

Spatial Keyword Range Search on Trajectories

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Big Trajectory Data

- ◆ Wireless Sensors with Global Position System
- ◆ Driving Route Track Record
- ◆ Social Network with Location-based Service

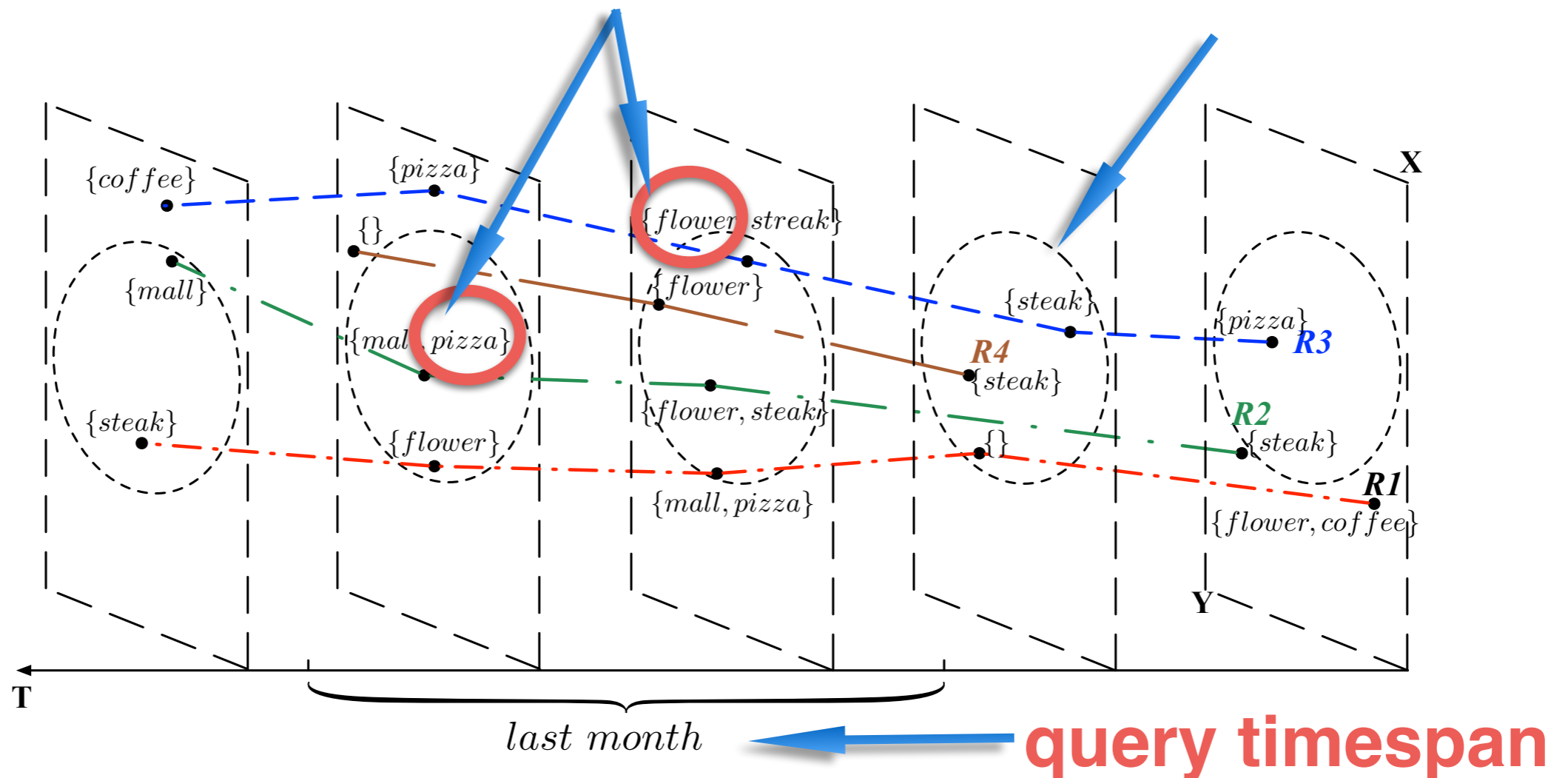


Motivation

It is meaningful to search trajectories based on **three** aspects, *i.e.*, spatio, temporal, textual.

query keywords

query range



Related Work

◆ Activity Trajectory Similarity Query (ATSQ)

[23] Zheng, K., Shang, S., Yuan, N. J., and Yang, Y. Towards efficient search for activity trajectories. In Data Engineering (ICDE), 2013 IEEE 29th International Conference on (2013), IEEE, pp. 230–241.

◆ Top- k Spatial Keyword Query (TkSK) on trajectory

[6] Cong, G., Lu, H., Ooi, B. C., Zhang, D., and Zhang, M. Efficient spatial keyword search in trajectory databases. arXiv preprint arXiv:1205.2880 (2012).

Problem Statement

Spatial Keyword Range search on Trajectories (SKRT):

Given a query *region*, a *timespan* and a set of *keywords*, we aim to retrieve trajectories that go through this *region* during query *timespan*, and contain *all* the query *keywords*.

