

Time Parameterized Queries in Spatio Temporal Databases

Instructor: Cyrus Shahabi

Outline

- Introduction
- Related work
- TP Queries
 - TP Window Query
 - TP K Nearest Neighbor Query
 - TP Join Query
- Conclusion



Introduction

- Conventional Queries
- Continuous Queries
- Time Parameterized Queries



Conventional Queries

- These are the traditional ‘*instantaneous*’ queries that are evaluated only once to return a single result .
- Are these type of queries reliable in dynamic environments?
 - No!
- Why?
 - The results of a conventional query may be invalidated very soon due to the movements of objects and queries
 - E.g., Which are my nearest gas stations now?



Continuous Queries

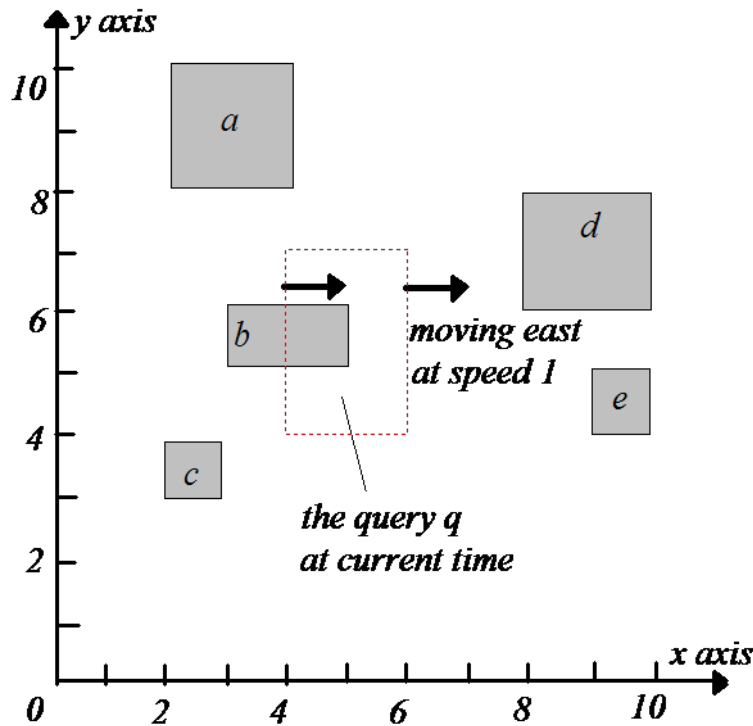
- What is Continuous Queries?
 - Moving objects only
 - Moving queries only
 - Both
- How to deal with such queries? Updates!
 - When to update?
- Due to their dynamic nature , the result of any query is strongly related to the **temporal** context.



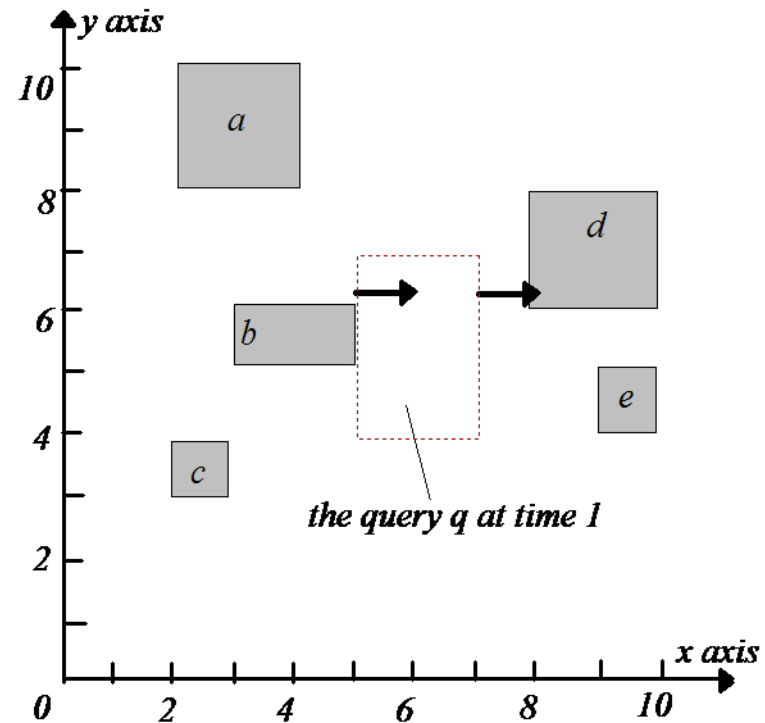
Time Parameterized Queries

- Time Parameterized queries (TP) , whenever a query is issued , a TP returns
 - The **actual result** that satisfies the corresponding spatial query.
 - The **validity period**/expiration time of the result.
 - The **change** that cause the expiration of the results.

