Towards Adaptive Trajectory Data Management: Modelling, Accessing, Distributing, and Query Optimization in Distributed Database

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Msc Thesis In Geomatics

Towards Adaptive Trajectory Data Management: Modeling, Accessing, Distributing, and Query Optimization in Distributed Database





Introduction

1 Introduction

- What are the trajectories?
- What are the difficulties?
- What are the DDBMSs?

2 Q1: Modelling

3 Q2: Accessing

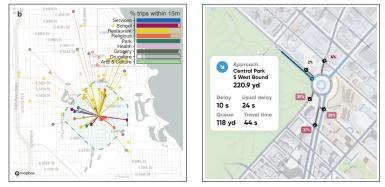




Trajectory Applications



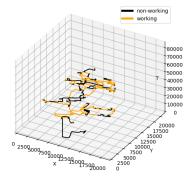
- \blacksquare Urban planning \rightarrow 15-minute city quantification.
- **Traffic management** \rightarrow **Traffic condition identification.**



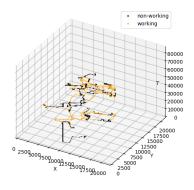
(a) 15-minute City Quantification, adopted (b) Traffic Condition Identification, from Abbiasov et al. (2024) adopted from TomTom



- Humans' understanding → Continuous sequences.
- **Data records acquired** \rightarrow **Discrete points.**



(a) Continuous Sequences

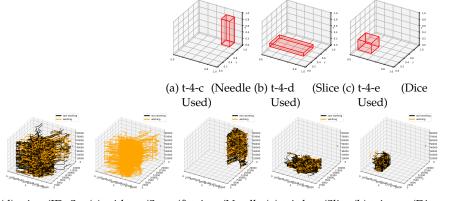


(b) Discrete Points

Trajectory Operations



Selection by ID + Selection by range.

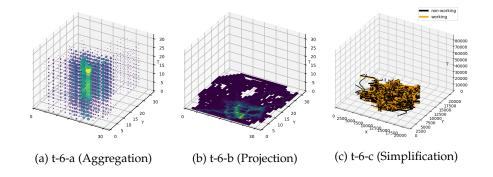


(d) t-4-a (ID Se- (e) t-4-b (State (f) t-4-c (Needle (g) t-4-d (Slice (h) t-4-e (Dice lection) Selection) Selection) Selection) Selection)

Trajectory Operations (Continued)



■ Aggregation + Projection + Simplification.



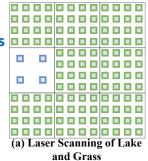
Trajectory Properties

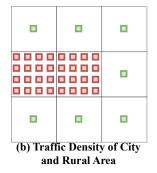


Conclusion

High frequency: Huge volume.

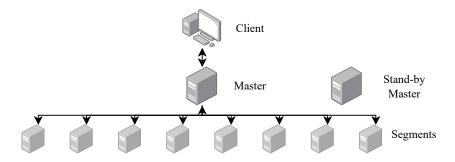
- High cardinality: Numerous entries.
- High dimensionality: Integration of space, time and semantics.
- High heterogeneity: Uneven distribution.





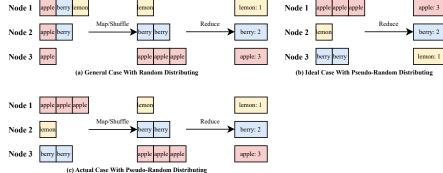


Scalability (Speed-up and Scale-up) and Localization (Cluster data and localize computations)



Experiment setting: 5 virtual machines (1 master, 4 nodes), each node has 2 segments.