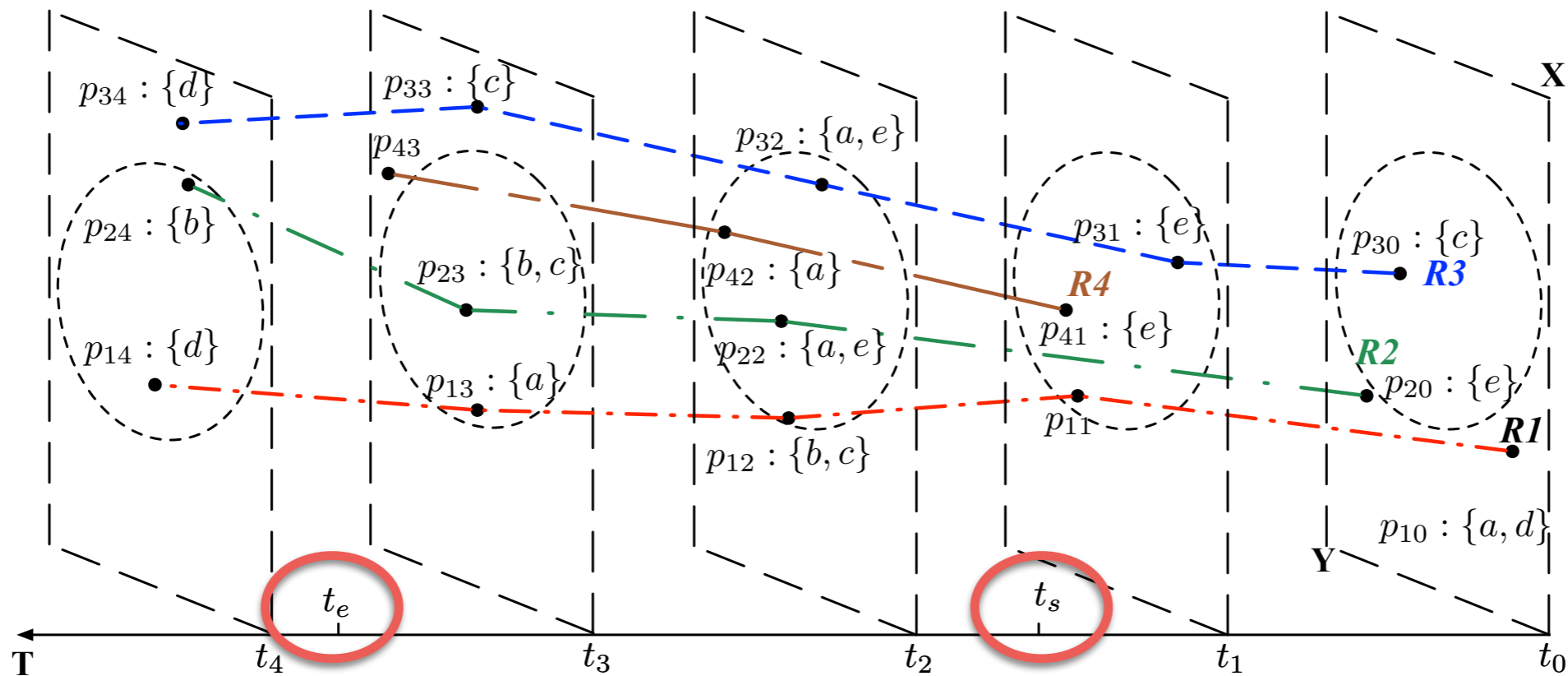
Example

$Q.R$: the space within dotted circle

$$Q.T = [t_s, t_e]$$

$$Q.\Phi = \{a, c\}$$

Trajectories Retrieved:
R1, R2



Inverted Octree (IOC-Tree)

- ✦ **inverted index**
- ✦ **octree**
- ✦ **morton code**
- ✦ ***signature* technique**

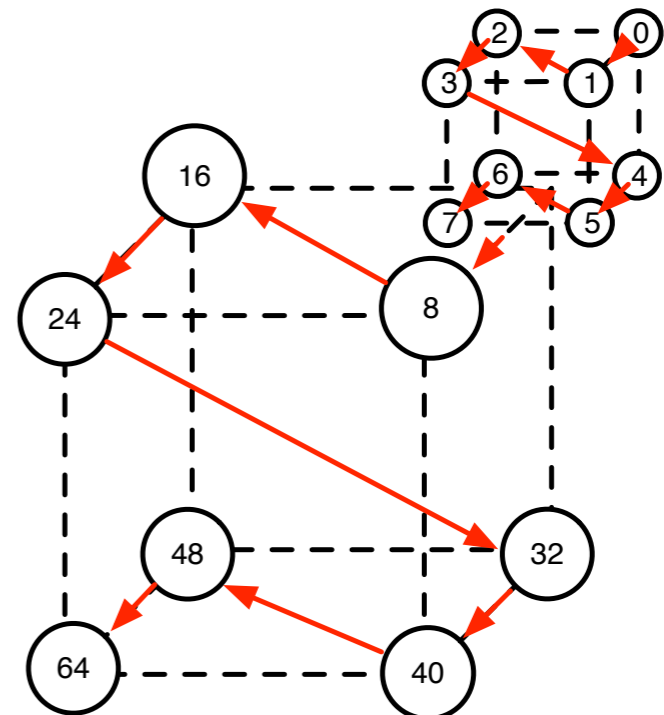
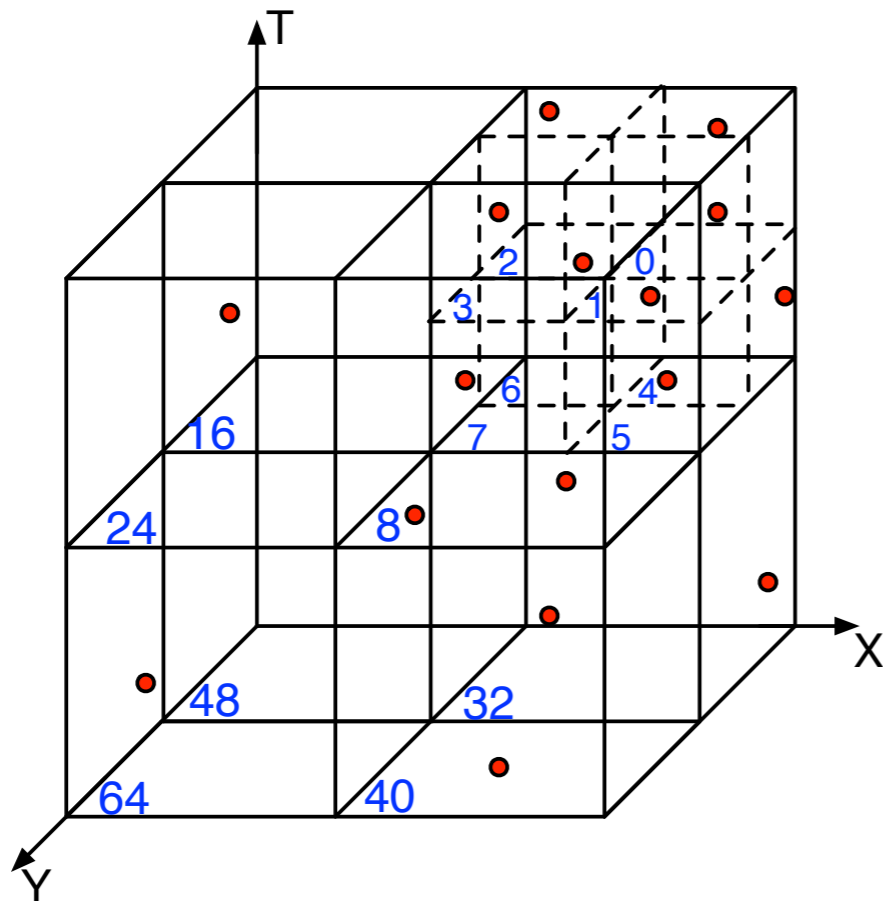
Inverted Index

WHY follow keyword-first-pruning ?