Time Parameterized Queries



- Result={b}
- Conventional Query



- At time 1 b would be the nearest neighbor, after that time the results expire and d would be the new nearest neighbor
- Time Parameterized Query



Time Parameterized Queries

- Instead of Updates:
 - Here the objects dynamic behaviour does not necessarily require updates , but can be stored as a function of time using appropriate indexes.
- All TP queries could be reduced to:
 - Some form of NN search
- Could be applied in
 - Moving object
 - Moving query
 - Both



Related Work

- Time Parameterized R-tree (TPR tree)
 - TPR-tree is an extension of an R-tree that can answer prediction queries on dynamic objects.
- Branch and Bound (BaB) Algorithm
 - Using R-tree for NN queries
 - Bound: Min-dist, minmax-dist
 - Different searching manner:
 - Depth first (DF) traversing
 - Best first (BF) traversing

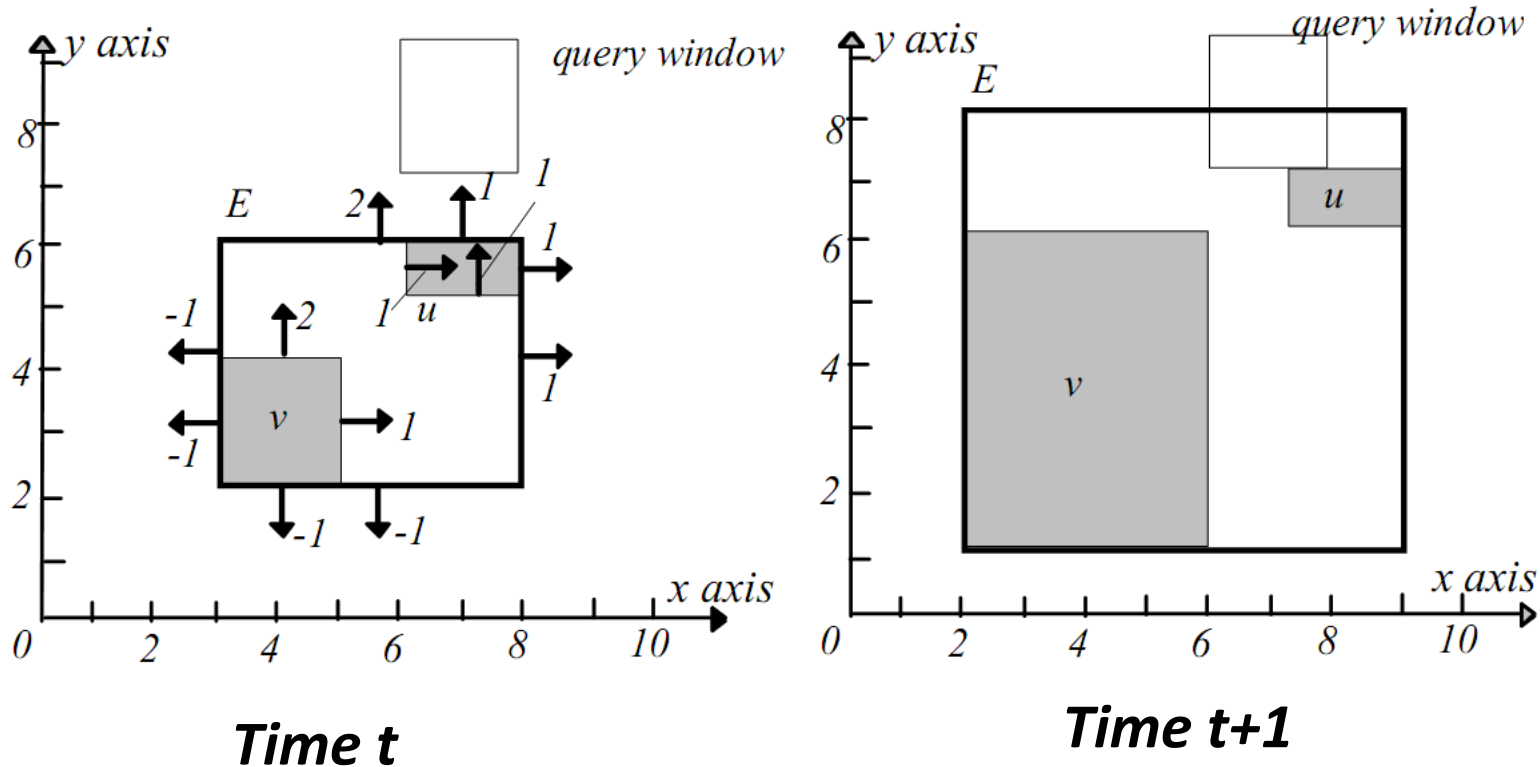


Time Parameterized tree

- Key Features:
 - A dynamic object is represented with an MBR that bounds its extents at the current time and each dynamic object has a velocity vector.
 - Future MBRs are not stored explicitly but computed based on current extents and velocity vectors
 - As in a traditional R-tree the extents are such that the MBR tightly encloses all entries in the node at current time.

