

# 6D SLAM with Cached kd-tree Search

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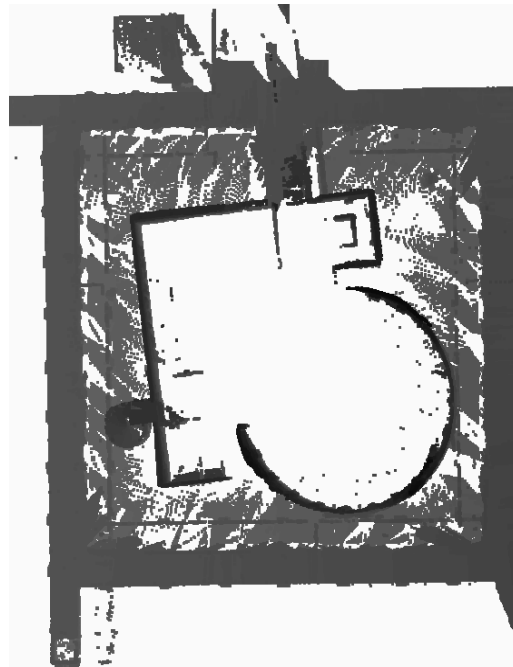
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Institute of Computer Science

Knowledge Based Systems Research Group

Joint work with Kai Lingemann, Joachim Hertzberg





This talk is about 3D mapping using six degree of freedom.





# Contents

- State of the Art in Robotic Mapping
- 6D SLAM with Scan Matching
- Approximate and Exact Data Association
- 3D Mapping Examples



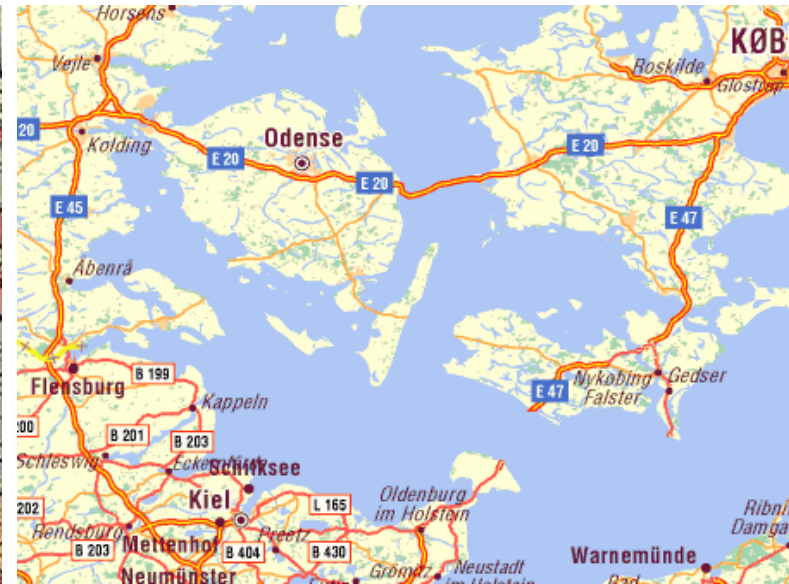
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# Simultaneous Localization and Mapping

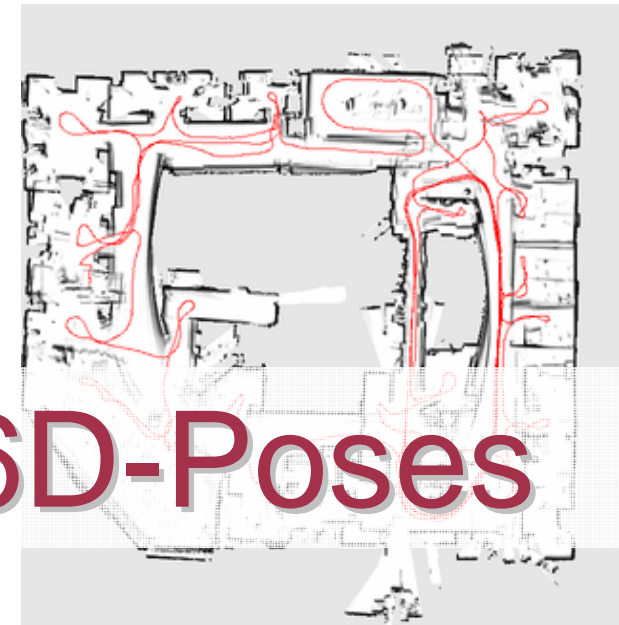
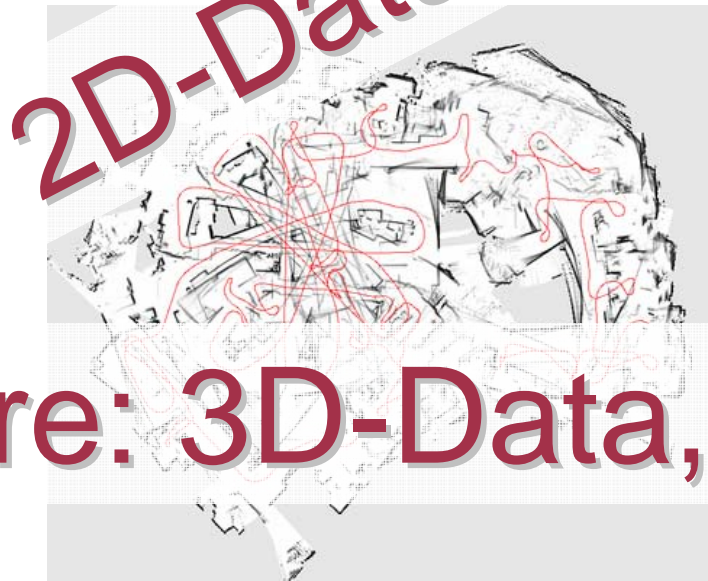
- If one knows the pose (position and orientation) of a mobile robot precisely, then the sensor readings can be used to build a map.
- Unfortunately, pose measurements are always imprecise ☹
- The pose of a robot is easy to compute from sensor readings, given a map.



# State of the Art in Robotic Mapping (1)



- Laser scanner are the state of the art sensors for metrical environment mapping
- Mapping based on scan matching (Lu, Milios)
- Probabilistic theory of mapping using uncertain motion and sensor models (Kalman-Filter, Maximum Likelihood Estimation, Expectation Maximization).

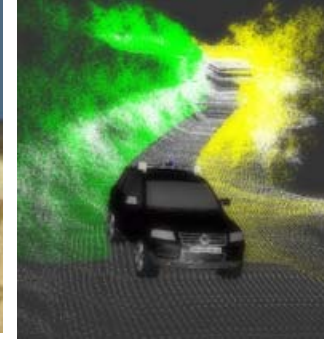


⇒ Here: 3D-Data, 6D-Poses

Related work: Freiburg (Burgard),  
UW (Fox),  
Stanford (Thrun)

# State of the Art in Robotic Mapping (2)

		Dimensionality of pose representation	
		3D	6D
Sensor data	2D		
	3D		





# The Mobile Robot Kurt3D



- Kurt3D is a lightweight (25 kg)
- Two 90W (200W) motors, 48 NiMH a 4500mAh, C167 Microcontroller, CAN Controller, Centrino Notebook

- Indoor/Outdoor versions available
- main Sensor:  
3D scanner ⇒ 3D data, 6D poses





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# The ICP-Algorithm

**Scan registration** Put two independent scans into one frame of reference

**Iterative Closest Point** algorithm [Besl/McKay 1992]

For prior point set  $M$  (“model set”) and data set  $D$

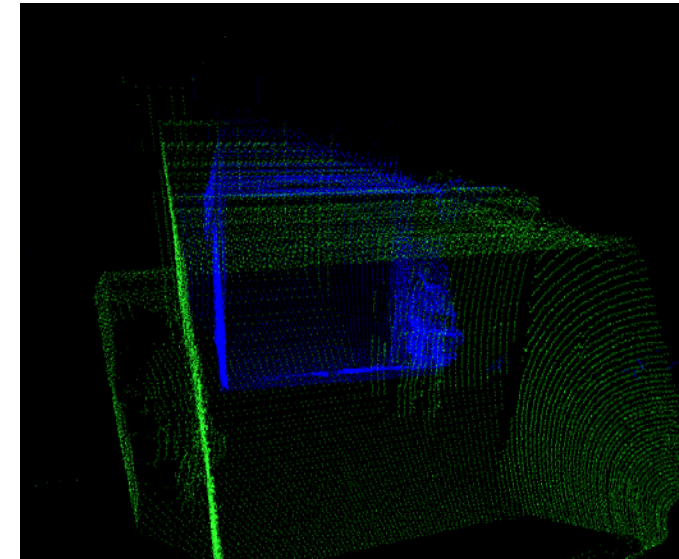
1. Select point correspondences  $w_{i,j}$  in  $\{0,1\}$
2. Minimize for rotation  $\mathbf{R}$ , translation  $\mathbf{t}$

$$E(\mathbf{R}, \mathbf{t}) = \sum_{i=1}^{N_m} \sum_{j=1}^{N_d} w_{i,j} \|\mathbf{m}_i - (\mathbf{R}\mathbf{d}_j + \mathbf{t})\|^2$$

