

# CONTINUOUS NEAREST NEIGHBOR SEARCH

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# OVERVIEW

- Introduction
- Preliminary & Related Work
- Continuous k-Nearest Neighbor Query(CkNN)
  - Definition
  - Problem Characteristics
  - R-tree algorithm
  - Query analysis
  - Complex CNN extension
- Experiments
- Discussion and Conclusion

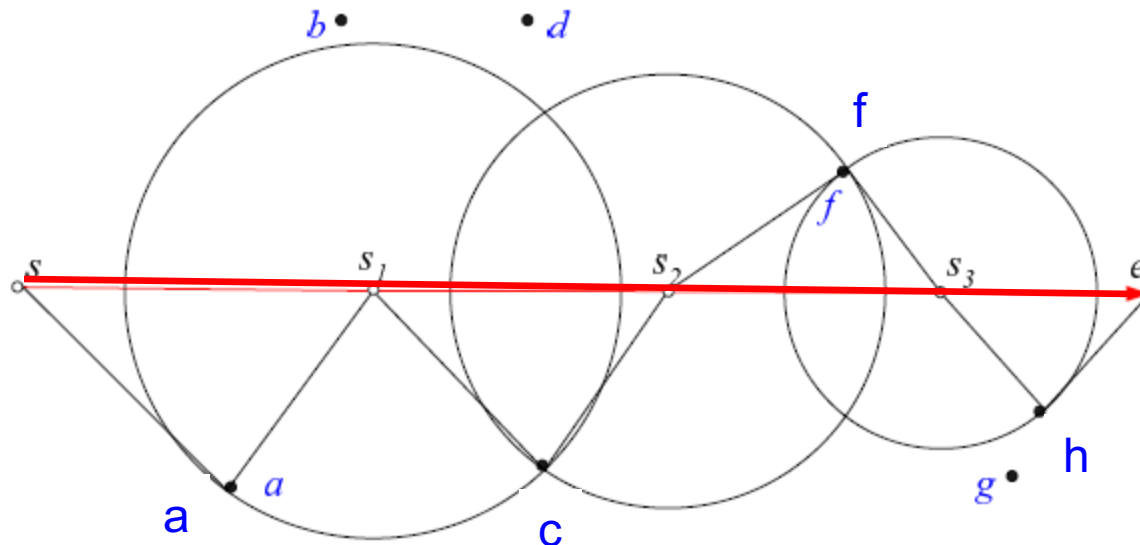
# INTRODUCTION

- Continuous Nearest Neighbor



- Why called "continuous"?
  - Nearest neighbor of every points in the trajectory

# PRELIMINARY -- CONTINUOUS NEAREST NEIGHBOR

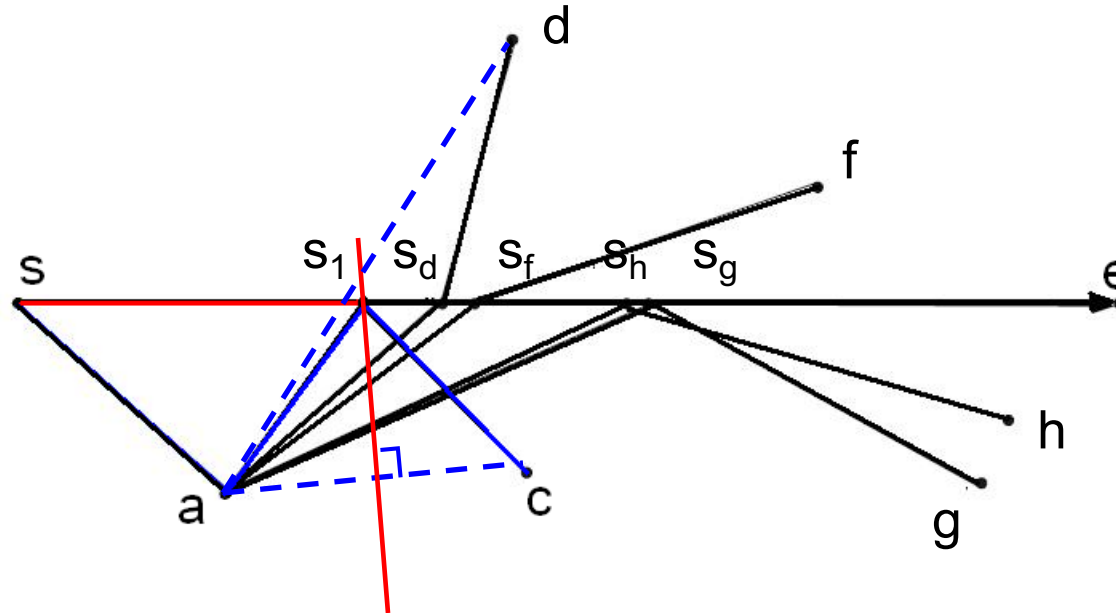


- Data: A set of points ( $P=\{a,b,c,d,f,g,h\}$ )
- Query: A line segment  $q=[s, e]$
- Result: The nearest neighbor (NN) of every point on  $q$ .
- Result representation:  $\{ \langle a, [s, s_1] \rangle, \langle c, [s_1, s_2] \rangle, \langle f, [s_2, s_3] \rangle, \langle h, [s_3, e] \rangle \}$

