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INSTITUT FÜR INFORMATIK
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FÜR DATENBANKSYSTEME



Indexing Uncertain Spatio-Temporal Data

Tobias Emrich, Hans-Peter Kriegel, Matthias Renz, Andreas Züfle (LMU)
Nikos Mamoulis (HKU)



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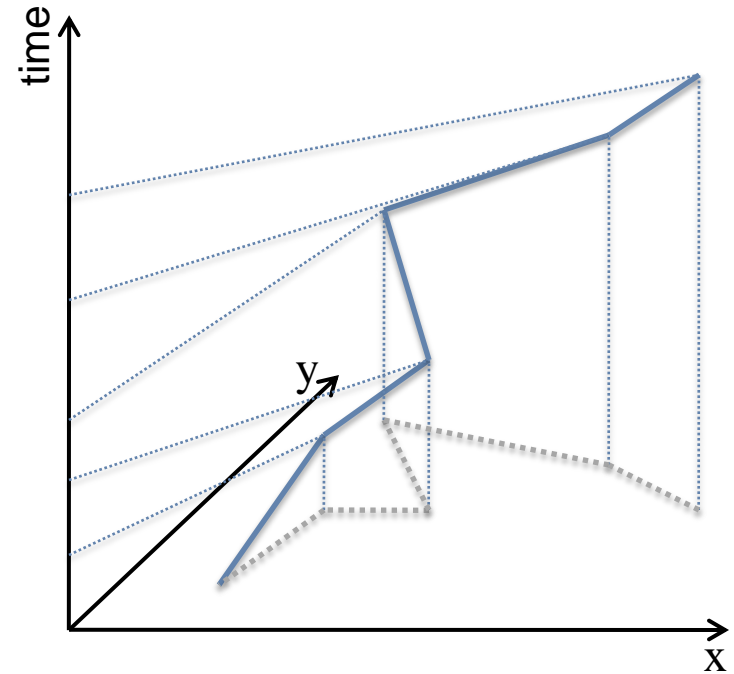
Agenda

- Spatio–Temporal Data
- Uncertain Spatio–Temporal Data
- Modeling Uncertain Spatio–Temporal Data
- Indexing Uncertain Spatio–Temporal Data
- Probabilistic Pruning
- Experiments

What is (certain) Spatio-Temporal Data?

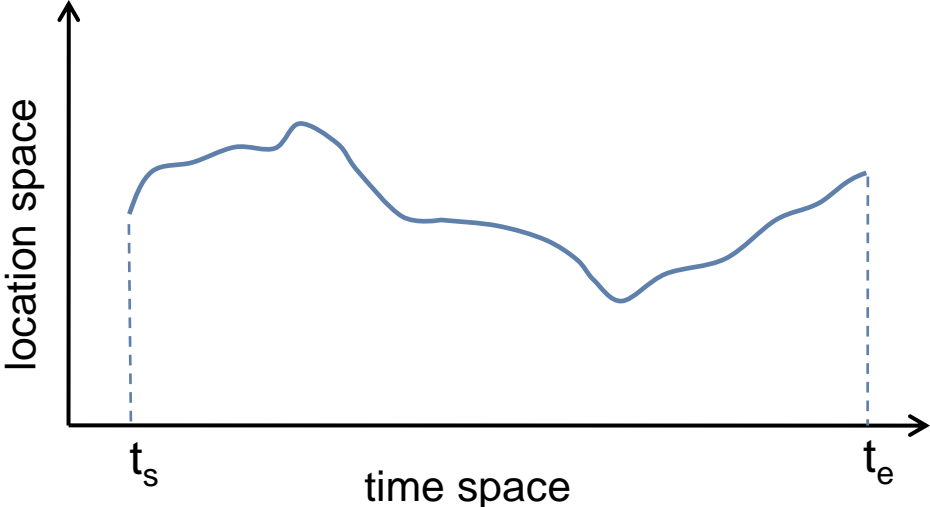
- Trajectory data from
 - vehicles
 - mobile users
 - animals
 - ...

- Can be represented treating time as an additional spatial dimension



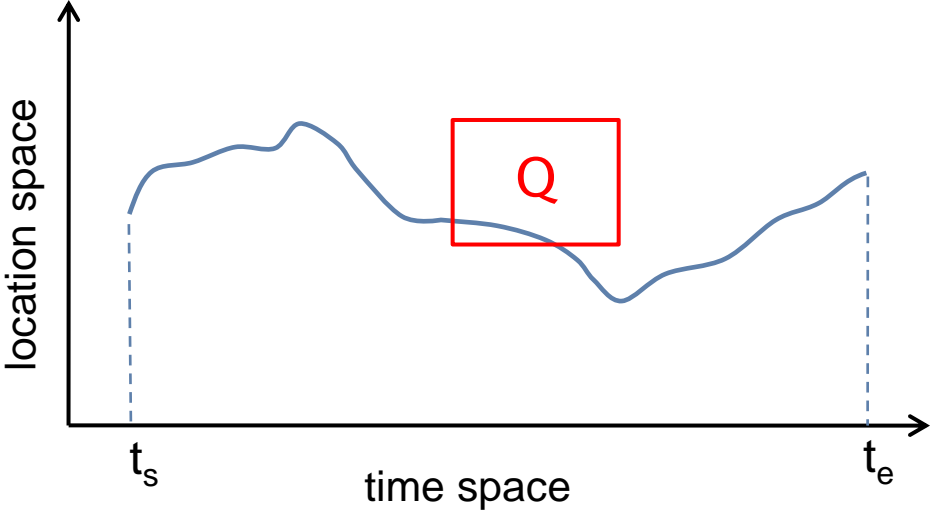
What is (certain) Spatio-Temporal Data?

- A spatio-temporal database stores triples (oid, time, loc)
- In the best case, this allows to look up the location of an object at any time



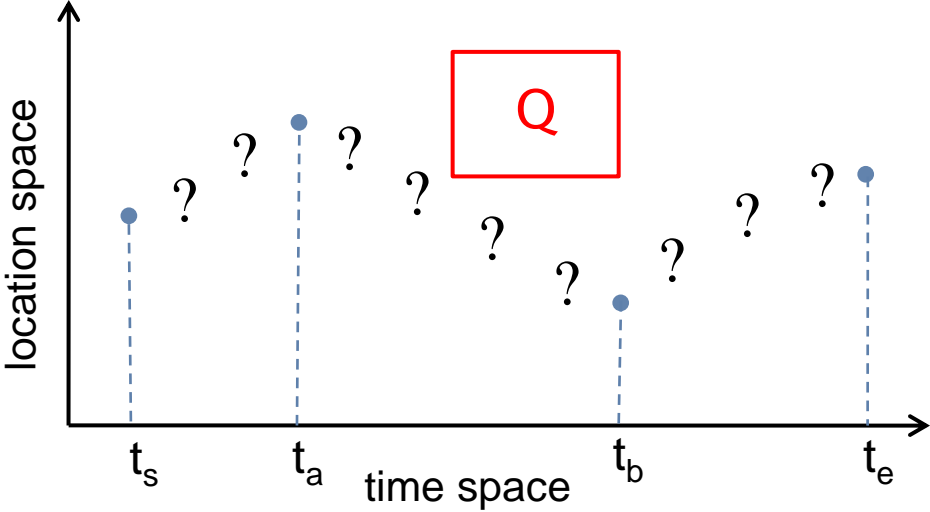
What is (certain) Spatio-Temporal Data?

- A spatio-temporal database stores triples (oid, time, loc)
- In the best case, this allows to look up the location of an object at any time
- Allows to answer queries such as „Return objects that intersects some spatial window within some time interval“.



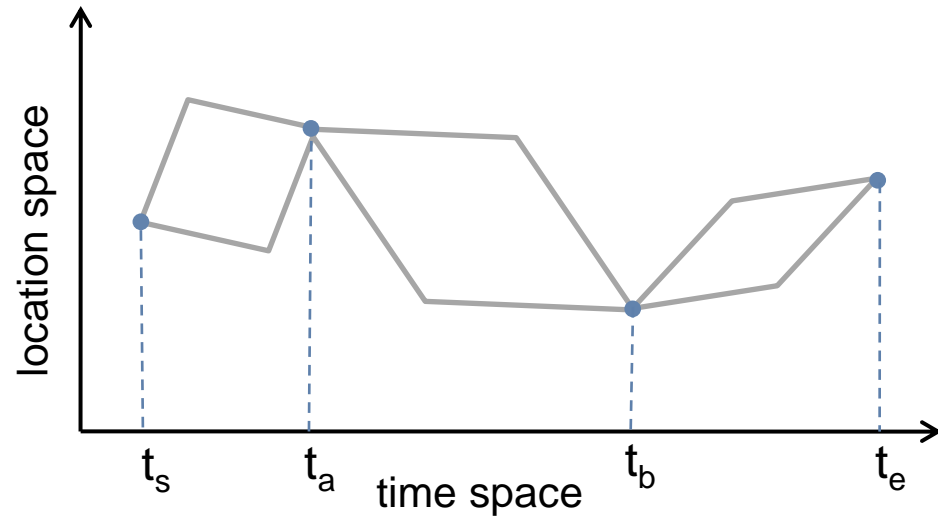
What is Uncertain Spatio-Temporal Data?

