Manage the Data from Indoor Spaces: Models, Indexes & Query Processing

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Overview

- 1. Outlines
- 2 2. Indoor Space Models & Applications
- 3. Indoor Data Cleansing
- 4. Indoor Movement Analysis
- 5. Appendix

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Aims

- To give a brief review introduction to *indoor data* management techniques.
- To review a series of works in this field, including their proposed models, indexes and algorithms.
- To discuss how to bring those advanced theoretical contents into practice.

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About This Work...

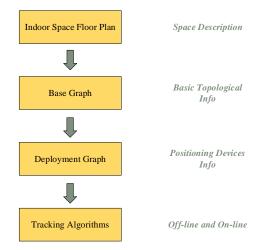
Graph Model Based Indoor Tracking. [2] C. S. Jensen, H. Lu, and B. Yang.

- Published in year 2009, MDM conference.
- A pioneering work that introduces base graph model to indoor data management.
- Detailed tracking algorithms are designed for RFID-based positioning.
- Easy to understand, with comprehensive concepts.

Motivation

- We are spending most of our time in indoor spaces
 - Office building, University, Shopping Centers, etc.
- We cannot use GPS-based tracking indoor movements
 - Indoor navigation and route guidance (museum)
 - Flow analysis
 - \bullet how do people use the indoor space \to important in pricing of advertisement space in store rental
- We can use other technology...
 - Wi-Fi, Infrared, Bluetooth or RFID
 - This paper is focusing on RFID, since it is now mature and effortless
 - RFID tags are cheap and RFID reader are expensive

Idea

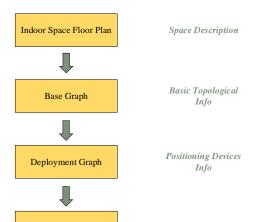


Off-line and On-line

2.1 Graph Model Based Indoor Tracking

Tracking Algorithms

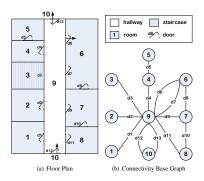
Idea



Goal: Improve indoor tracking accuracy from a data management perspective, to capture where a particular object can be at a particular time.

Base Graph Model

By capturing the essential connectivity and accessibility, **Base Graph** describes the topology of a floor plan of a possibly complex indoor space.

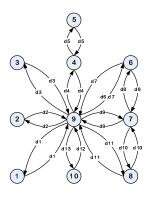


Connectivity Base Graph

- a labeled, undirected graph.
 - $G_{conn} = \{V, E_d, \Sigma_{door}\}$
 - V: each separate partition is represented as a vertex
 - E_d : each door is captured as an edge
 - Σ_{door} : a set of edge labels that represent connections

Base Graph Model

Accessibility Graph is constructed to represent the movement permitted by doors or connections.



Accessibility Graph

a labeled, directed graph.

•
$$G_{accs} = \{V, E, \Sigma_{door}, l_e\}$$

• V: the set of vertices

• E: the set of directed edges, i.e.,
$$E = \{ \langle v_i, v_j \rangle | v_i, v_j \in V \land v_i \neq v_j \}$$

• l_e : a function that maps edges to subsets of the set of doors, i.e., $l_e: E \to 2^{\Sigma_{door}}$

Base Graph Model

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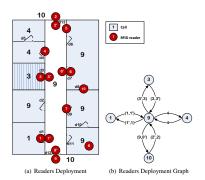
$$Doors: \Sigma_{door} \to Line\ Segments$$
 (2)

RFID Deployment Graph Model

- RFID based proximity analysis
 - RFID readers deployment may cover only part of the space, or it may be capable of only detecting some movements in the space.
 - assume that all RFID readers have disjoint activation ranges.
- Types of RFID readers
 - Partitioning Readers partition the indoor space into cells in the sense that an object cannot move from one cell to another without being observed.
 - **Presence Readers** simply observe the presence(and non-presence) of tags in their activation ranges.

RFID Deployment Graph Model

Vertices represent cells. A directed edge indicates that one can move from one cell to another without entering other cells, which is detected by a corresponding partitioning reader.



RFID Deployment Graph

- a labeled, directed graph.
 - $G_{RFID} = \{C, E_r, \Sigma_{reader}, l_e\}$
 - C: the set of the vertices
 - E_r : An edge is an ordered pair $\langle c_i, c_j \rangle$ of distinct vertices from C
 - l_e maps an edge to a partitioning reader (pair), i.e.,

$$E_r \to 2^{\Sigma_{reader}} \cup 2^{\Sigma_{reader} \times \Sigma_{reader}}$$

RFID Deployment Graph Construction

Each cell created by partitioning readers corresponds to one or more base graph partitions.

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Mapping
$$1: \Sigma_{reader} \to \{(loc, range, flag) | loc \in R^2 \land range \in (0, d_{max}] \land flag \in PAR, PRE\}$$
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A mapping of readers to the cells that their activation ranges intersect is introduced:

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Mapping 2:
$$\Sigma_{reader} \to 2^C$$
 (5)

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- 2.2 Scalable Continuous Range Monitoring of Moving Objects in Symbolic Indoor Space

Motivation

Scalable Continuous Range Monitoring of Moving Objects in Symbolic Indoor Space. [3]

2.3 Probabilistic Threshold k Nearest Neighbor Queries over Moving Objects in Symbolic Indoor Space

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Probabilistic Threshold k Nearest Neighbor Queries over Moving Objects in Symbolic Indoor Space. [4]

2.4 Spatio-temporal Joins on Symbolic Indoor Tracking Data

Spatio-temporal Joins on Symbolic Indoor Tracking Data

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2.5 A Foundation for Efficient Indoor Distance-aware Query Processing

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2.6 Efficient Distance-aware Query Evaluation on Indoor Moving Objects

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Efficient Distance-aware Query Evaluation on Indoor Moving Objects. [2]

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The End. Thanks:)