Time Parameterized tree

- Different edge velocities will cause the object to grow or shrink with time.





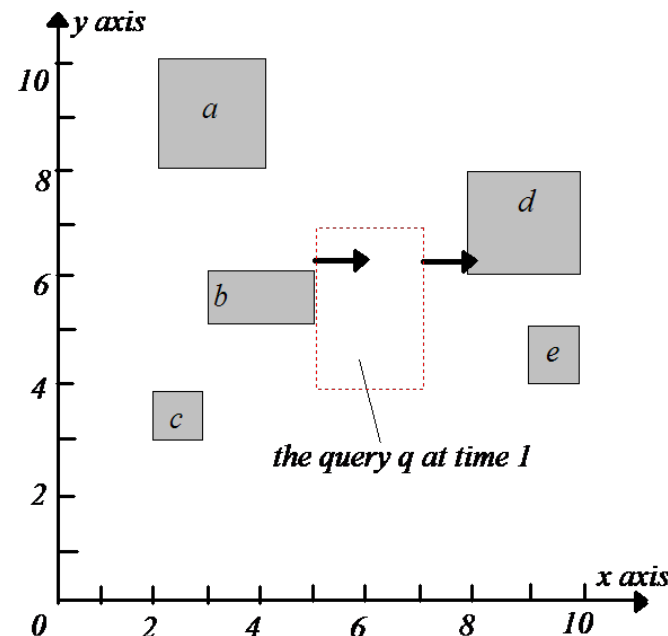
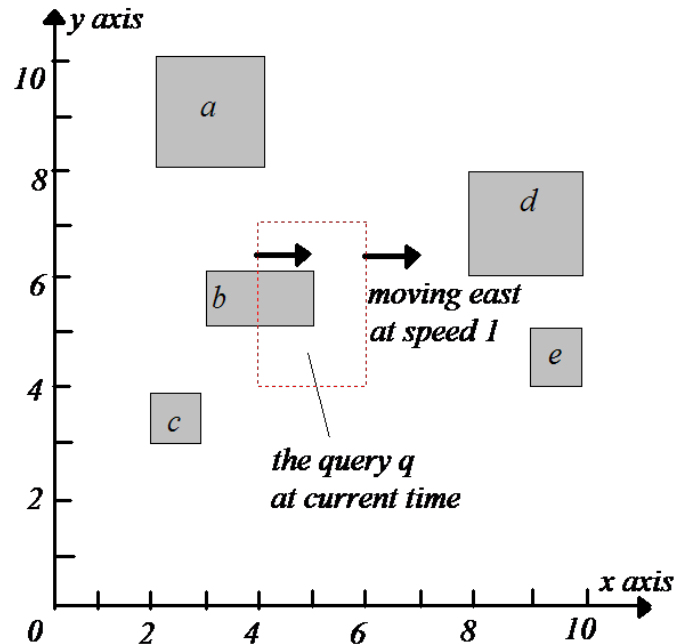
Time Parameterized(TP) queries

- Introduction
- TP Window Query
- TP K Nearest Neighbor Query
- TP Join Query



Time Parameterized(TP) queries

- Intro: Output form $\langle R, T, C \rangle$
 - R- set of objects satisfying the query (current result)
 - T- is the expiry time of the results
 - C- set of objects that will affect R at time T ($\langle T, C \rangle$ - TP component)

