

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

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March 13, 2016

Overview

1. Outlines
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.

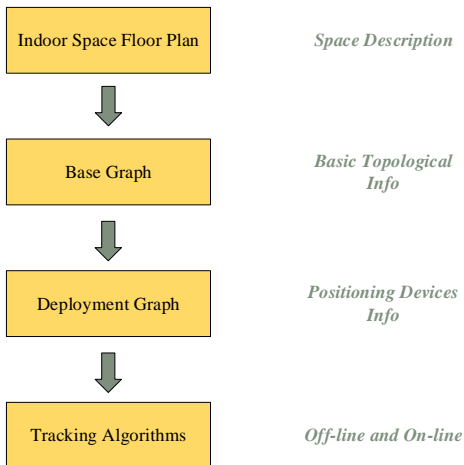
- Published in year 2009, *MDM* conference.
- A pioneering work that introduces base graph model to indoor data management.
- Detailed tracking algorithms are designed for RFID-based positioning.
- Easy to understand, with comprehensive concepts.

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
 - how do people use the indoor space → 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

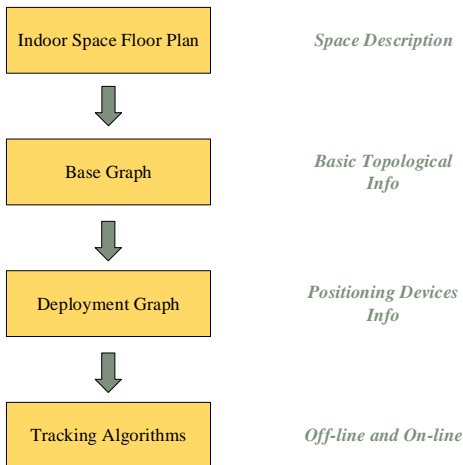
2.1 Graph Model Based Indoor Tracking

Idea



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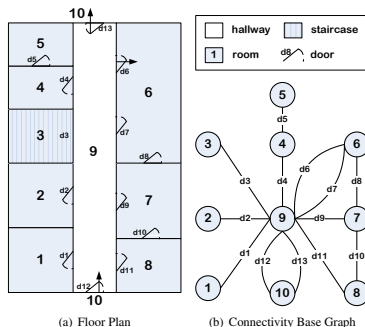


Goal: Improve indoor tracking accuracy from a data management perspective, to capture where a particular object can be at a particular time.

2.1 Graph Model Based Indoor Tracking

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

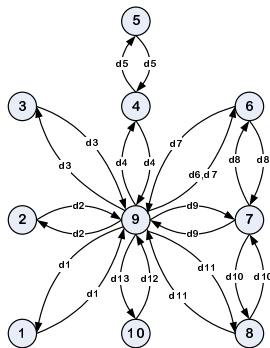
a labeled, undirected graph.

- $G_{conn} = \{V, E_d, \Sigma_{door}\}$
- 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

2.1 Graph Model Based Indoor Tracking

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.,
 $E = \{\langle v_i, v_j \rangle | v_i, v_j \in V \wedge v_i \neq v_j\}$
- l_e : a function that maps edges to subsets of the set of doors, i.e.,
 $l_e : E \rightarrow 2^{\Sigma_{door}}$

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The *Doors Mapping* is defined as:

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

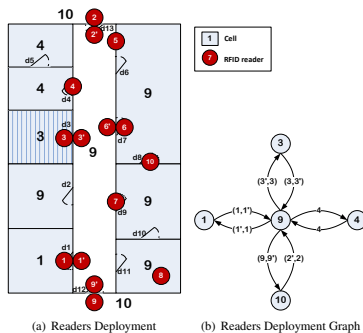
RFID Deployment Graph Model

- 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.

2.1 Graph Model Based Indoor Tracking

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.,
 $E_r \rightarrow 2^{\Sigma_{reader}} \cup 2^{\Sigma_{reader} \times \Sigma_{reader}}$

RFID Deployment Graph Construction

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

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$$\begin{aligned} \text{Mapping } 1 : \Sigma_{reader} \rightarrow \{ & (loc, range, flag) | loc \in R^2 \wedge \\ & range \in (0, d_{max}] \wedge flag \in PAR, PRE \} \end{aligned} \quad (4)$$

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

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

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

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

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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Efficient Distance-aware Query Evaluation on Indoor Moving Objects

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The End. Thanks :)