3. Iterate 1. and 2.

SVD-based calculation of rotation

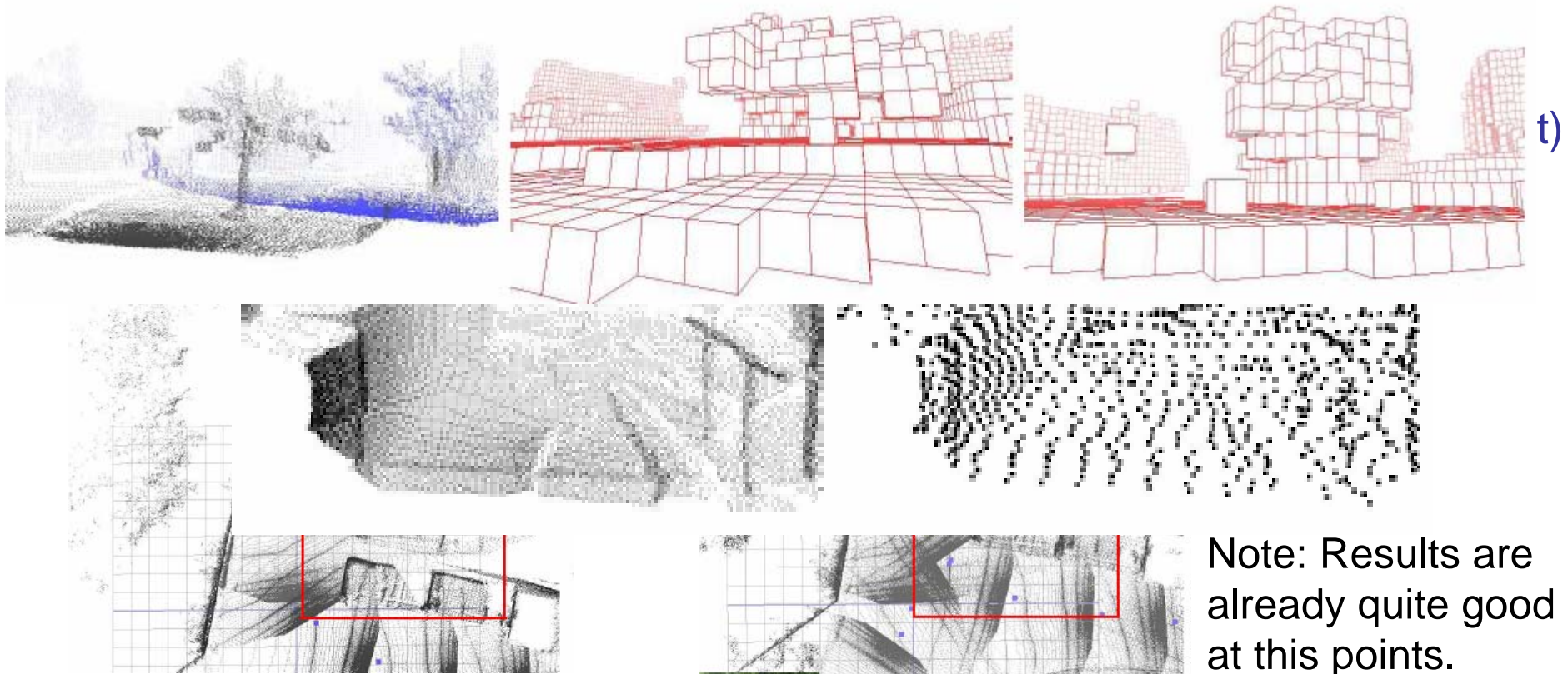
- works in 3 translation plus 3 rotation dimensions  
⇒ 6D SLAM with closed loop detection and global relaxation.





