



DITA: Distributed In-Memory Trajectory Analytics

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Motivation

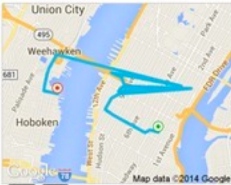
Trajectory data is getting bigger and bigger

UBER

DECEMBER 11, 2014

\$293.00

Thanks for choosing Uber, Megan



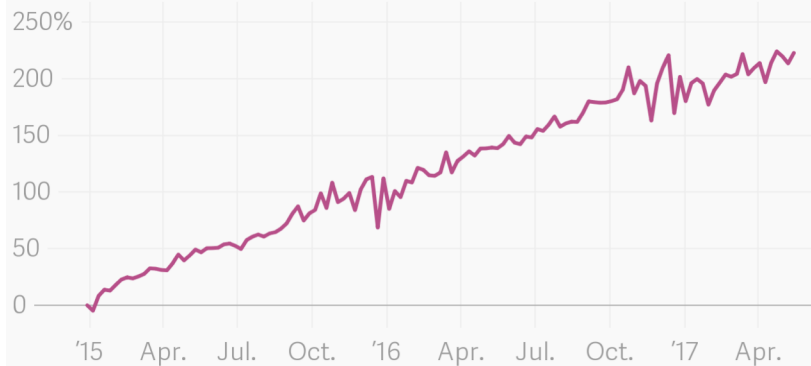
FARE BREAKDOWN

Base Fare	7.00
Distance	54.40
Time	36.41
Normal Fare	\$97.81
Surge x2.8	176.06
Subtotal	\$273.87
NJ Surchage LT Toll (?)	20.00
Rounding Down	-0.87
CHARGED	\$293.00

YOU'VE EARNED 2X POINTS
MEMBERSHIP REWARDS®

You rode with Adnan
Issued on behalf of Weiser
Ride provided by 002598 Hinter

Growth in weekly Uber users since January 2015

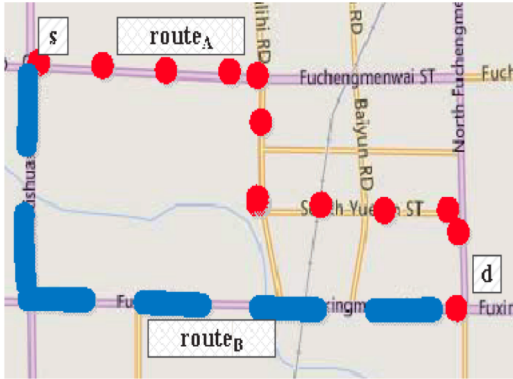


△ T L △ S | Data: Second Measure. Anonymized credit-card purchases from millions of US consumers.