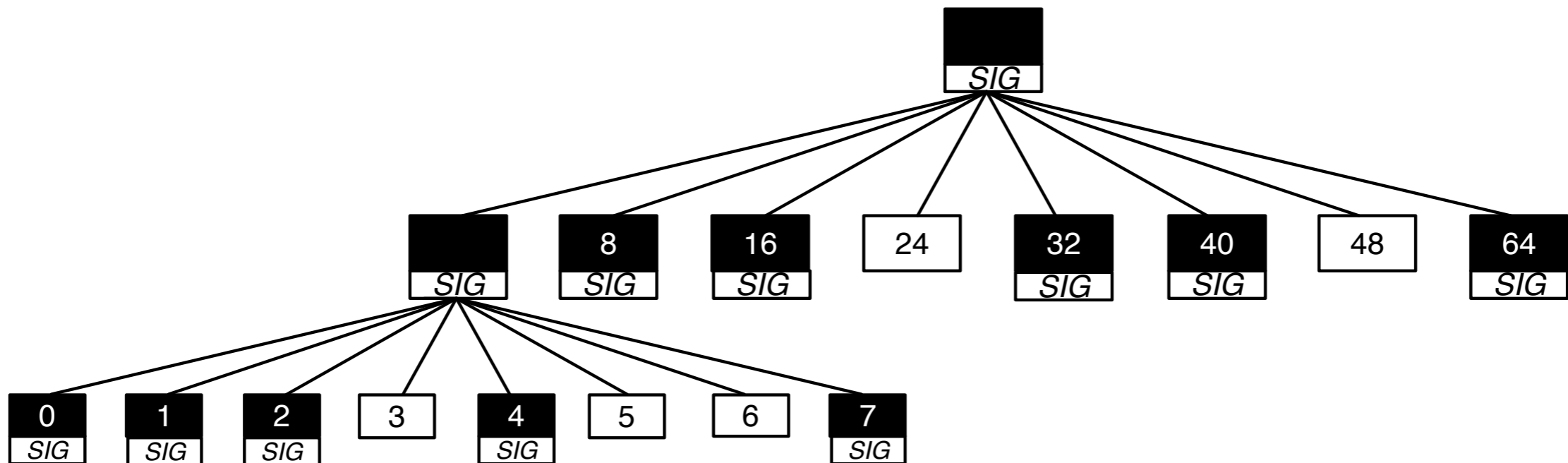
Octree & Morton Code

- ✦ octree: 3D analog of quadtree
- ✦ morton code: two nodes with high spatio-temporal proximity are assigned to the same page in the secondary storage



Signature

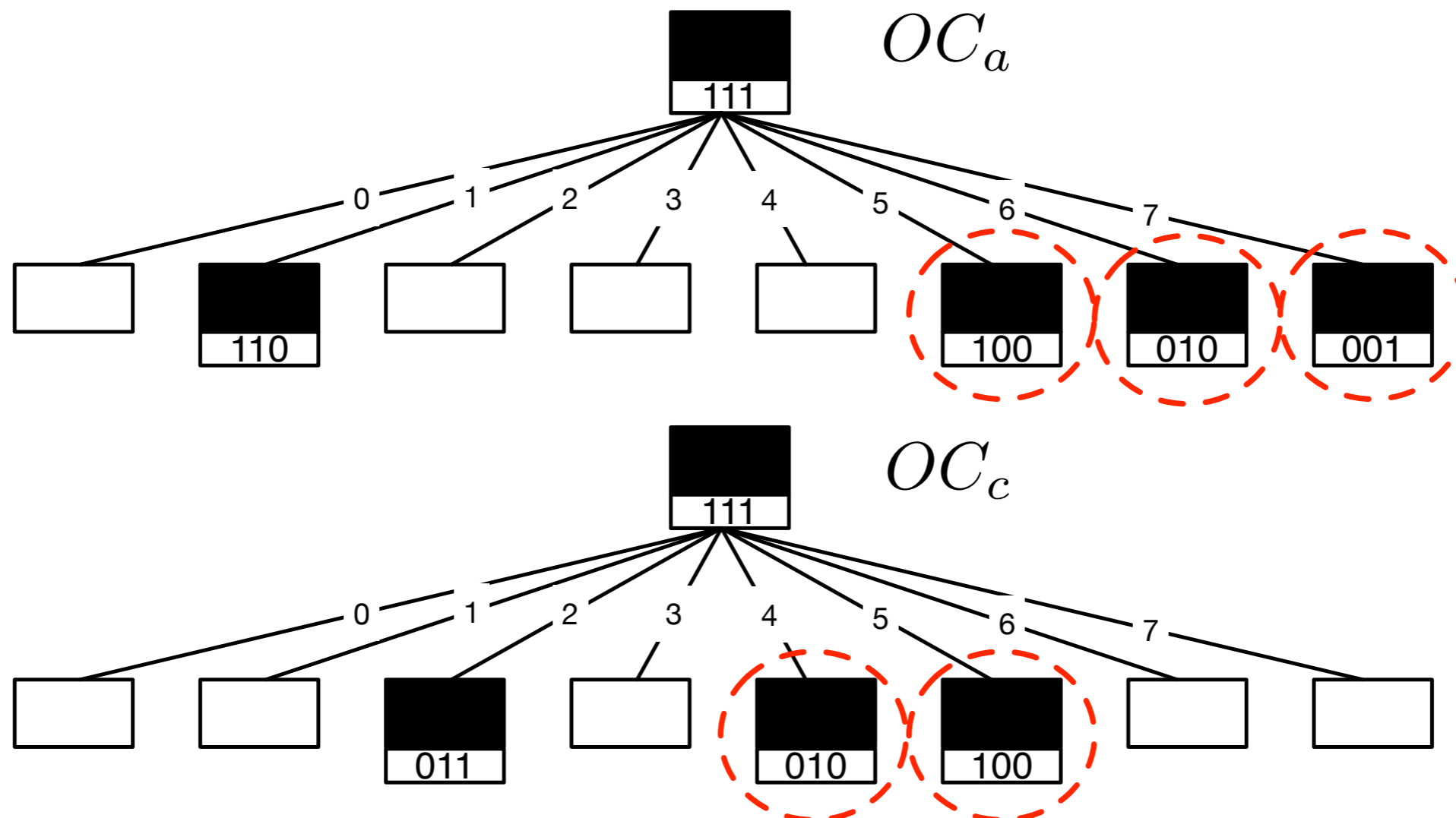
- ✦ for each node in octree, a *signature* is maintained to **summarize** the identifications of the trajectories that go through the corresponding spatio-temporal region.