# 6D SLAM Algorithm Overview

1. Find a heuristic pose update using octrees [KI 2005].



4. Global relaxation.

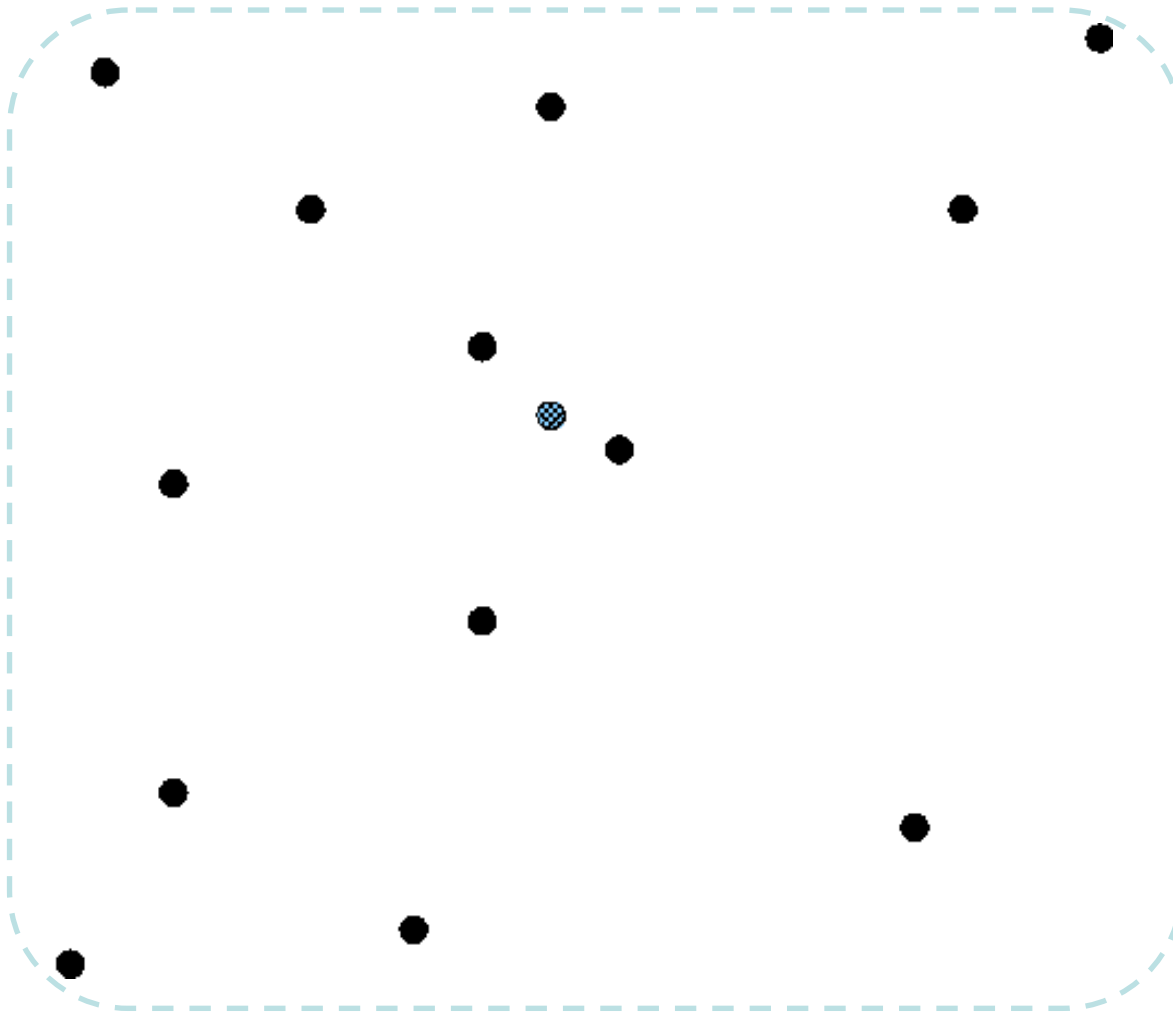


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- The 3D Laser Scanner and Kurt3D
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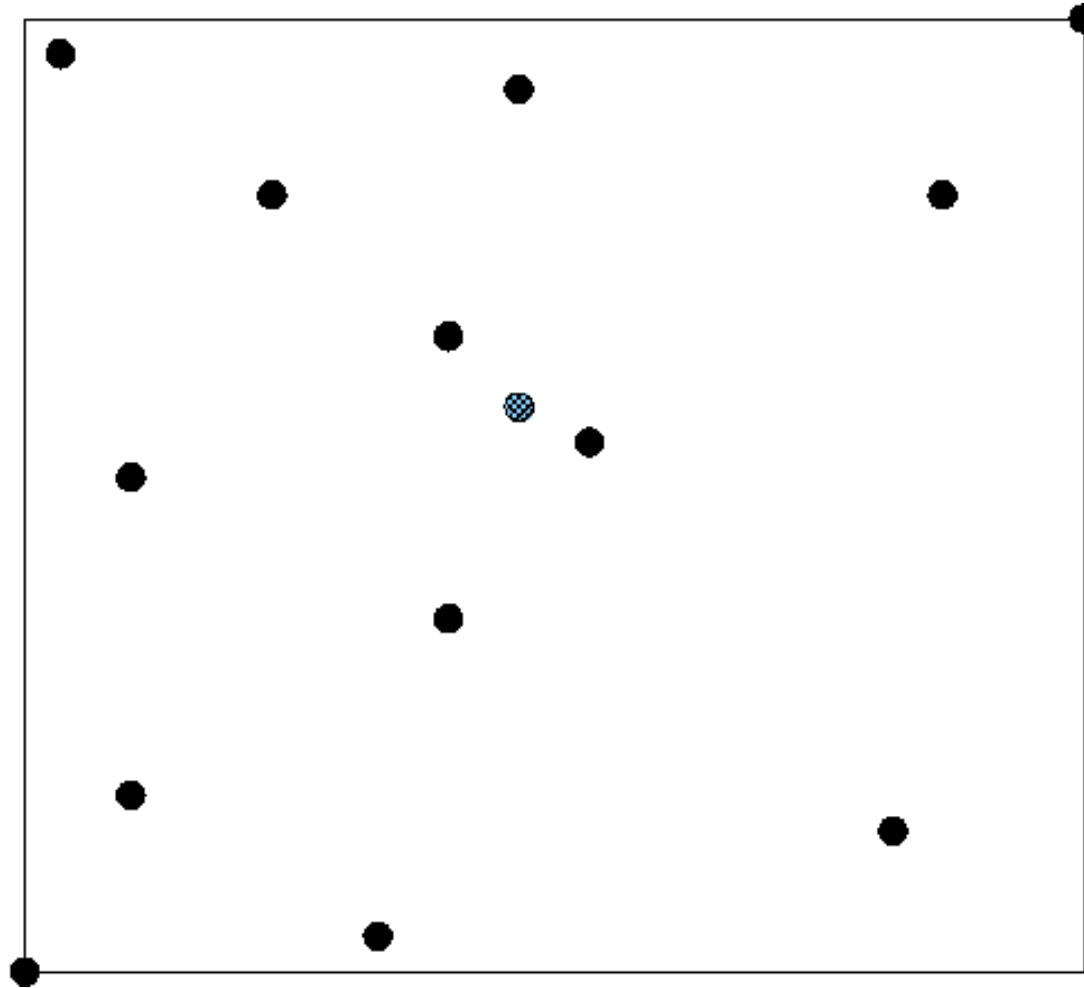
# Nearest Neighbor Search



- Given: Set of points.
- Building  $k$ D-tree for implementing a fast search,  **$\log n$**  time
- Idea: Recursively splitting the region into two equal partitions.