- In most applications, this data is not complete
 - Delays between GPS signals
 - RFID sensors located only in certain locations
 - Wireless sensor nodes sending infrequently to preserve power
 - Geo-application check-ins



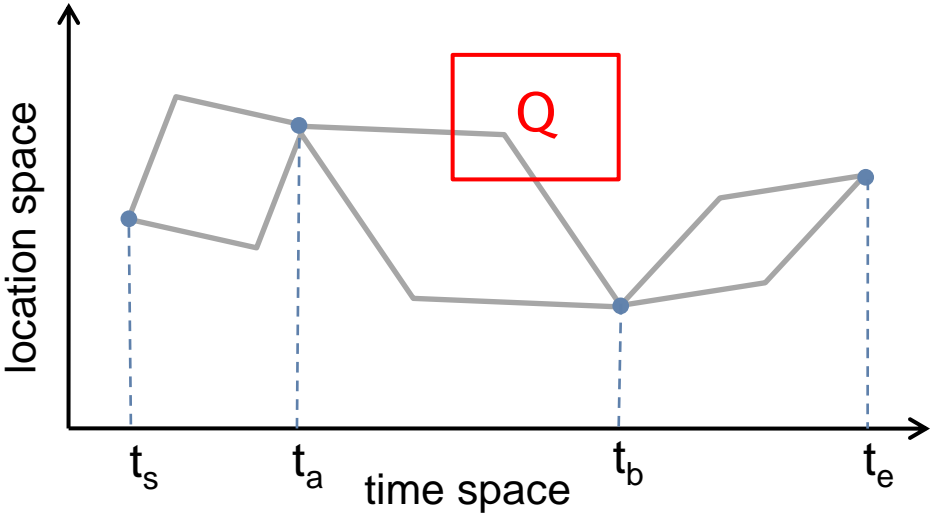
What is Uncertain Spatio-Temporal Data?

- Existing works
 - Bound the set of possible (location,time) pairs of an object between **observations** by using spatio-temporal approximations (**diamonds**)
 - e.g. by modeling knowledge about maximum speed



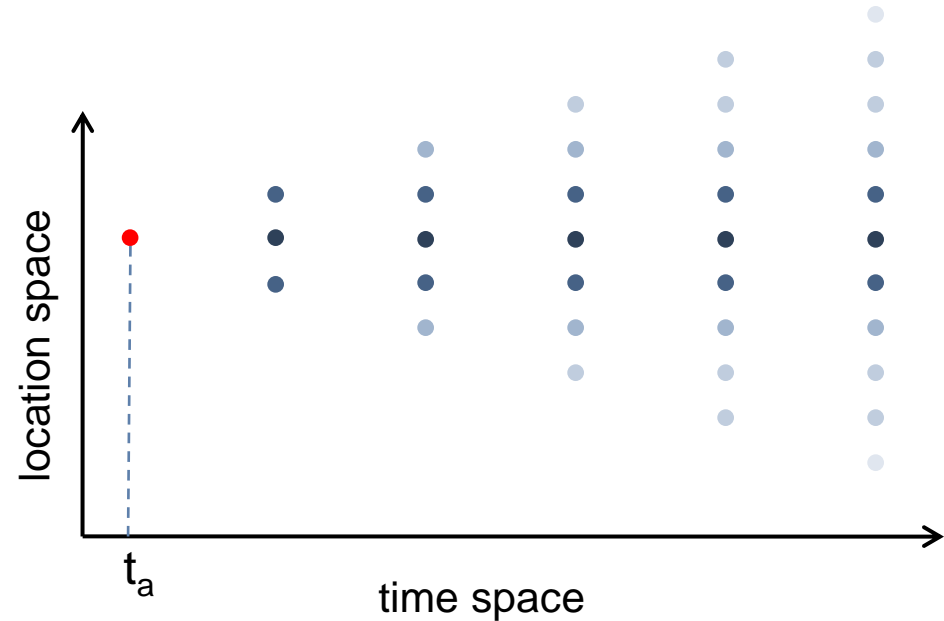
What is Uncertain Spatio-Temporal Data?

- Existing works
 - Bound the set of possible (location,time) pairs of an object between **observations** by using spatio-temporal approximations (**diamonds**)
 - e.g. by modeling knowledge about maximum speed
 - Allows to make statements like „its possible that o intersects some query window Q“
 - **But how likely is this event?**
„What is the probability of the object traveling through Q?“



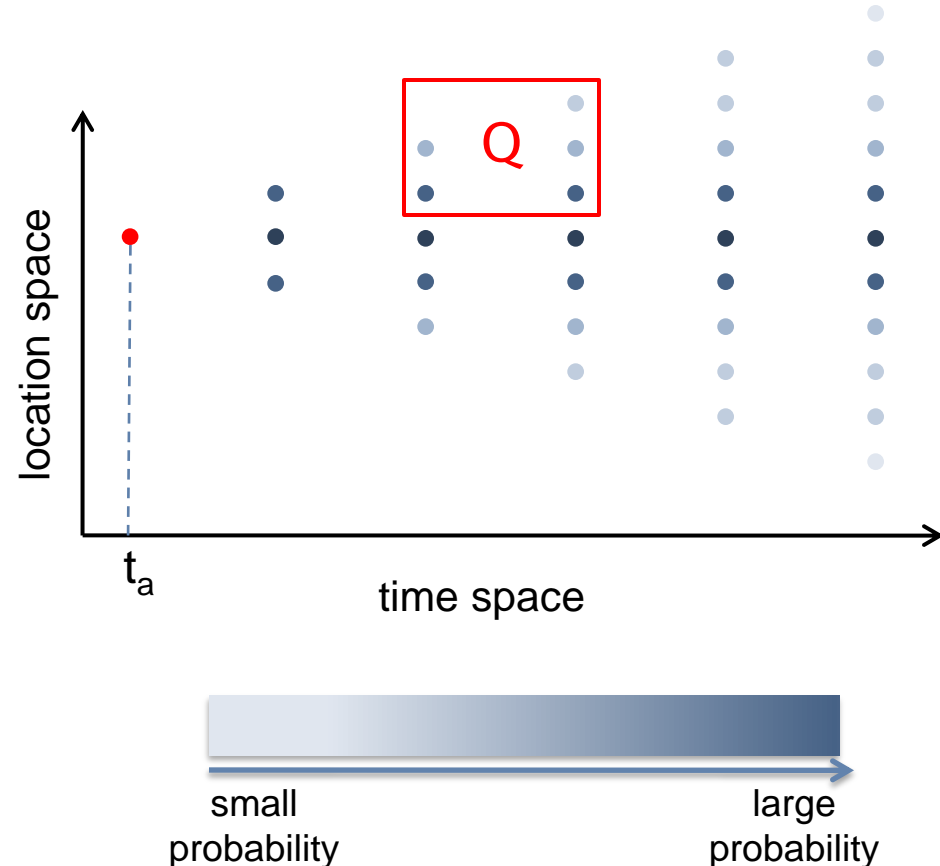
Modeling Uncertain Spatio-Temporal Data

- Using a more powerful model [1]
- The position of an object o at some time t is a random variable
- The trajectory of o follows a stochastic process, i.e. a family of random variables $o(t)$