IOC-Tree of Example

Table 2. Distribution of Trajectory Points

Node#	0	1	2	3	4	5	6	7
Points	p_{11}	p_{10}	p_{30}	p_{20}	p_{23}	p_{12}	p_{22}	p_{42}
		p_{31}	p_{41}		p_{24}	p_{13}	p_{32}	p_{43}
					p_{33}	p_{14}	p_{34}	



Algorithm

Main Idea: Prune as many trajectories as possible based on spatio, temporal, textual info by using IOC-Tree

- ✦ **Prune & Verification Strategy**
- ✦ **Three types of octree nodes:**
 - a) ones locate outside the range,**
 - b) fully-covered nodes,**
 - c) partially-covered nodes.**

Pruning

- ◆ Procedure *Prune* prunes non-promising nodes based on IOC-Tree
- ◆ *STRangeFilter*: only explore *black* nodes(i.e., non-empty nodes)
- ◆ deals with nodes level by level among related octrees

Procedure Prune(Q, L)

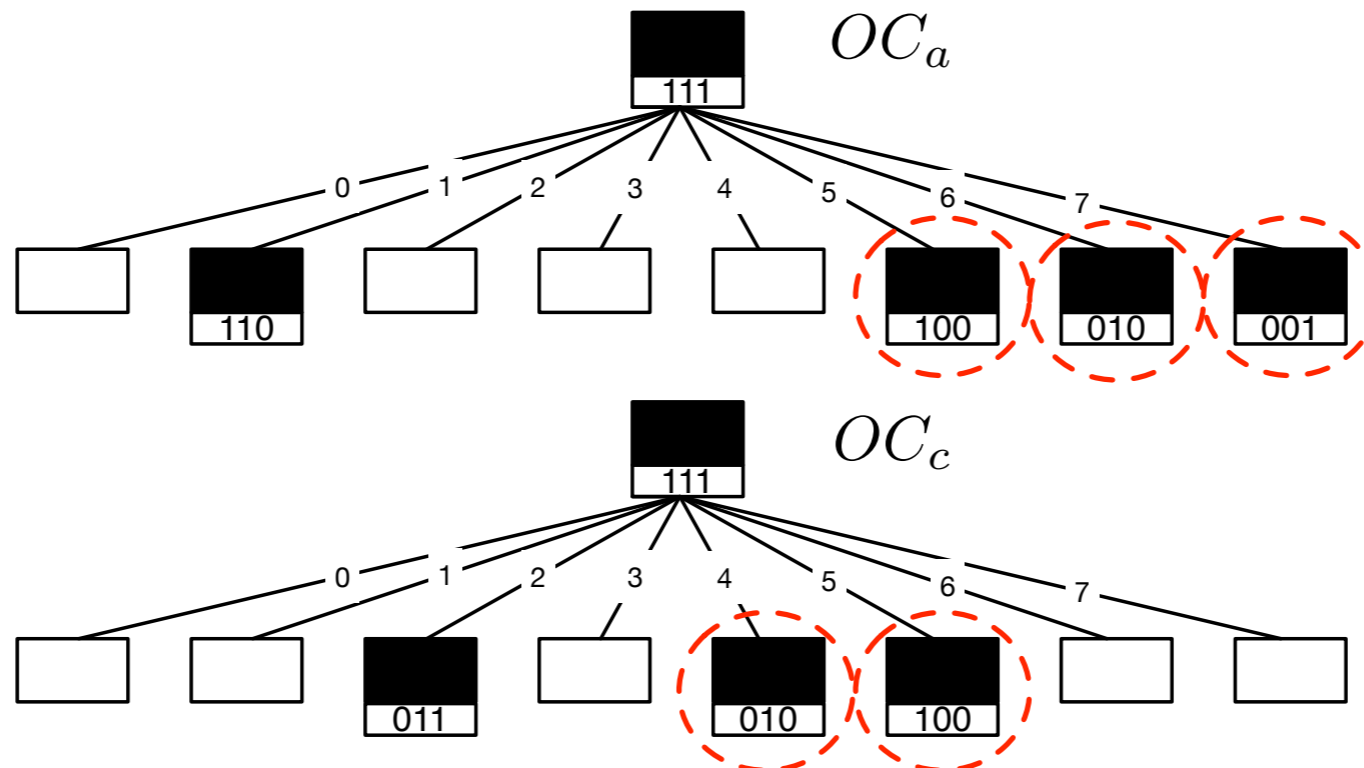
```
1 while  $L \neq \emptyset$  do
2   STRangeFilter( $Q.R, Q.T, L$ );
3   foreach  $k_i \in Q.\delta$  do
4      $SIG_i =$  bitwise-OR of signatures of nodes  $v \in L$  from  $OC_i$ ;
5     foreach node  $v \in L$  from  $OC_i$  do
6       foreach  $SIG_j$  where  $j \neq i$  do
7         SignatureCheck( $v, SIG_j$ );
8     foreach node  $v \in L$  that survive from the signature test do
9       Suppose  $v$  comes from  $OC_j$ ;
10      if  $v$  is a fully covered leaf node then
11         $\lfloor$  add  $v$  into  $CN_j^f$ ;
12      else if  $v$  is a partially covered leaf node then
13         $\lfloor$  add  $v$  into  $CN_j^p$ ;
14      else if  $v$  is non-leaf node then
15        foreach child node  $v'$  of  $v$  do
16           $\lfloor$  if  $v'$  is not a white node then
17             $\lfloor$  put  $v'$  into  $L$ ;
18       $\lfloor$  delete  $v$  from  $L$ ;
19 foreach  $k_i \in Q.\delta$  do
20    $\lfloor$  determine  $TR(CN_i^f)$  and  $TR(CN_i^p)$ ;
```

Verification

- ◆ Procedure *Verification* aims at validating candidate trajectories
- ◆ Further validation should load corresponding cell on the disk

Procedure Verification(Q, \mathcal{A})

```
1  $\mathcal{A} \leftarrow CT^f$ ;  
2 foreach trajectory  $Tr \in CT^p$  do  
3   find out a keyword set  $\Psi = \{k_i | Tr \in \mathcal{TR}(CN_i^f)\}$  ;  
4    $\Omega \leftarrow Q.\Phi - \Psi$ ;  
5   foreach  $\mathcal{TR}(CN_j^p)$  where  $k_j \in \Omega$  do  
6     if  $Tr \in \mathcal{TR}(CN_j^p)$  and  $LoadAndJudge(Tr, \mathcal{TR}(CN_j^p))$  then  
7        $\mathcal{A} \leftarrow \mathcal{A} \cup Tr$  ;
```



Extension

- ✦ **Spatial Keyword Range search on Trajectories with Order-sensitive keywords (SKRTO)**

Experiment

- ◆ Implemented in C++

- ◆ Windows 7

 - Intel i5 CPU(3.10GHz)

 - 8 GB main memory

- ◆ Three Real Trajectory Datasets:

 - a) two Foursquare check-in datasets from LA and NY

 - b) one geo-tagged tweets dataset

- ◆ Two Baselines:

 - a) GAT (ICDE'13)

 - b) B^{ck}-tree (arxiv'12)

Dataset Statistics

Table 3. Dataset Statistics

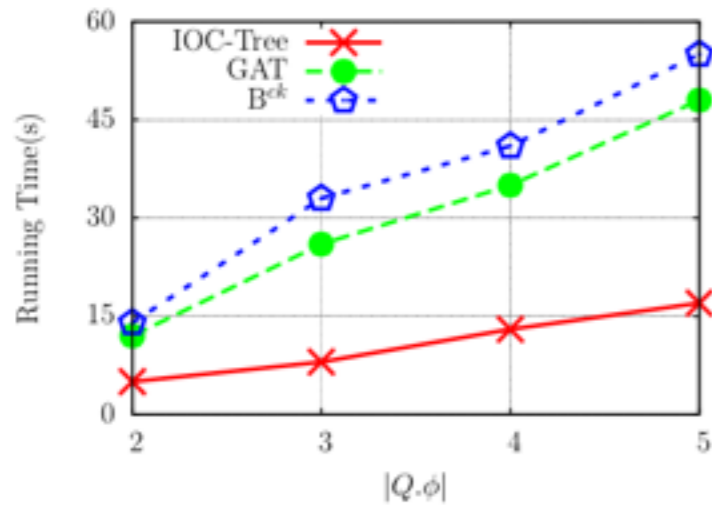
	LA	NY	TW
#trajectory	31,553	49,022	214,834
#location	215,614	206,416	1,287,315
#tag	3,175,597	3,068,401	28,645,905
#distinct-tag	100,843	89,665	1,852,141

Experimental Settings

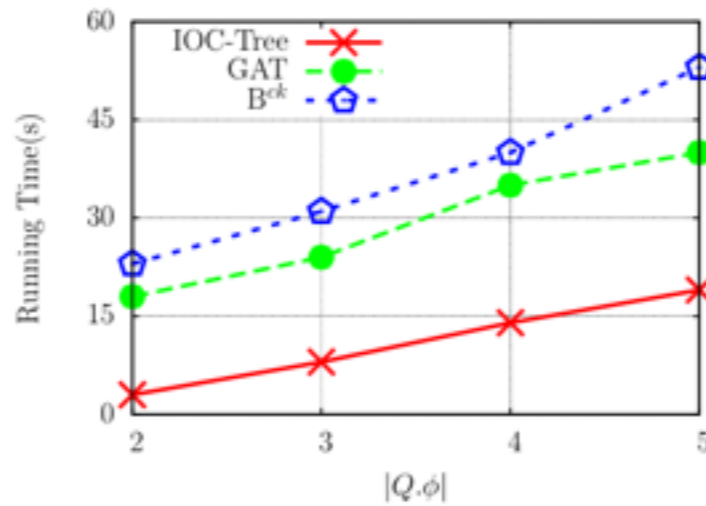
Table 4. Experimental Settings

	$ Q.\phi $	$ Q.T $ (month)	$\delta(Q.R)$ (km)
LA, NYC	2, <u>3</u> , 4, 5	1, 2, <u>3</u> , 4, 5	10, 20, <u>30</u> , 40, 50
TW	2, <u>3</u> , 4, 5	0.5, 1, <u>1.5</u> , 2, 2.5	5, 10, <u>15</u> , 20, 25

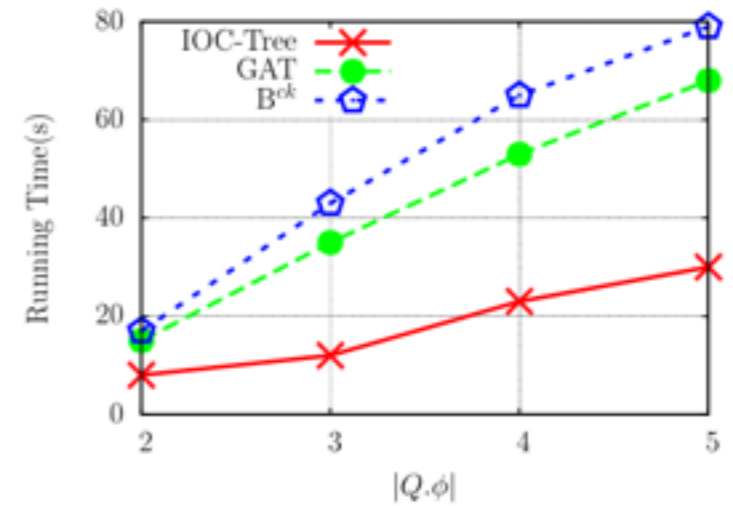
Varying number of query keywords



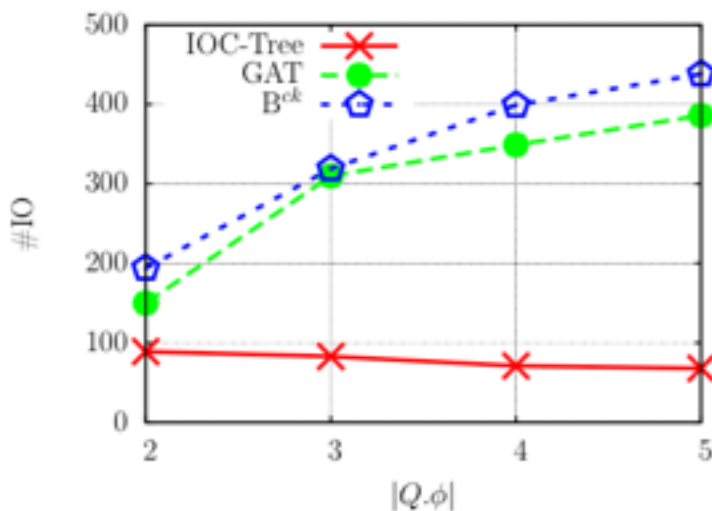
(a) *SKRT* on LA



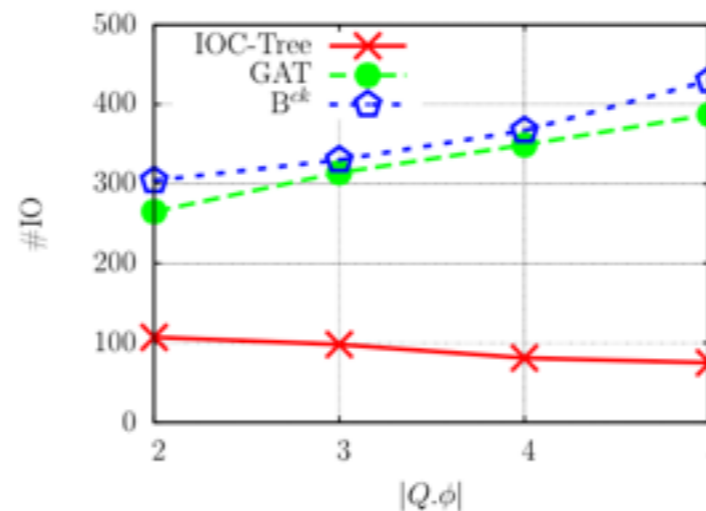
(b) *SKRT* on NYC



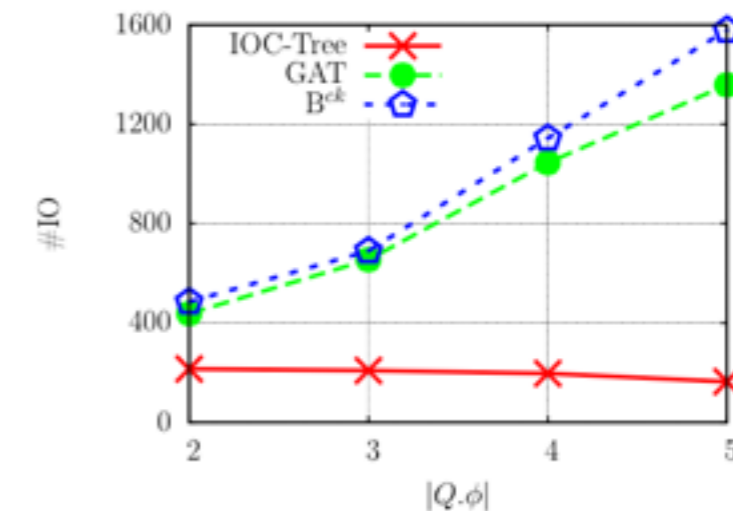
(c) *SKRT* on TW



(a) *SKRT* on LA

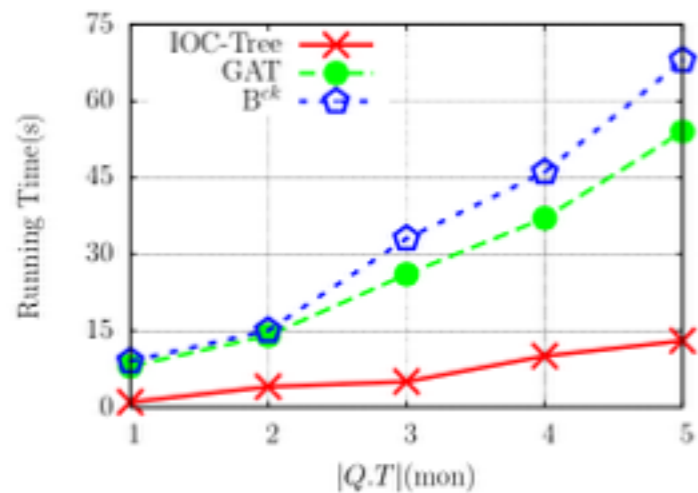


(b) *SKRT* on NYC

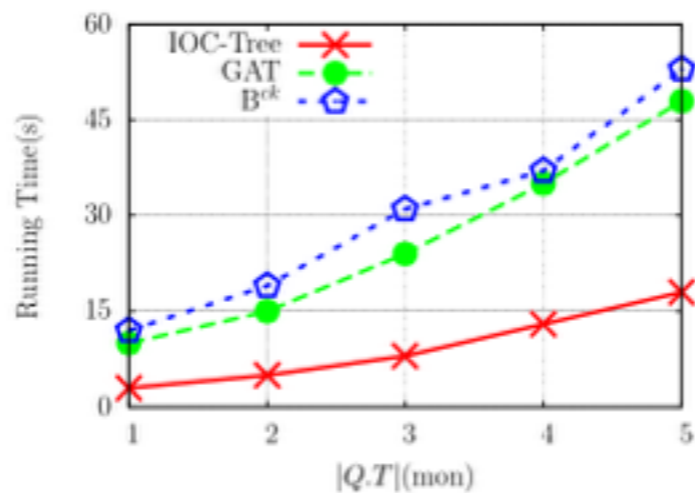


(c) *SKRT* on TW

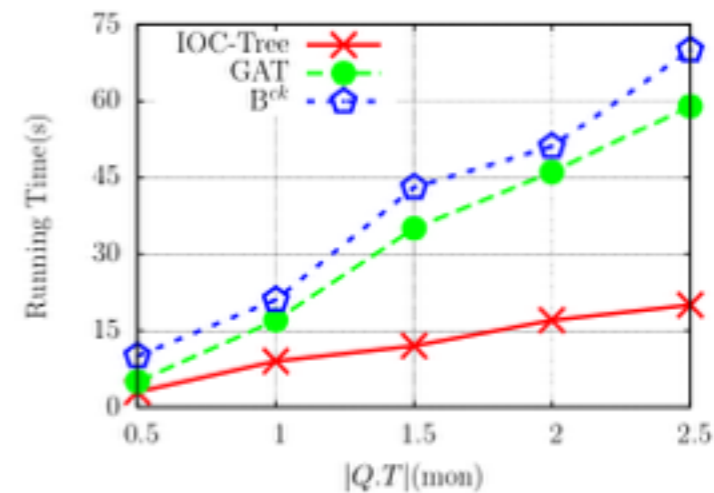