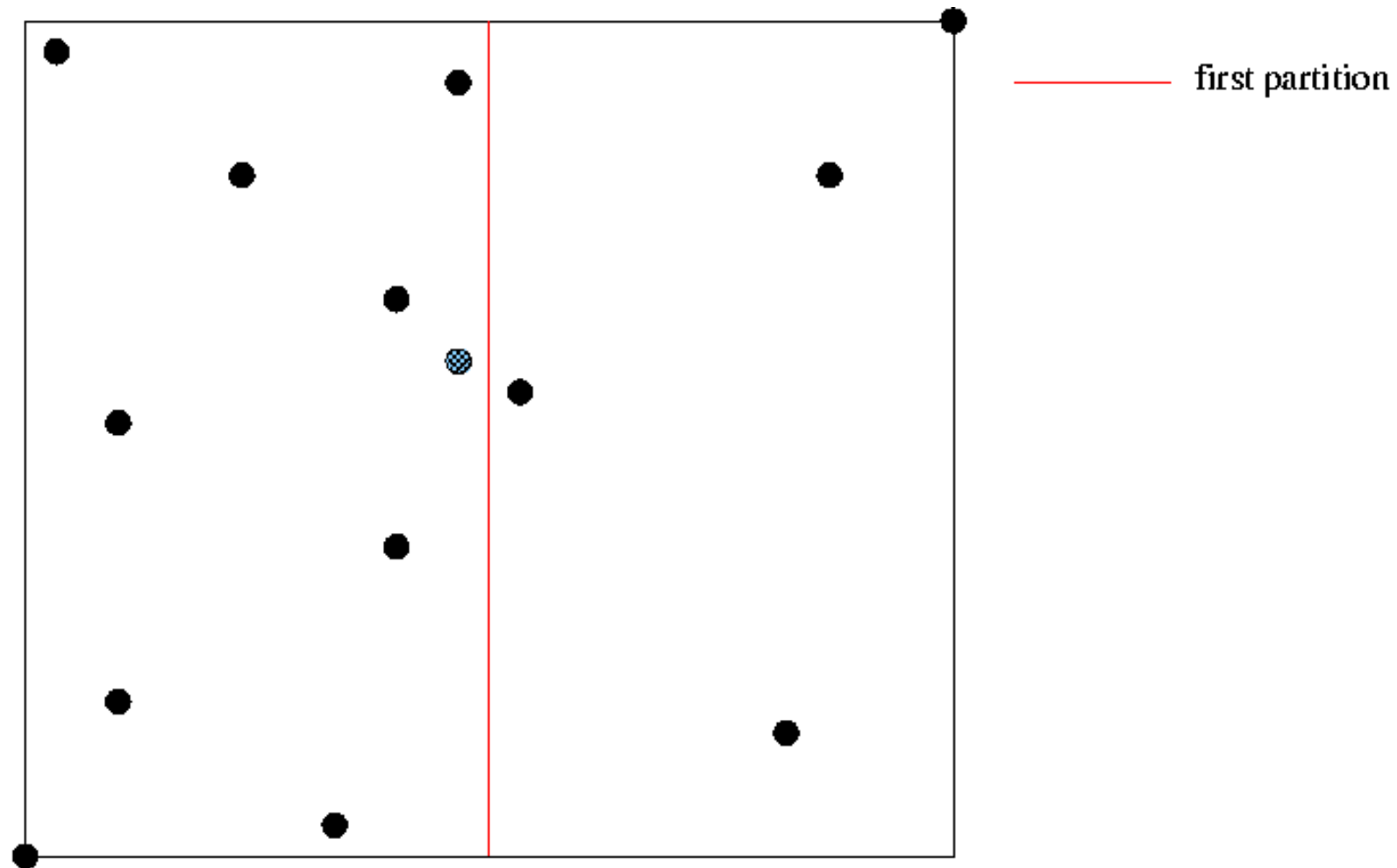
# Nearest Neighbor Search

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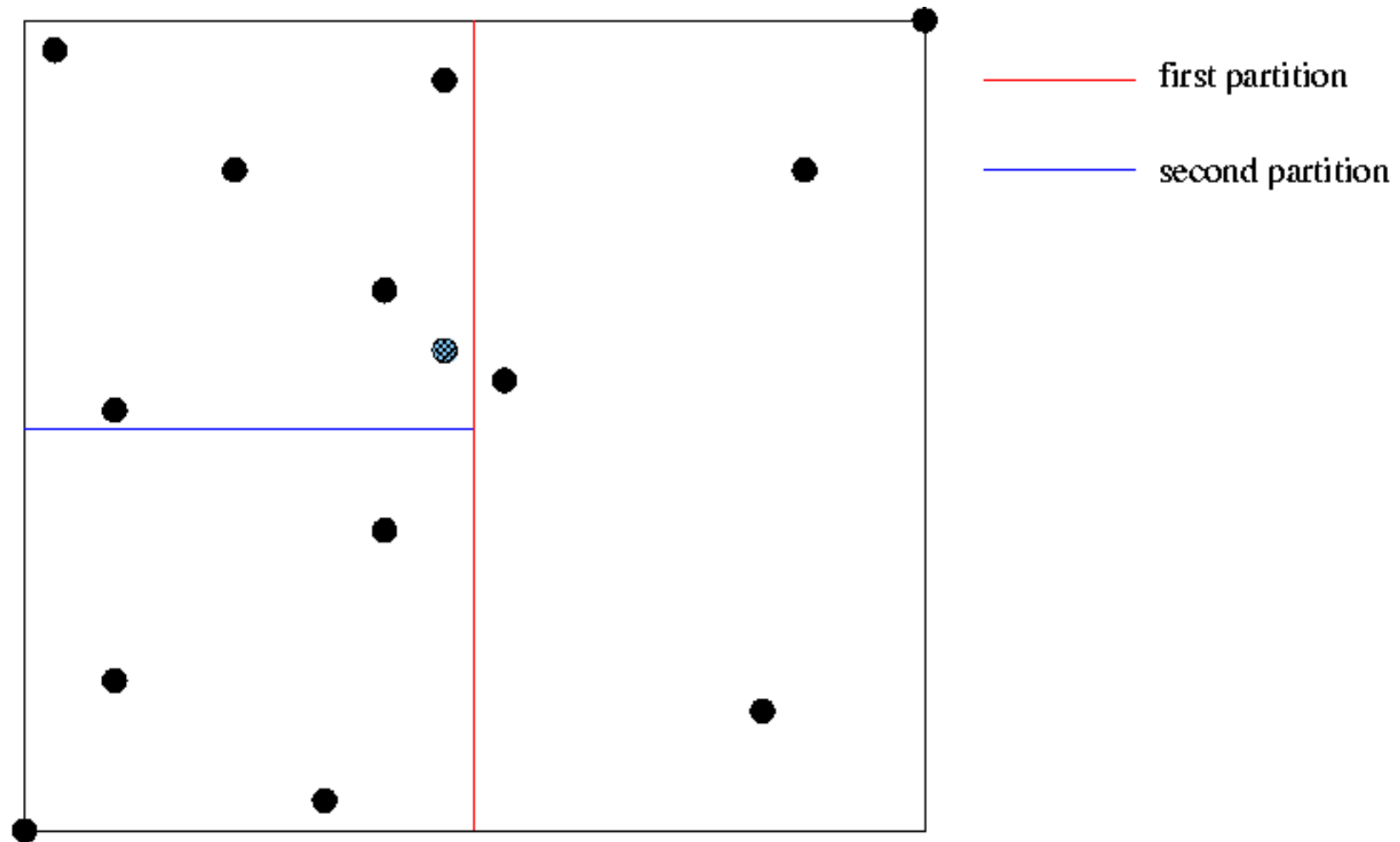




