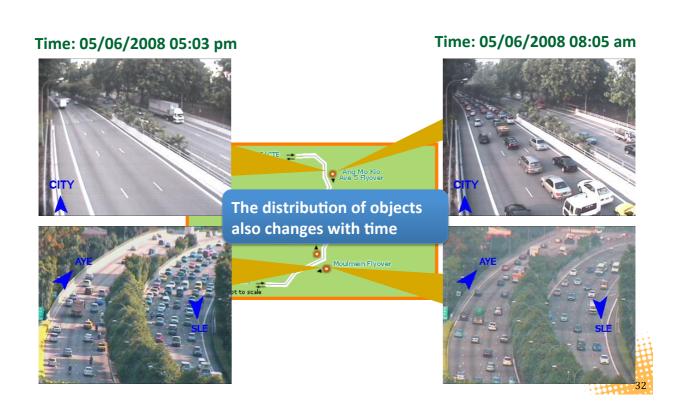
Diversity in time





Diversity in space & time





Back to technical stuff ...



- Assume we have stored trajectories (i.e., the recorded locations of a moving object over time)
- Main Question: how can we approximate a trajectory?
 - □ Like, e.g. MBRs for (static) spatial objects?
 - □ Then, we could build indices upon the approximations

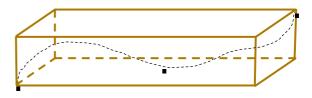




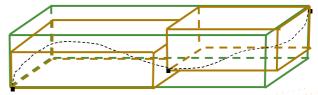
Two approaches: native vs. parametric space



- Typically approximate using MBRs; then index these MBRs
 - □ we can use R-trees etc. ³
 - □ 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 something better?
 - Smart "partitioning" for MBRs [Hadjieleftheriou et al. 2002]





MOD indexing techniques



- Indexing the past (= trajectories)
 - unconstrained movement : **TB-tree** [Pfoser et al. 2000]
 - network-constrained movement: 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**^x-tree [Jensen et al. 2004], **ST**²**B**-tree [Chen et al. 2008]
- (Hybrid solution for) Indexing the past & present
 - □ **R**^{PPF}-tree [Pelanis et al. 2006]
- We focus on historical tracks of moving objects (trajectories) ->
 concentrate on the 1st 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

 t1

 t1

 t1

 t2

 t3

 t4

 t5

 t6

 t7

 t8

 t9

 t10

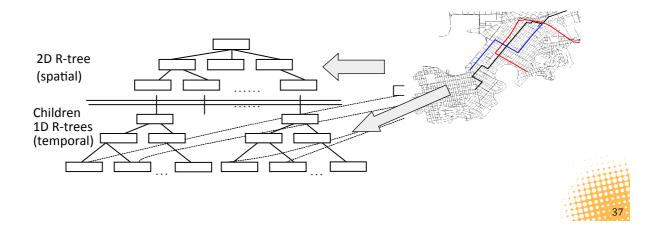
 t11

 t12

FNR-tree



- (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
 - Objective: spatial and temporal dimensions to be considered as firstclass citizens.
 - □ Technically: data types, indices, query processing & optimization strategies for trajectories of moving objects
- Current MOD solutions:
 - □ State-of-the-art MODs (Secondo, Hermes, etc.) vs.
 - SDBMS simulated to work as MODs (e.g. PostGIS)



Moving Objects Database Systems



- A few prototype MOD engines
 - SECONDO by Güting et. al. @ Hagen
 - PLACE by Aref et al. @ Purdue
 - □ **HERMES** by Pelekis et. al. @ Piraeus
- A quick note on PostGIS solution:
 - PostGIS 2.0 new features include 3D/4D indexing
 - □ A trajectory can be simulated by a 3D path (= sequence of 3D points)







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



SECONDO Architecture

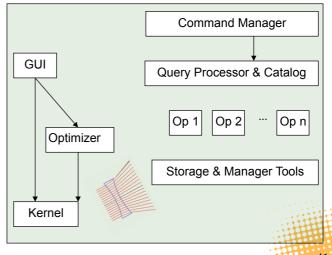


- Generic GUI independent of data models
 - The interface includes command prompt and is extensible by a set of different viewers

□ The core functionality is the optimization of conjunctive queries, i.e.,

producing an efficient query plan

 On top of the query optimizer, there is a SQL-like language in a PROLOG notation

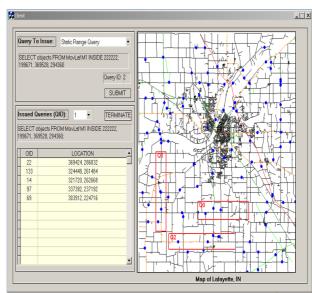






http://www.cs.purdue.edu/place/

- Continuous evaluation of queries over spatio-temporal data streams
- Shared execution among concurrent continuous queries
- Built on top of PREDATOR database system
- Incremental evaluation of continuous queries
- Spatio-temporal query operators