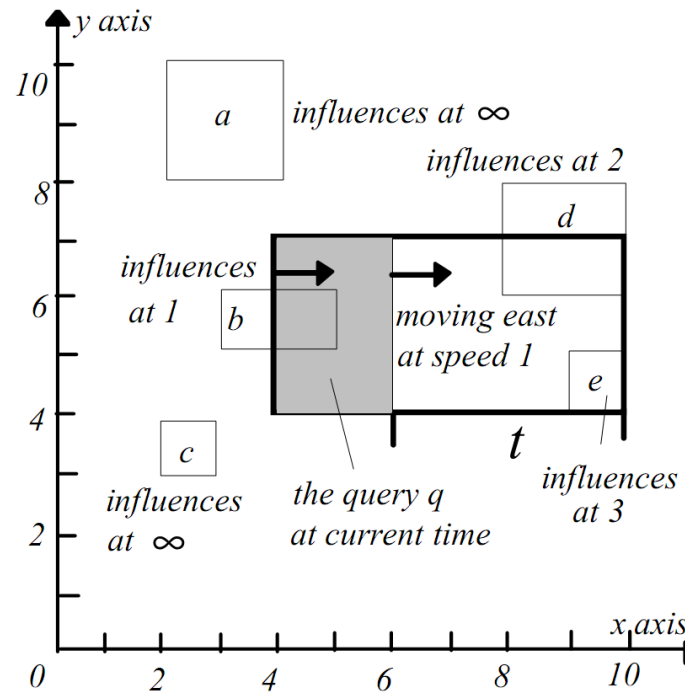


Result: $\langle \{b\}, 1, \{b\} \rangle$



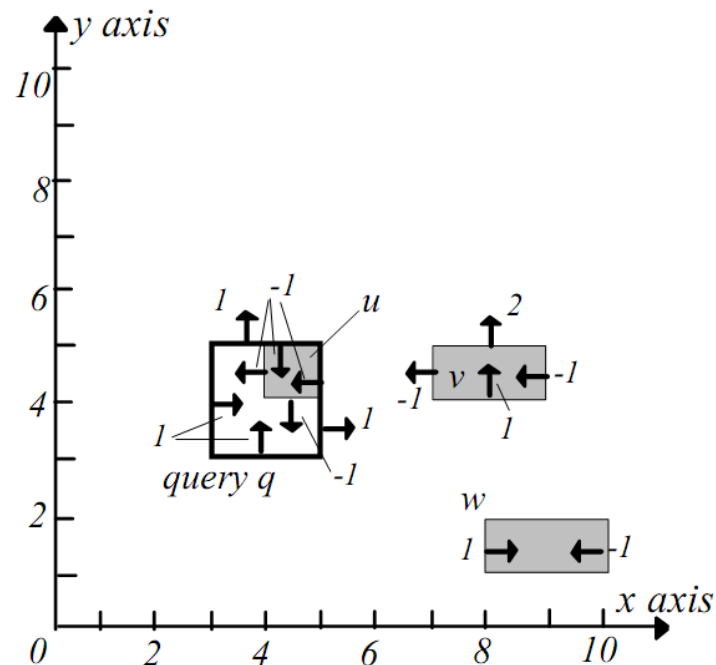
Time Parameterized(TP) queries

- Influence Time
 - Some objects influence the query at current time, but not in the future
 - Some objects are not currently in the result, but they may influence the query in the future.



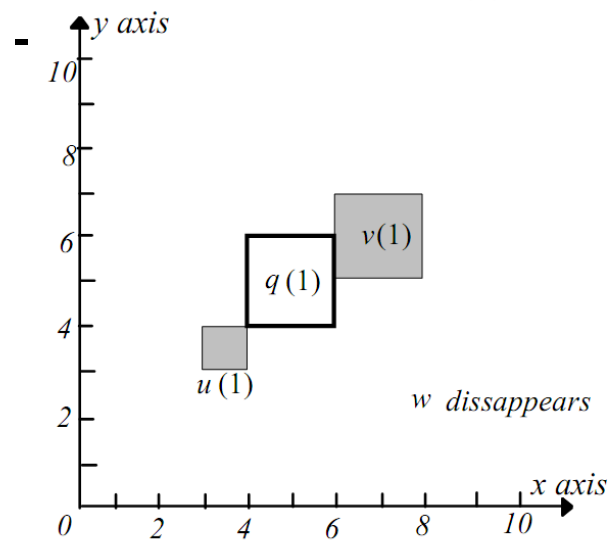
TP window query

- To find the influence time $T_{INF}(o,q)$ of an object o , with the query window q , we need the intersection period $[T_S, T_E)$ during which o will intersect q .
- At time 0:

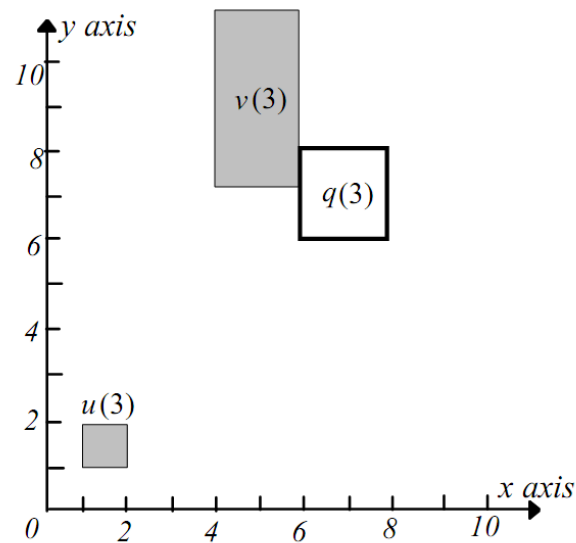


TP window query

- What is the intersection time for u, v, w ?
 - $u \rightarrow [0,1)$
 - $v \rightarrow [1,3)$