# Nearest Neighbor Search

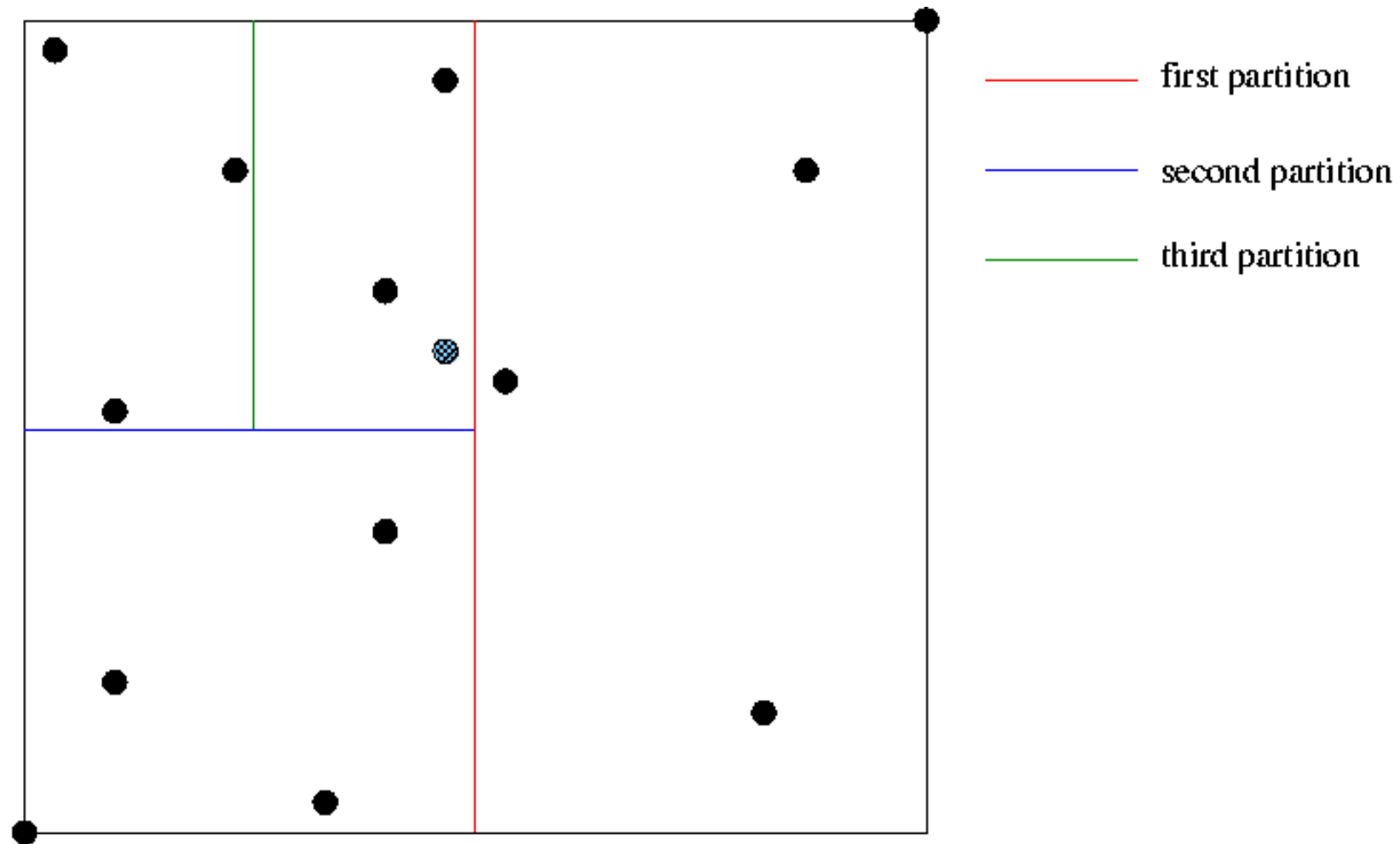


# Nearest Neighbor Search

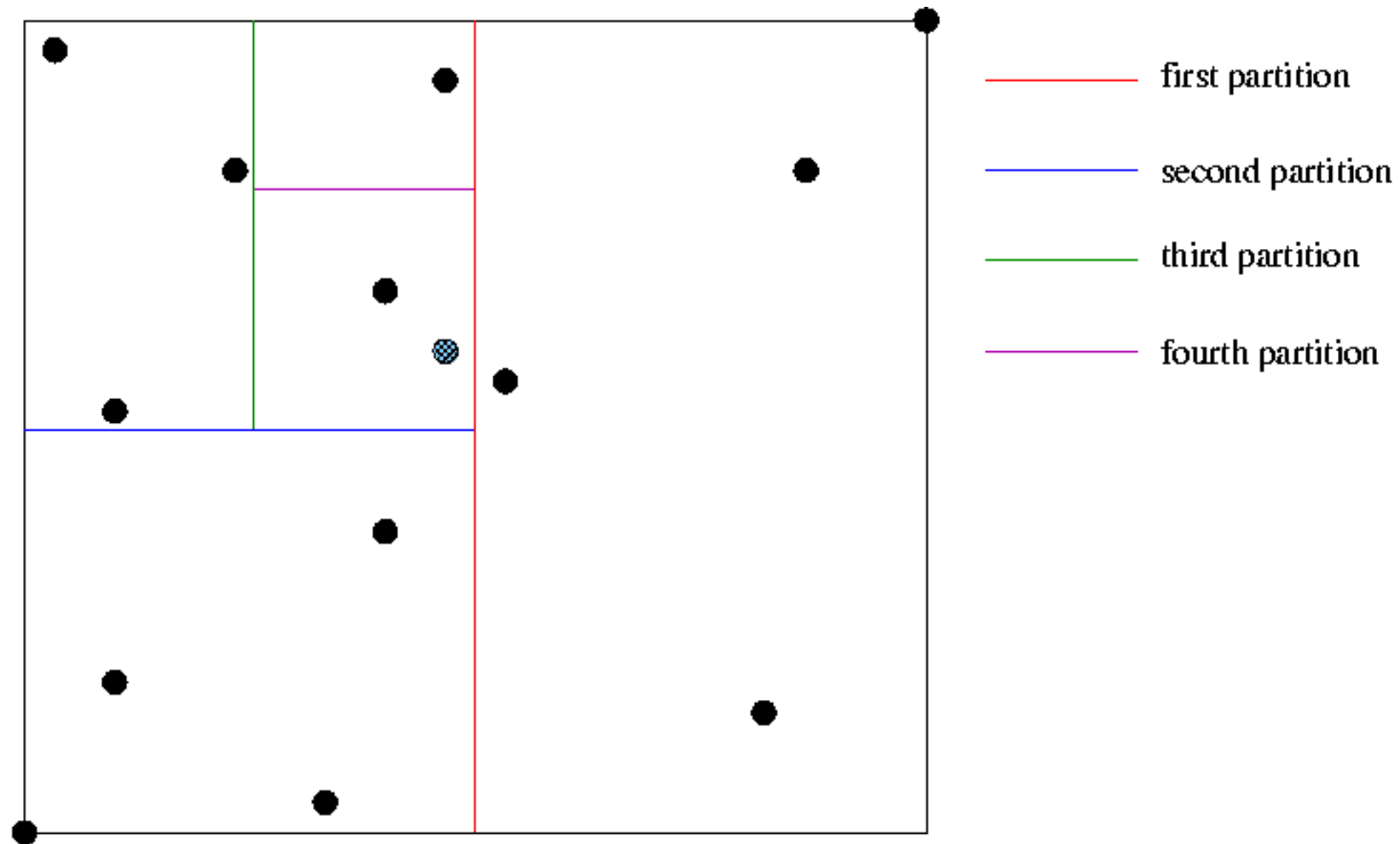




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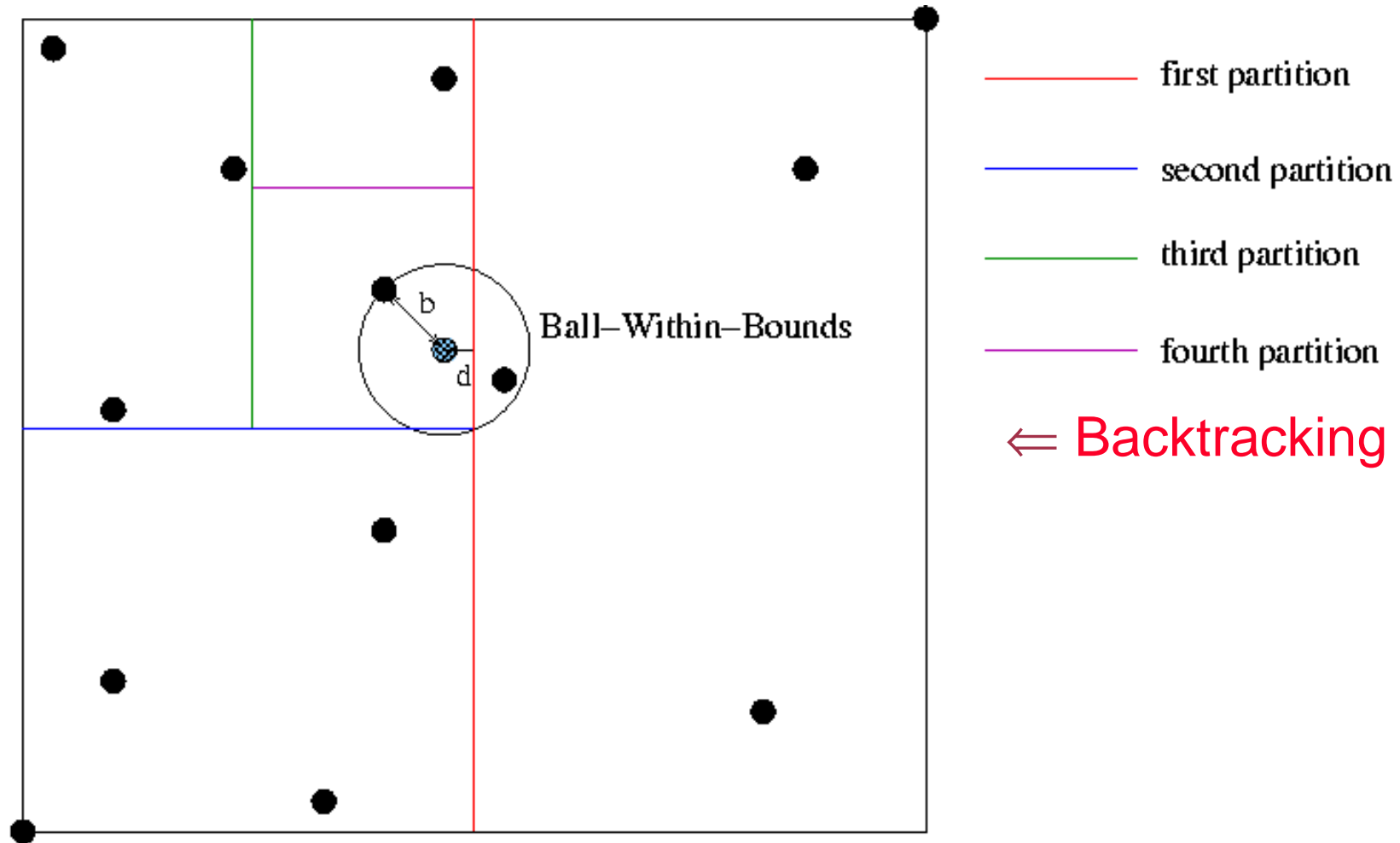


# Nearest Neighbor Search

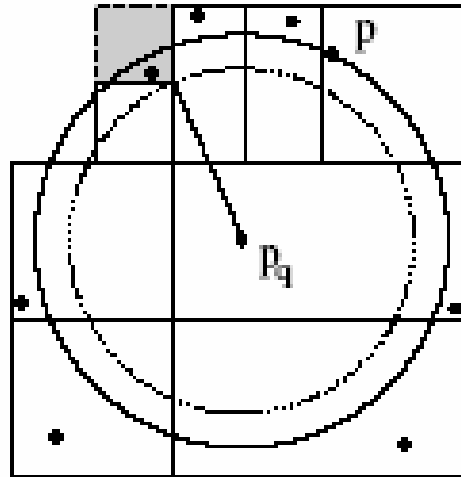




# Nearest Neighbor Search



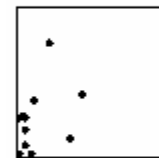
# Approximate Nearest Neighbor Search



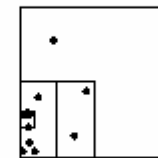
- Consider the query point  $p_q$
- The approximate search is discontinued if the distance to the unanalyzed leaves is larger than

$$\|p_q - p\| / (1 + \varepsilon)$$

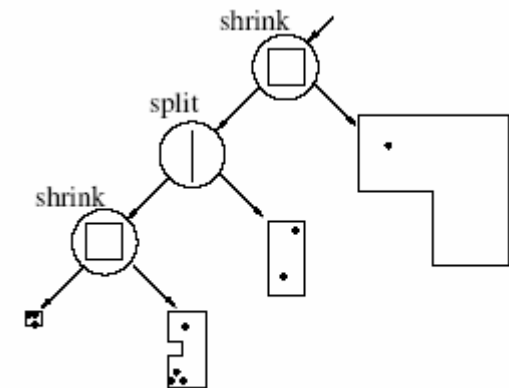
- Approximate closest point is the input for ICP
- Extreme case: Return closest point in leaf as closest point.
- Alternatively use box decomposition trees.