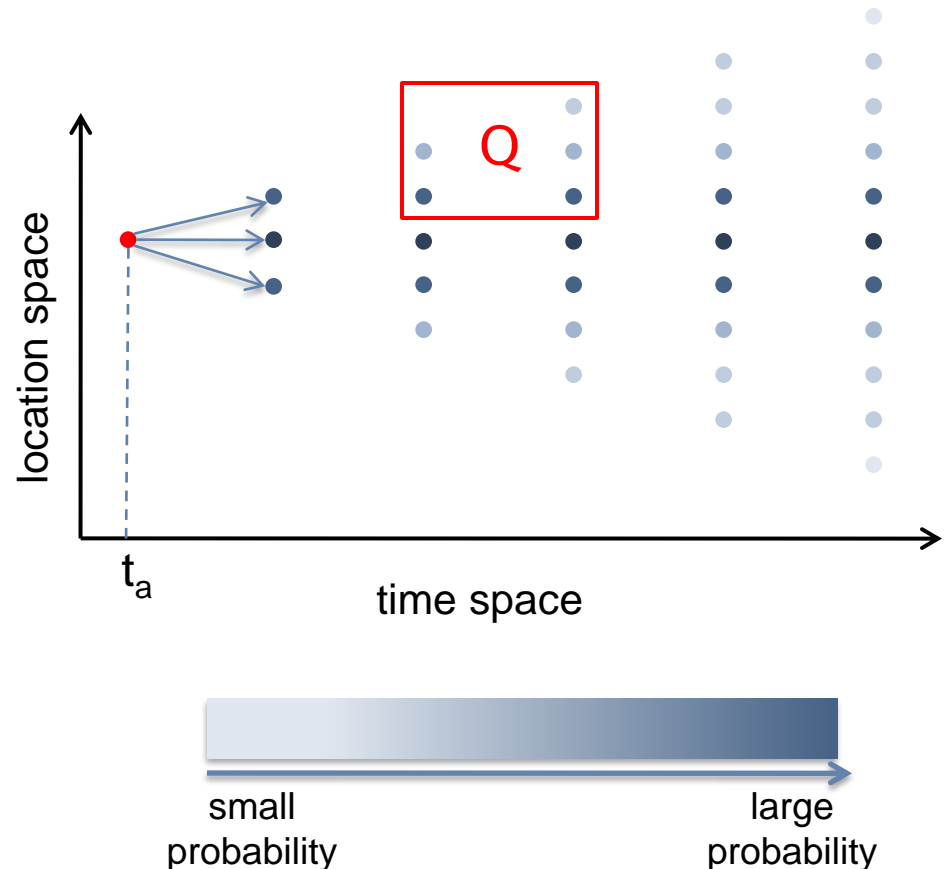
Modeling Uncertain Spatio-Temporal Data

- Using a more powerful model [1]
- The position of an object o at some time t is a random variable
- The trajectory of o follows a stochastic process, i.e. a family of random variables $o(t)$
- Given a predicate Φ , the event that o satisfies Φ is a random event.



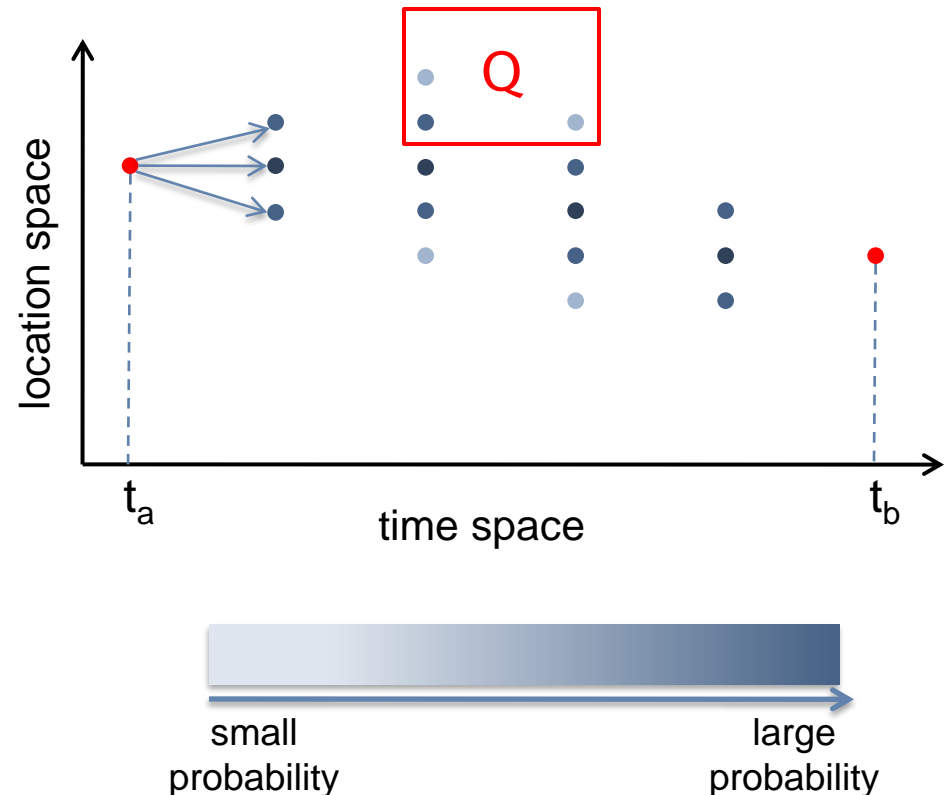
Markov Chain Model for UST data

- Assumes discrete state space S and discrete time space T
- Given the position of an object o at time $t=i$, the position at $t=i+1$ is conditionally independent of $t=i-1$
- Transition probabilities stored in a (sparse) $|S| \times |S|$ matrix $M(o,t)$, called transition matrix
- $M(o,t)[i,j]$ is the probability that object o will transition to state j at time $t+1$, given o is located at state i at time t



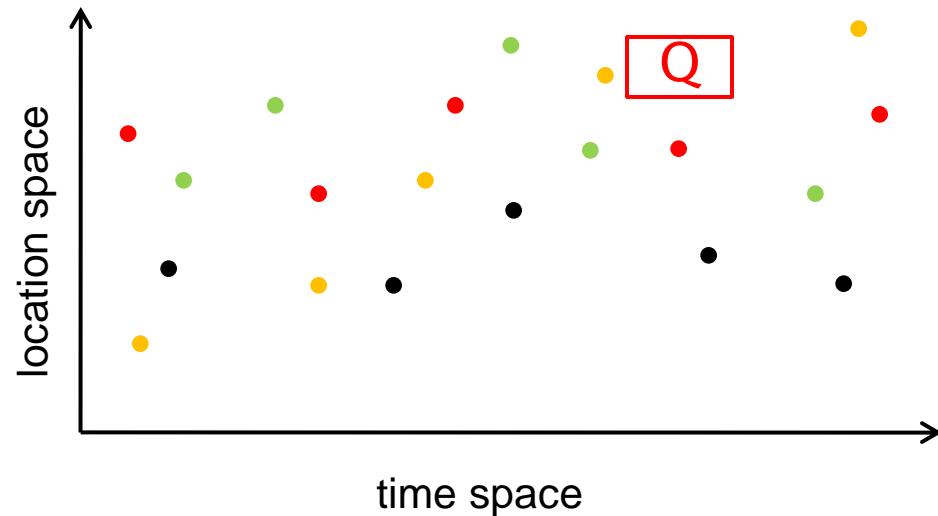
Markov Chain Model for UST data

- Incorporation of additional observations possible using Bayesian inference
- Each possible path can be associated with a probability
- Several probabilistic window queries possible
- Use sparse matrix operations for efficient implementation
- Details can be found in [1]



Indexing Uncertain Spatio-Temporal Data

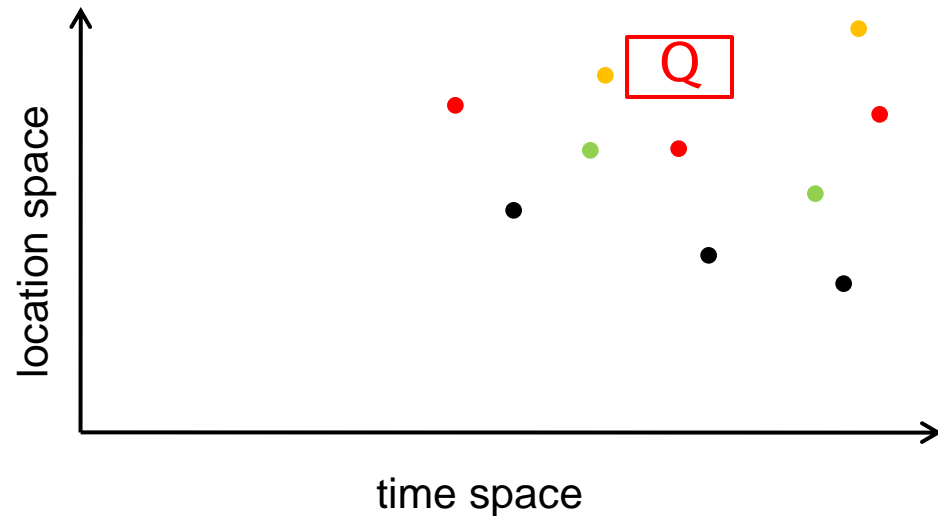
- Large number of objects and observations in a database
- Checking each pair of successive observations requires too much computation
- How to prune as many objects and observations as possible during query evaluation?



“Which objects intersect Q with a probability of at least τ ?”