# RELATED WORK – SAMPLING

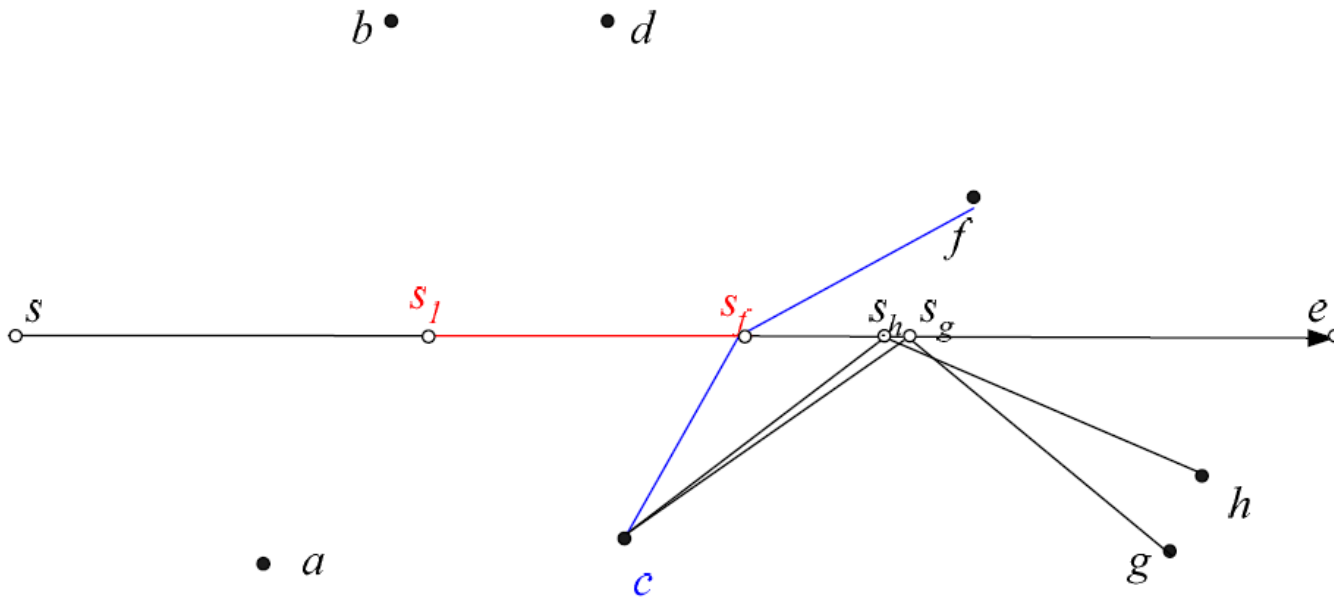
- Try to convert the continuous-NN to point-NN
  - Every point on the line -> unlimited points
  - Sampling
- Drawback:
  - Sample Rate: low -> incorrect
  - Sample Rate: high -> overhead (still cannot guarantee accuracy)

# RELATED WORK – TP NN (CONT.)



- Step 1: Find the NN of the start point  $s$ , i.e., point  $a$ .
- Step 2: Use the TP technique: find the first point on the line segment ( $s_1$ ) where there is a change in the NN (i.e., point  $c$ ) will become the next NN – result:  $\langle a, [s, s_1], c \rangle$
- Can be thought as conventional NN query, where the goal is to find the point  $x$  with the minimum  $dist(s, sx)$

# RELATED WORK – TP NN (CONT.)



- Step 3: Perform another TP NN to find:
- Starting from  $s_1$ , the smallest distance we need to travel for the current NN (i.e.,  $c$ ) to change
- Repeat this until we finish the entire segment.

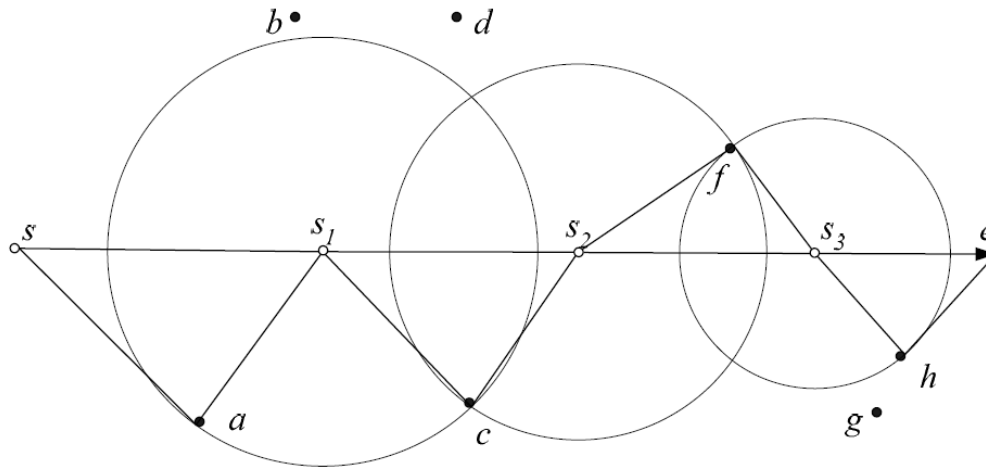


# RELATED WORK – TP NN (CONT.)

- Not only NN, but support k-NN
- Still overhead:  $n$  (= split points) times NN queries, multiple scans of database