(a)



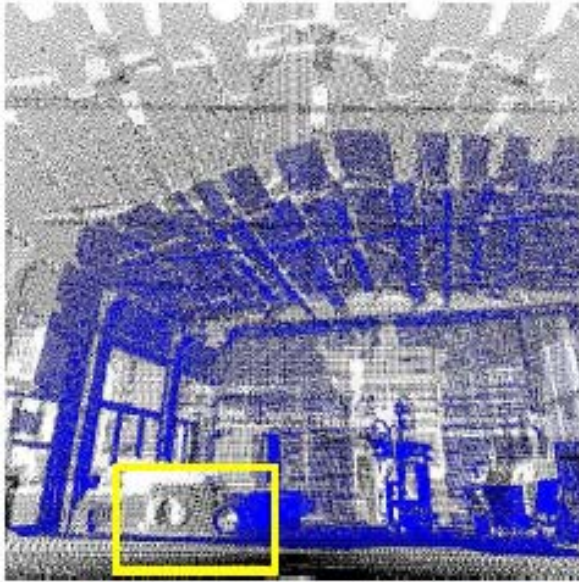
(b)



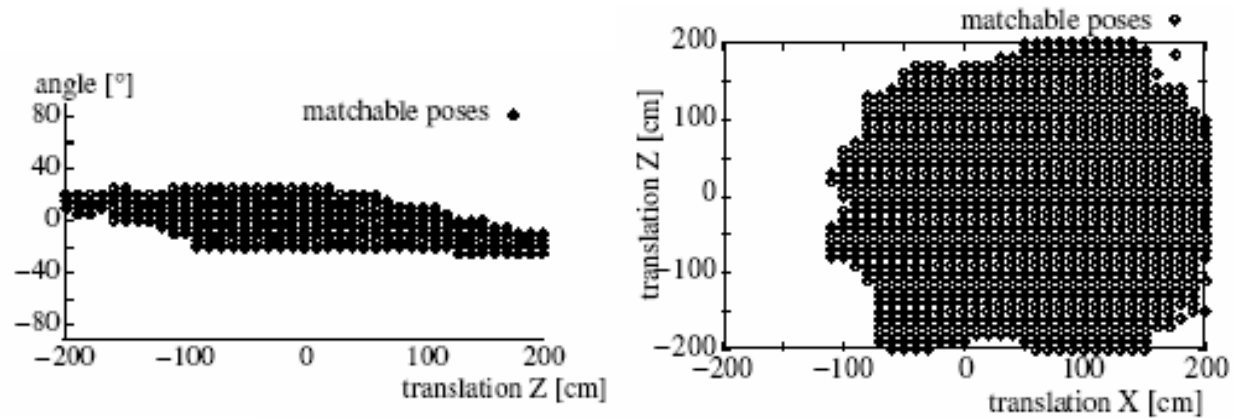
(c)



# Evaluation of the ICP

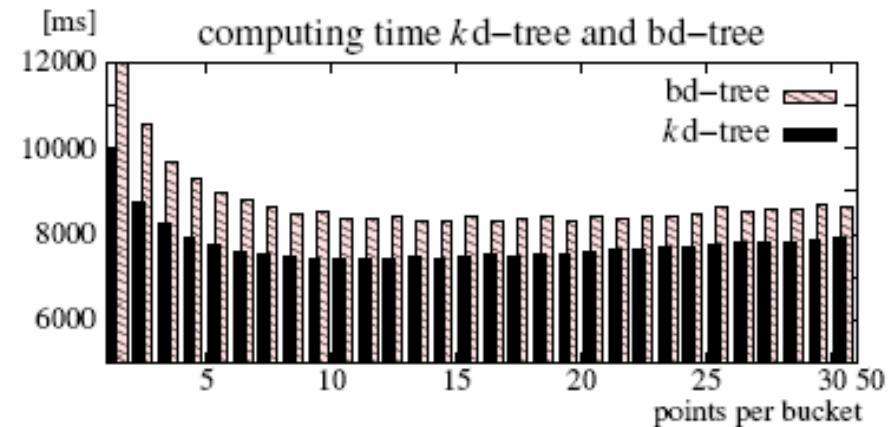


- Systematically record poses that lead to a correct match.

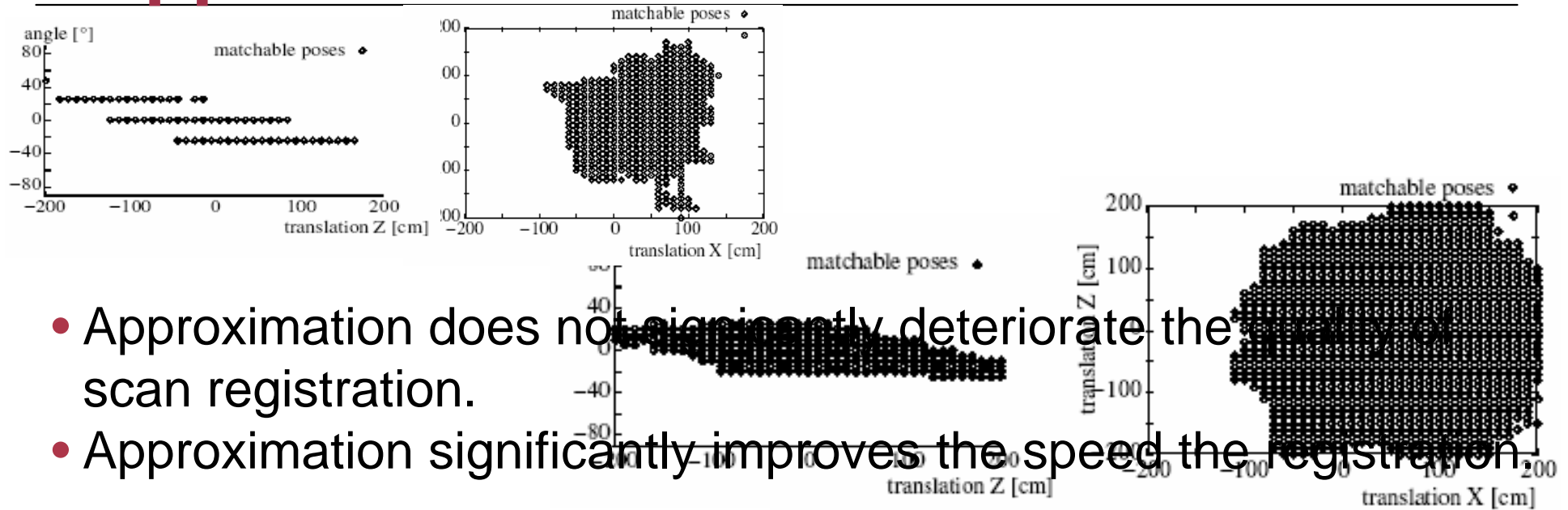


- Record scan matching time.

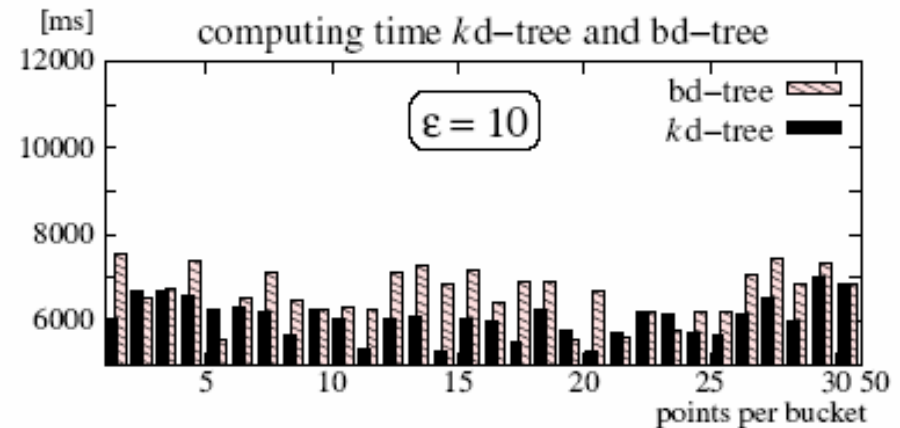
⇒ *kd*-trees outperform *bd*-trees



# Approximation Effects to ICP



- Approximation does not significantly deteriorate the quality of scan registration.
- Approximation significantly improves the speed of the registration.