(b) At time 1



(c) At time 3

TP window query

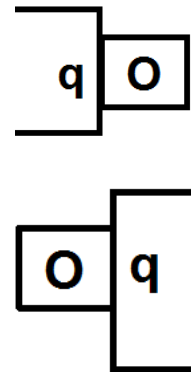
- For the i -th dimension:
 - Notation:
 - Object: MBR- $[o_{iL}, o_{iR}]$ Velocity- $[o.V_{iL}, o.V_{iR}]$
 - Query: MBR- $[q_{iL}, q_{iR}]$ Velocity- $[q.V_{iL}, q.V_{iR}]$
- Disappearance Time ($o.T_{iDSP}$)
 - $o.T_{iDSP} = (o_{iR} - o_{iL}) / (o.V_{iL} - o.V_{iR})$
 - The intersection between query and objects should happen before they disappear.
 - The influence time $T_{inf}(o, q)$ should be no later than
 - $\text{Min}(o.T_{DSP}, q.T_{DSP})$

TP window query

- Core task:
 - Find intersects $[T_s, T_e)$ between object and query
 - Object o and query intersects if and only if they intersect along all dimensions

Computation: (for dimension i)

- If objects do not intersect then:
 - T_{iLR} : Time the leftmost point of object meet the rightmost point of query
 - T_{iRL} : Time the rightmost point of object meet the leftmost point of query
 - $T_{is} = \min(T_{iLR}, T_{iRL})$
 - $T_{ie} = \min(\max(T_{iLR}, T_{iRL}), o.T_{DSP}, q.T_{DSP})$
- If objects already intersect then $T_{is} = 0$, and $T_{ie} = \min(T_{iLR}, T_{iRL}, o.T_{DSP}, q.T_{DSP})$
- $[T_s, T_e) = \cap [T_{is}, T_{ie})$





TP window query

- Important Notation:
 - Intersection Period: $[T_s, T_e)$: the time that object and query intersect
 - Influence time $T_{INF}(o, q)$:
 - If o does not currently intersect the query
 - $T_{INF}(o, q) = T_s$
 - If o is currently intersecting the query
 - $T_{INF}(o, q) = T_e$
 - The expiry time of the current result is the minimum influence time of all objects

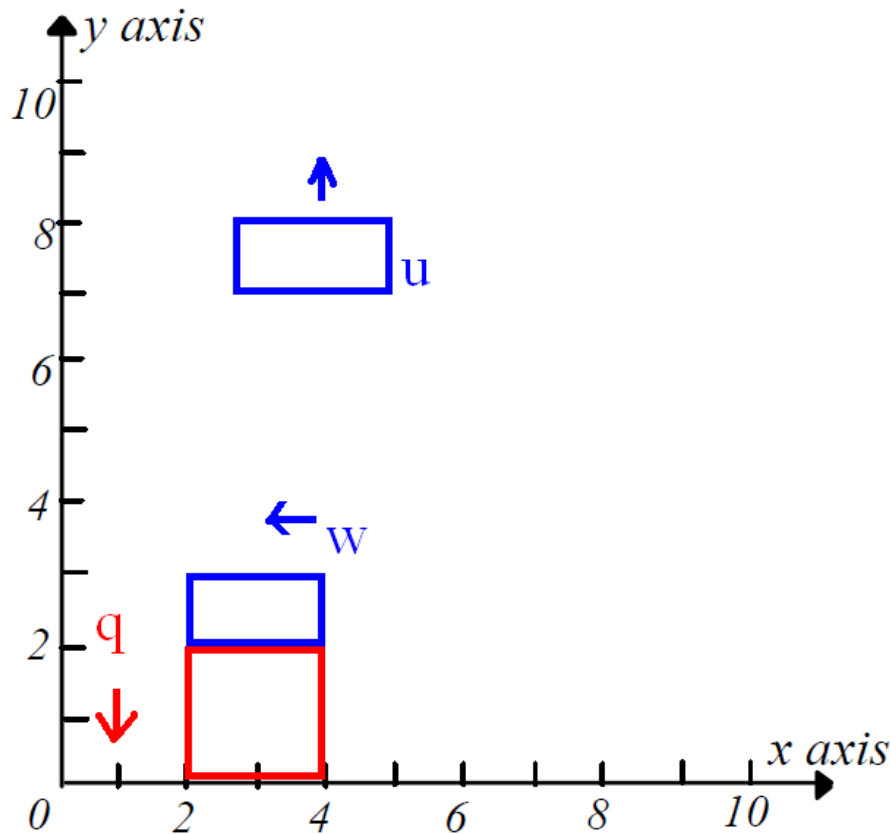


Window Query processing

- Both Best-FS and DFS can be used for processing TP queries.
- Algorithm:
 - Start with $R = \text{Null}$, $T = \text{infinity}$, $C = \text{Null}$
 - For each object o
 - If o satisfy q , then $R = R \cup \{o\}$
 - If $T_{\text{INF}}(o, q) < T$
 - $C = \{o\}$
 - $T = T_{\text{INF}}(o, q)$
 - Else if $T_{\text{INF}}(o, q) = T$
 - $C = C \cup \{o\}$

Example:

- Assuming their velocity is 1.



Intersect time:

$u: [0, 0.5)$

$v: [inf, inf)$

$w: [1, 3)$

Query Result for time
 $[0, 3)$?

$\langle \{u\}, 0.5, \{u\} \rangle$

$\langle \{\}, 1, \{w\} \rangle$

$\langle \{w\}, 3, \{w\} \rangle$

$t = 3$

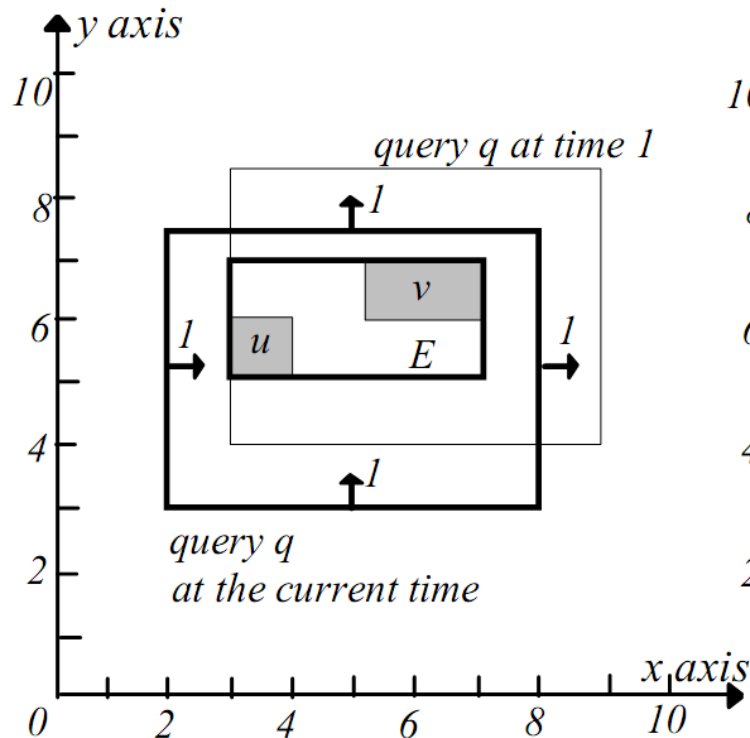


TP window query – Intermediate Node

- Influence time of the intermediate Entry E corresponds to the minimum possible influence time of any object in the subtree of E
- If E does not intersect with q , then $T_{INF}(E, q)$ is the time E starts intersecting with q – because this is also the earliest time when any of the objects inside E can intersect (influence) q

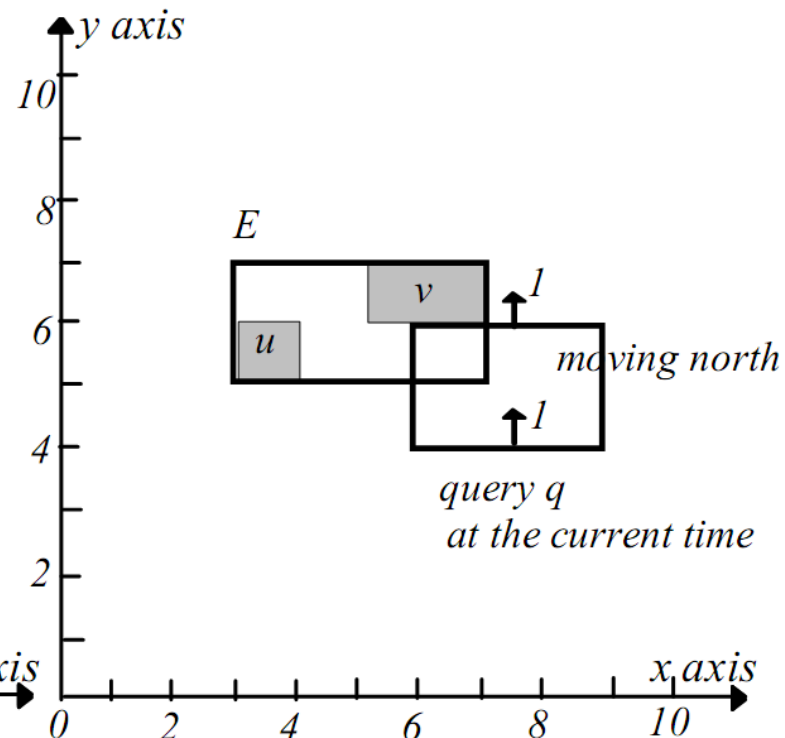
TP window query -- Intermediate Node

- If E intersects q , then two cases:



(a) E is contained in q

$T_{INF}(E,q)$ = the time that E starts to partially intersect q in the future



(b) E partially intersects q

$T_{INF}(E,q) = 0$, because some object inside E (e.g., v) may influence the result as soon as the query move

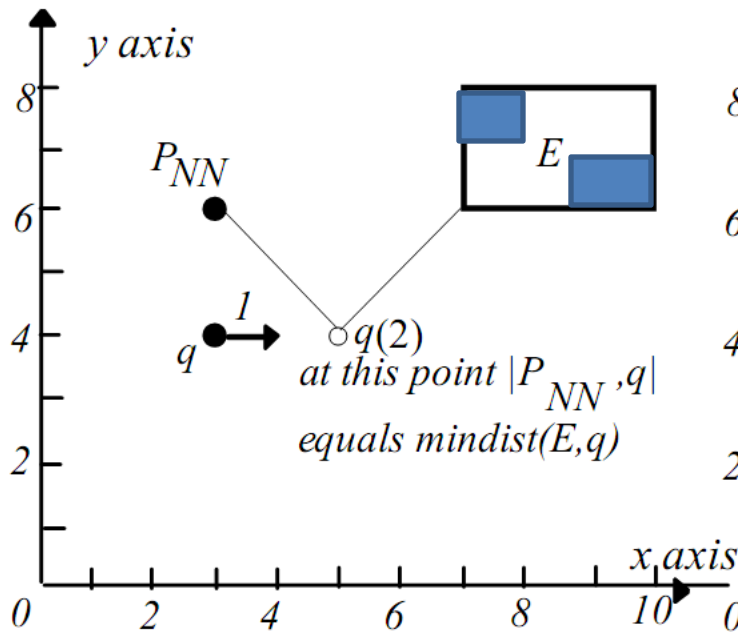


TP NN query

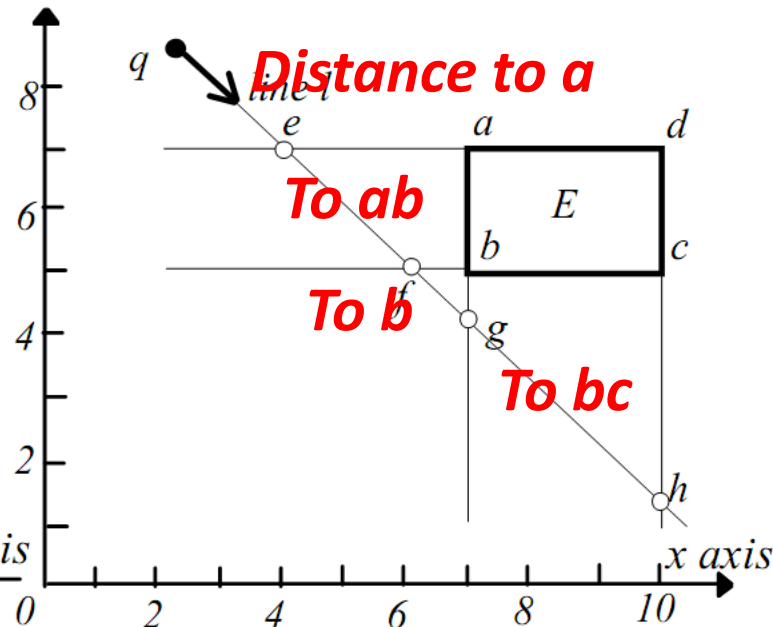
- Start with single NN
 - Influence time is no longer intersection period.
 - Instead, it is the time the object becomes the NN result.
- T_{INF} is the minimum t that satisfies:
 - $\text{Dist}(o(t), q(t)) < \text{Dist}(P_{NN}(t), q(t))$
 - P_{NN} is the current NN of q .

TP NN query

- Influence time of the Intermediate Entry E
 - Use mindist as the dist between q and E



(a) Example of $T_{\text{INF}}(E, q)$

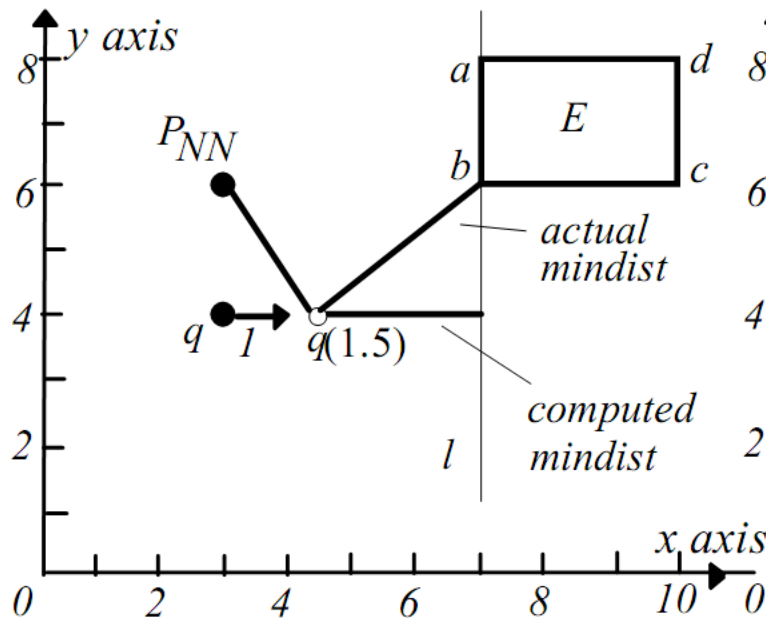


(b) Different cases of mindist

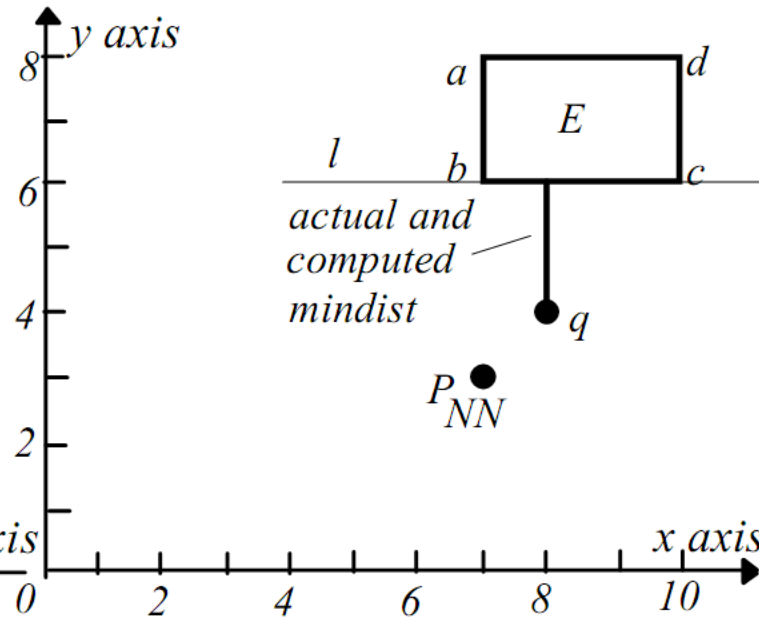
The computation of $\text{mindist}(E(t), q(t))$ depends on the relative positions of E and q .

TP NN query

- Simpler Calculation (underestimates *mindist* (to ensure the correctness of BaB algorithms)).



If mindist to a point, then distance to farther edge (ab) containing the point



If mindist to an edge, then mindist at current time

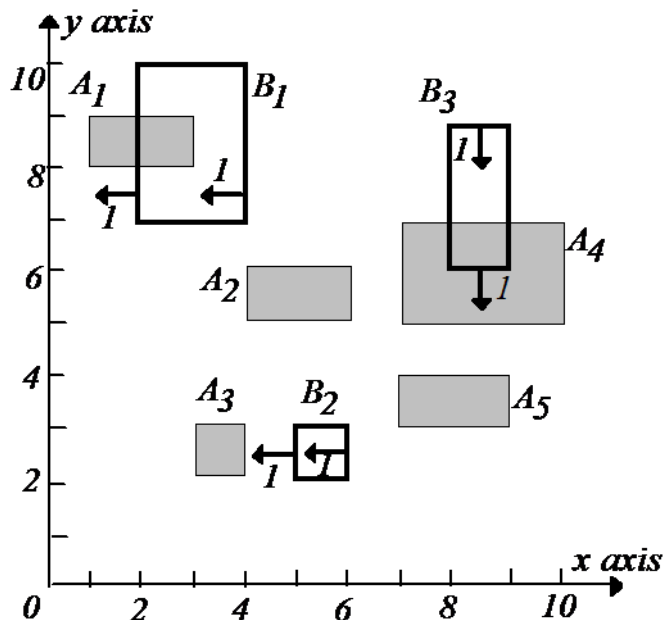


TP join query

- A Join query returns all pairs of objects from two datasets that satisfy some spatial condition. (e.g., Intersection)
- A join result can change in the future when,
 - A pair of objects in the current result ceases to satisfy the join condition
 - New objects start satisfying the join condition
- TP join can be regarded as a closest pair query by treating $T_{INF}(o1,o2)$, instead of $T_{INF}(q,o)$

TP join query

- Influence time of object pairs



	A_1	A_2	A_3	A_4	A_5
B_1	3	∞	∞	∞	∞
B_2	∞	∞	1	∞	∞
B_3	∞	∞	∞	4	2

- The expiry time is the minimum influence time (i.e., $T_{INF}(A_3, B_2) = 1$).



Conclusion

- Introduction of the novel concept of time-parameterized queries.
- Techniques for transforming the most common spatial queries to their TP counterparts.
- Development of efficient processing methods.

- Drawbacks?



References

- Tao, Y. & Papadias, D. [Time-parameterized queries in spatio-temporal databases](#) .
SIGMOD Conference, 2002, 334-345.A
presentation by Penny Pan in csci587 Fall'2010

Practice

- Queries / objects are aligned with integer coordinates
 - All arrows mean 1 grid/time velocity
- Question:
 - What is the Influence time of all objects?
 - What is the Query result Using TP strategy during time $[0,3)$?