2 Billion Uber trips by 06/2016
62 Million Uber trips in 06/2016

Motivation

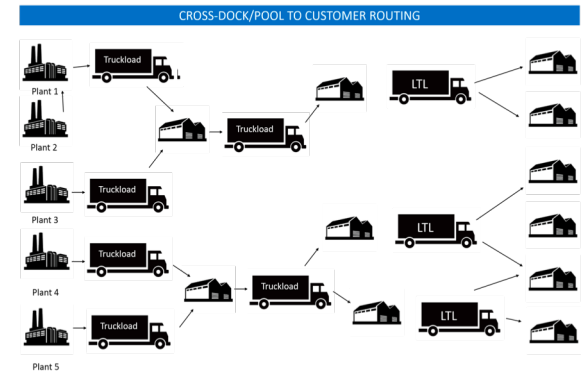
Applications of trajectory analytics



Trajectory Recommendation



Road Planning



Transportation Optimization

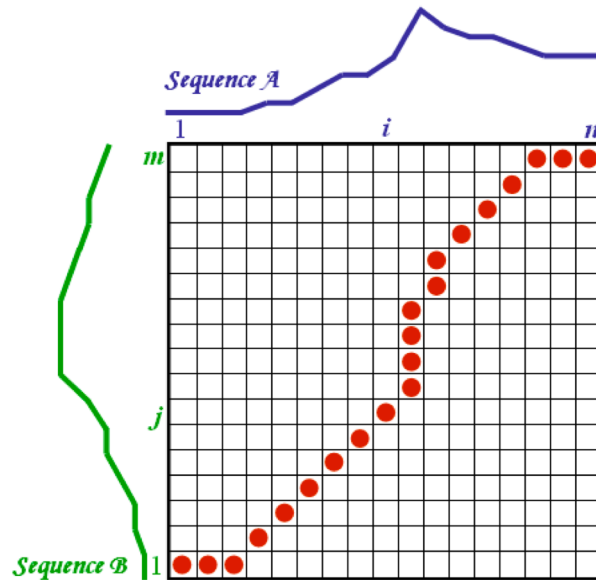
Motivation

Existing systems are limited in a number of ways

- Data locality
- Load balance
- Easy-to-use interface
- Versatility to support various trajectory similarity functions
 - Non-metric ones: DTW, LCSS, EDR
 - Metric ones: Frechet

Background

- Trajectory: a sequence of multi-dimensional points
 - E.g., $(1, 2) \rightarrow (2, 3) \rightarrow (3, 4) \rightarrow (5, 5)$
- Distance Function between trajectories (e.g., Dynamic Time Warping)



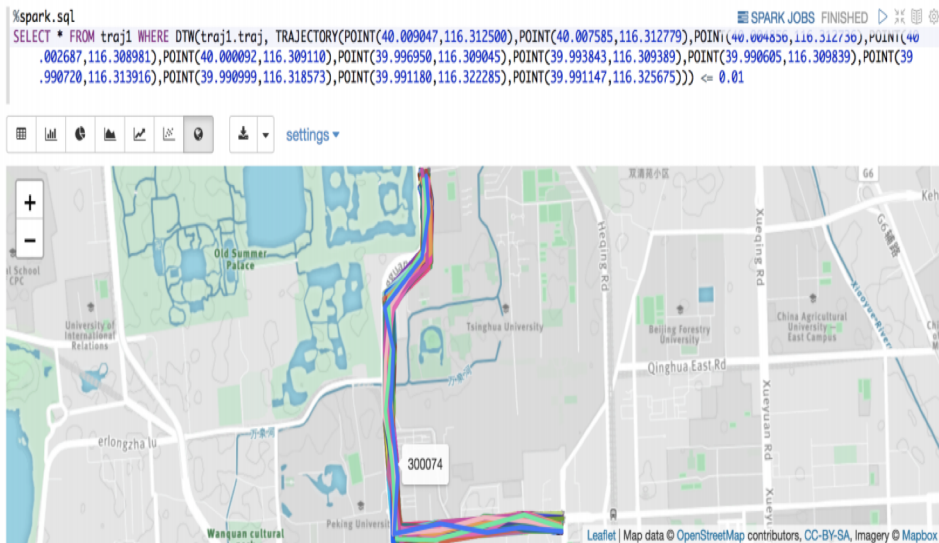
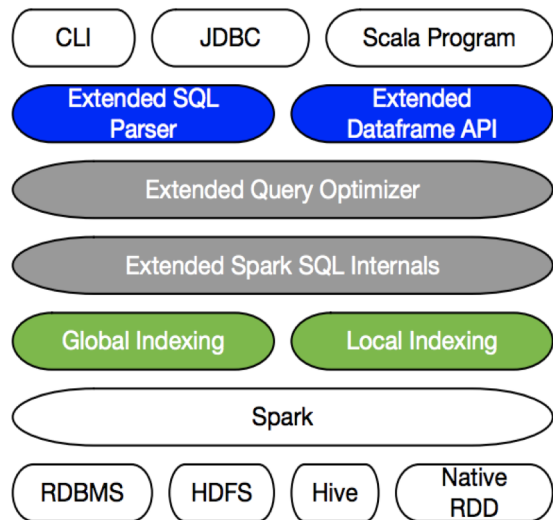
Background

Trajectory Similarity

Given two trajectories T and Q , a trajectory-based distance function f (e.g., DTW), and a threshold τ , if $f(T, Q) \leq \tau$, we say that T and Q are similar.