Indexing Uncertain Spatio-Temporal Data

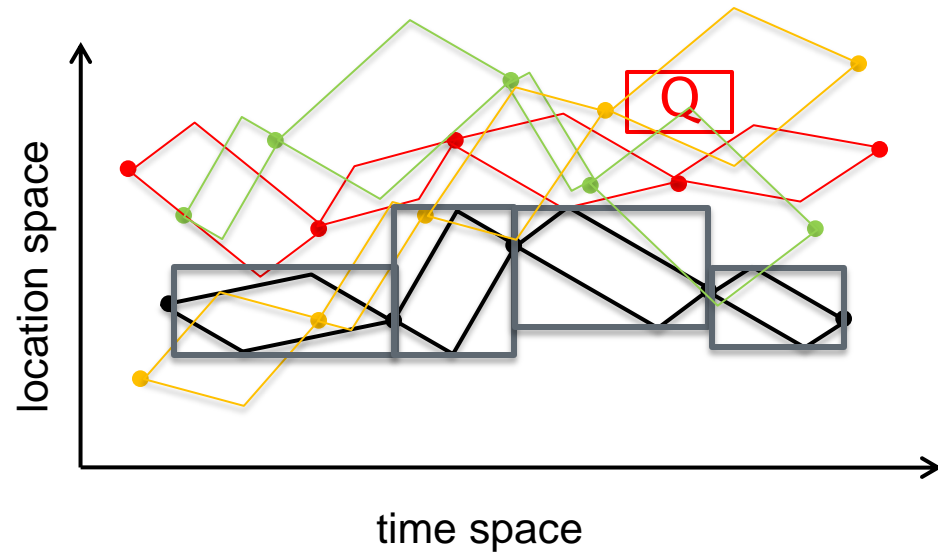
- Large number of objects and observations in a database
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- How to prune as many objects and observations as possible during query evaluation?
- Temporal Pruning?



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Indexing Uncertain Spatio-Temporal Data

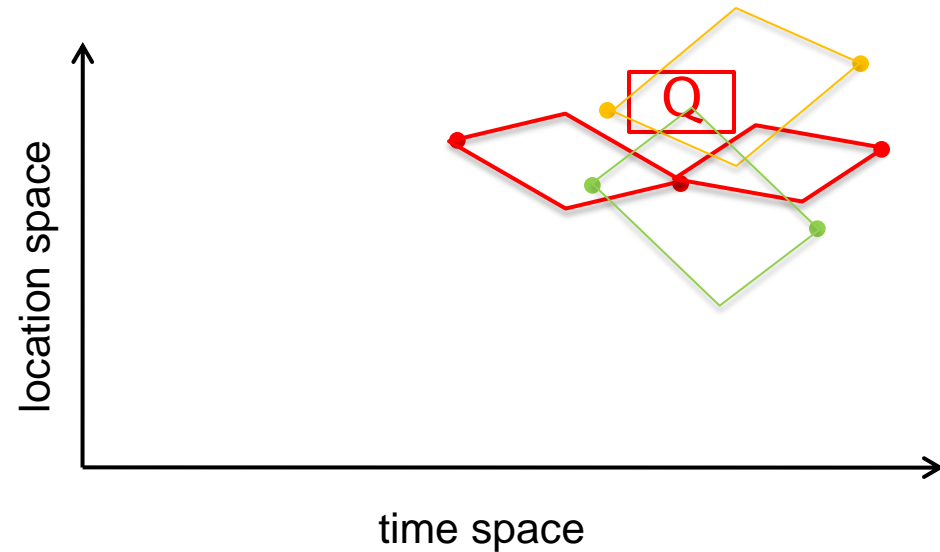
- Index possible positions in (location + time) in an R*-Tree
- Apply multistep filter
 - R-Tree Filter



“Which objects intersect Q with a probability of at least τ ?”

Indexing Uncertain Spatio-Temporal Data

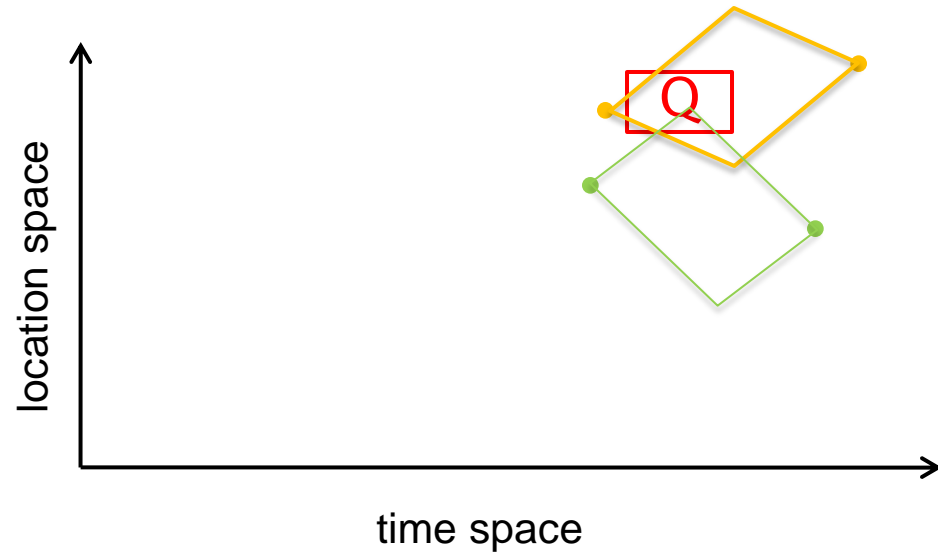
- Index possible positions in (location + time) in an R*-Tree
- Apply multistep filter
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 - Diamond Filter



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Indexing Uncertain Spatio-Temporal Data

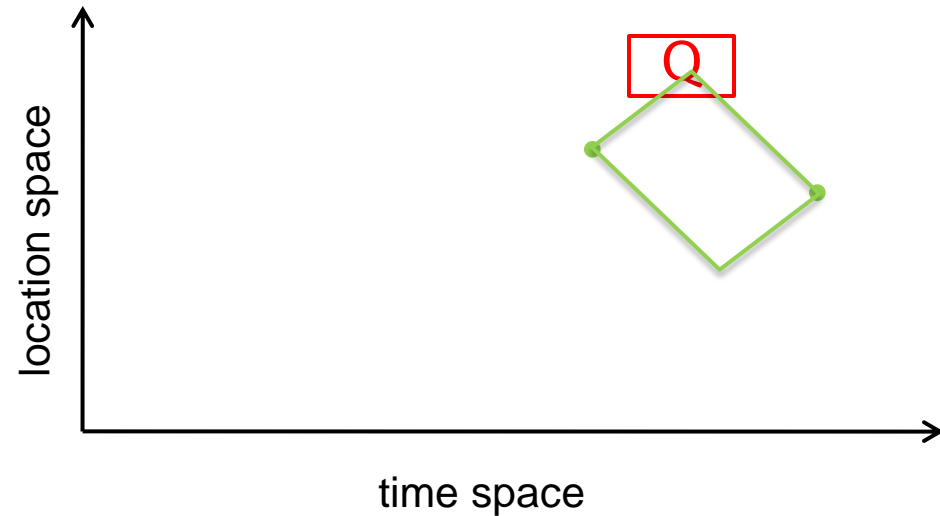
- Index possible positions in (location + time) in an R*-Tree
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 - True Hit Detection



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Indexing Uncertain Spatio-Temporal Data

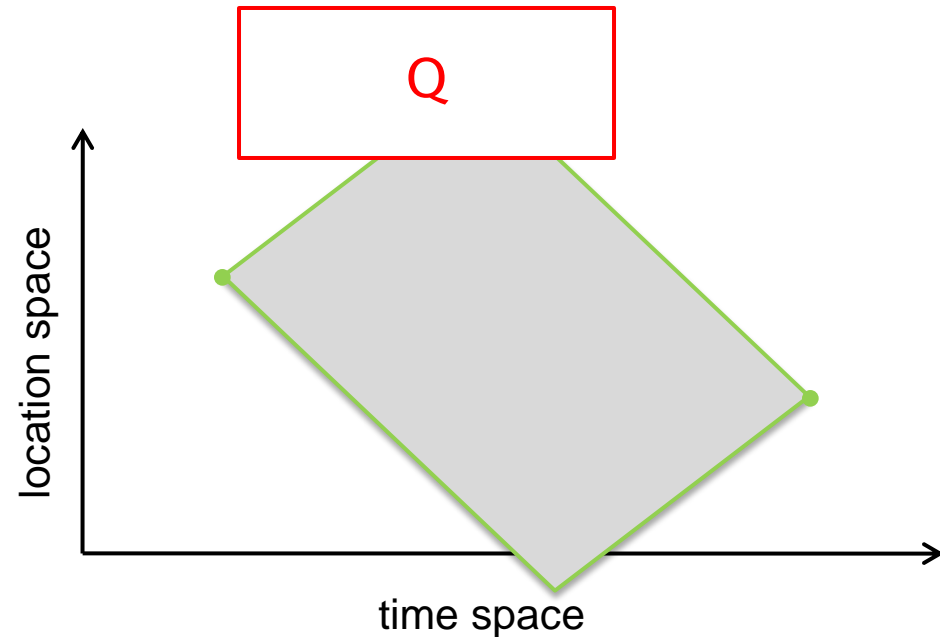
- Index possible positions in (location + time) in an R*-Tree
- Apply multistep filter
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 - Diamond Filter
 - True Hit Detection
 - Probabilistic Pruning



“Which objects intersect Q with a probability of at least τ ?”

