

Edge-Based Locality Sensitive Hashing for Efficient Geo-Fencing Applications

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Outline

- ❖ Application and motivation of geo-fencing
 - Pairing a point with polygon: INSIDE/WITHIN
 - Crossing number algorithm and its scalability problem
- ❖ Proposed edge-based LSH algorithm
 - R-tree for pre-filtering
 - LSH for INSIDE, plus probing for WITHIN
 - Simple but effective and efficient
- ❖ Experimental results
- ❖ Conclusions

Motivation & Concepts: INSIDE

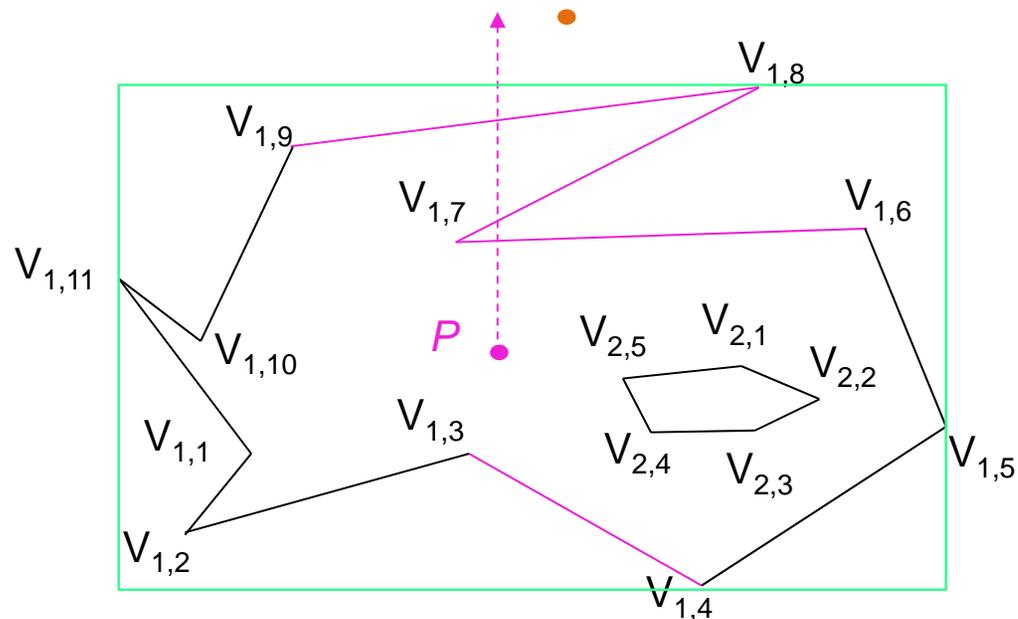
❖ Basis: Well-known crossing number algorithm

- Inside **iff** number of intersections == odd
- Requires checking each edge \Rightarrow **inefficient**

❖ Enhancements:

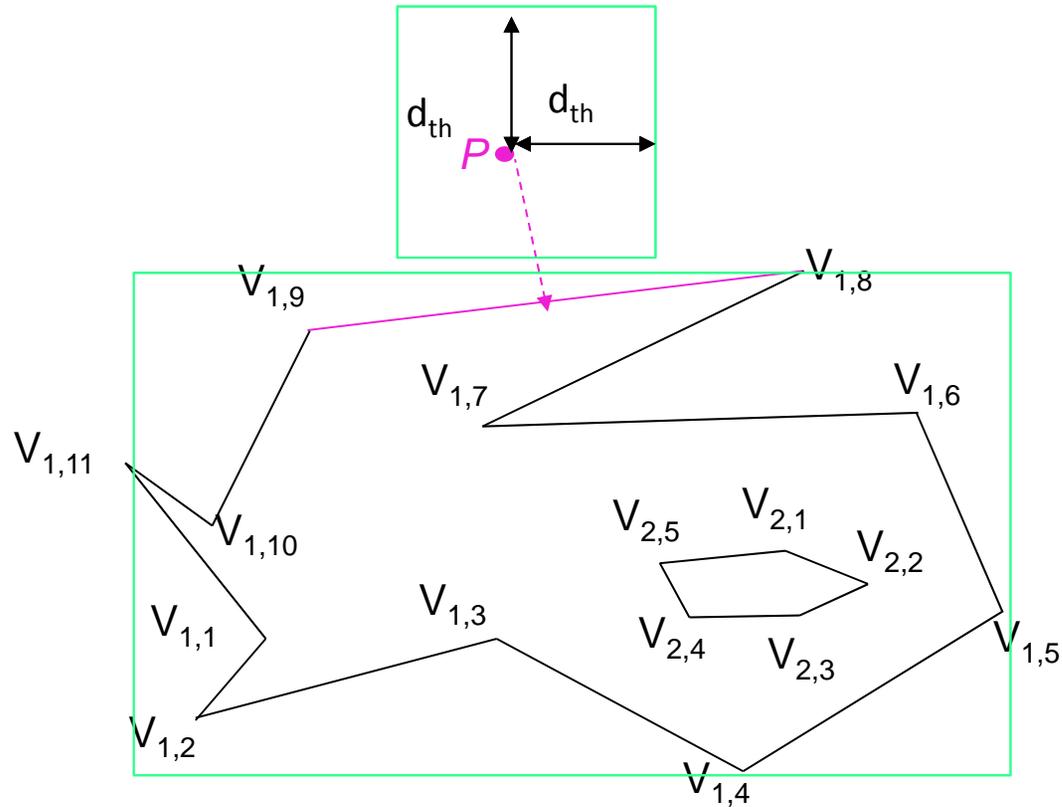
- 1) Exploiting MBR for pre-filtering
- 2) Locality-sensitive hashing (LSH) for further acceleration

MBR: minimum bounding rectangle

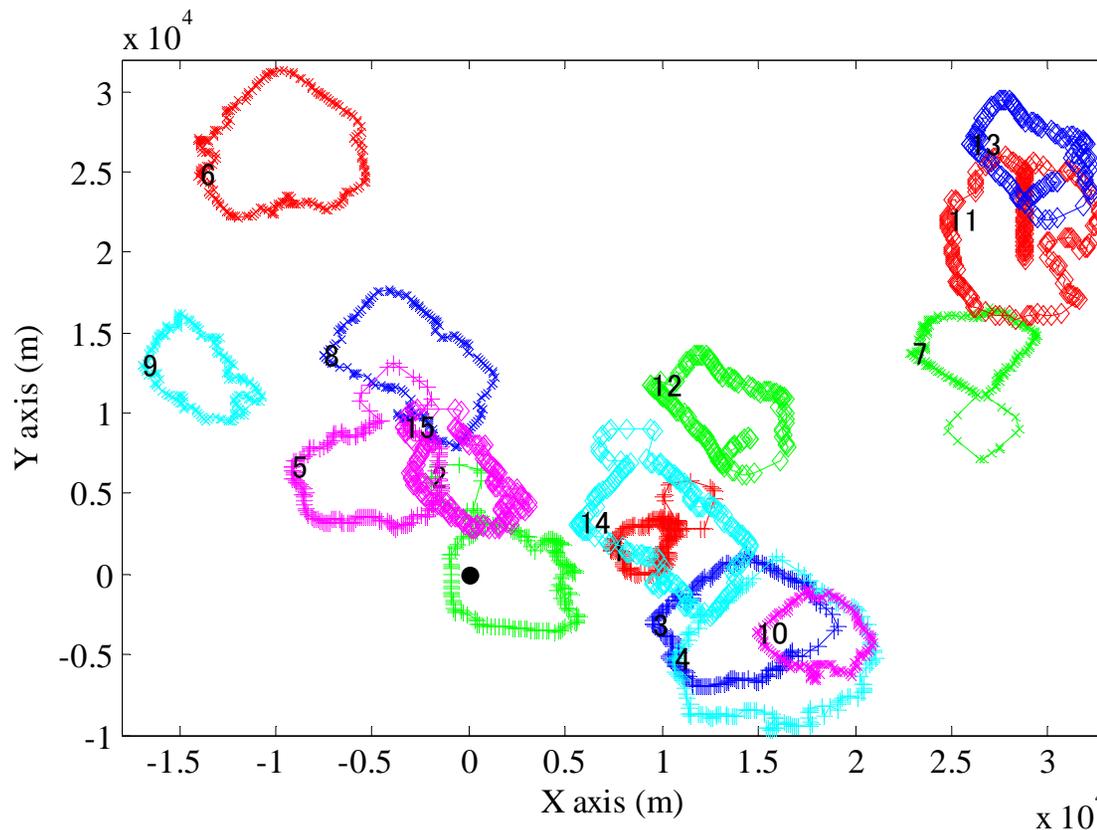


Motivation & Concepts: WITHIN

- ❖ Case: Point P outside of MBR but within a distance of d_{th}
 - A rectangle centered at P , edge length being two times d_{th}
 - If **no overlap** \Rightarrow point surely not WITHIN distance



Scalability: Polygons, Points, Edges



edges of 15 polygons

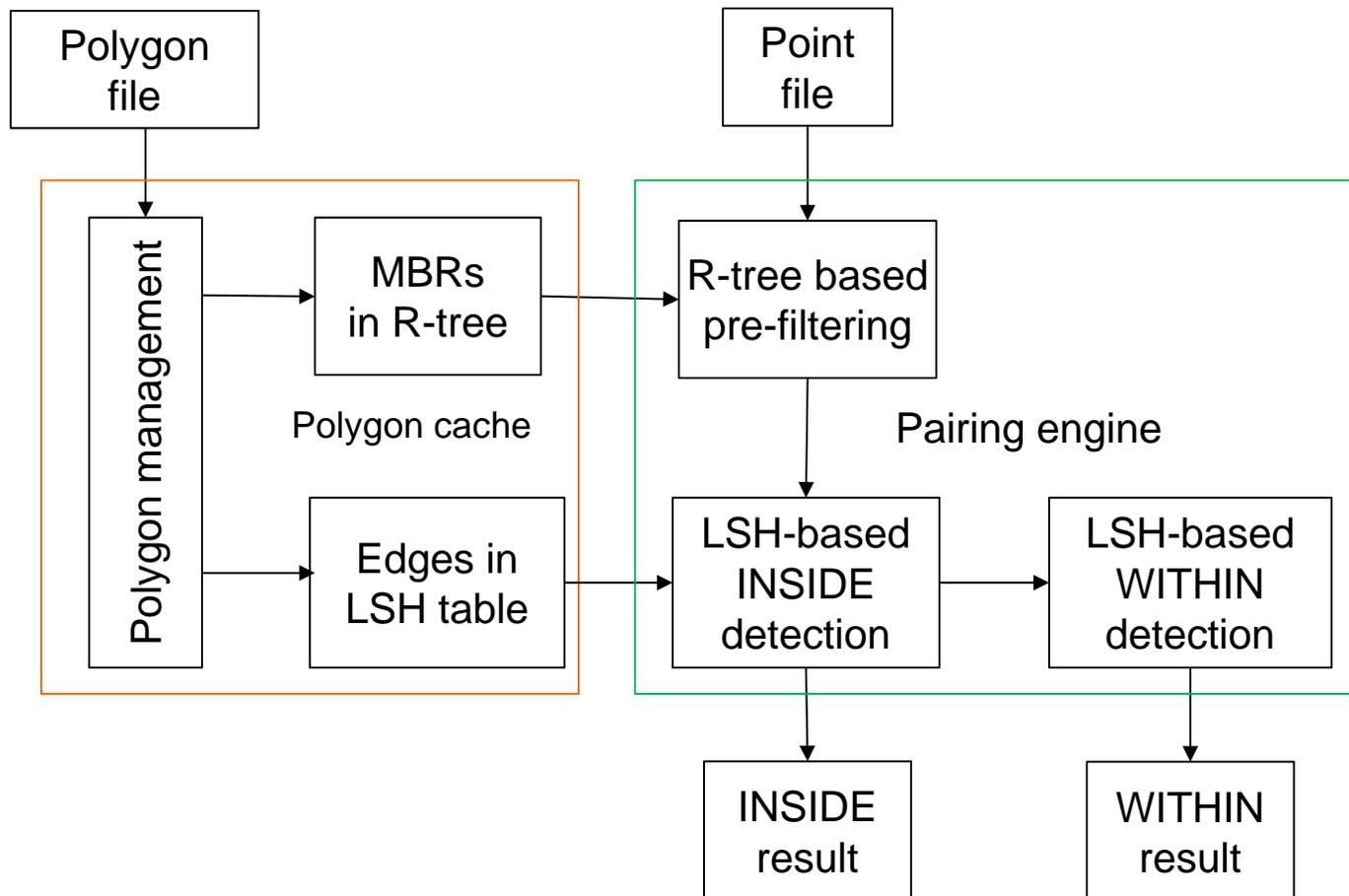
1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
285	255	235	196	264	250	240	239	226	226	242	153, 15	152, 20	250	217

⇒ Problem: scalability with the number of edges

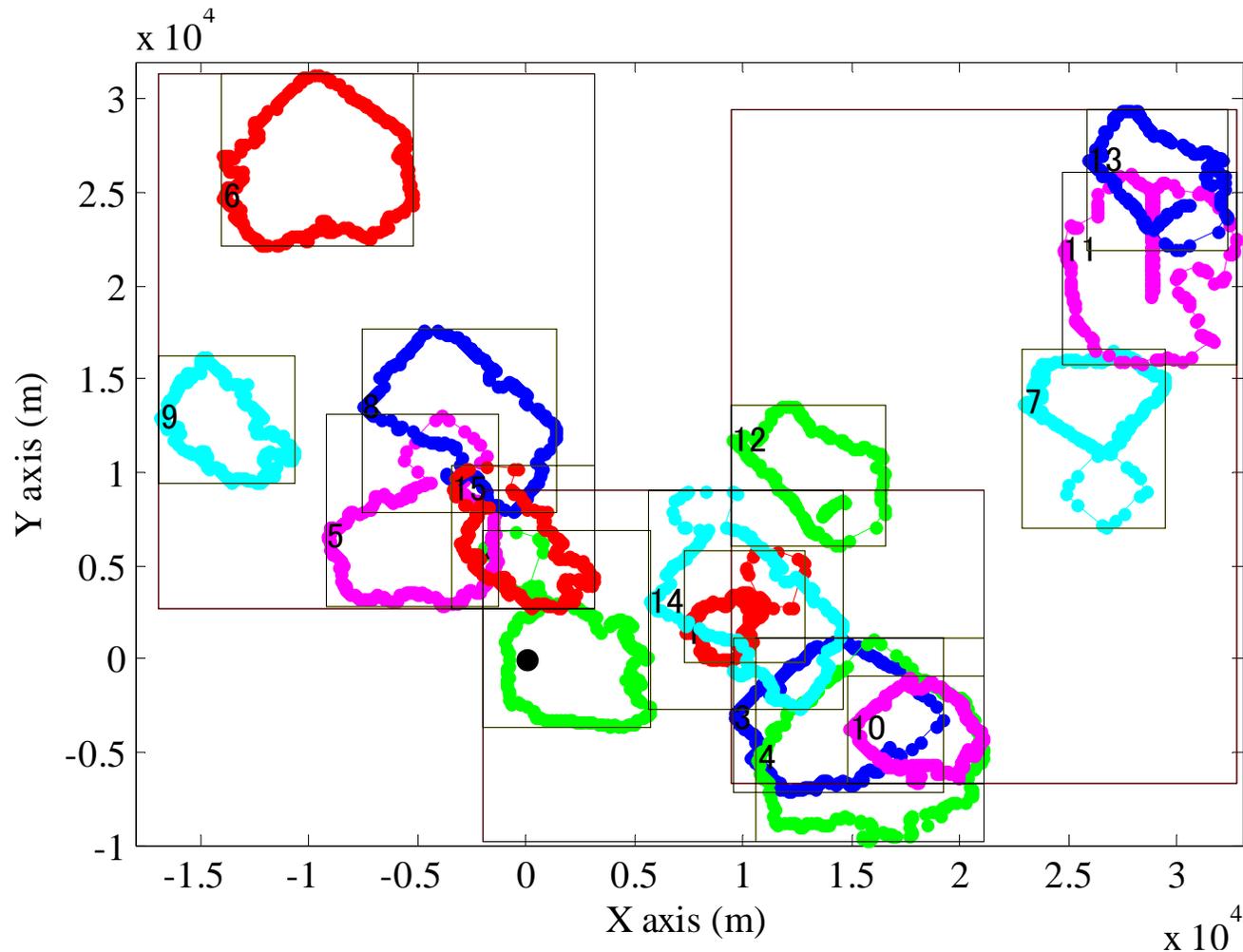
Efficient Geo-Fencing: Framework

❖ Two stages

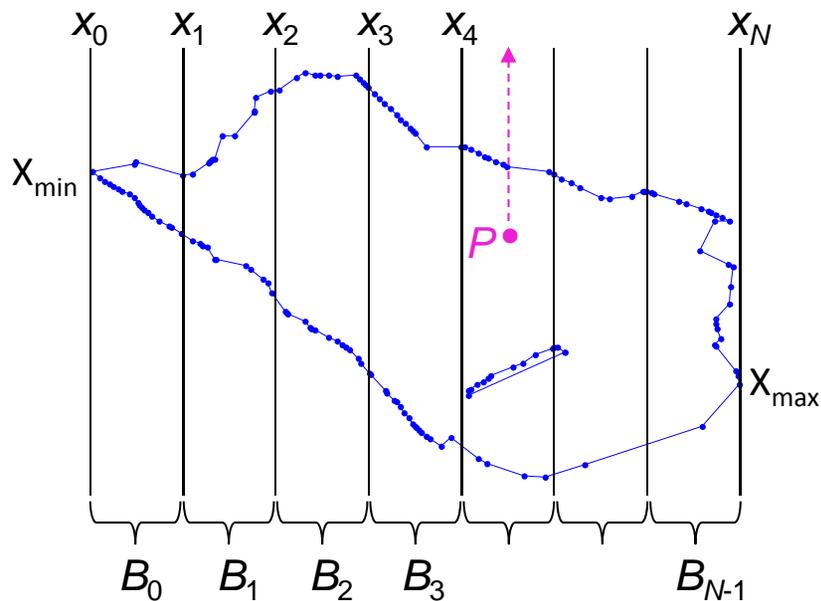
- 1) R-tree-based pre-filtering
- 2) LSH adapted to crossing number algorithm



Efficient Geo-Fencing: R-tree Based Pre-filtering



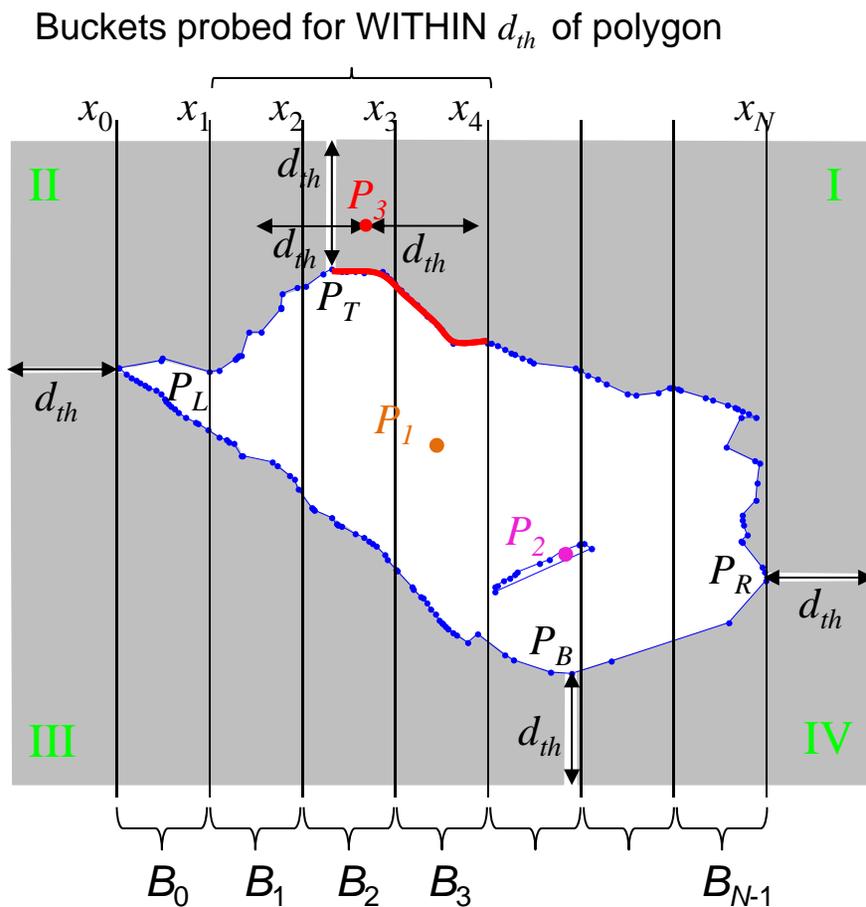
Efficient Geo-Fencing: INSIDE Detection



Hash table of edges

- ❖ A separate hash table for each polygon
- ❖ A fixed number of buckets, N , for each hash table
- ❖ Hash function
 - $T = (X_{\max} - X_{\min}) / N$
 - $\text{HashKey}(x) = \text{int}((x - X_{\min})/T)$
- ❖ An edge $(x_1, y_1) - (x_2, y_2)$ stored in buckets from key_1 to key_2 ,
 - $\text{key}_1 = \text{HashKey}(x_1)$,
 - $\text{key}_2 = \text{HashKey}(x_2)$