Overview of System

- Built on Spark SQL
- Support SQL and DataFrame
- Filter-verification framework

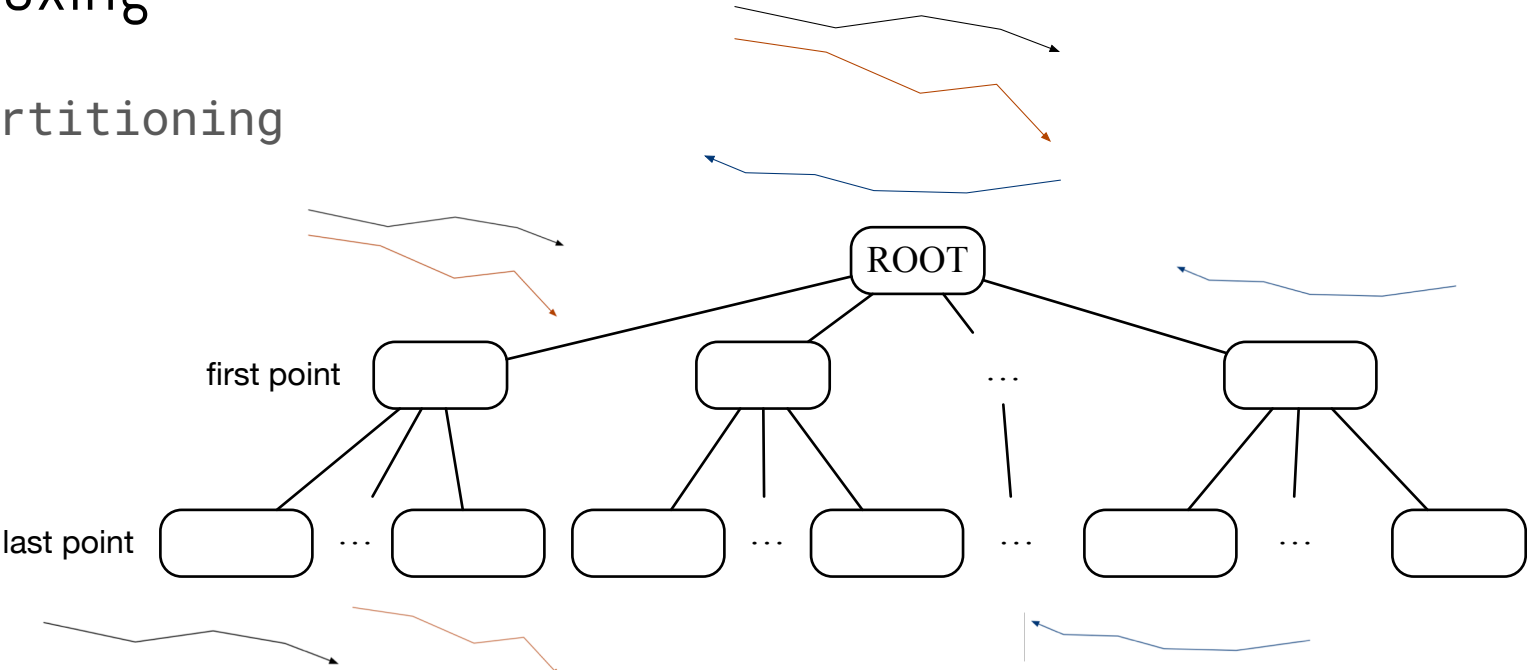


Overview of Methods

- Index
 - Partitioning
 - Global and Local Index
- Trajectory Similarity Search
 - Filter (global + local)
 - Verification
- Trajectory Similarity Join
 - Cost Models
 - Division-based Load Balancing