# Cached kd-tree Search (1)

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- In scan registration approximation is often considered critical  $\Rightarrow$  exact method is needed

- Recall ICP:

For prior point set  $M$  (“model set”) and data set  $D$

1. Select point correspondences  $w_{i,j}$  in  $\{0,1\}$

2. Minimize for rotation  $\mathbf{R}$ , translation  $\mathbf{t}$

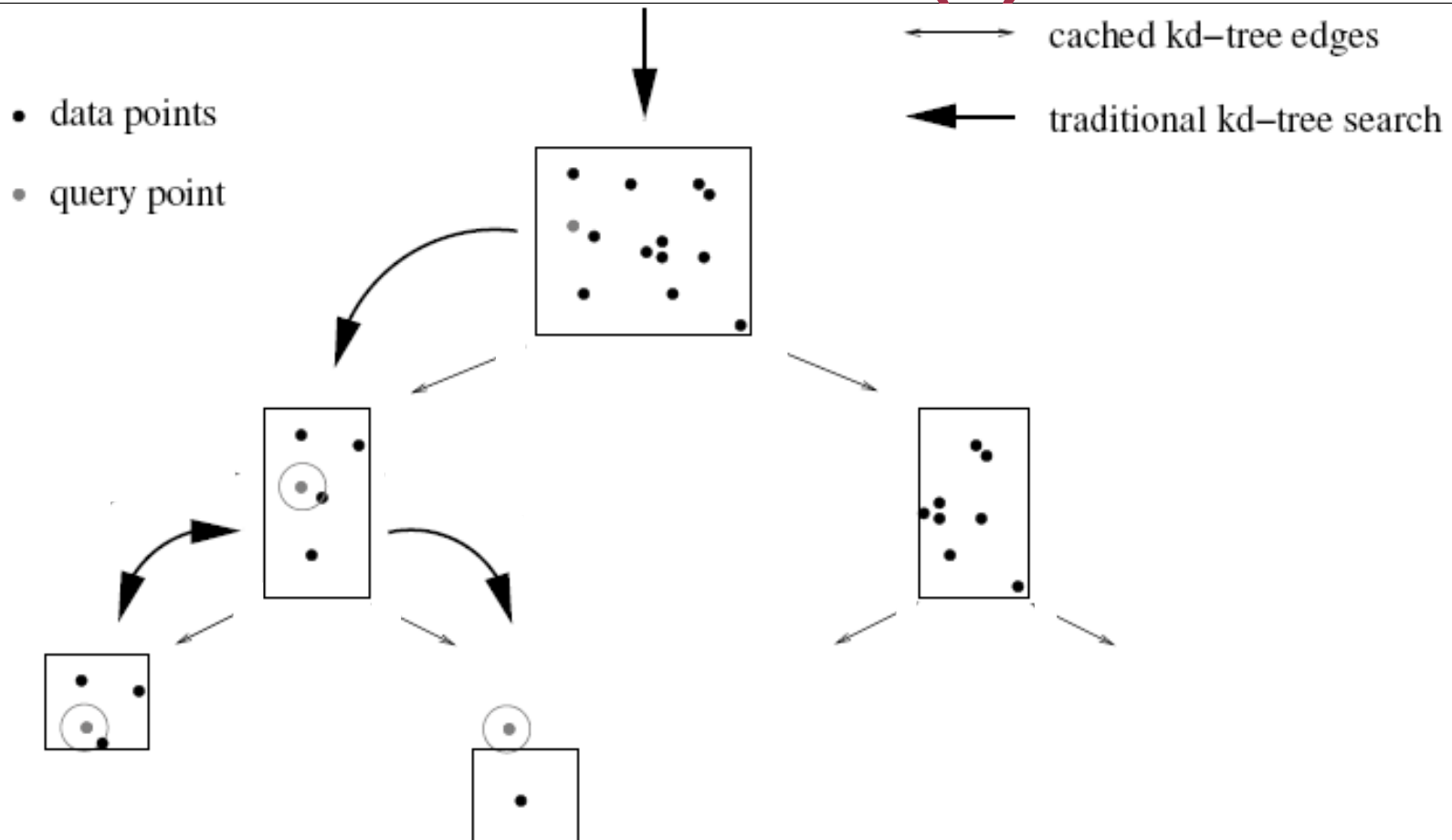
$$E(\mathbf{R}, \mathbf{t}) = \sum_{i=1}^{N_m} \sum_{j=1}^{N_d} w_{i,j} \|\mathbf{m}_i - (\mathbf{R}\mathbf{d}_j + \mathbf{t})\|^2$$

3. Iterate 1. and 2.

- Idea: Make use of the iterative structure of the algorithm
  - Save closest Points in a vector  $\mathbf{v}$
  - In addition, we save a pointer to the leaf, where the closest point has been found
  - The kd-tree contains pointers to predecessor nodes