Probabilistic Pruning

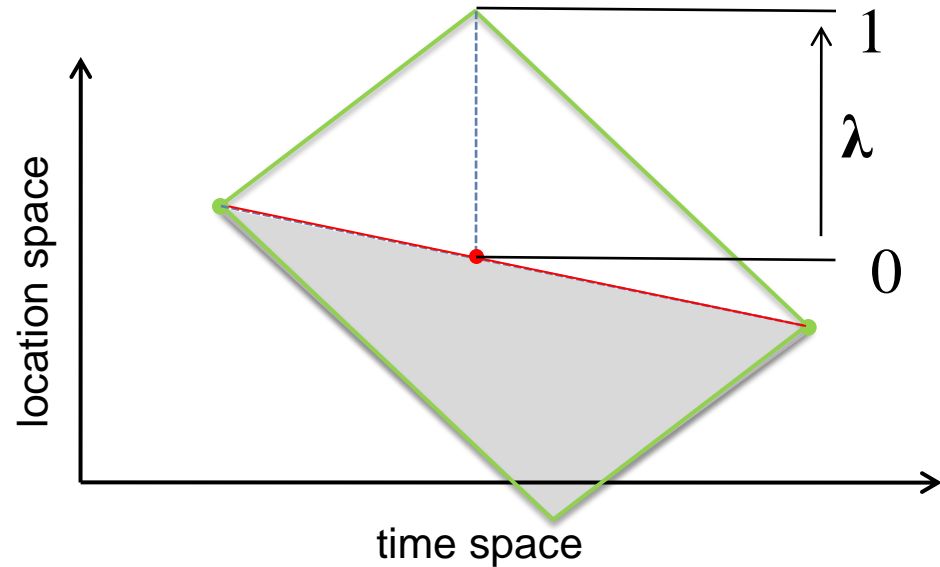
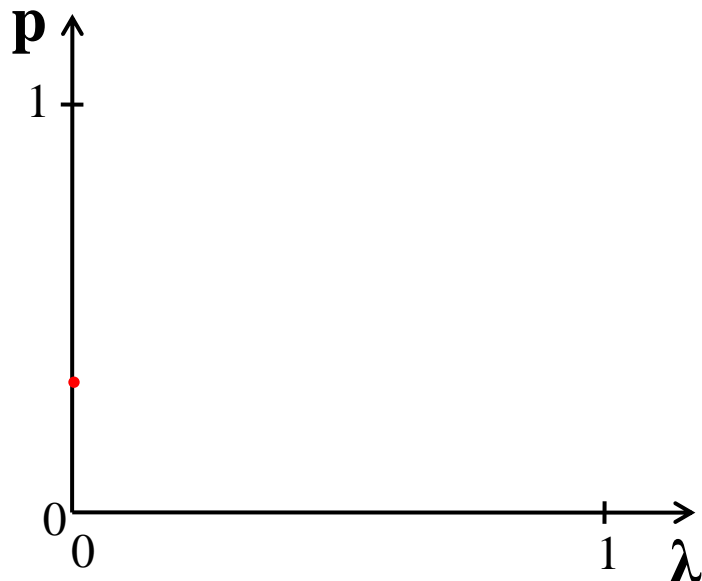
- Lower bound the probability of the object (o) to move through the gray area only $P_{LB}(o \text{ in gray})$
- Then $P_{UB}(o \text{ intersects } Q) = 1 - P_{LB}(o \text{ in gray})$
- If $P_{UB}(o \text{ intersects } Q) < \tau$
 $\Rightarrow o$ can be pruned
- How obtain $P_{LB}(o \text{ in gray})$ without computing it?



“Which objects intersect Q with a probability of at least τ ?”

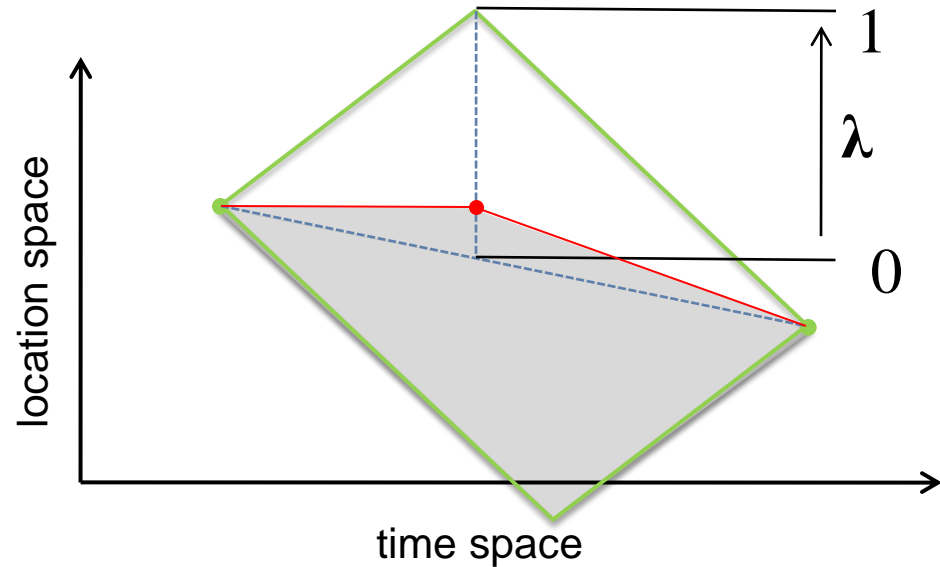
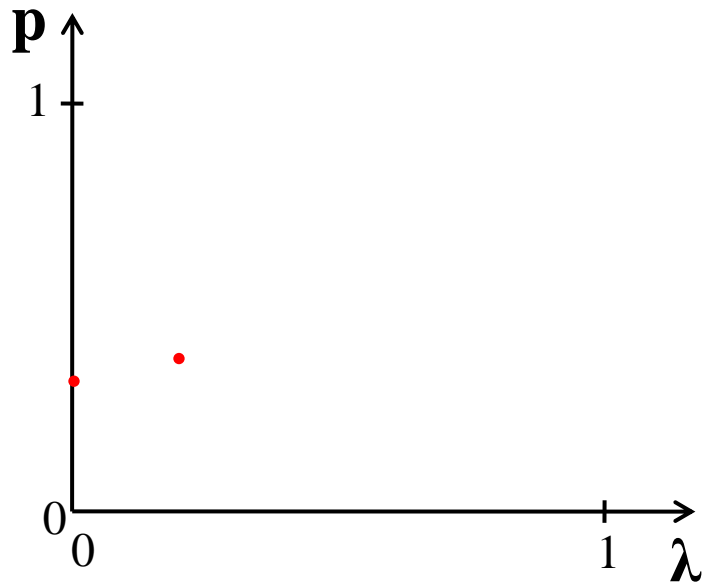
Probabilistic Pruning

- Offline computation:
 - Precompute probability for θ to stay in sub-diamonds
 - Simple variation of the window query for each sub-diamond