Indexing

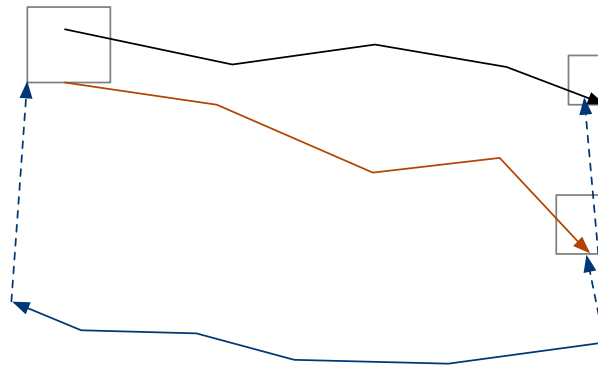
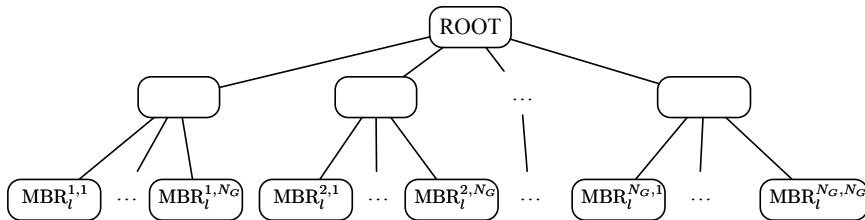
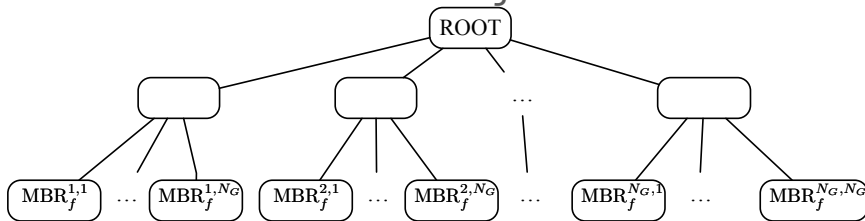
Partitioning



Indexing

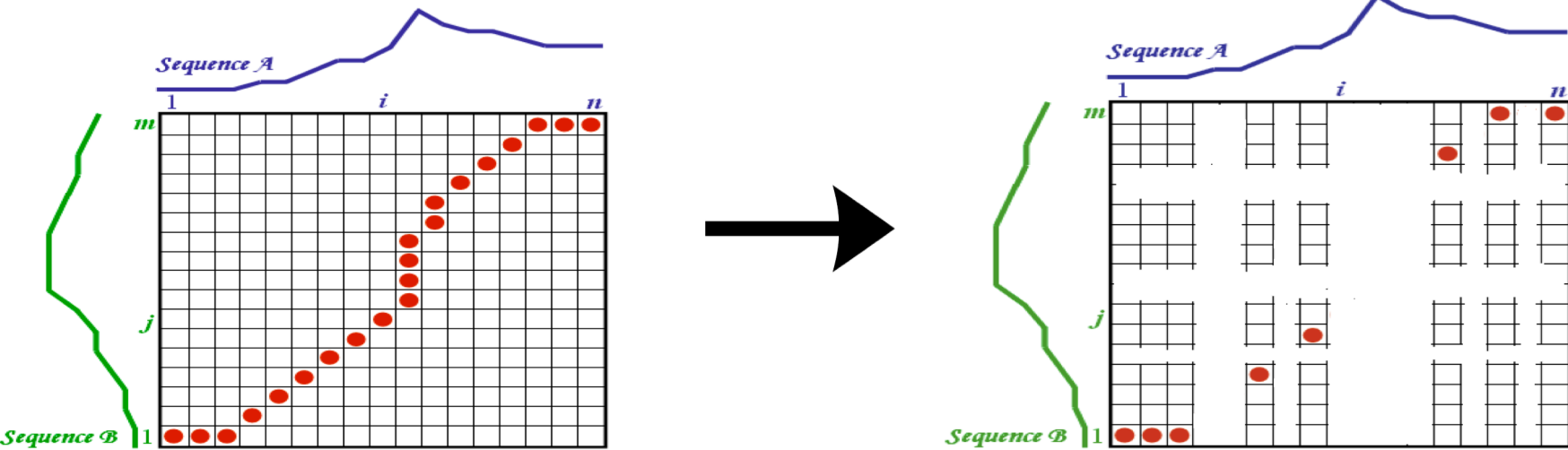
Global Index

- If $\text{MinDist}(q, \text{MBR}) \leq \tau$, then for any $q \in \text{MBR}$, $\text{Dist}(p, q) \leq \tau$
- If $\text{MinDist}(q, \text{MBR}_f) + \text{MinDist}(q, \text{MBR}_l) > \tau$, then the partition (f, l) doesn't have trajectories similar with q



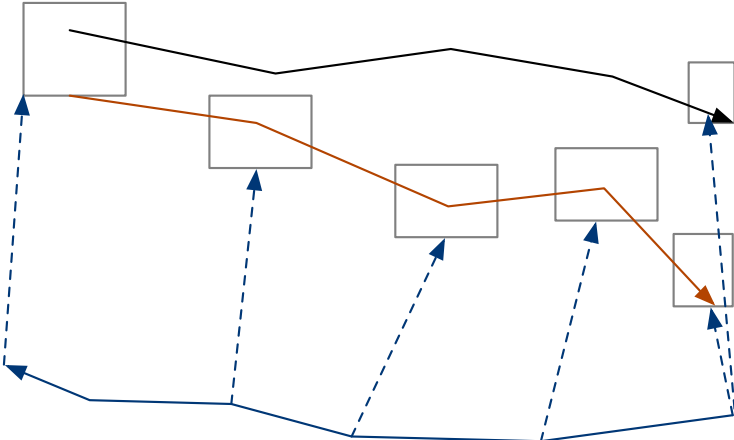
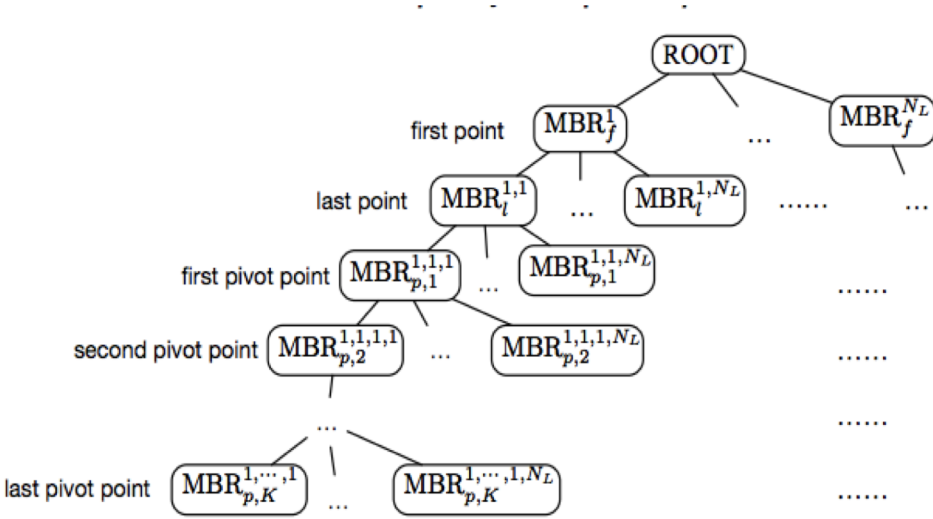
Indexing

- Pivot Point Based Distance Estimation



Indexing

Local Index

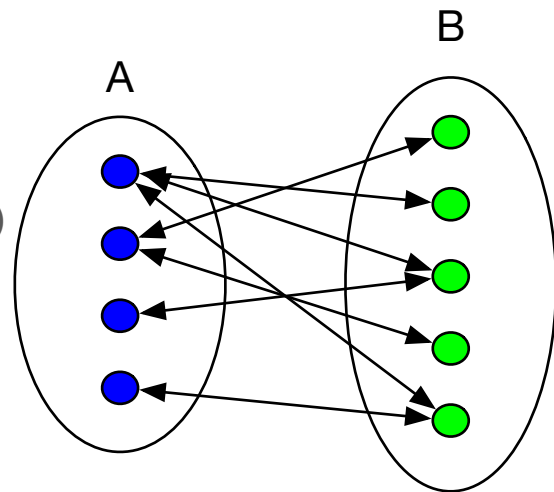


Trajectory Similarity Search

- **Basic Idea**
 - Global Pruning: find relevant partitions
 - Local Search: find similar trajectories