Efficient Geo-Fencing: WITHIN Detection



❖ LSH with multi-probing

- Inside polygon • P_1
- Inside inner ring • P_2
- Outside outer ring • P_3

❖ Optimization P_3

- Range of a point
- Divide outer area into 4 ranges
- Only check edge in the same range with the point

Geo-Fencing: Evaluation Setup

❖ Training dataset

- Two point files: Point500 (39,289 instances), Point1000 (69,619 instances)
- Two polygon files: Poly10 (30 instances), Poly15 (40 instances)
- Ground truth available (different combinations of inputs and predicates)

❖ Two predicates

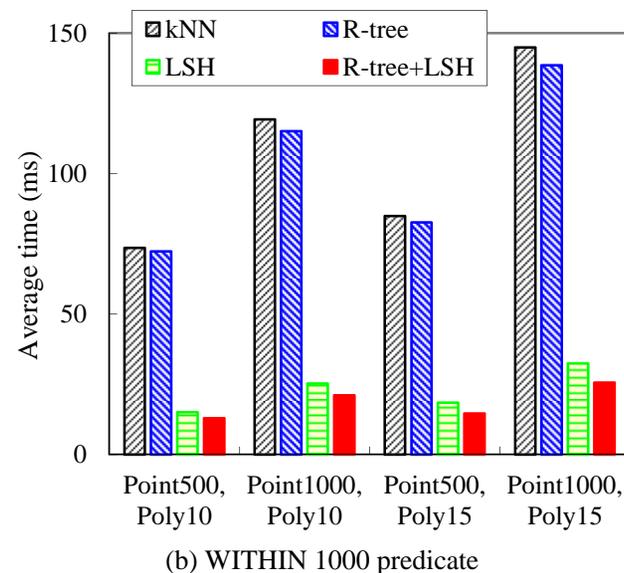
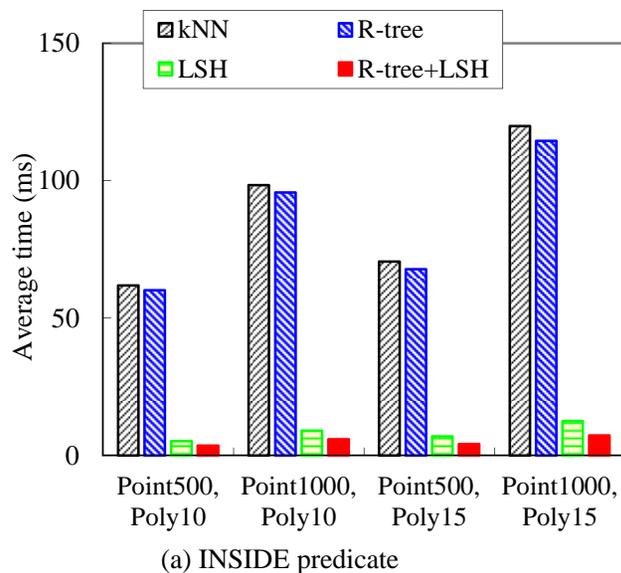
- INSIDE & WITHIN 1000
- Execution times without overhead (file I/O, data conversion)
- Accuracy & efficiency (4 methods)

❖ Environment

- A laptop PC (Intel Core i5 CPU, 64-bit Windows 7)

Geo-Fencing: Example Experiments

- ❖ 100% accuracy with test set
- ❖ Running time without system overhead
 - Measured via Windows `QueryPerformanceCounter()`: 100 runs
 - **LSH+R-tree**: Execution speed-up by 970% for INSIDE and by 370% for WITHIN



Other Optimizations

- ❖ Execution profiling showed that I/O processing required considerable time
 - Large amounts of text data needed to be read
- ❖ Therefore we applied several I/O optimizations
 - Reading data in larger blocks, not line-by-line
 - Writing pairing results in large blocks
 - Optimized number conversion: text-float to binary-float
 - Multi-threading
- ❖ Batch processing
 - Multiple points at a time, find candidate pairs for each polygon
 - Precise pairing for each polygon (CPU cache optimization)

Conclusions

- ❖ Different levels of approximation
 - Polygon as MBR: pre-filtering via R-tree
 - Edges in bucket: LSH
- ❖ LSH table per polygon
 - Compatible with R-tree
 - Fixed number of buckets, less affected by the distribution and shapes of polygons
- ❖ Simple, effective and efficient
 - 100% accuracy, high speed

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Thank You – Q&A



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