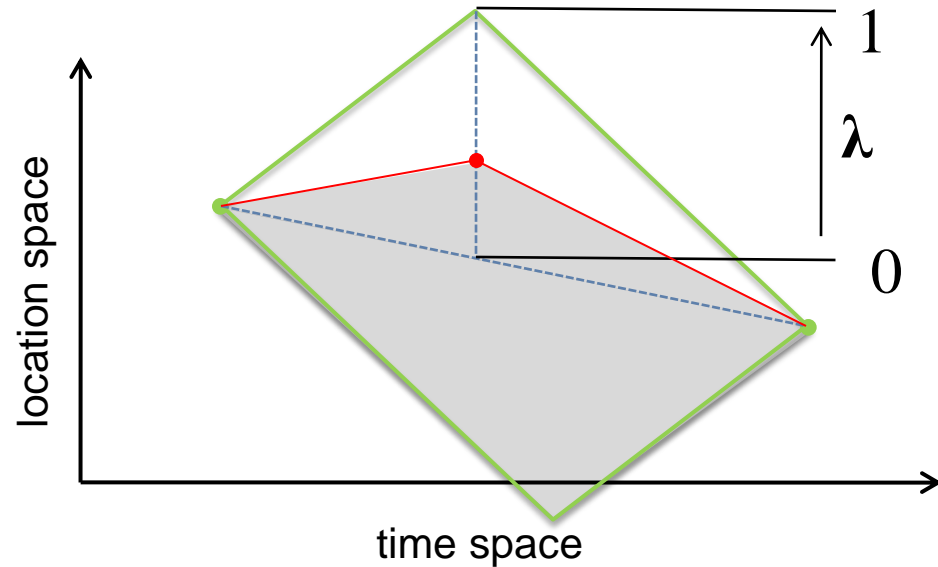
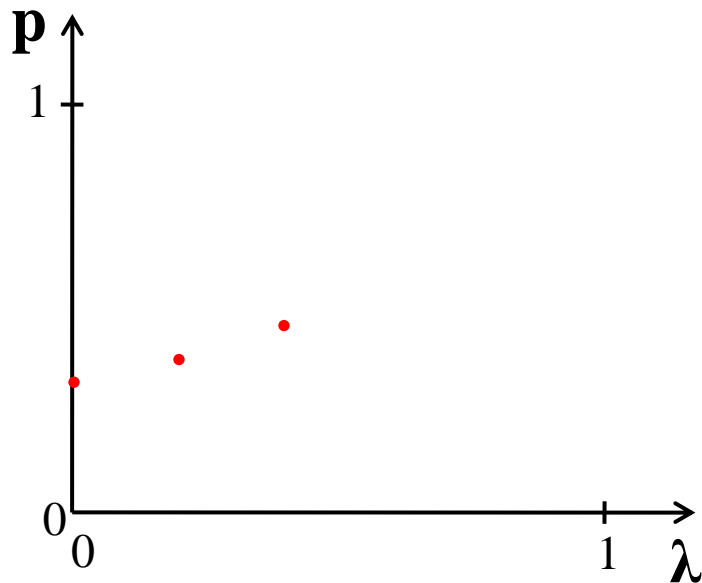
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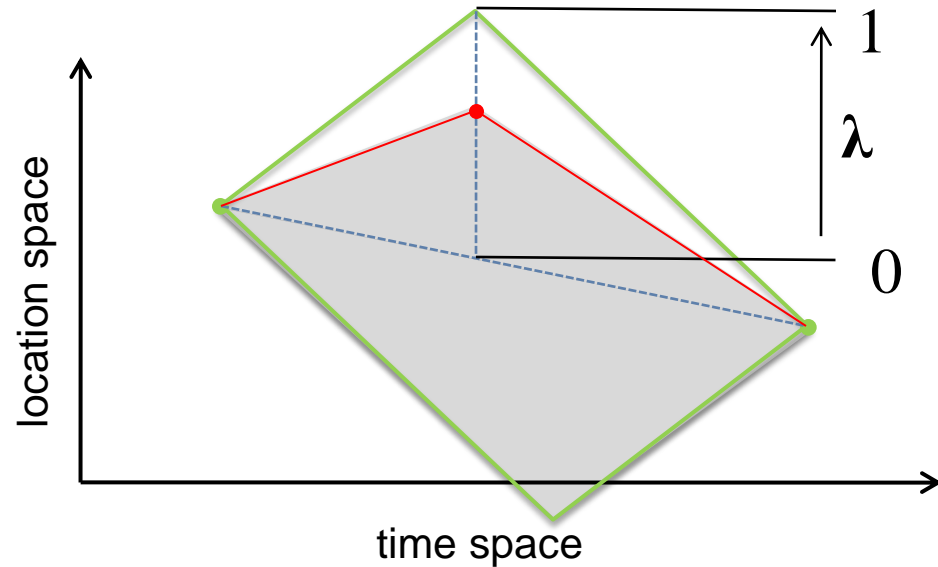
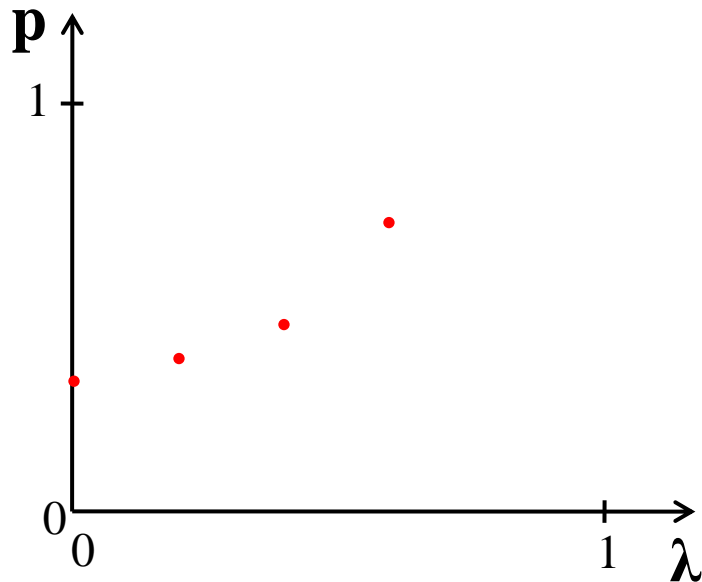
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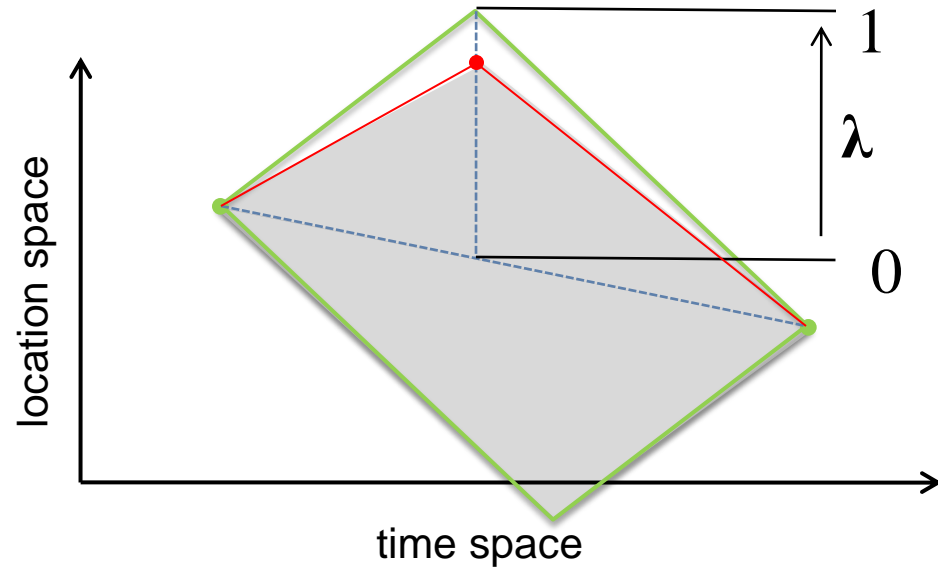
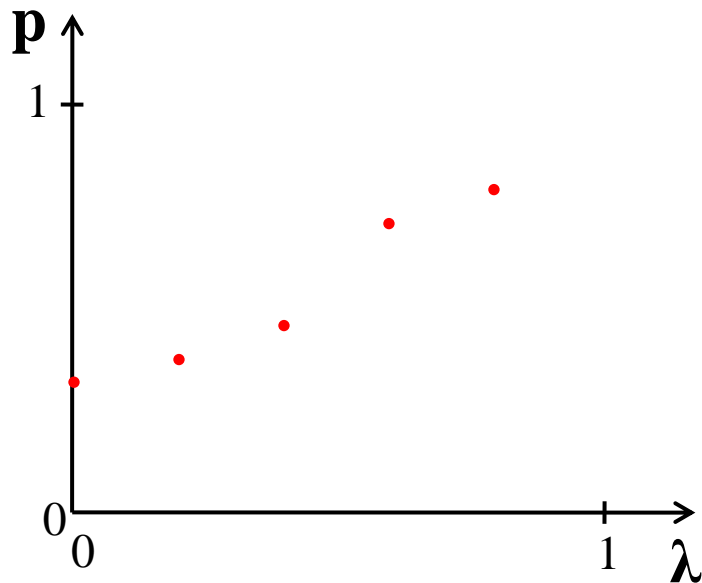