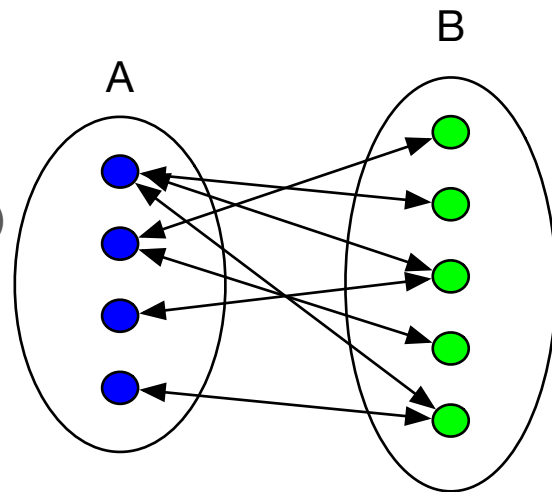
Trajectory Similarity Join

- Cost Models
 - Join Graph
 - Weight of edges (a->b)
 - a sends candidate trajectories to b
 - Transmission cost of a (data transmitted)
 - Computation cost of b (candidate pairs)
 - Built by Sampling



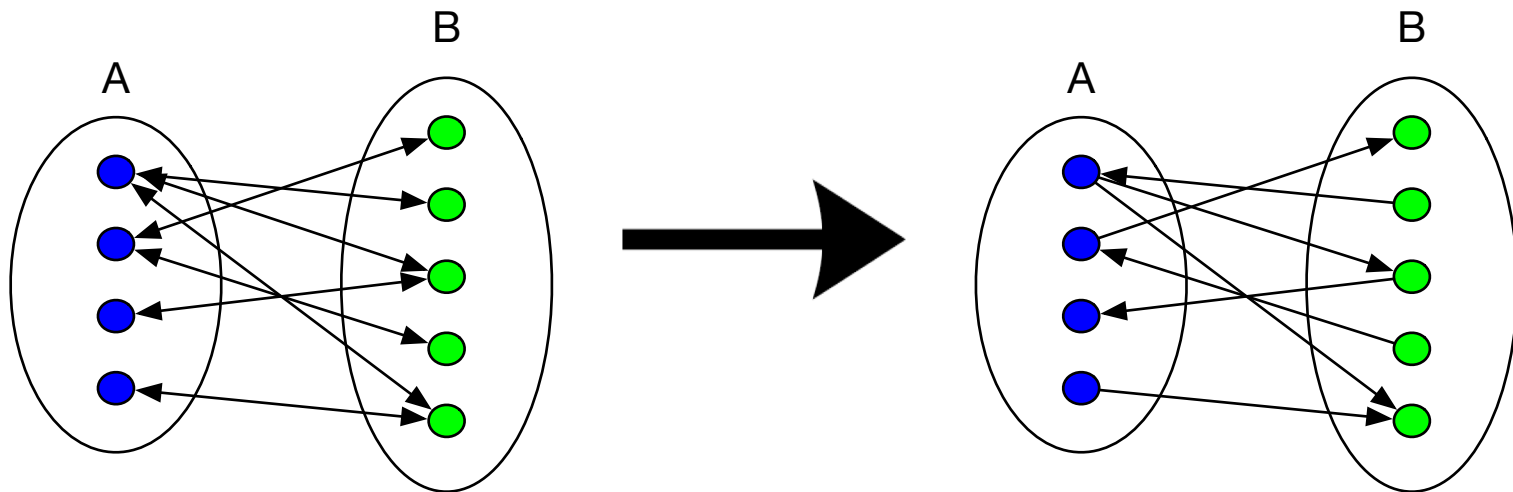
Trajectory Similarity Join

- Cost Models
 - Join Graph
 - Weight of edges (a->b)
 - a sends candidate trajectories to b
 - Transmission cost of a (data transmitted)
 - Computation cost of b (candidate pairs)
 - Built by Sampling
 - Goal: minimize the maximum total cost



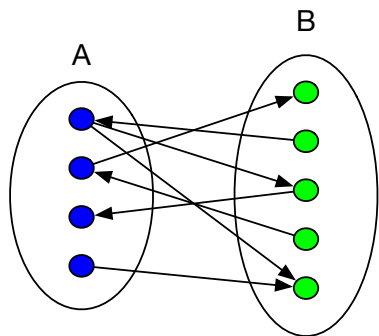
Trajectory Similarity Join

- Graph Orientation

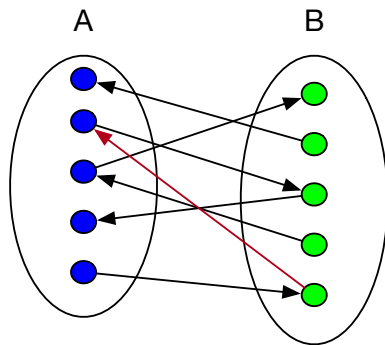


Trajectory Similarity Join

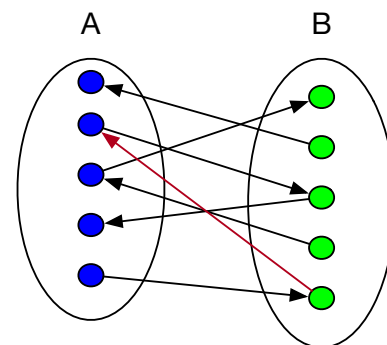
Greedy Algorithm



Initialize



Find partition with largest total cost



Repeat

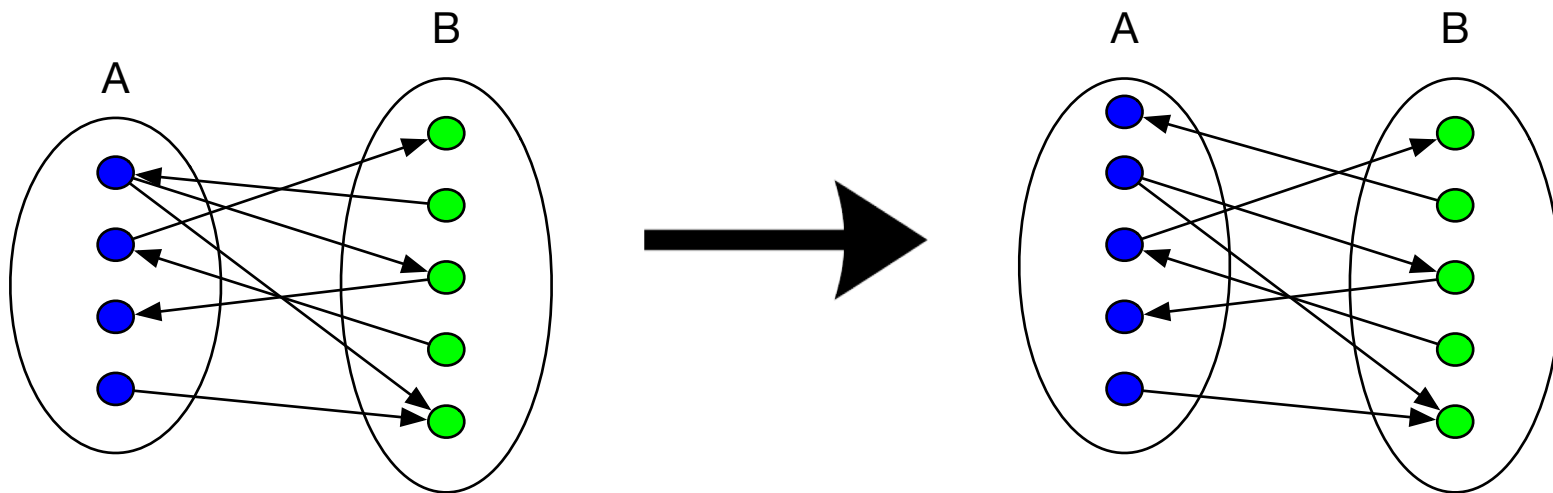
Trajectory Similarity Join

- Limitation of Graph Orientation
 - It is greedy
 - Doesn't work well for partitions with inherently huge cost

