

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

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# 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.

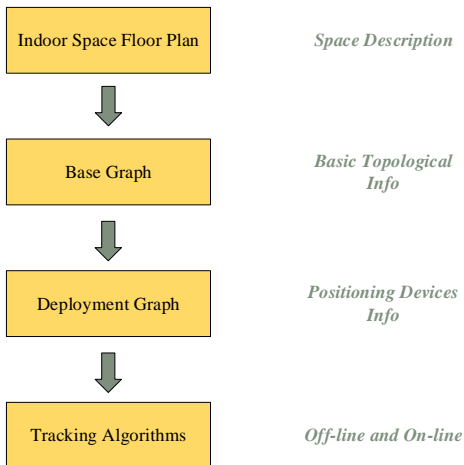
- Published in year 2009, *MDM* conference.
- A pioneering work that introduces base graph model to indoor data management.
- Detailed tracking algorithms are designed for RFID-based positioning.
- Easy to understand, with comprehensive concepts.

# 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

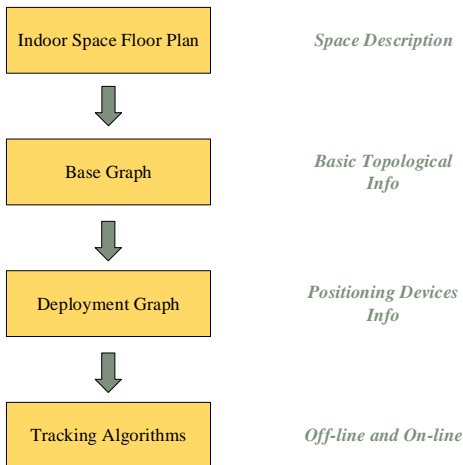
# Idea





## 2.1 Graph Model Based Indoor Tracking

# Idea

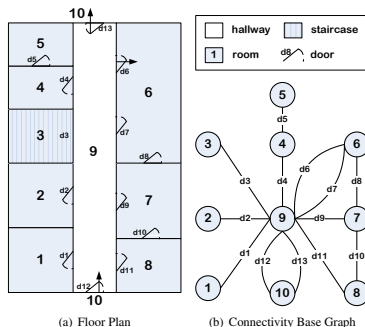


**Goal:** Improve indoor tracking accuracy from a data management perspective, to capture where a particular object can be at a particular time.

## 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

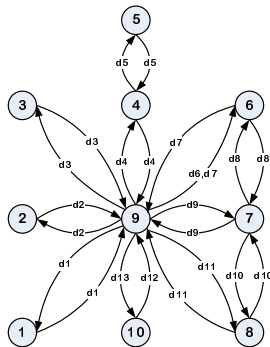
a labeled, undirected graph.

- $G_{conn} = \{V, E_d, \Sigma_{door}\}$
- $V$ : each separate partition is represented as a vertex
- $E_d$ : each door is captured as an edge
- $\Sigma_{door}$ : 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:

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The *Doors Mapping* is defined as:

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



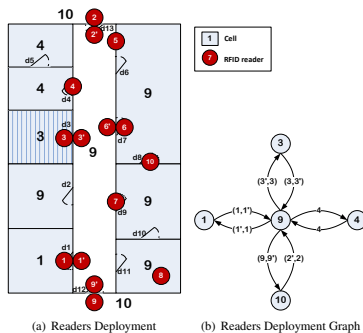
# RFID Deployment Graph Model

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



# Motivation

Scalable Continuous Range Monitoring of Moving Objects in  
Symbolic Indoor Space. [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. [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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## 2.6 Efficient Distance-aware Query Evaluation on Indoor Moving Objects

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## References



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The End. Thanks :)