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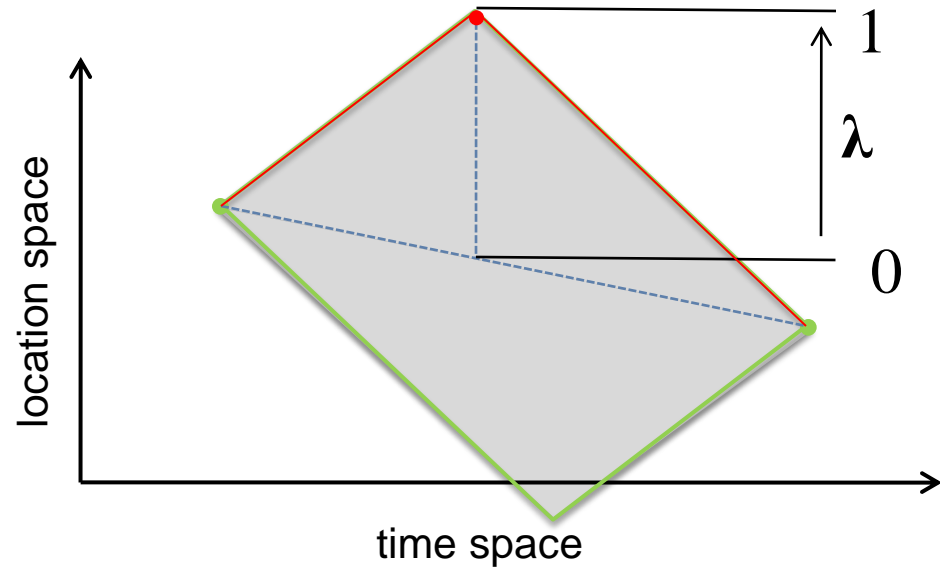
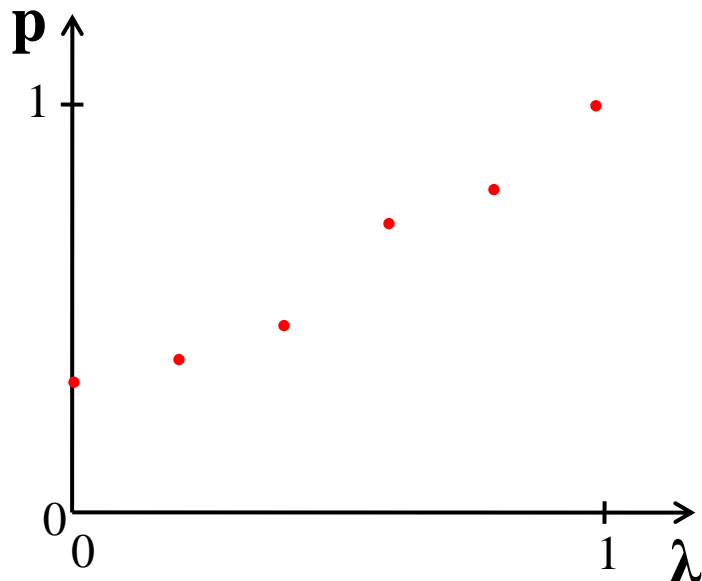
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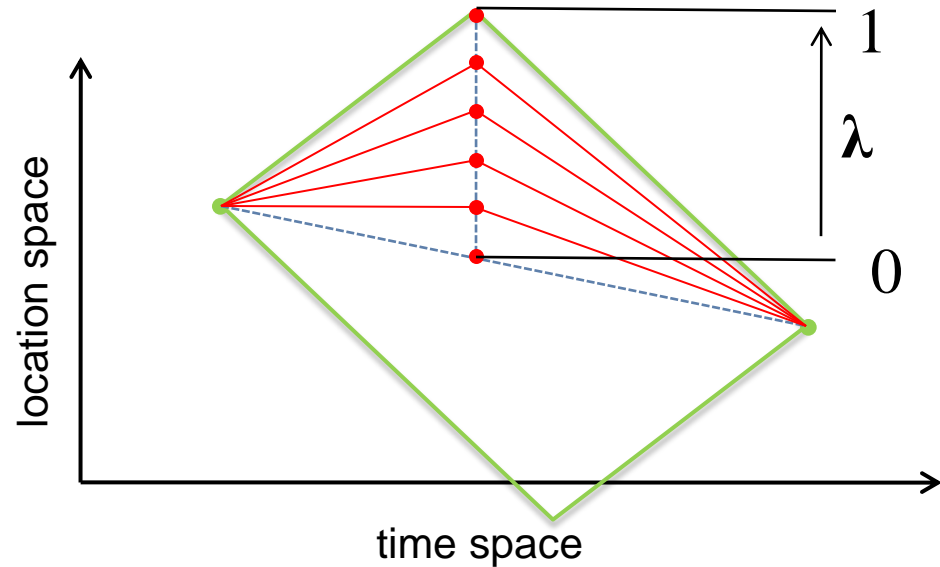
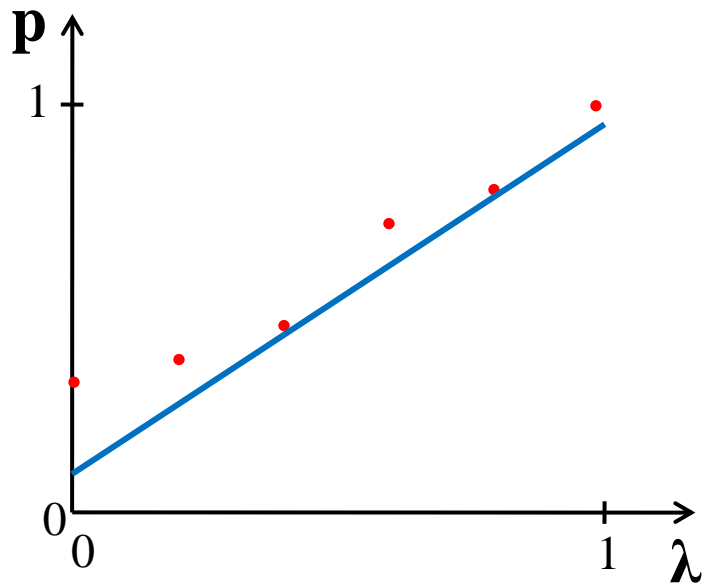
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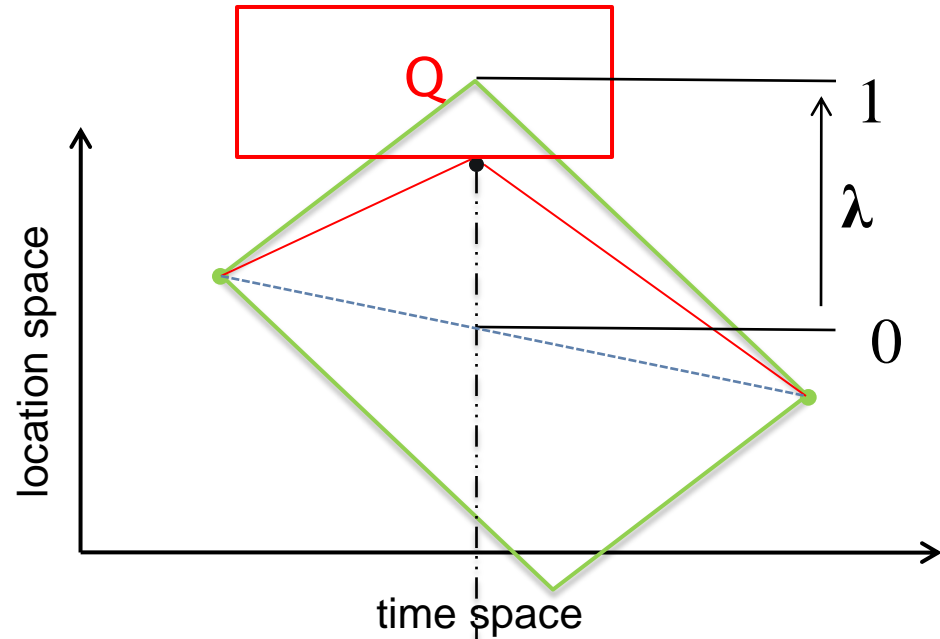
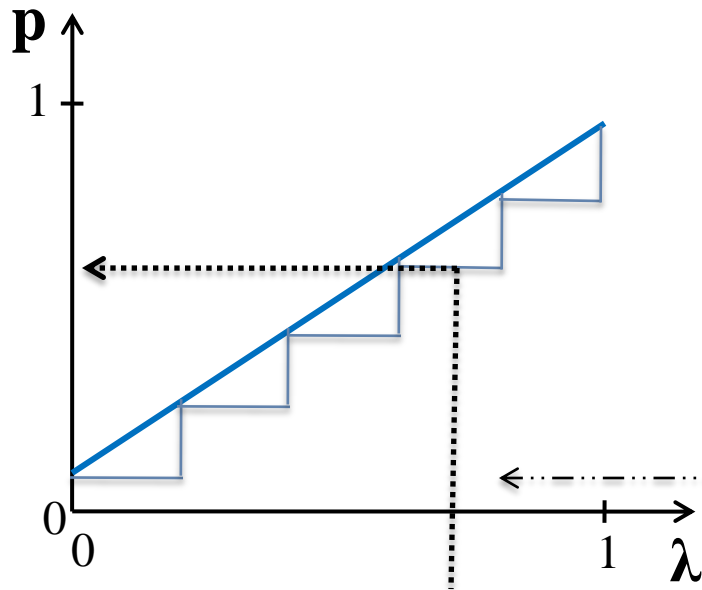
Probabilistic Pruning