 MapReduce partitions data into key-value pairs, processes them in parallel mappings and then aggregates the results through local computation in the reduce step. Introduction Q1: Modelling Q2: Accessing Q3: Distributing Conclusion

Distributed Products





Feature	Traditional Database	Hadoop	Spark	MPP Database
Volume	GB-TB	PB-EB	TB-PB	TB-PB
Robustness	High	High	High	Medium
Scalability	Low	High	High	High
Latency	Medium	High	Low	Very Low
Throughput	Low	High	High	Medium
Data Type	Structured	All	All	Structured

Introduction Q1: Modelling Q2: Accessing Q3: Distributing Conclusion

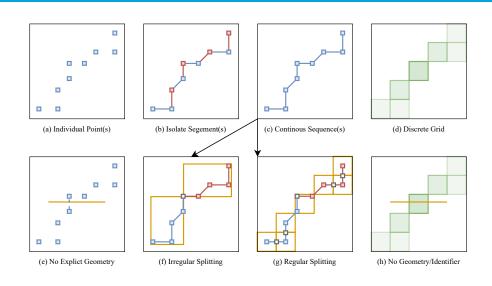
Q1: Modelling

 2 Q1: Modelling
 How to model trajectory?
 Does the Above Methods Work? -Compression

3 Q2: Accessing







Reasons for Sequence-based Model



The sequence is better! **More supported** Reconstruct D operations. Ó Smaller entries cardinality.

 Higher compression potential.

(a) Individual Points

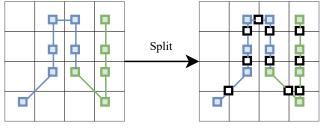
(b) Entire Trajectory

Solutions for Unstructured Nature



Split by semantics.

 Split by spatio-temporal cube.

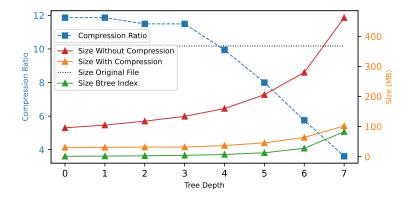


(b) Entire Trajectory

(c) Splitted Sequences

Introduction Q1: Modelling Q2: Accessing Q3: Distributing Conclusion Does the Above Methods Work? - Compression #UDelft

Further subdivision of the space would decrease the compression ratio.



Introduction Q1: Modelling Q2: Accessing Q3: Distributing Conclusion

Q2: Accessing

1 Introduction



3 Q2: Accessing

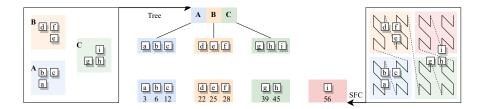
- How to access trajectory?
- Does the Above Methods Work? -Selection







- **R**-tree \rightarrow Adaptive but complex and costs more storage.
- Space filling curve → Rigid but simple and corresponds to the nature of modelling.



Uneven Distribution

Q1: Modelling

Introduction

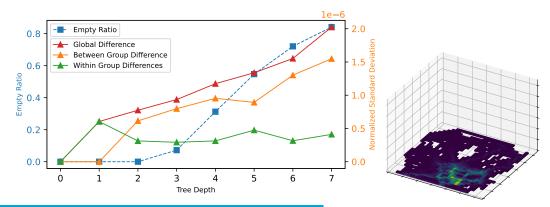


Conclusion

Q3: Distributing

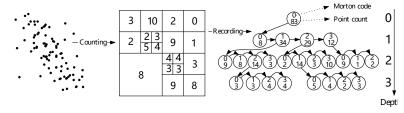
- Further subdivision of the space would increase the empty ratio.
- Further subdivision of the space would increase the global difference (unevenness).

Q2: Accessing





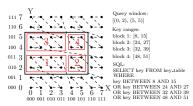
- Histogram Tree: Adaptive octree.
- Space-filling-curve (Morton): Represent a record and use b-tree for indexing.



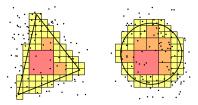
Adopted from Liu, 2022



Recursively partitioning the extent of data according to SFC regions to match different query geometries, for selecting data in the table.



(a) Executing a window query on a uniformly distributed 2D point set based on Morton encoding

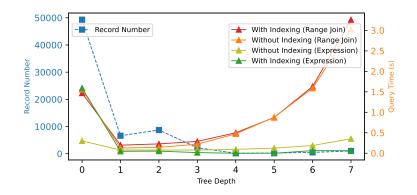


(b) Querying with a triangle and a circle: false positive points in boundary cells will be filtered out by a second filter

Adopted from Liu, 2022



Selection performance is first increasing then decreasing with the further subdivision of the space.