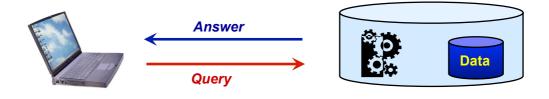




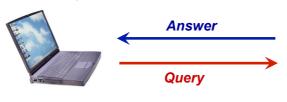
Snapshot vs. Continuous Query Processing

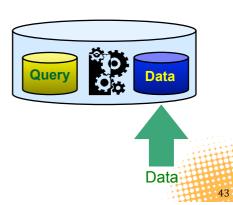


Traditional (Snapshot) Queries



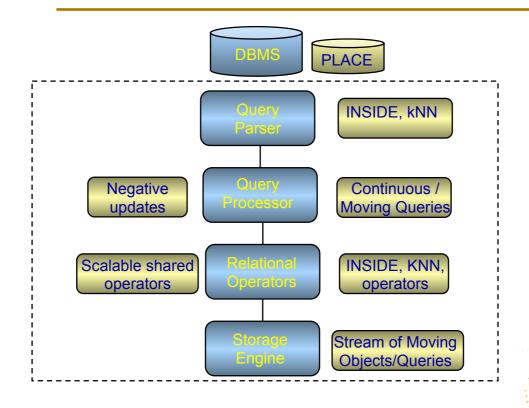
Continuous Queries





PLACE Architecture





Extended SQL Syntax in PLACE



SELECT select_clause
FROM from_clause
WHERE where_clause
INSIDE inside_clause
kNN knn_clause
WINDOW window_clause

- Stationary query: (x_1, y_1, x_2, y_2)
- Moving query: ('M', OID, width, length)

- Stationary query: (k, x, y)
- Moving query: ('M', OID, k)



The Hermes MOD engine



- A palette of data types
 - ... methods and indexes
- ... on top of an extensible DBMS
 - Oracle Spatial (1)
 - PostgreSQL (2)



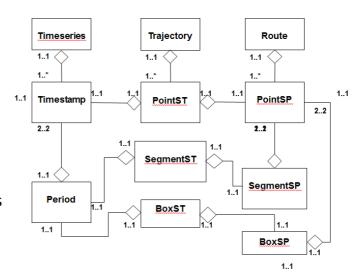
(1) URL: http://infolab.cs.unipi.gr/hermes/

(2) URL: http://hermes-mod.java.net/

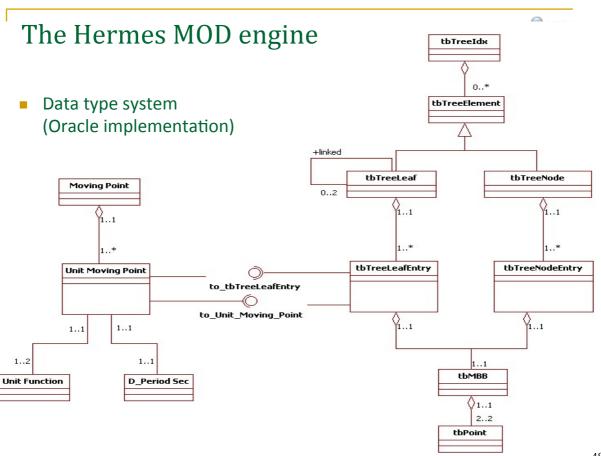




- Data type system (Postgres implementation)
 - Temporal datatypes
 - Timeseries; Timestamp; Period
 - Spatial datatypes
 - Route; PointSP; SegmentSP; BoxSP
 - Spatio-temporal datatypes
 - Trajectory; PointST; SegmentST; BoxST

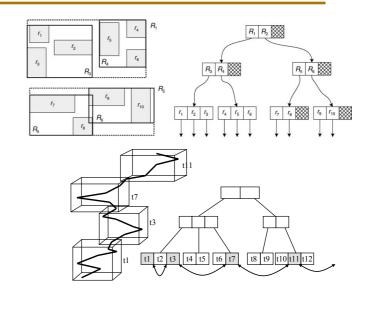








- Indexing support
 - 3D R-trees (in both implementations)
 - TB-trees (in Oracle implementation only)
 - options for indexing a trajectory (both are supported by Hermes):
 - a single MBB for the entire trajectory
 - one MBB per trajectory segment





The Hermes MOD engine



- Methods supported
 - □ Find the (1D) **timestamp** (**temporal period**) of a trajectory ...
 - ... when located at a given spatial point (within a given area, resp.)
 - □ Find the (2D) **location** (route) of a trajectory ...
 - ... at a given timestamp (during a given temporal period, resp.)
 - □ Find the (3D) **sub-trajectory** of a trajectory ...
 - ... spanning during a given temporal period (or within a given spatial area)
 - □ Find the **speed** (or **acceleration**) of a trajectory ...
 - ... at a given timestamp





- Methods supported (cont.)
 - □ Find the **enter-leave points** of a trajectory ...
 - ... w.r.t. a given spatial area
 - Apply spatio-temporal compression on a trajectory
 - ... w.r.t. state-of-the-art trajectory compression algorithm (TD-TR)
 - □ Find the **Potential Area of Activity** (PAA) of a trajectory ...
 - ... between (sequences of) sampled locations
 - etc. etc. methods



The Hermes MOD engine



- Queries supported
 - Timeslice queries
 - Range queries, 3 variations:
 - spatial only vs. temporal only vs. spatio-temporal range window
 - Nearest Neighbor (NN) queries, 2 variations:
 - w.r.t. a reference location (spatial point or area)
 - w.r.t. a reference trajectory
 - Pair-wise trajectory similarity queries
 - ... w.r.t. popular and state-of-the-art trajectory similarity functions:
 - □ Manhattan (L_1), Euclidean (L_2), Chebyshev (L_∞), ...
 - □ LCSS, DTW, ERP, EDR, DISSIM, ...



- Queries supported (cont.)
 - Trajectory cluster analysis queries:
 - ... w.r.t. state-of-the-art trajectory clustering algorithms (T-Optics, S2T-Clustering
 - OD-matrix queries, 2 variations:
 - origin/destination pairs be spatial or spatio-temporal ranges





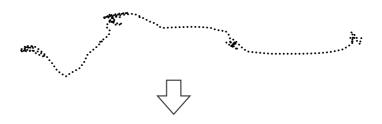
The next wave: semantic trajectories



From raw to "semantic" trajectories



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



meaningful mobility tuples <place, time_{in}, time_{out}, tags>



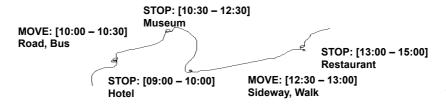
- Semantic Trajectory: $T=\{e_{first},...,e_{last}\}$
- Episode: e_i = (STOP | MOVE, t_{from} , t_{to} , place, tag)



From raw to "semantic" trajectories



- Detection of homogenous fractions of movement,
 - Trajectory is reconstructed as a sequence of episodes (stops/moves)
- 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?





InfoLaB Offline construction of semantic trajectories (Yan et al. ESWC '10, EDBT '11) output semantic MARKET trajectory Semantic Annotation • spatial join (region) map-matching (line) • HMM (point) Layer structured trajectory S1() · velocity-based Trajectory · density-based Structure · orientation Layer spatiotemporal trajectory • raw GPS gap Trajectory trajectory Identification time interval Layer · spatial extent cleansed GPS feeds Data outlier removal **Preprocess** · kernel smoothing Layer compression original input GPS feeds

Summary



Summary on Mobility Data Management



- From (stationary) spatial to moving object databases
- Advances in ...
 - Modeling and representation (at logical database level)
 - Moving point, trajectory etc. data types
 - Indexing and query processing (at physical database level)
 - 3D R-trees, TB-trees, etc.
 - Coordinate- vs. trajectory- based queries
- A few prototype MOD engines are already out there ...
 - state-of-the-art: SECONDO and HERMES (offline mode), PLACE (streaming mode)



Questions









Reading list



Mobility data modeling & MOD engines



- de Almeida, V.T. et al. (2006) <u>Querying Moving Objects in SECONDO</u>.
 Proceedings of MDM.
- □ Behr, T. and Güting, R.H. (2005) <u>Fuzzy Spatial Objects: An Algebra Implementation in SECONDO</u>. Proceedings of ICDE.
- Cao, H. and Wolfson, O. (2005) <u>Nonmaterialized Motion Information in Transport Networks</u>. Proceedings of ICDT.
- □ Chen, C.X. and Zaniolo, C. (2000) <u>SQLST: A Spatio-Temporal Data Model and Query Language</u>. Proceedings of ER.
- Cheng, R. et al. (2004) <u>Efficient Indexing Methods for Probabilistic Threshold</u> <u>Queries over Uncertain Data</u>. Proceedings of VLDB.
- Dieker, S. and Güting, R.H. (2000) <u>Plug and Play with Query Algebras:</u> <u>SECONDO – A Generic DBMS Development Environment</u>. Proceedings of IDEAS.



Mobility data modeling & MOD engines



(cont.)

- Güting, R.H. et al. (2000) <u>A Foundation for Representing and Querying Moving Objects</u>. 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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