- Offline computation:
 - Storing all these probabilities is of course memory inefficient
 - Approximation with a linear function using linear programming



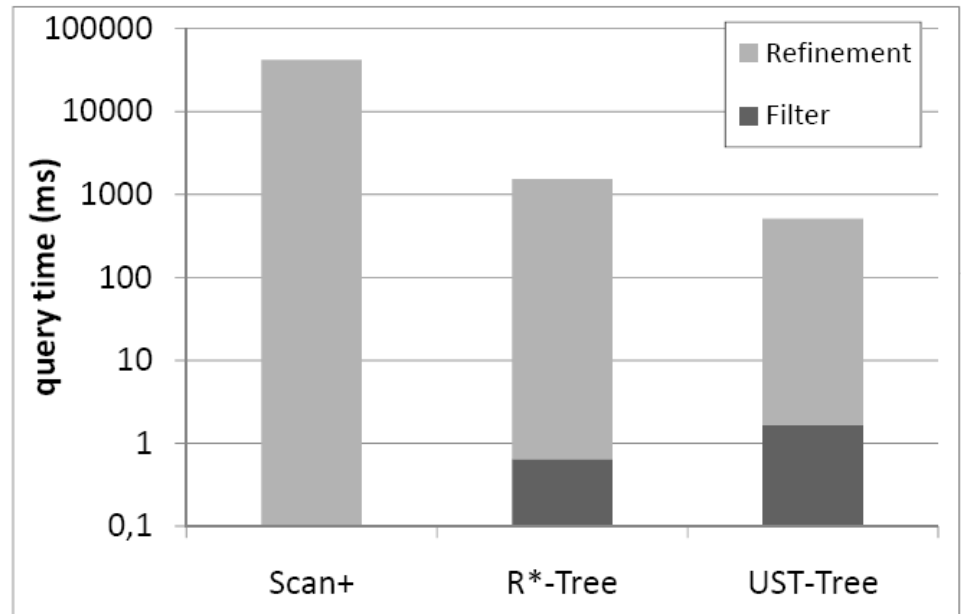
Probabilistic Pruning

- During Query Processing:
 - Construct largest possible sub-diamond
 - Lookup of $P_{LB}(o \text{ in sub-diamond})$



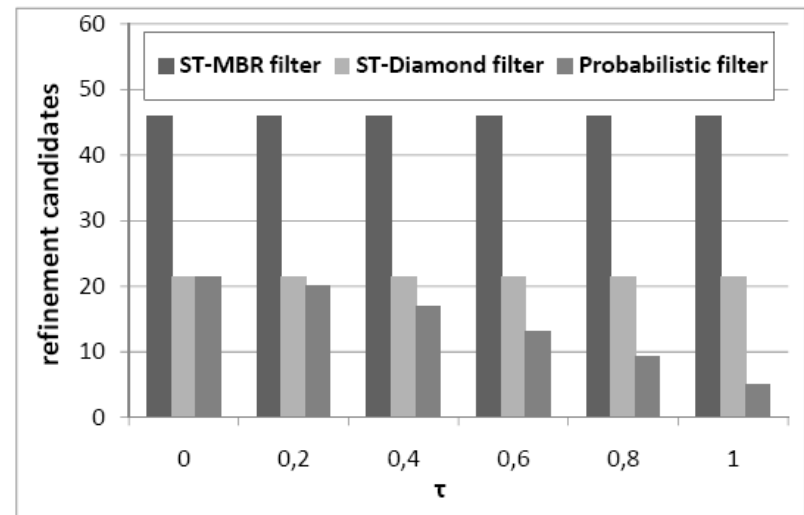
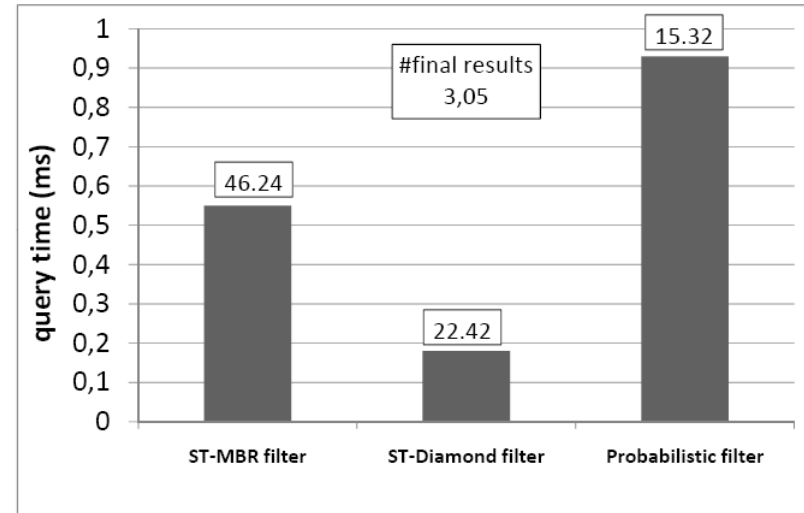
Experiments

- Indexing UST data yields orders of magnitude over a scan based method (with temporal pruning)
- Using diamond and probabilistic filtering (UST-Tree) speeds up the R*-Tree by factor 3-4
- Setting: 1000 objects á 100 observations and 10000 possible locations in 2D



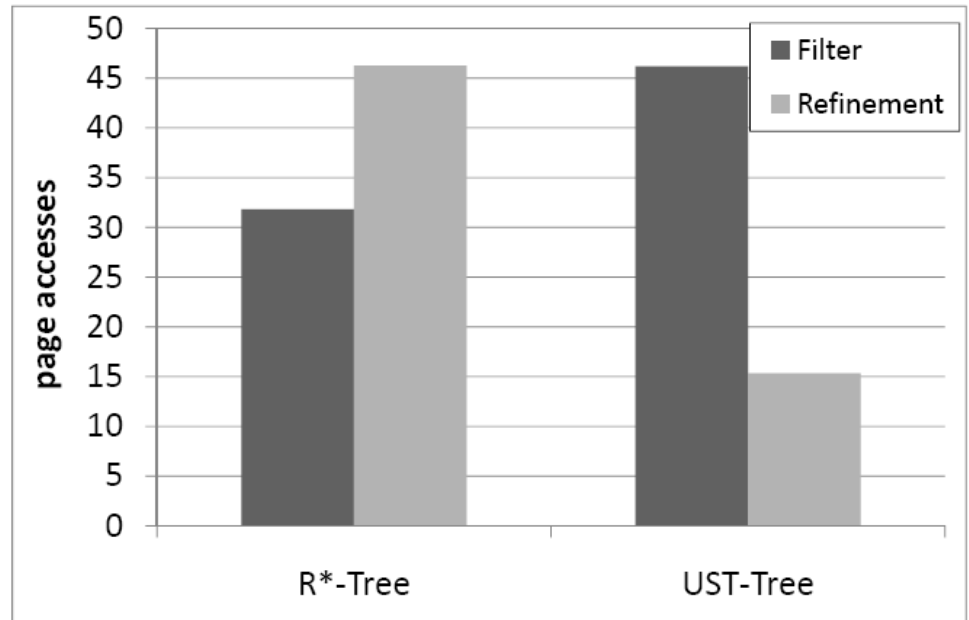
Experiments

- All Filters can be computed efficiently in contrast to the verification step (~500 ms in this example)
- Effectiveness of probabilistic filter is dependent on the threshold τ



Experiments

- Probabilistic spatio-temporal window queries are usually CPU-bound
- I/O-cost of UST-Tree are higher in the filter step since more information is stored in the leaves (\Rightarrow higher tree)
- I/O-cost of UST-Tree are lower in the refinement step since less candidates have to be loaded from disk



Summary

- Indexing based on the Markov-Chain Model for UST data
 - Yields several orders of magnitudes over scan-based method
 - Yields 3-4 times speedup over straightforward indexing
- Techniques for spatio-temporal and probabilistic pruning are applicable to other models for UST data
- Techniques are extendable to the multi-dimensional case
- More special cases are treated in the paper
 - Other window queries
 - Query window overlapping several diamonds of the same object



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Thank you for listening!



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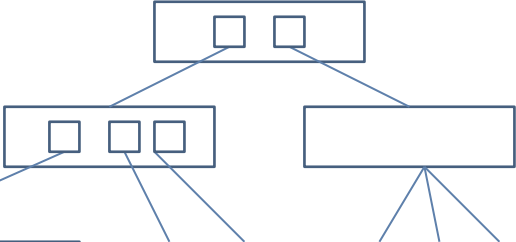


Related Work

- [1] T. Emrich, H.-P. Kriegel, N. Mamoulis, M. Renz, and A. Züfle. *Querying uncertain spatio-temporal data*. In Proceedings of the 28th International Conference on Data Engineering (ICDE), Washington, DC, 2012.

Index Entries at Leaf Level:

directory levels:



leaf level:

