

Scalable Management of Trajectories and Context Model Descriptions

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Motivation

Context-awareness and context-aware computing

“... the ability of a mobile user’s applications to discover and react to changes in the environment they are situated in.”

(Schilit and Theimer, 1994)

Examples:

- Adapt ring volume to ambient noise
- Navigation of blind persons

➔ Advancements are closely connected to **miniaturization** of sensing, computing, and communication hardware

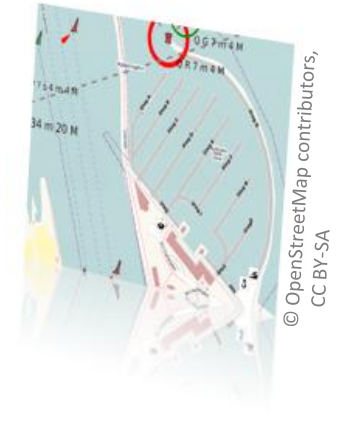
Motivation (2)

Proliferation of **sensing** technologies

- Cameras, satellites, weather stations, WSNs, GPS receivers, RFID readers, **smartphones**, ...

Comprehensive **models** of physical environment

- Differ in geographic extent, aspect, granularity, dynamism, ...
- Variety of commercial and non-commercial providers
 - Examples: TeleAtlas, Google, OpenStreetMap, CWOP



➔ Huge **potential** for context-aware applications

- Ability to select context information from different providers
- Trend: Millions of **context models** shared by large number of applications

Problem Statement

How to provide efficient access to the immense amounts of distributed dynamic context information

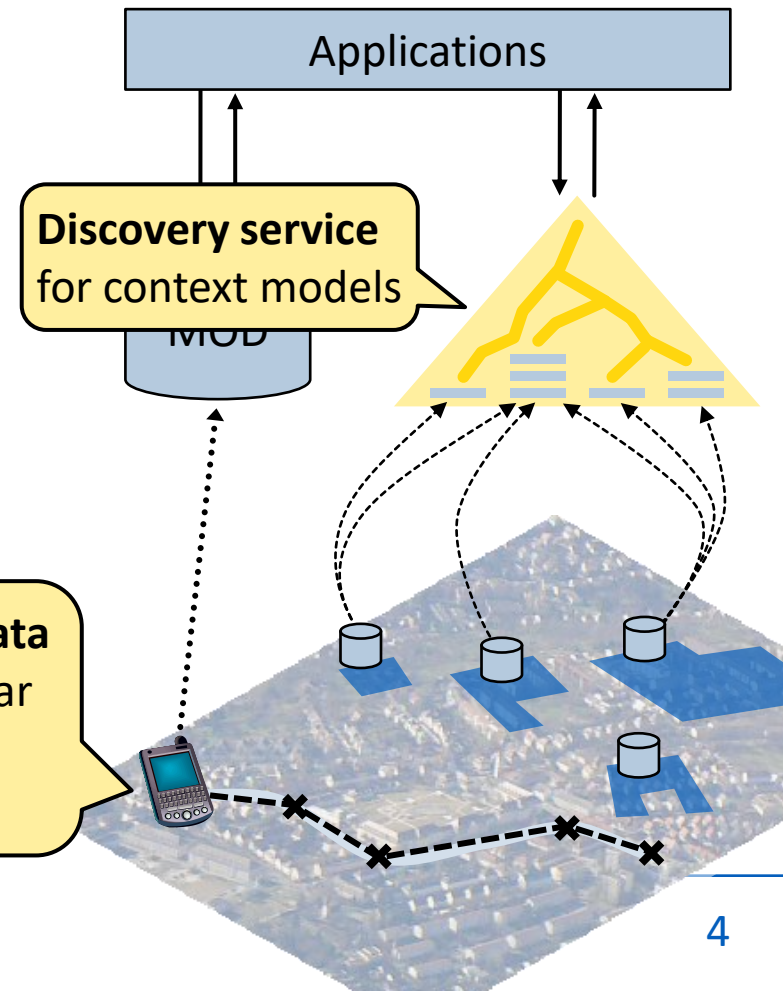


1. Formal describing of context models
2. Indexing of context model descriptions

3. Efficient real-time access
4. Distributed in space-partitioned

Large amounts of **trajectory data**

- E.g. $3 \cdot 10^7$ GPS records per year
- High communication cost
- Huge storage consumption



Overview

- Motivation and problem statement
- Efficient real-time trajectory tracking
 - Formal problem description
 - Related work
 - Connection-Preserving Dead Reckoning (CDR)
 - Generic Remote Trajectory Simplification (GRTS)
 - Evaluation and conclusion
- Distributed indexing of space-partitioned trajectories
- Summary



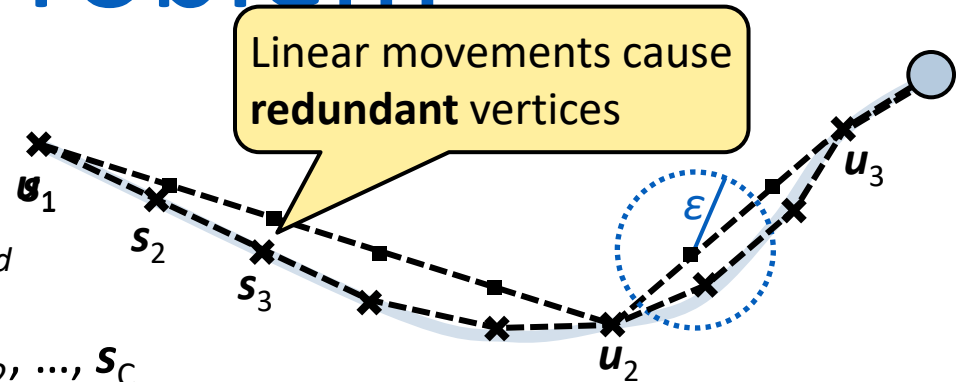
Tracking: Formal Problem

Actual trajectory $\mathbf{a}(t)$ is function $\mathbb{R} \rightarrow \mathbb{R}^d$

Sensed trajectory $\mathbf{s}(t)$ with vertices $\mathbf{s}_1, \mathbf{s}_2, \dots, \mathbf{s}_C$

- Attribute $\mathbf{s}_i.p$ denotes position at time $\mathbf{s}_i.t$
- Differs from $\mathbf{a}(t)$ due to δ_{sense} and movement during T_{sense}