Varying query timespan



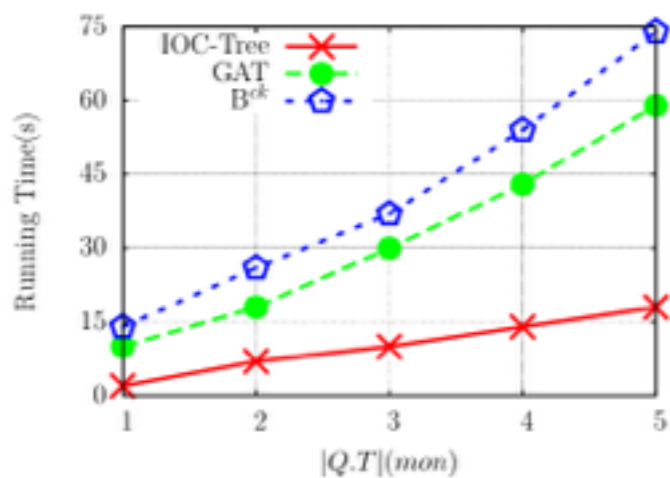
(a) *SKRT* on LA



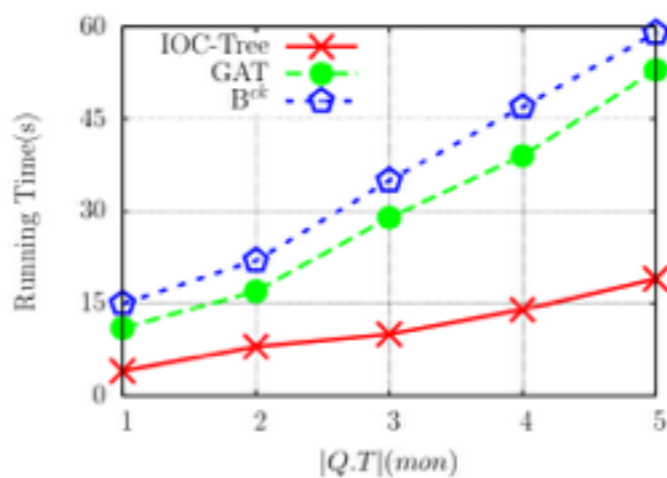
(b) *SKRT* on NYC



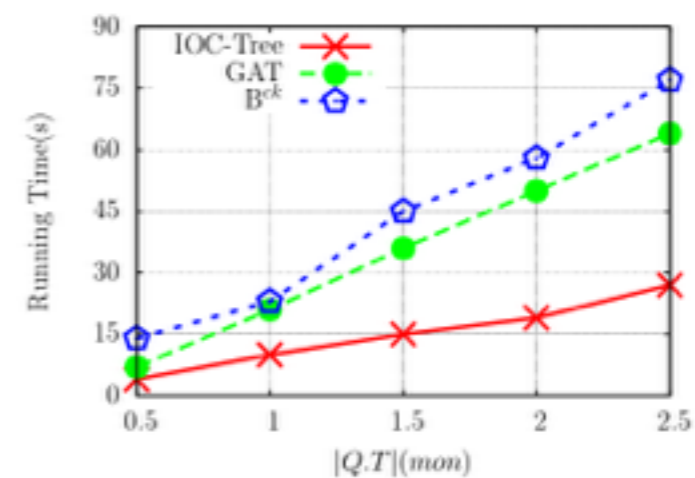
(c) *SKRT* on TW



(a) *SKRTO* on LA

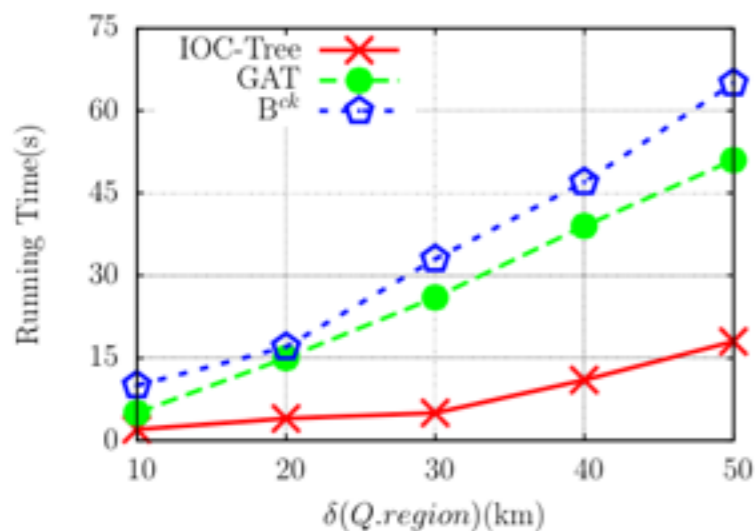


(b) *SKRTO* on NYC

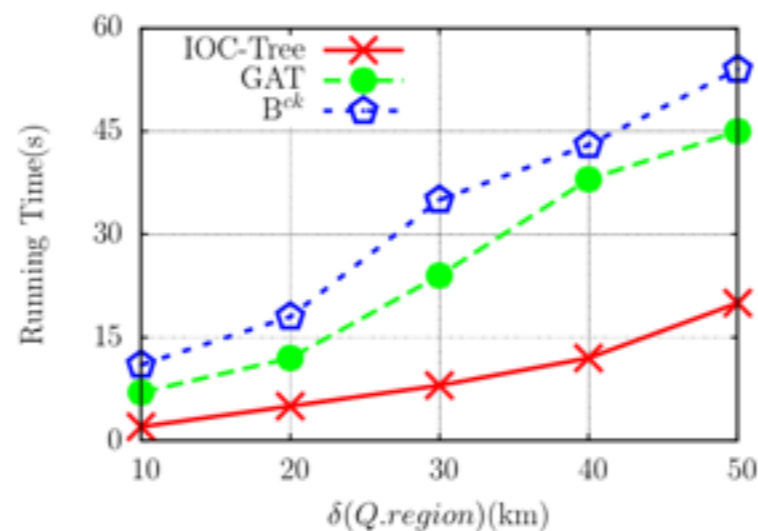


(c) *SKRTO* on TW

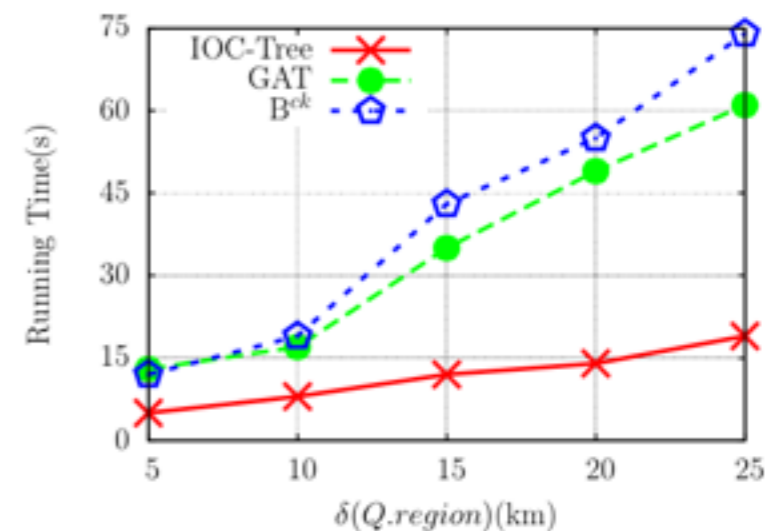
Varying query range



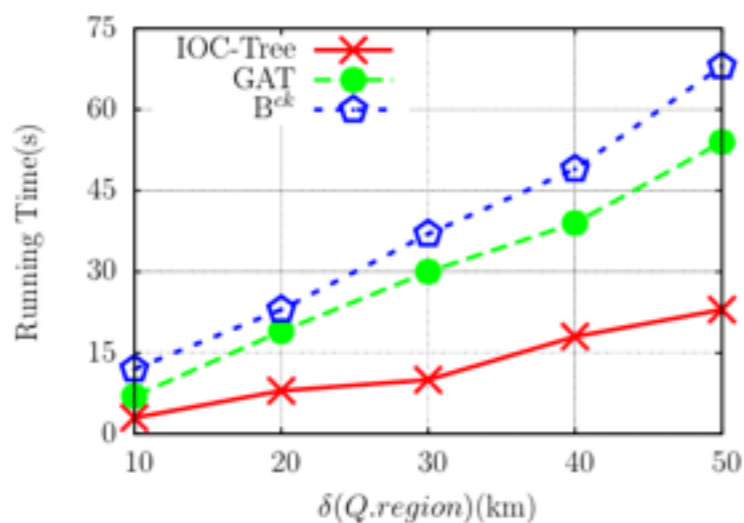