# CKNN - DEFINITION



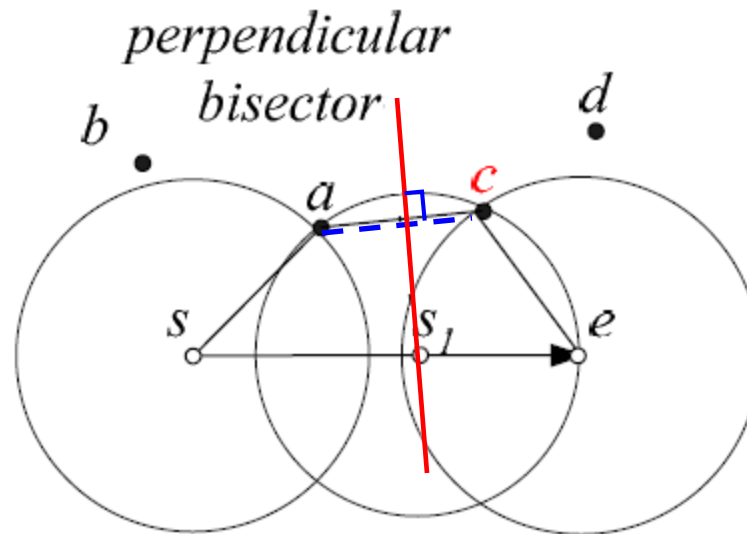
- Goal: Find all **split points** (as well as the corresponding NN for each segment) with a **single** traversal.
- **Split List (SL)**: The set of split points (including  $s$  and  $e$ ).
- Each split point  $s_i \in \text{SL}$  and all points in  $[s_i, s_{i+1}]$  have the same NN, denoted as  $s_i.\text{NN}$  (e.g.,  $s_1.\text{NN}$  is  $c$ , which is also the NN for all points in interval  $[s_1, s_2]$ )
- $s_i.\text{NN}$  (e.g.,  $c$ ) **covers** point  $s_i$  ( $s_1$ ) and interval  $[s_i, s_{i+1}]$  ( $[s_1, s_2]$ ).
- **Vicinity Circle (VC)**: The circle that centers at split point  $s_i$  with radius  $\text{dist}(s_i, s_i.\text{NN})$

# CKNN – PROBLEM CHARACTERISTICS

- Lemma 1: Given a split list  $SL \{s_0, s_1, \dots, s_{|SL|-1}\}$ , and a new data point  $p$ , then:  $p$  covers some point on query segment  $q$  **if and only if**  $p$  covers a split point.

Analyzing the first data point “a” (in alphabetical order)

Analyzing “b”: not in VC of  $s$  and  $e$ , hence no point on  $[s,e]$  closer to  $b$  than  $a$



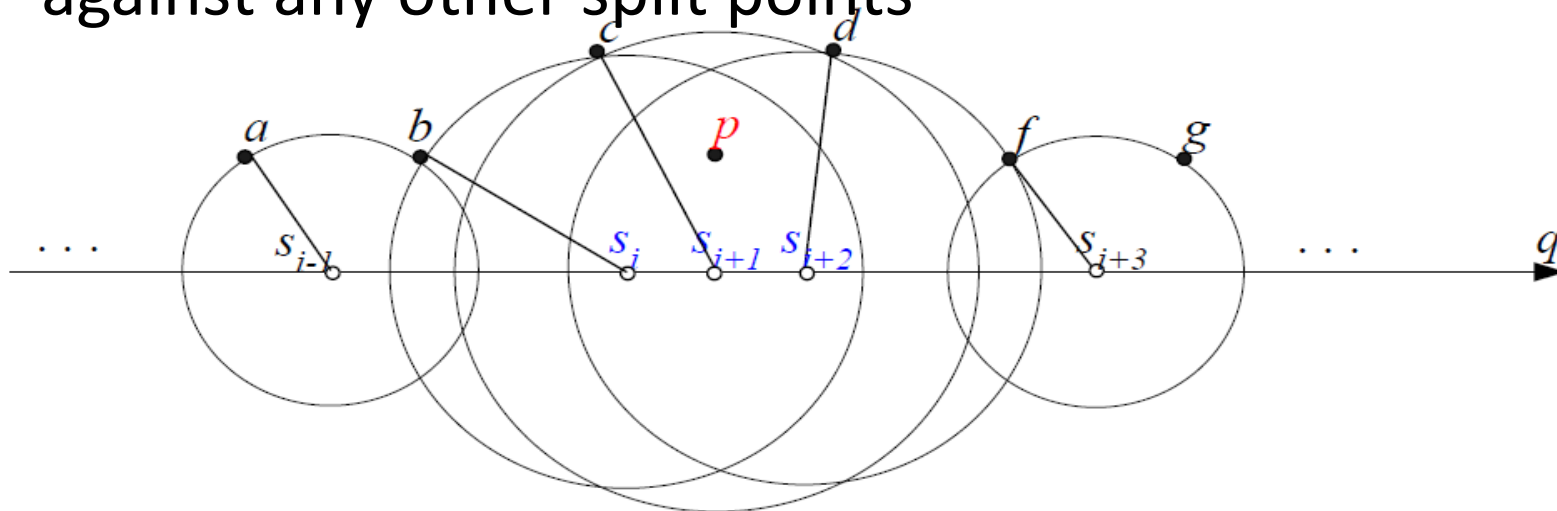
Analyzing “c”: in VC of  $e$ , hence: Creating a new split point...

$d$  not in any VC (note that it was in VC of  $e$  before adding  $c$ )

Result:  $\{ \langle a, [s, s_1] \rangle, \langle c, [s_1, e] \rangle \}$

# CKNN - PROBLEM CHARACTERISTICS

- Lemma 2: (Covering Continuity)
  - The split points covered by a point  $p$  are continuous.
  - Namely, if  $p$  covers split point  $s_i$  but not  $s_{i-1}$  (or  $s_{i+1}$ ), then  $p$  cannot cover  $s_{i-j}$  (or  $s_{i+j}$ ) for any value of  $j > 1$ .
  - Below:  $p$  cover  $s_i, s_{i+1}$  and  $s_{i+2}$  ( $p$  falls in their vicinity circles), but not  $s_{i-1}, s_{i+3}$ , so no need to check  $p$  against any other split points

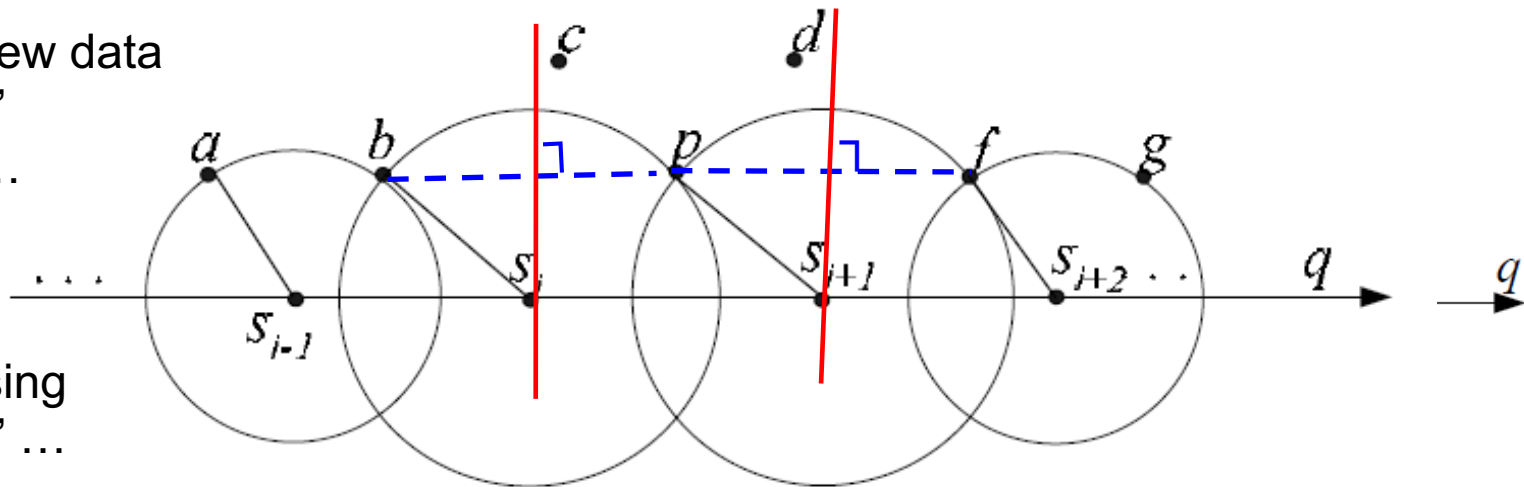


$$SL = \{s_{i-1} (.NN=a), s_i (.NN=b), s_{i+1} (.NN=c), s_{i+2} (.NN=d), s_{i+3} (.NN=f)\}$$

# CKNN - PROBLEM CHARACTERISTICS

- Finding new split points for b (after adding p):
  - b covers  $S_{i-1}$  and f covers  $S_{i+2}$ ; so we need to find the space that NN changes from b to p and then to f

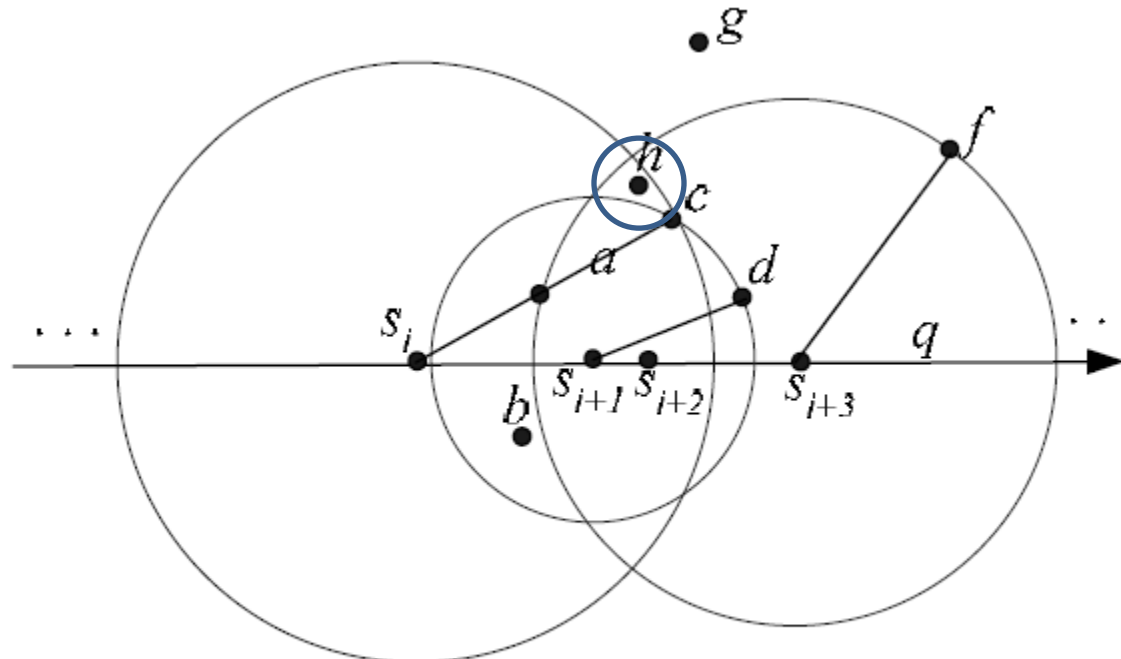
When new data point "p" arrives...



$$SL = \{s_{i-1} (.NN=a), s_i (.NN=b), s_{i+1} (.NN=p), s_{i+2} (.NN=f)\}$$
$$SL = \{s_{i-1} (.NN=a), s_i (.NN=b), s_{i+1} (.NN=c), s_{i+2} (.NN=d), s_{i+3} (.NN=f)\}$$

# CKNN - PROBLEM CHARACTERISTICS

- How about the k-NN?
- Lemma 1 : Fit || Lemma 2 : Cannot Fit
- Eg:
  - K=3



h covers  
 $s_i, s_{i+3}$   
But not  
 $s_{i+1}, s_{i+2}$

$$SL = \{s_i(.NN_{1-3} = a, b, c), s_{i+1}(.NN_{1-3} = a, b, d), s_{i+2}(.NN_{1-3} = a, c, d), s_{i+3}(.NN_{1-3} = c, d, f)\}$$

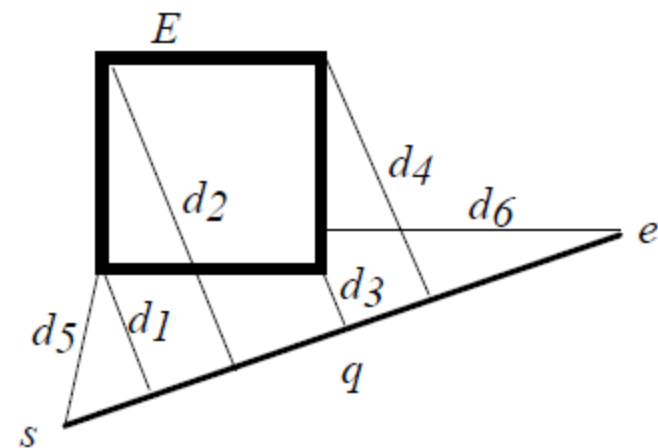
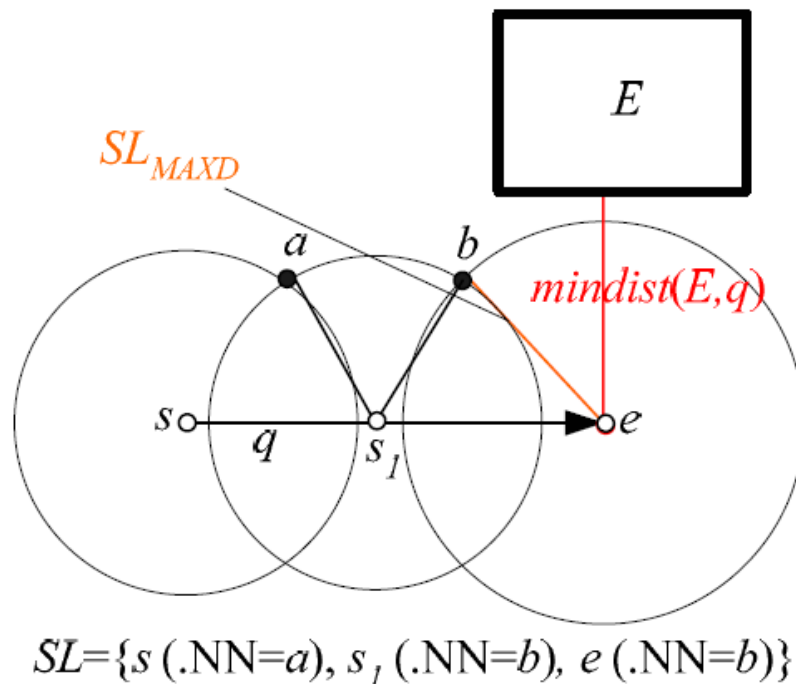


# CKNN – R-TREE ALGORITHM

- General key notes:
  - Use branch-and-bound techniques to prune the search space.
  - R-tree traverse principle:
    - When a leaf entry (i.e., a data point)  $p$  is encountered, SL is updated if  $p$  covers any split point (i.e.,  $p$  is a qualifying entry) – By Lemma 1.
    - For an intermediate entry, We visit its subtree only if it may contain any qualifying data point – Use heuristics.
  - Avoid accessing non qualified nodes

# R-TREE ALGORITHM – HEURISTIC 1

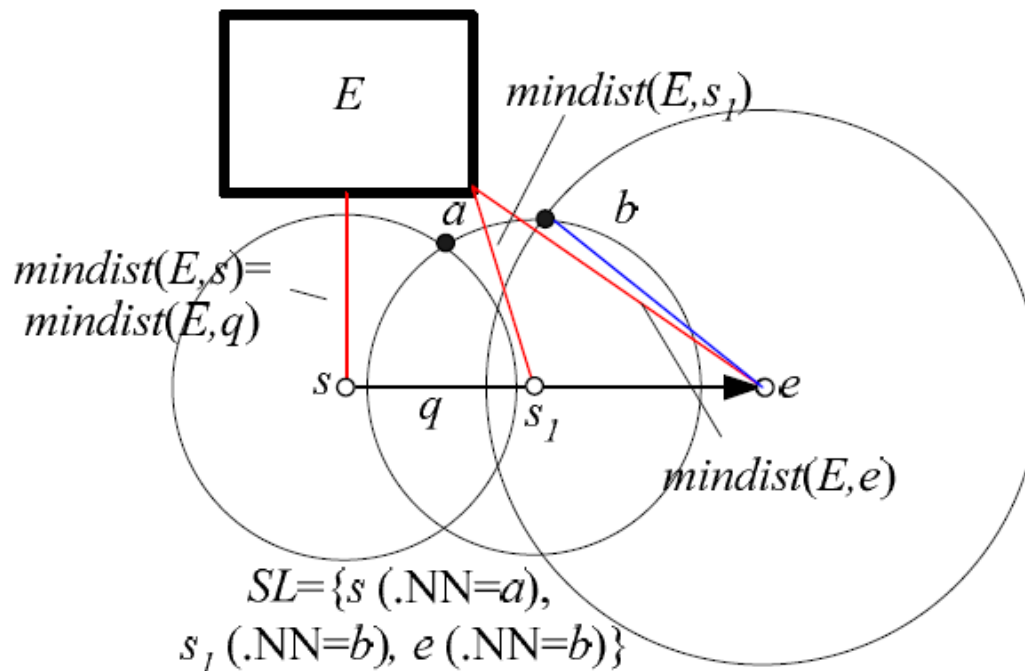
- Given an intermediate entry  $E$  and query segment  $q$ , the sub-tree of  $E$  may contain qualifying points only if  $\text{mindist}(E, q) < SL_{\text{MAXD}}$ , where  $SL_{\text{MAXD}}$  is the maximum distance between a split point and its NN.



Compute Mindist( $E, q$ )

# R-TREE ALGORITHM – HEURISTIC 2 (AFTER 1)

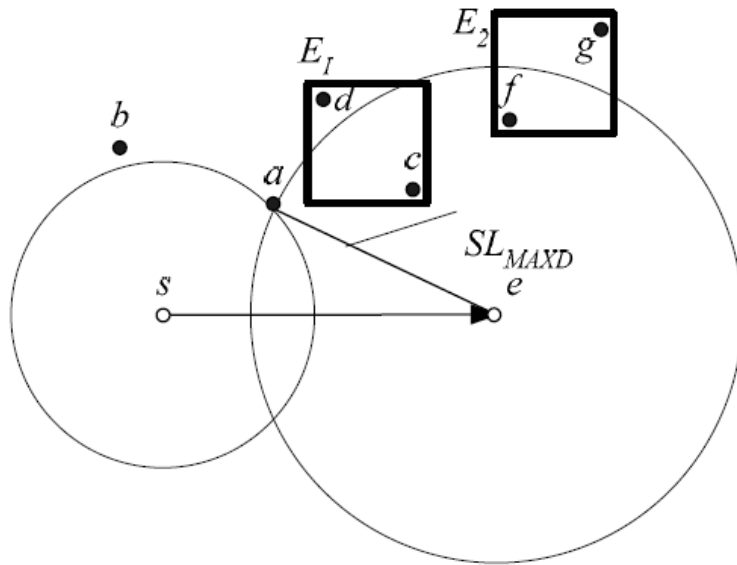
- Given an intermediate entry  $E$  and query segment  $q$ , the subtree of  $E$  must be searched **if and only if** there exists a split point  $s_i \in SL$  such that  $\text{dist}(s_i, s_i.NN) > \text{mindist}(s_i, E)$ .





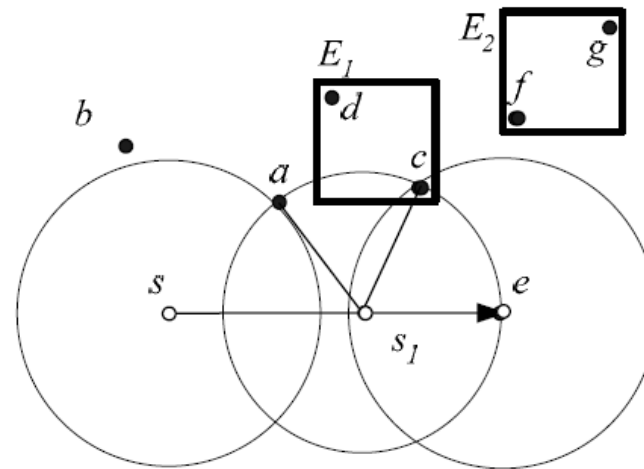
# R-TREE ALGORITHM – HEURISTIC 3 (ORDER)

- Entries (satisfying heuristics 1 and 2) are accessed in increasing order of their minimum distances to the query segment  $q$ .



$$SL = \{s (.NN=a), e (.NN=a)\}$$

Before processing  $E_1$

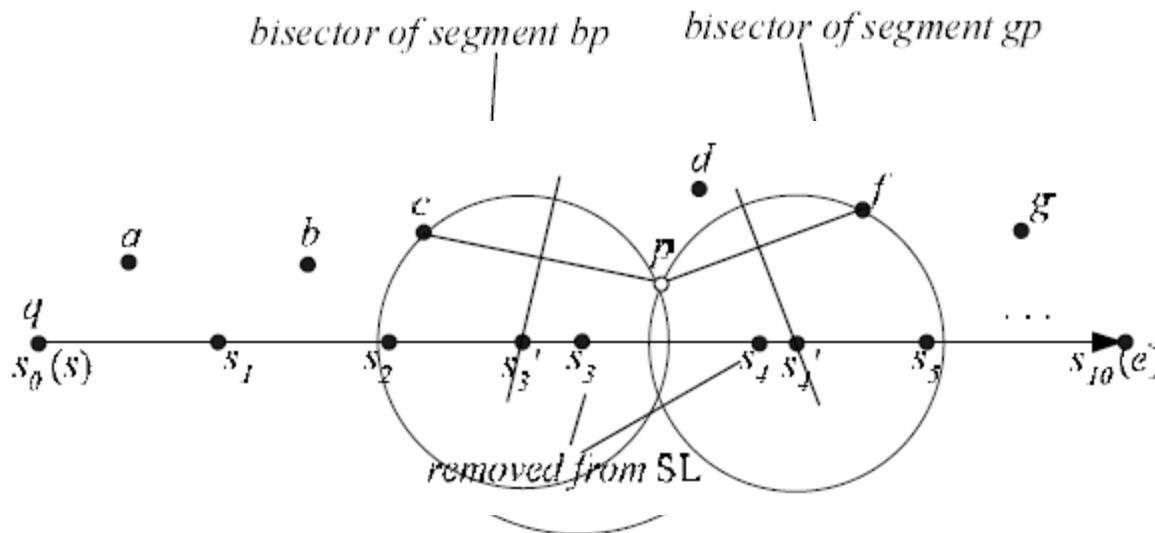


$$SL = \{s (.NN=a), s_1 (.NN=c), e (.NN=c)\}$$

After processing  $E_1$

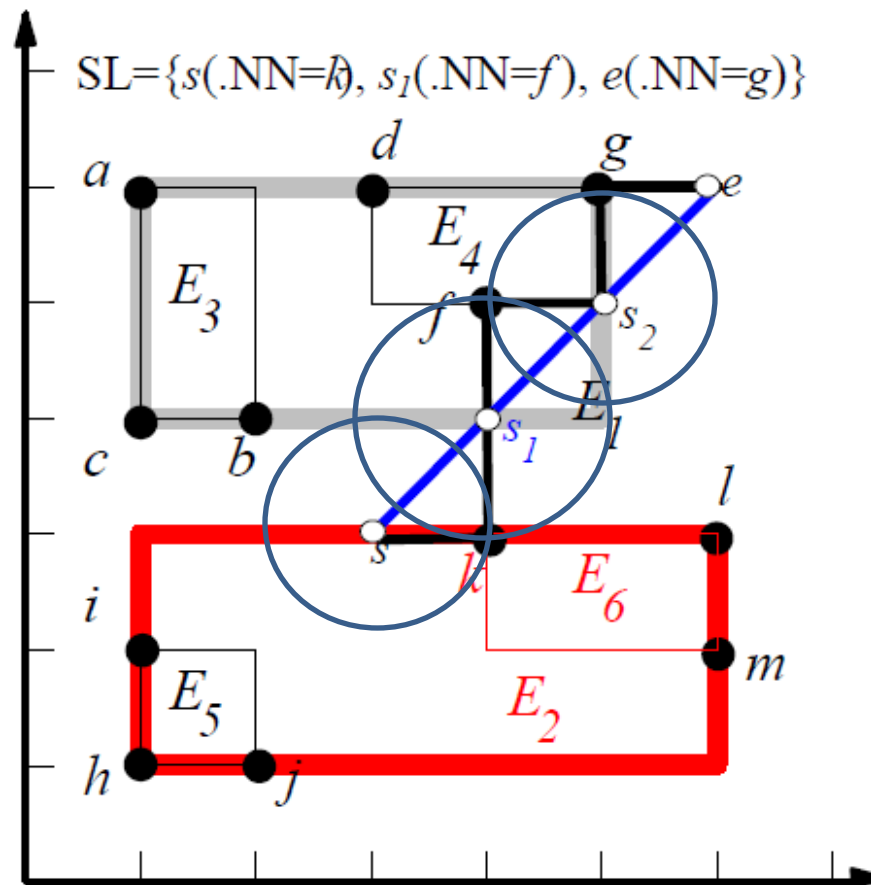
# R-TREE ALGORITHM – LEAF ENTRY

- Input: New entry  $p$ ,  $SL = \{s_1, \dots, s_{10}\}$ 
  - 1) retrieve the split points covered by  $p$
  - 2) update SL
- Binary search: 1)  $[s_0, s_{10}] \rightarrow s_5$  2)  $[s_0, s_5] \rightarrow s_2$ 
  - Using bisector to judge the direction



# CKNN – R-TREE ALGORITHM (EXAMPLE)

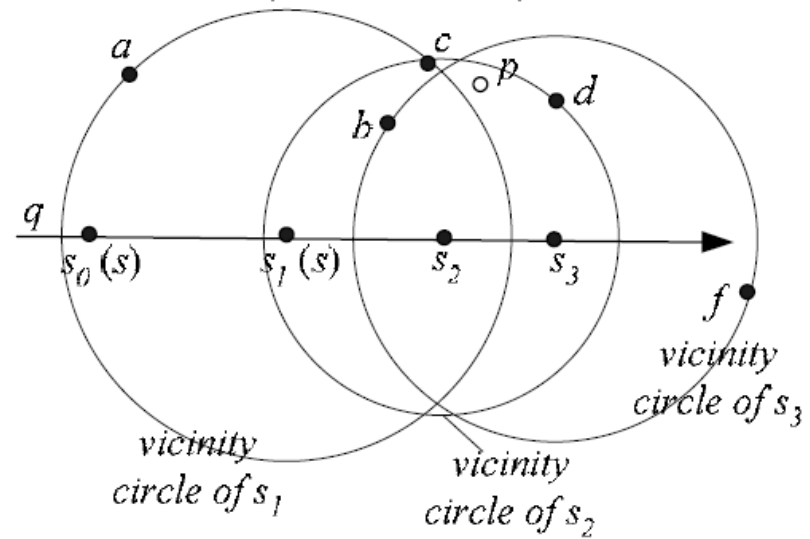
- Depth First (query segment:  $se$ )



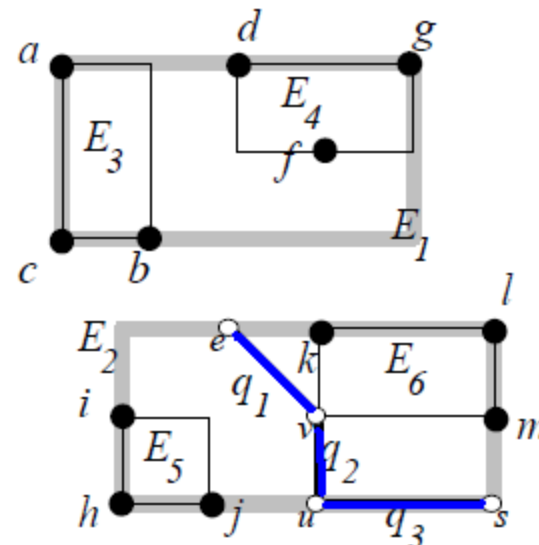
# OTHER CNN QUERY

- kCNN query (k=2)

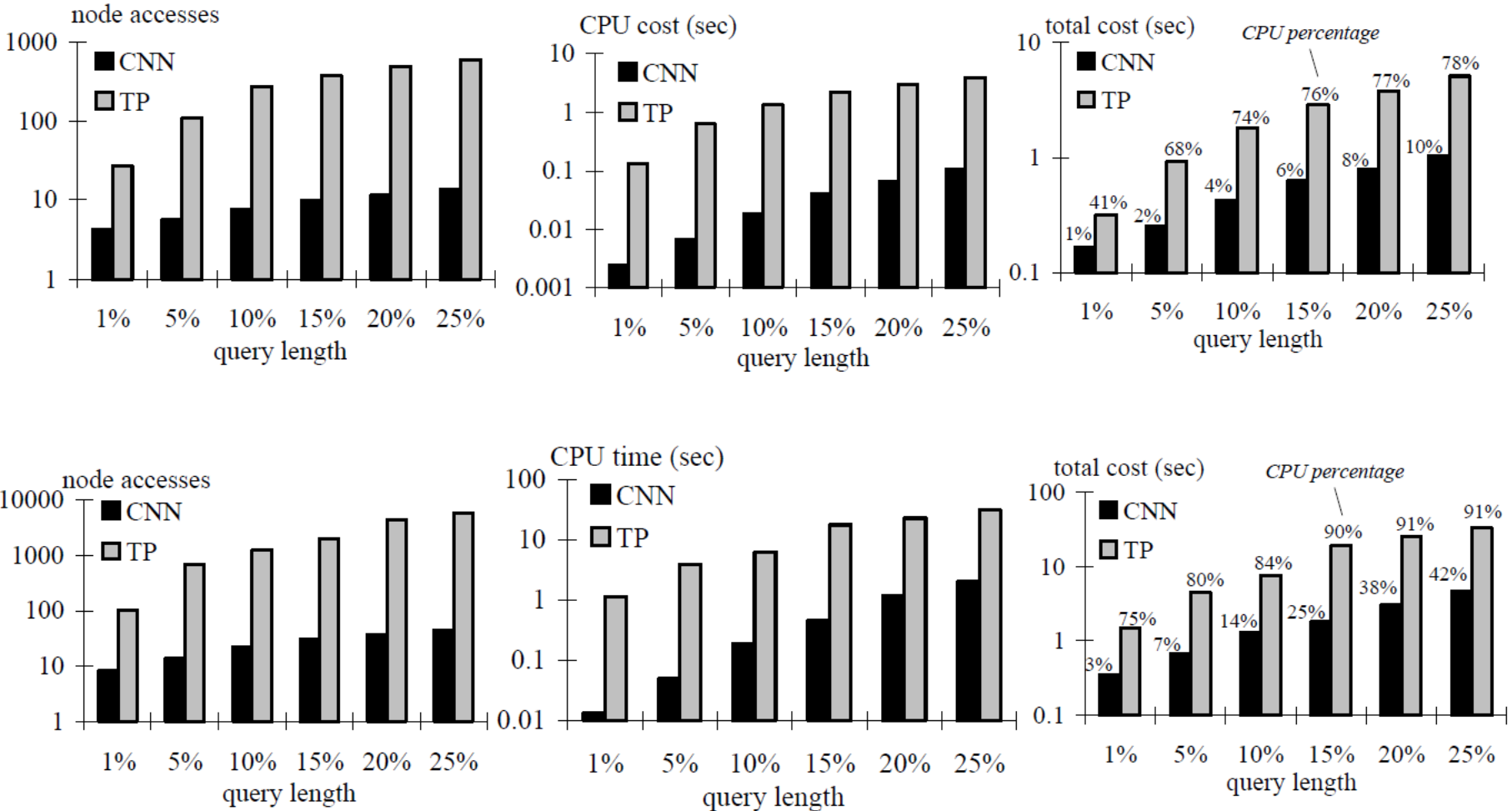
$$SL = \{s_0(.NN_{1,2}=a,b), s_1(.NN_{1,2}=b,c), s_2(.NN_{1,2}=b,d), s_3(.NN_{1,2}=d,f)\}$$



- Trajectory NN query (TNN)
  - $q_1 = [s,u]$
  - $q_2 = [u,v]$
  - $q_3 = [v,e]$
  - Each segment has a SL
  - Treated one by one



# EXP: PERFORMANCE VS QUERY LENGTH



# DISCUSSION AND CONCLUSION

- A fast algorithm for *C-kNN query*.
- Future work:
  - Rectangle data
  - Moving data points
  - Application to road networks (i.e., travel instead of Euclidean distance)



# References

- Tao, Y.; Papadias, D. & Shen, Q. Continuous Nearest Neighbor Search. VLDB, 2002, 287-298.
- A presentation by Penny Pan in csci587 Fall'2010

# Sample question