Simplified trajectory $\mathbf{u}(t)$ with vertices $\mathbf{u}_1, \mathbf{u}_2, \dots$



$$\frac{1}{2} a_{\max} \left(\frac{T_{\text{sense}}}{2} \right)^2$$

Definition: Efficient real-time trajectory tracking

- Goals: Minimize $|\mathbf{u}_1, \mathbf{u}_2, \dots|$ and communication cost
- Simplification constraint: $|\mathbf{u}(t) - \mathbf{a}(t)| \leq \epsilon \quad \forall t$
- Real-time constraint: $\mathbf{u}(t)$ is available $\forall t \in [s_1.t, t_C]$

Tracking: Related Work

Line simplification algorithms

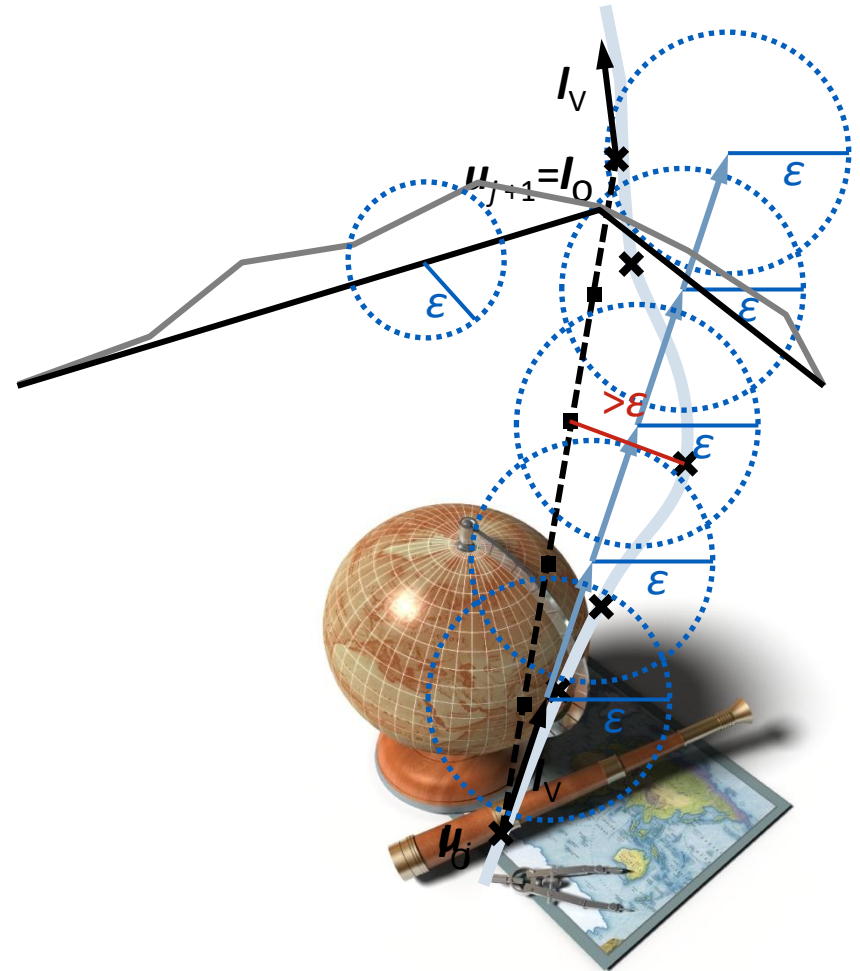
- E.g. Douglas-Peucker heuristic

➔ High communication cost,
no real-time behavior

Position tracking protocols

- Best mechanism: Dead Reckoning (DR)

➔ Linear DR with $\frac{1}{2}\epsilon$ allows for trajectory tracking [Trajcevski et al. 2006]



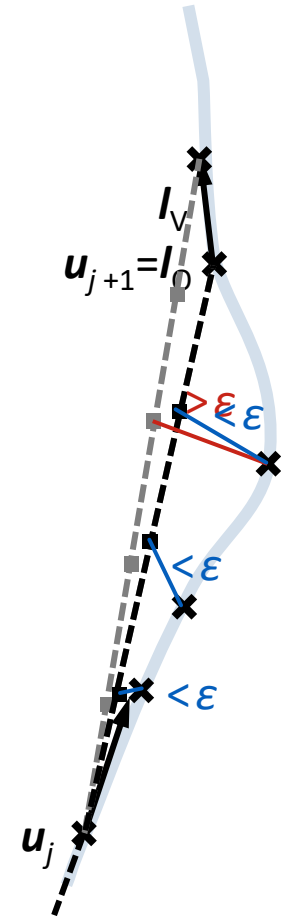
Tracking: CDR

Connection-preserving DR extends linear DR

- Object manages I_O , I_V , and **sensing history** \mathbb{S}
- Second update condition:
$$\exists s_i \in \mathbb{S} \text{ with } |\overline{I_O s_C}(s_i.t) - s_i.p| > \epsilon$$
- New prediction starts at **last** sensed position