Q1: Modelling

Q2: Accessing

Q3: Distributing

Conclusion

1 Introduction



Q3: Distributing

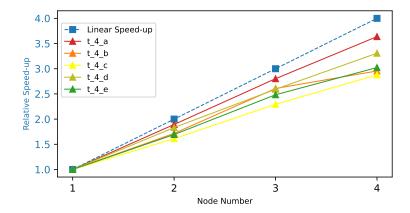
3 Q2: Accessing

- 4 Q3: Distributing
 - Does the Distributed DBMS Work?
 - Speed-up
 - Does the Distributed DBMS Work?
 - Scale-up
 - How to distribute trajectory?
 - Does the Above Methods Work? -Localization



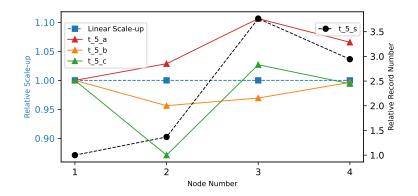
Introduction Q1: Modelling Q2: Accessing Q3: Distributing Conclusion
Does the Distributed DBMS Work? - Speed-up fullerit

■ Five operations are designed to test the speed-up (same problem sizes with increasing resources) and the result is positive.



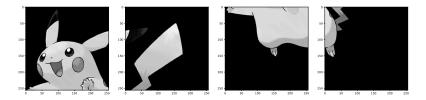
Introduction Q1: Modelling Q2: Accessing Q3: Distributing Conclusion
Does the Distributed DBMS Work? - Scale-up TUDelft

Three operations are designed to test the scale-up (increasing problem sizes with increasing resources) and the result is positive.

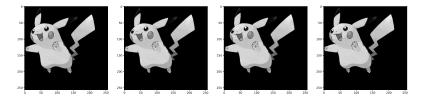


Distributing Strategy





Block-based: Fold twice and split.

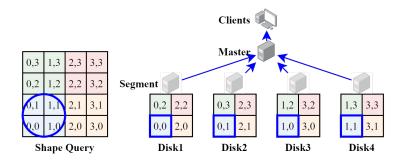


Sample-based: e.g. all the lower left pixels of 4 neighbours as a group.

Pseudo-Sampling Distributing

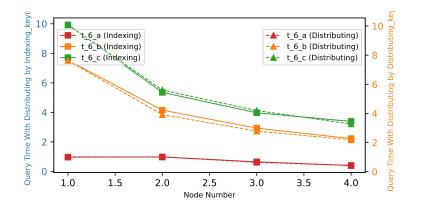


- Load-balancing: Block-based method leads to uneven data distribution.
- Localization: Random sampling leads to no locality being preserved.



Introduction Q1: Modelling Q2: Accessing Q3: Distributing Conclusion
Does the Above Methods Work? - Localization fullelft

Three operations (Aggregation, Projection and Simplification) are designed to test the localization but the result is negative.



Introduction Q1: Modelling Q2: Accessing Q3: Distributing Conclusion

Conclusion



2 Q1: Modelling

3 Q2: Accessing



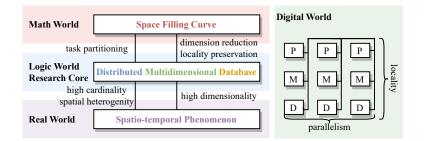


Conclusion



In conclusion, the above methods alleviate the difficulties mentioned.

- Modelling: Reduce cardinality.
- Accessing: Reduce dimensionality and alleviate uneven distribution.
- Distributing: Use parallelism to speed up and scale up.





The main lesson learnt from this thesis is the need to adapt the data properties and platform features.

Distribution Awareness

adaptive modelling (splitting)+adaptive accessing (indexing)+adaptive distributing (partitioning)+*Architecture adaptive modelling (splitting) adaptive querving (merging)

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(1)



However, there are still some limitations that should be done in future work.

- Realistic benchmarking: Not only in virtual machines.
- Workflow optimization: Adaptive splitting and range merging.
- Mathematical proofing: Not only by experiments.

Introduction

Q1: Modelling

Q2: Accessing

Q3: Distributing

Conclusion

THANK YOU!



Liu, H. (2022). nD-PointCloud Data Management: continuous levels, adaptive histograms, and diverse query geometries. A+ BE| Architecture and the Built Environment, (12), 1-206.