Trajectory Similarity Join

- Division-based Load Balancing

- Division unit: the 98% quantile of total cost
- For partitions whose total cost bigger than the division unit, we divide them into corresponding number of units



Experimental Results

- Setup

- 64 nodes with a 8-core Intel Xeon E5-2670 CPU and 24GB RAM
- Hadoop 2.6.0 and Spark 1.6.0
- Datasets

Table 2: Datasets

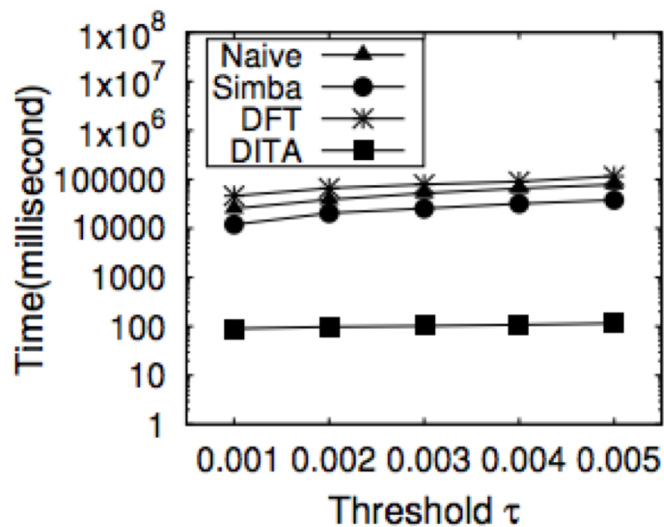
Datasets	Cardinality	AvgLen	MinLen	MaxLen	Size
Beijing	11,114,613	22.2	7	112	10.4GB
Chengdu	15,316,372	37.4	10	209	28GB
OSM(search)	141,236,563	113.9	9	3000	703 GB
OSM(join)	65,764,358	119.5	9	3000	312 GB

Experimental Results

- **Baseline Methods**
 - Naive
 - Simba (SIGMOD 2016)
 - DFT (VLDB 2017)

Experimental Results

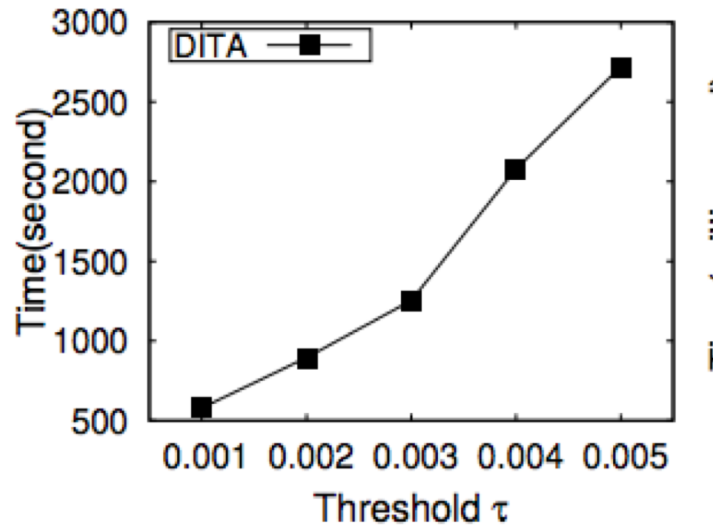
Search on Large Datasets (141M trajectories, 703GB)



(a) Search Time on OSM(search) (DTW)

Experimental Results

Join on Large Datasets (65M trajectories, 312GB)



) (b) Join Time on OSM(join) (DTW)

Conclusion

DITA: **D**istributed **I**n-memory **T**rajectory **A**alytics

- Support trajectory similarity search and join with SQL and DataFrame API
- Support most trajectory distance functions
- Filter-verification Framework
 - Global and Local Index
 - Optimizing Verification
- Experimental results show that DITA outperformed state-of-the-art approaches significantly
- Future Work