CDR_m limits computational cost by $|\mathbb{S}| \leq m$

- Compression approach to prevent periodic updates

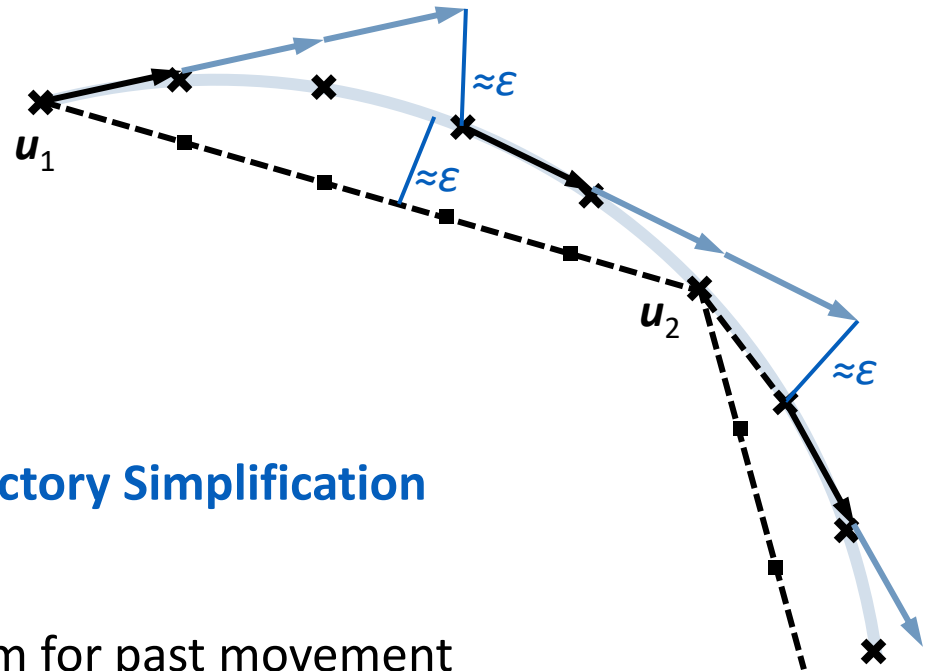


Tracking: GRTS

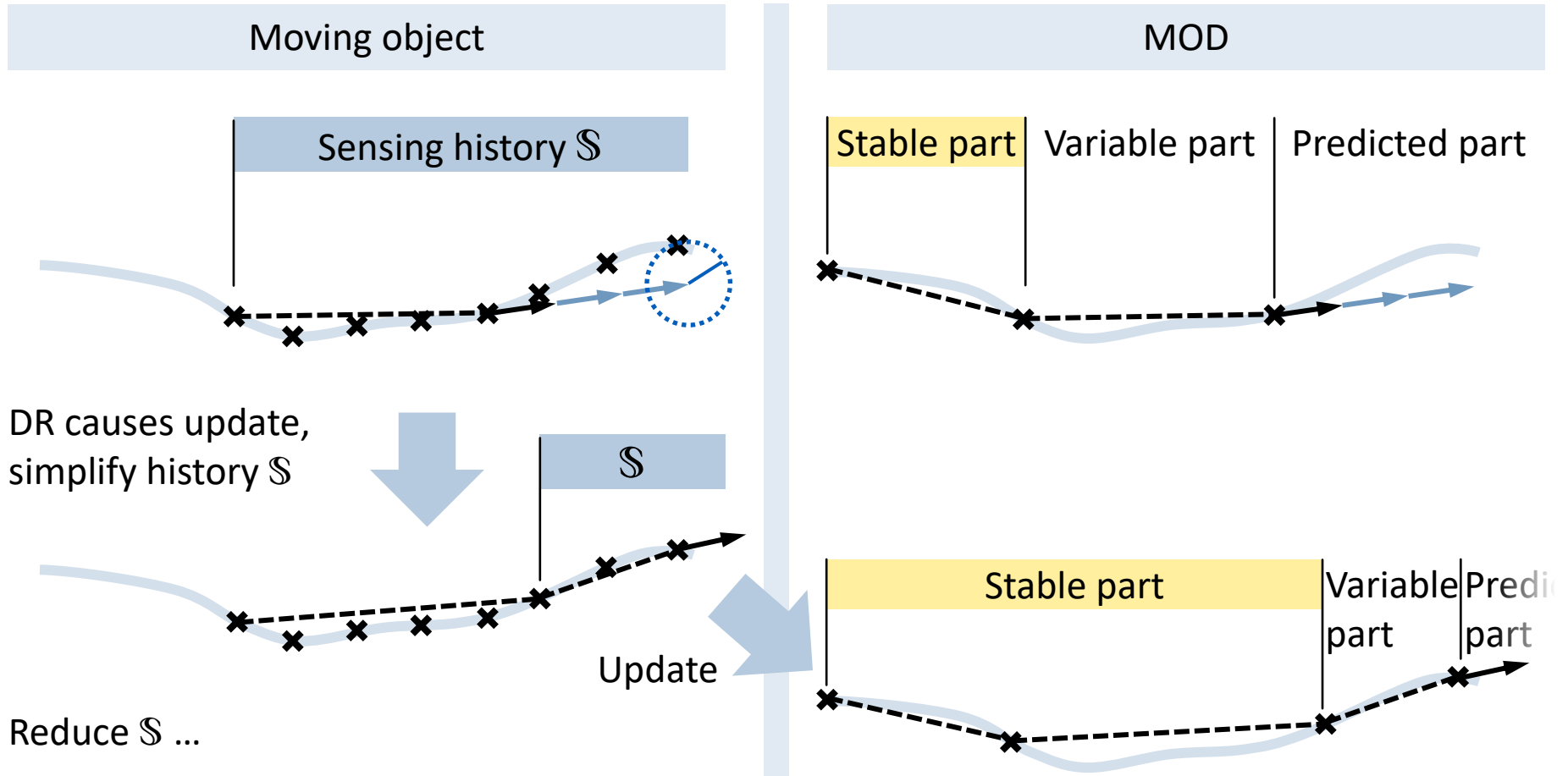
Tracking and simplification are **different concerns**

Basic approach of **Generic Remote Trajectory Simplification**

- DR to report latest movement
 - Arbitrary line simplification algorithm for past movement
 - Computational cost \leftrightarrow reduction efficiency
- ➔ Tracking and simplification must be synchronized!



Tracking: GRTS Protocol



Tracking: GRTS k/m Variants

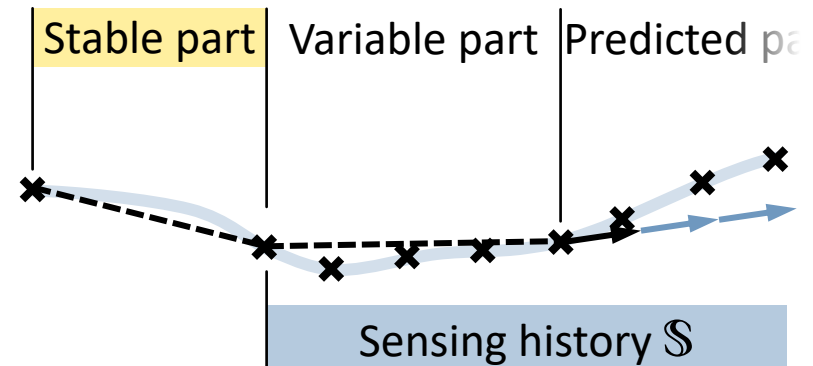
So far, not defined how to reduce \mathcal{S} ...

GRTS_k limits variable part to k line sections

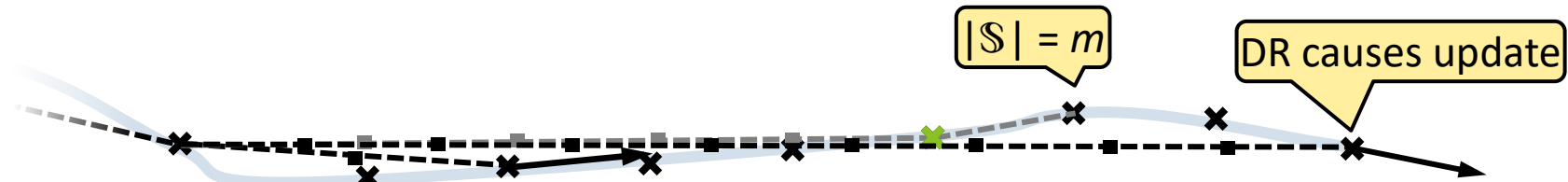
- Unlimited computational costs
- Only of theoretical interest

GRTS_m limits $|\mathcal{S}|$ to m

- Limits computational costs – essential for real-time behavior
- Adds vertex to $\mathbf{u}(t)$ at least every m sensing operations
- Compress \mathcal{S} if its size exceeds m



Tracking: GRTS mc Variant



s_2 with attributes
 $p = (48.42^\circ, 9.01^\circ)$
 $t = 2009-12-15\ 14:25:17.3$
 $\delta = 7m$

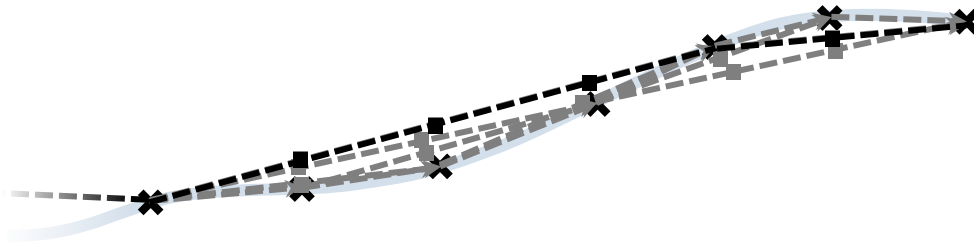
GRTS_{mc}

- $s_j \cdot \delta$ gives **maximum deviation** along line section from s_{i-1} to s_i
- Number of compressed positions should be small ($c = 1$ or 2)
- With uncompressed positions, $s_j \cdot \delta$ may represent varying δ_{sense}

Tracking: $\text{GRTS}_*^{\text{Sec}}$ and $\text{GRTS}_*^{\text{Opt}}$

$\text{GRTS}_*^{\text{Opt}}$ – with optimal simplification algorithm [Imai and Iri 1988]

- Reduces simplification to shortest-path problem



- Segmentation by k or m influences reduction efficiency

$\text{GRTS}_*^{\text{Sec}}$ – with Section Heuristic [e.g. Meratnia and de By 2004]

- Online algorithm enables per-sense simplification
- Proposed improved version, optimizing $|\mathcal{S}|$

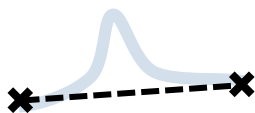
Tracking: Evaluation Setup

Comparing CDR variants and GRTS* realizations to ...

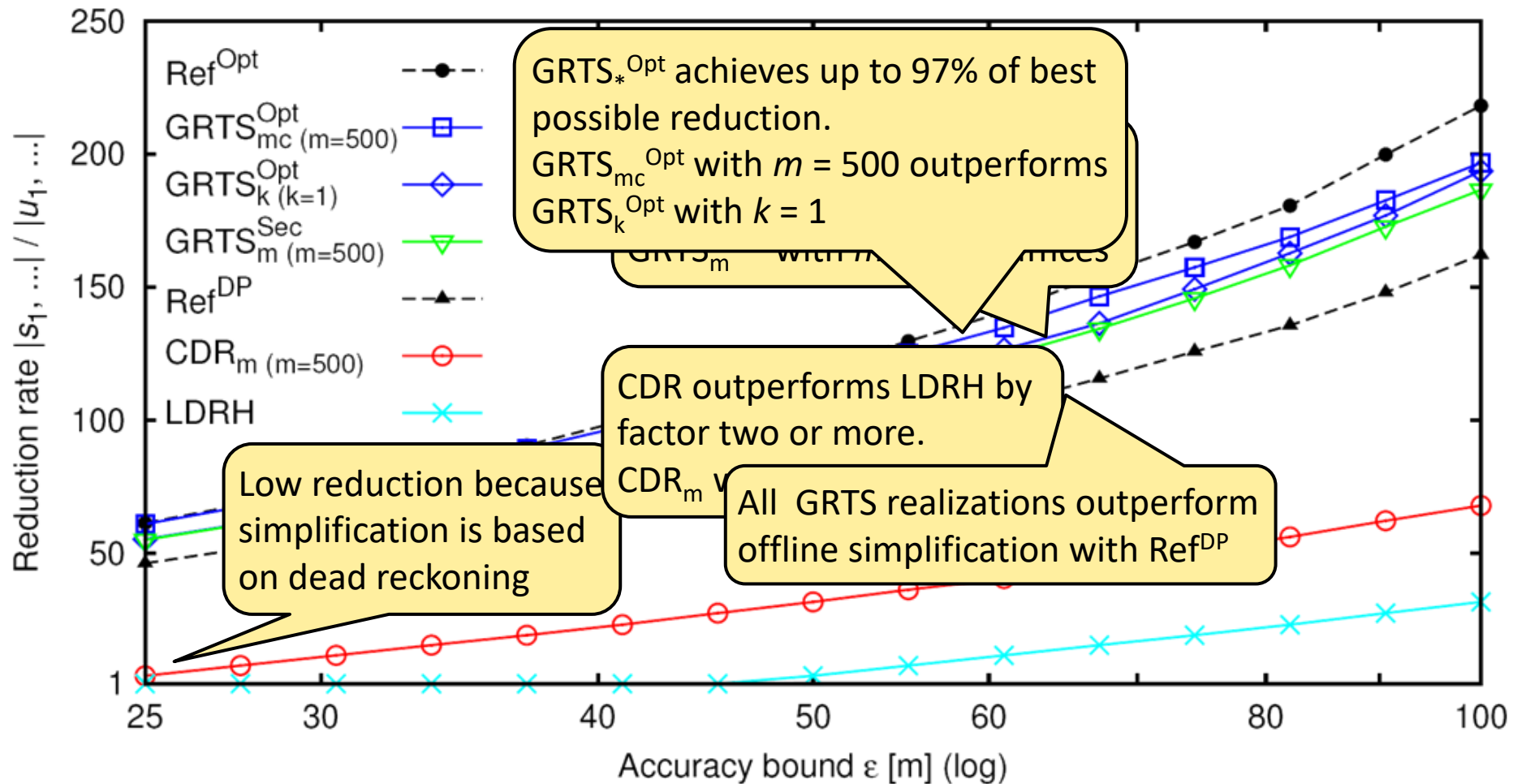
- Linear DR with $\frac{1}{2}\varepsilon$ (LDRH)
- Optimal offline simplification (Ref^{Opt})
- Douglas-Peucker algorithm (Ref^{DP})

Simulated with **real** GPS traces from OpenStreetMap

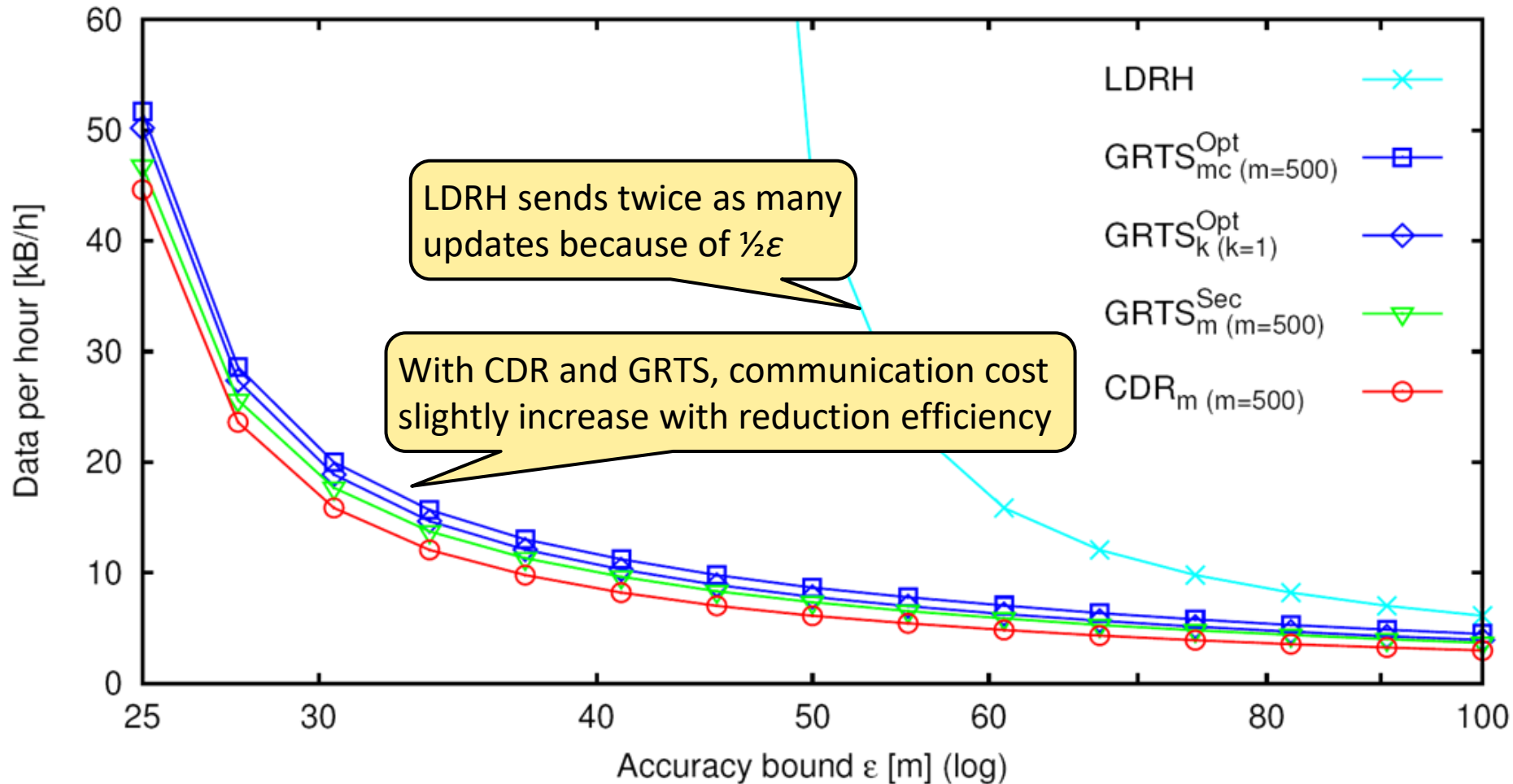
- **> 1.2 million** sensed positions, i.e. **> 330h** of data
- Sensor inaccuracy of $\delta_{\text{sense}} = 7.8 \text{ m}$
- Sensing period $T_{\text{sense}} = 1 \text{ s}$
- Maximum acceleration $a_{\text{max}} = 10 \text{ m/s}^2$


$$\delta_{\text{sense}} + \frac{1}{2} a_{\text{max}} \left(\frac{T_{\text{sense}}}{2} \right)^2 = 9.1 \text{ m}$$

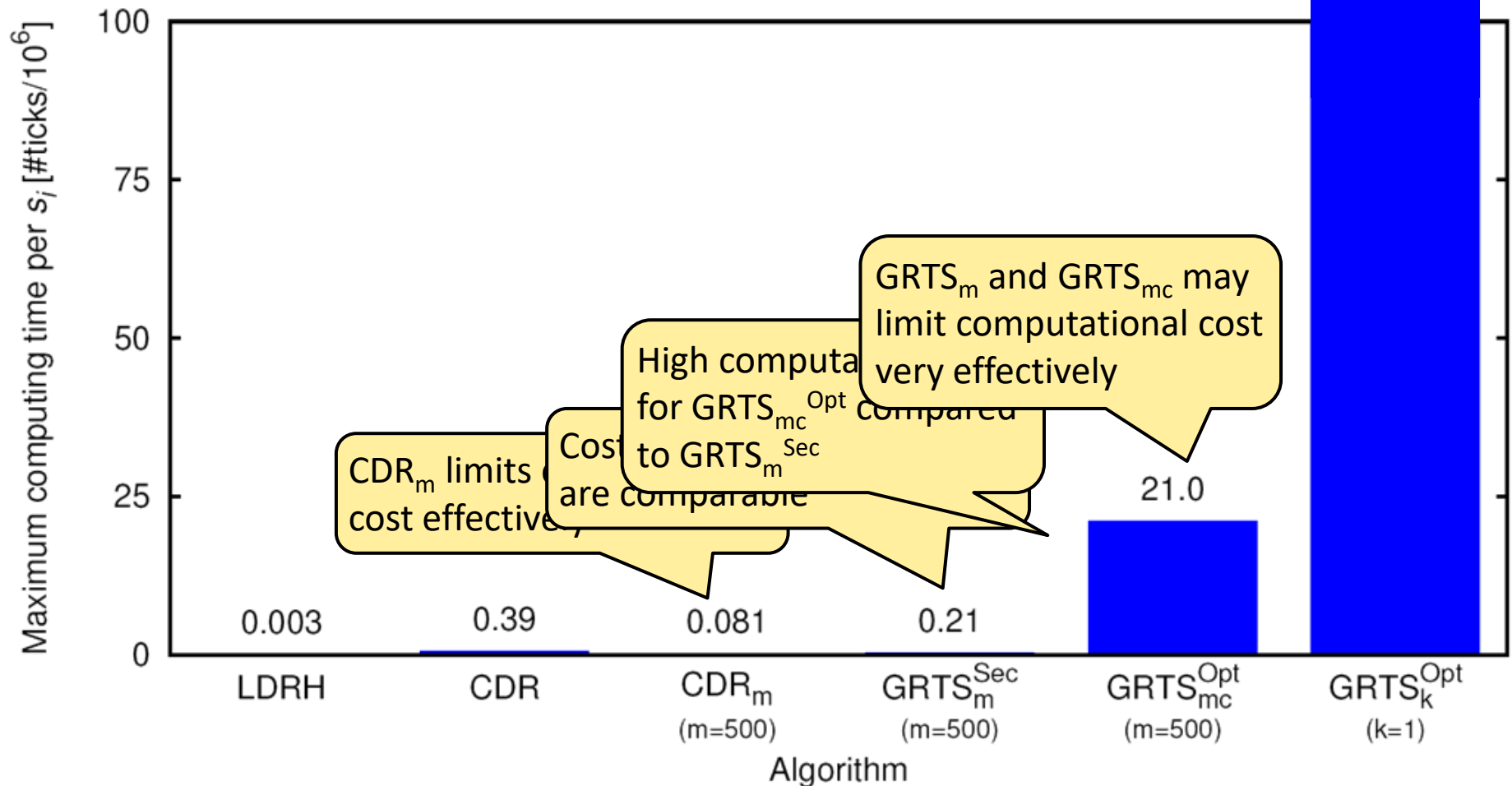
Tracking: Reduction Efficiency



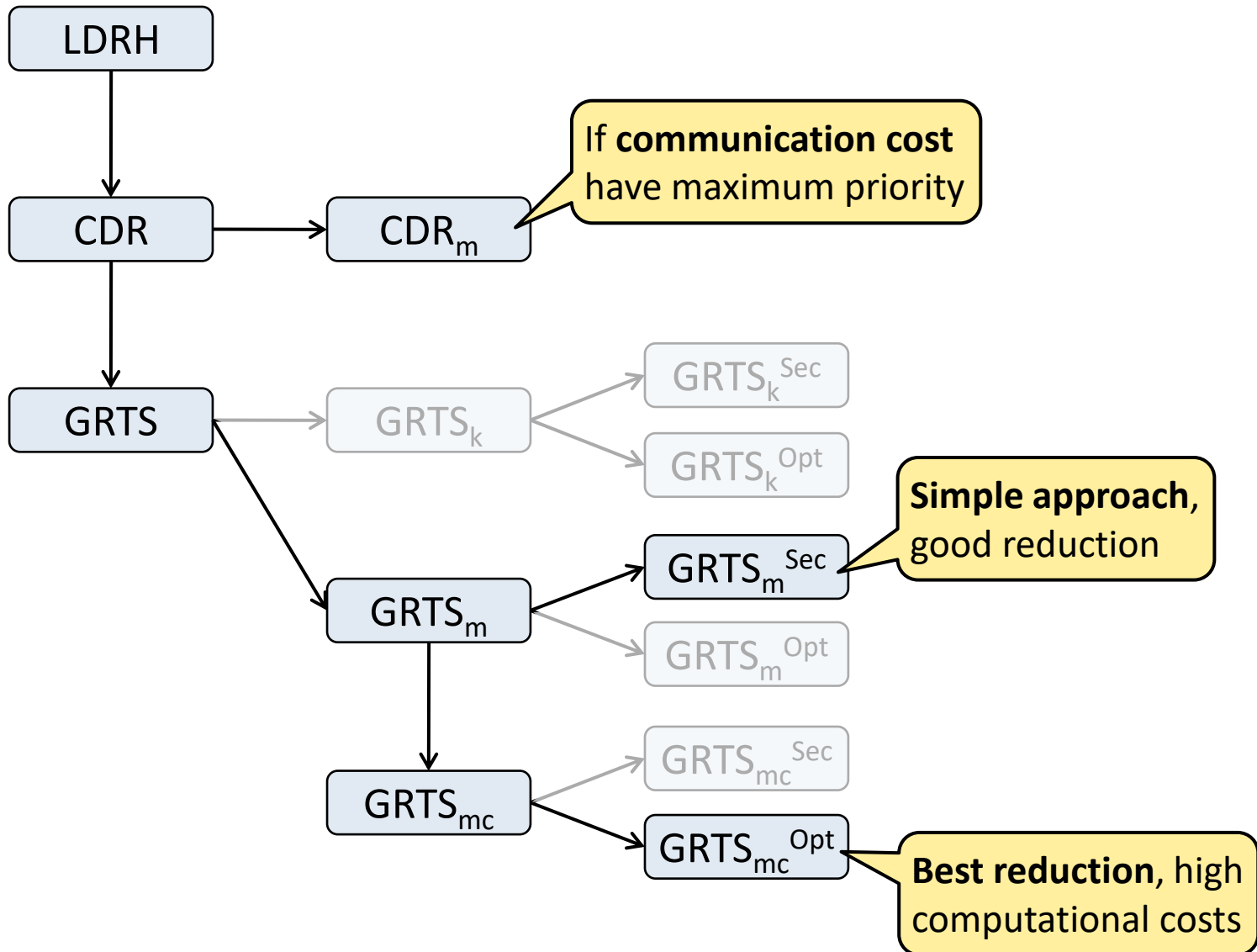
Tracking: Communication



Tracking: Computing Time

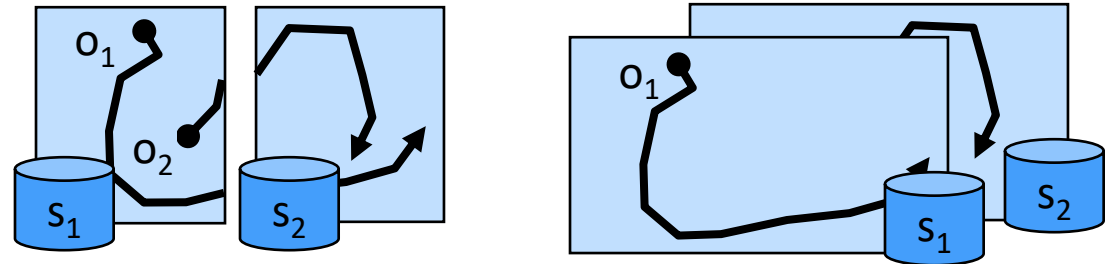


Tracking: Conclusion



Distributed Indexing of Trajectories

Managing large number of mobile objects requires **distributed** MOD



	Spatial partitioning	Object-based partitioning
Coordinate-based queries	✓	-
Trajectory-based queries	?	✓
Update-aware distribution	✓	

How to determine and route to relevant servers efficiently?

- **Coordinate-based** query: “Which objects were located in R during $[t_1, t_2]$?”
- **Trajectory-based** query: “What distance covered object o_2 during $[t_3, t_4]$?”

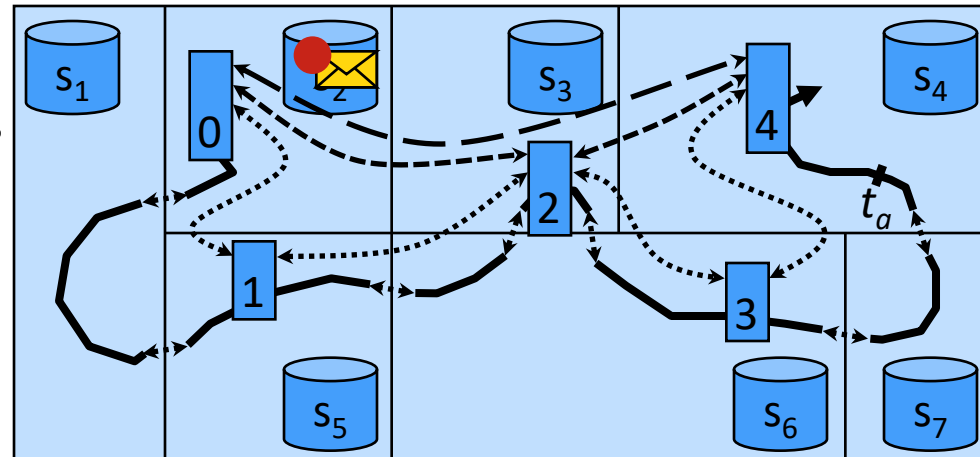
Distributed Indexing of Trajectories (2)

Trajectory-based routing along queried trajectory can take many hops

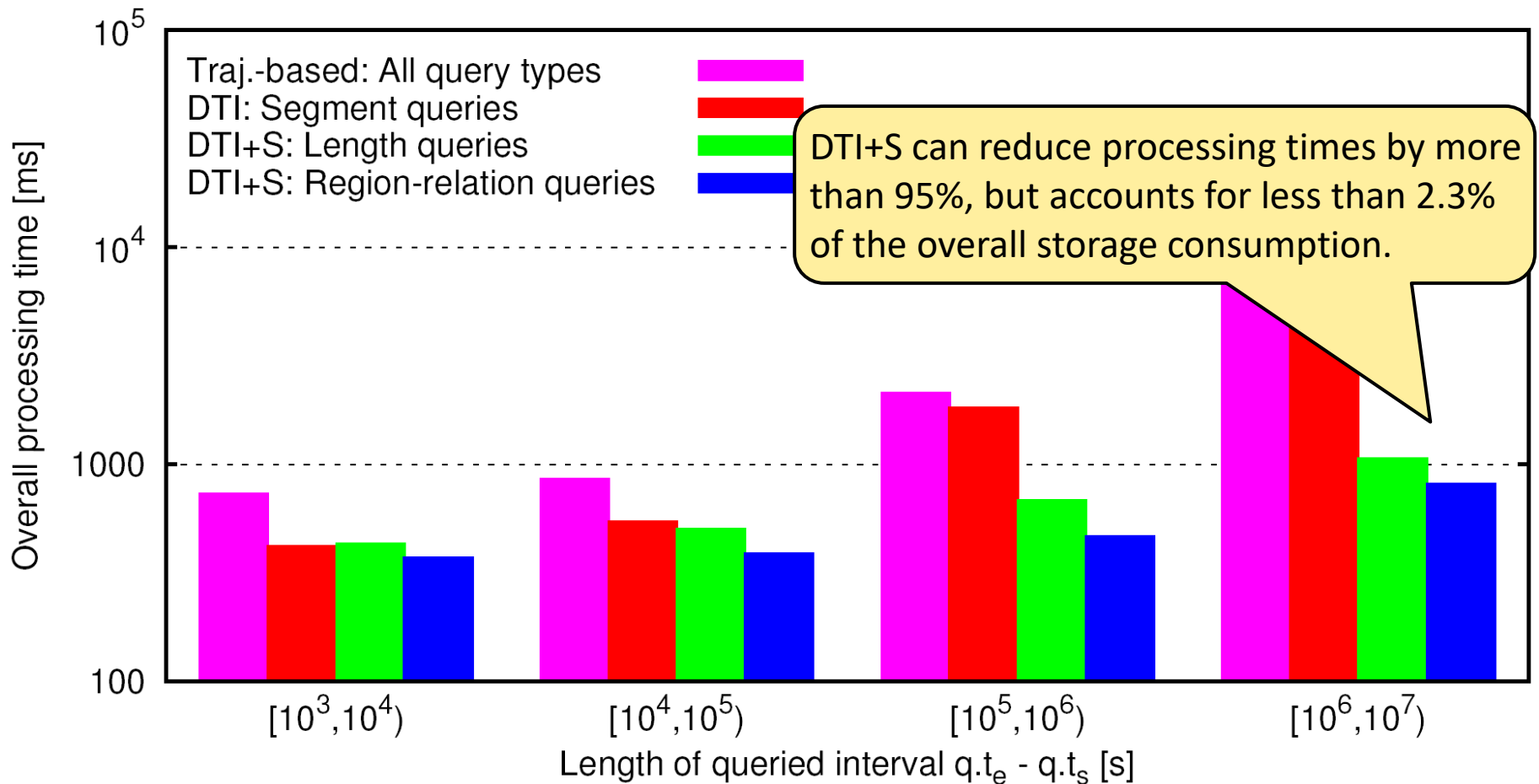
- Depends on queried time span and trajectory route