(a) *SKRT* on LA



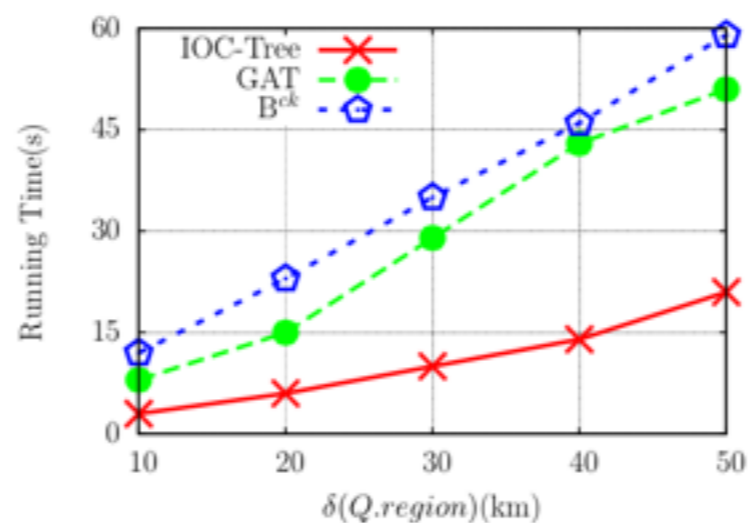
(b) *SKRT* on NYC



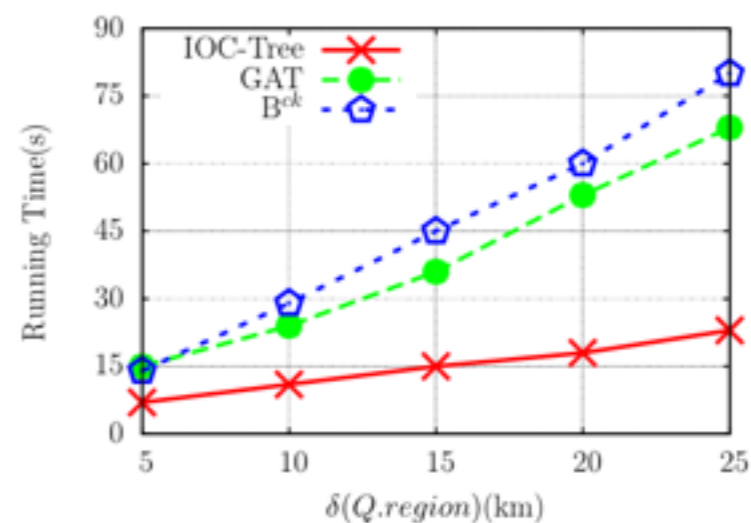
(c) *SKRT* on TW



(a) *SKRTO* on LA



(b) *SKRTO* on NYC



(c) *SKRTO* on TW

Conclusion

- ✦ **Propose spatial keyword range search on trajectories**
- ✦ **Design IOC-Tree structure and query processing algorithm**
- ✦ **Extensive experiments on real datasets confirm the efficiency of our techniques**

Thank you!