Solution: **Distributed Trajectory Index** (DTI)

- DTI scheme creates independent index (overlay) for each trajectory
- Enables efficient routing by pointers to temporally distant positions
- **DTI+S** also stores **summaries** (speed, MBR, ...) on segments
- Linear storage consumption



Distributed Indexing of Trajectories (3)



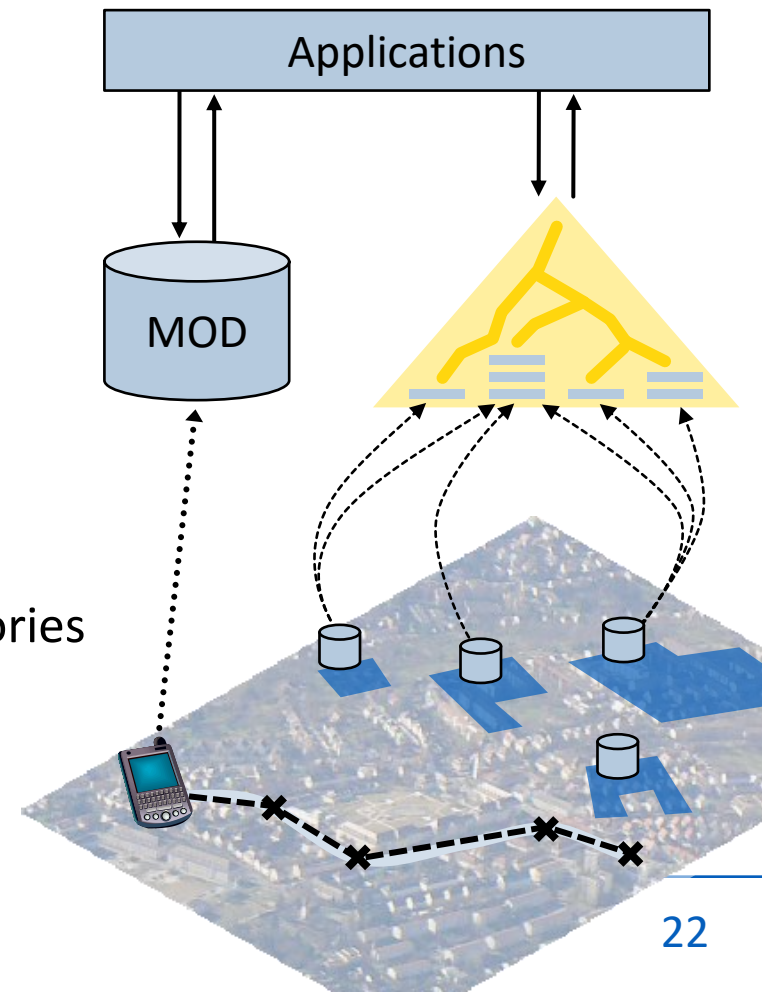
Summary

Proliferation of sensing technologies constitutes huge potential for context-aware applications

How to provide efficient access to the immense amounts of distributed dynamic context information



1. CDR and GRTS for real-time trajectory tracking
2. DTI for scalable distributed indexing of trajectories
3. Formalism and SDC-Tree for describing and indexing of context models (*see thesis*)



Thank you for your attention!

Selected Publications:

- Ralph Lange, Frank Dürr, Kurt Rothermel: “Indexing Source Descriptions based on Defined Classes”. In: *Proc. of 14th IDEAS*. Montreal, QC, Canada. Aug 2010.
 - Ralph Lange, Frank Dürr, Kurt Rothermel : “Efficient Tracking of Moving Objects using Generic Remote Trajectory Simplification”. In: *Proc. of 8th PerCom Workshops*. Mannheim, Germany. Mar 2010.
 - Ralph Lange, Harald Weinschrott, Lars Geiger, André Blessing, Frank Dürr, Kurt Rothermel, Hinrich Schütze: “On a Generic Uncertainty Model for Position Information”. In: *Proc. of 1st QuaCon*. Stuttgart, Germany. Jun 2009.
 - Ralph Lange, Tobias Farrell, Frank Dürr, Kurt Rothermel: “Remote Real-Time Trajectory Simplification”. In: *Proc. of 7th PerCom*. Galveston, TX, USA. Mar 2009.
 - Ralph Lange, Frank Dürr, Kurt Rothermel: “Scalable Processing of Trajectory-Based Queries in Space-Partitioned Moving Objects Databases”. In: *Proc. of 16th ACM GIS*. Irvine, CA, USA. Nov 2008.
 - Ralph Lange, Frank Dürr, Kurt Rothermel: “Online Trajectory Data Reduction using Connection-preserving Dead Reckoning”. In: *Proc. of 5th MobiQuitous*. Dublin, Ireland. Jul 2008.
 - Tobias Farrell, Ralph Lange, Kurt Rothermel: “Energy-efficient Tracking of Mobile Objects with Early Distance-based Reporting”. In: *Proc. of 4th MobiQuitous* . Philadelphia, PA, USA. Aug 2007.
- See www.lange-ralph.de for complete list