# Cached kd-tree Search (2)

- data points
- query point

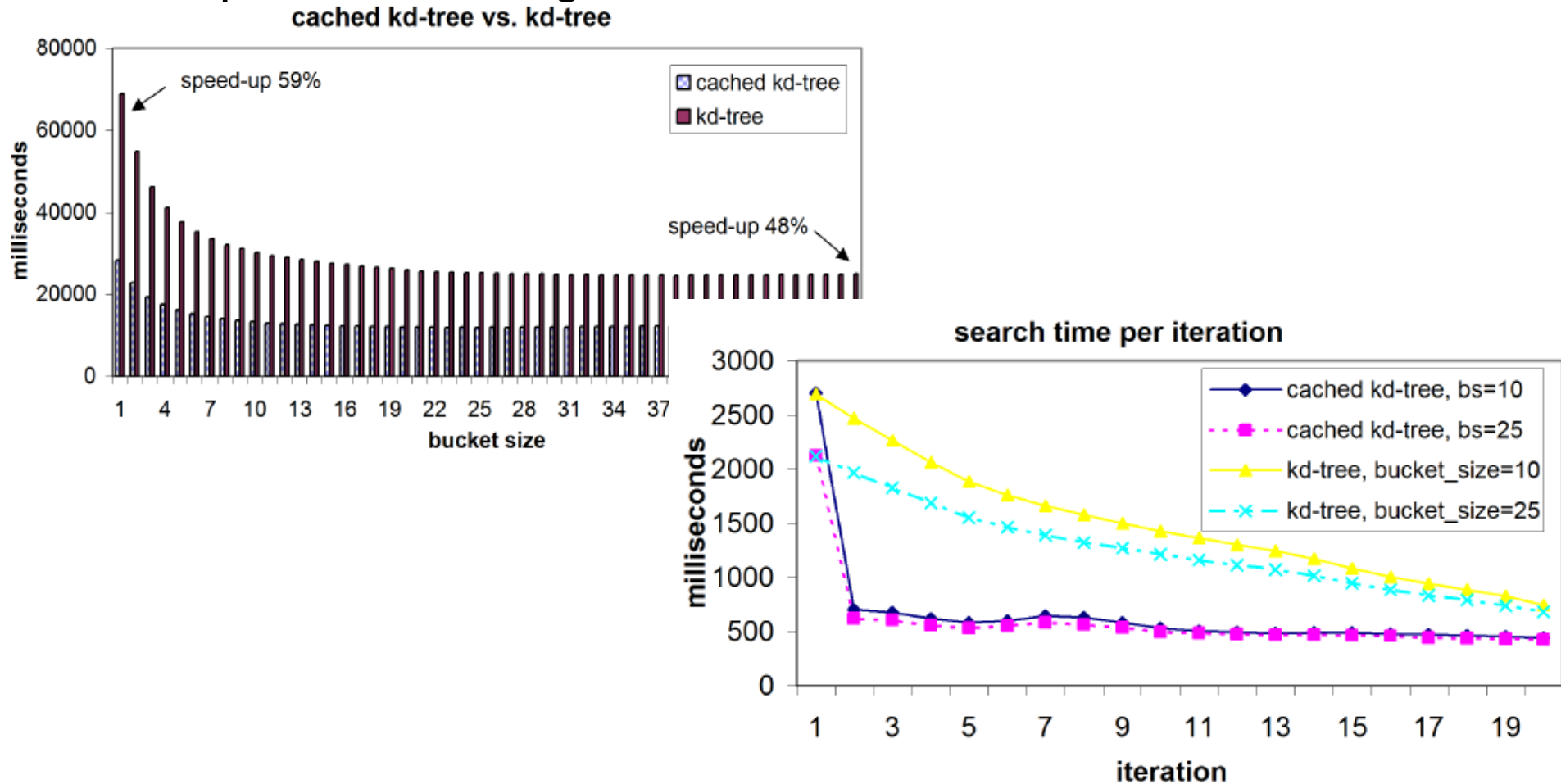


vector of point pairs  $v$



# Cached kd-tree Search (3)

- Comparison running times



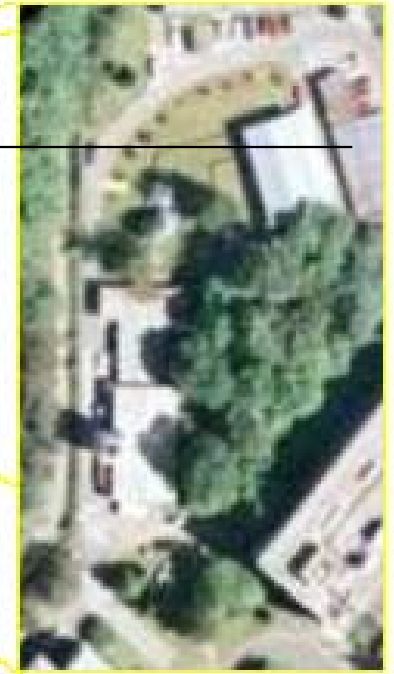
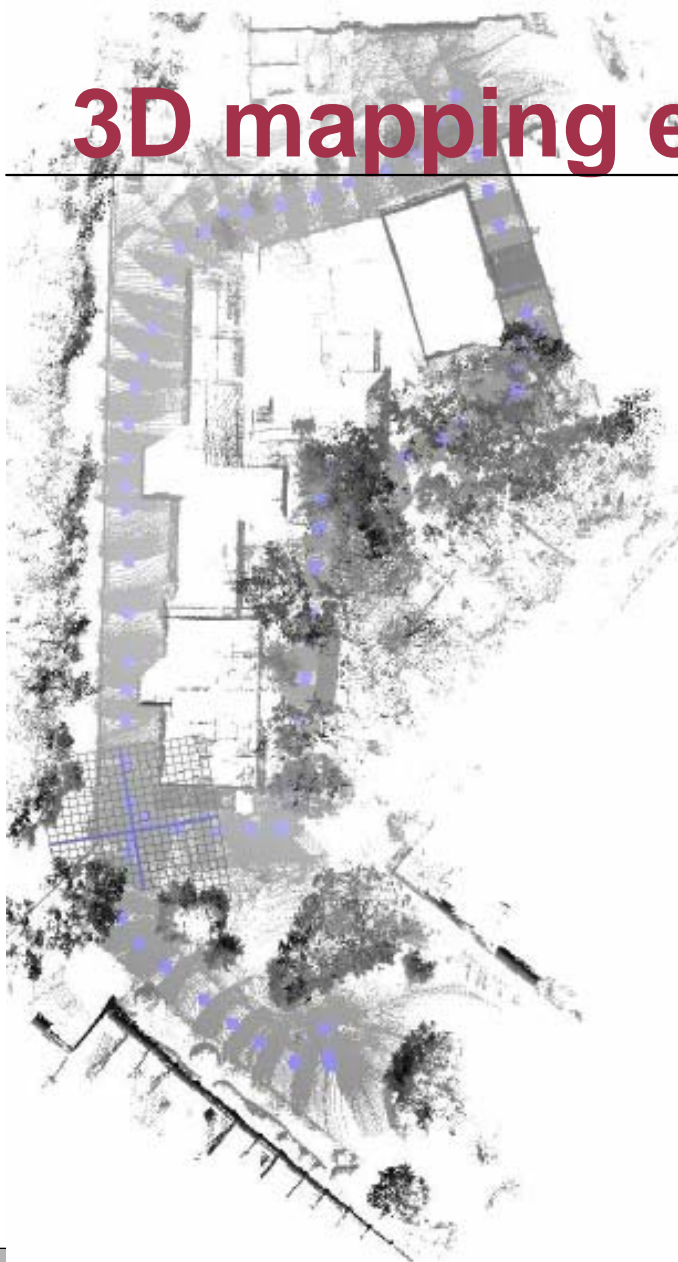


# Contents

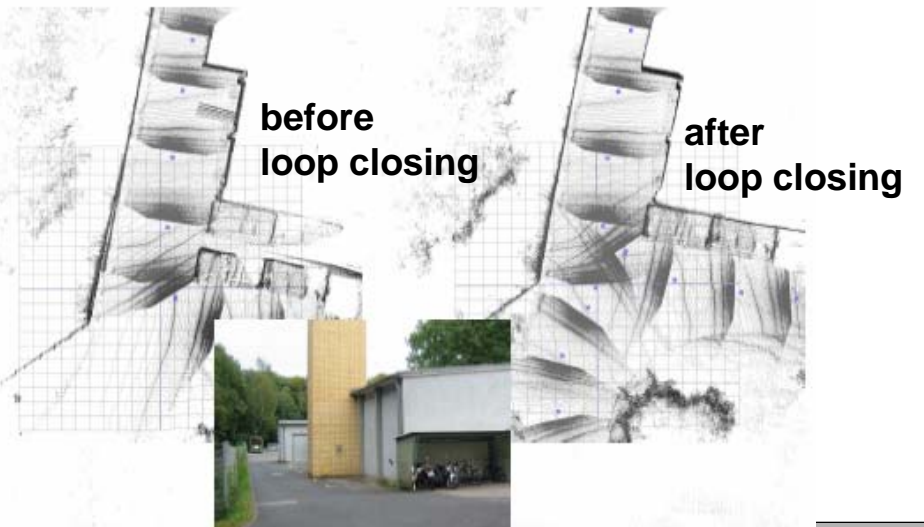
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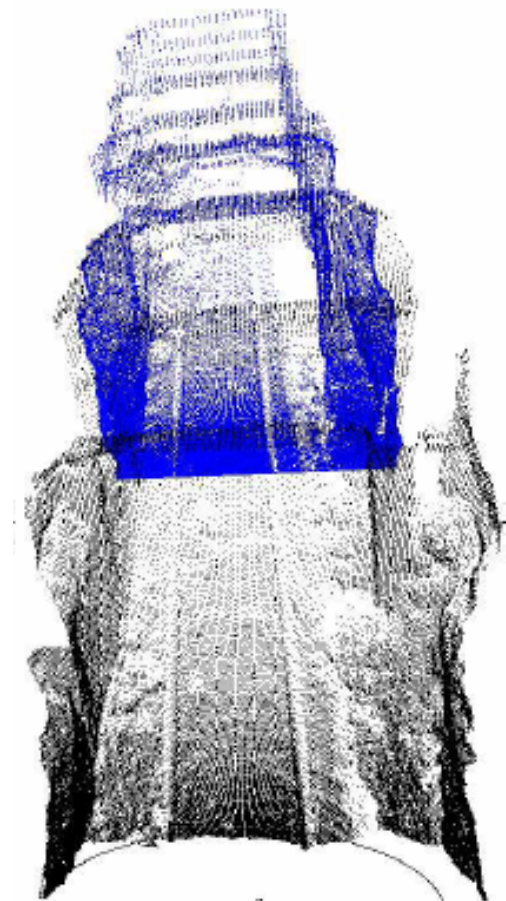
# 3D mapping examples



**(Video)**



# Autonomous Mine Inspection (CMU)



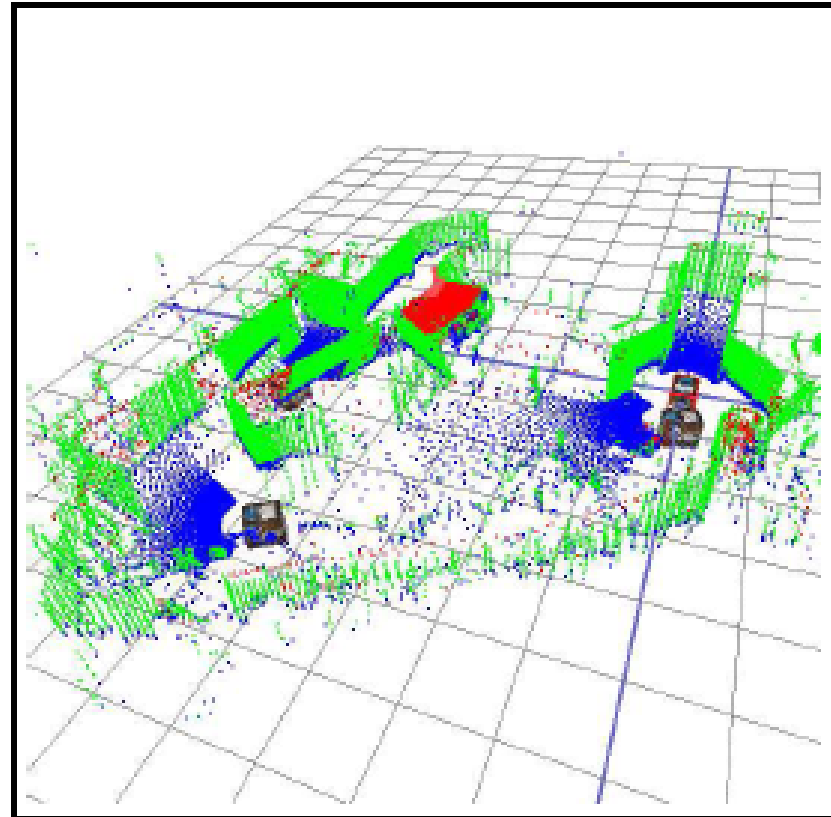
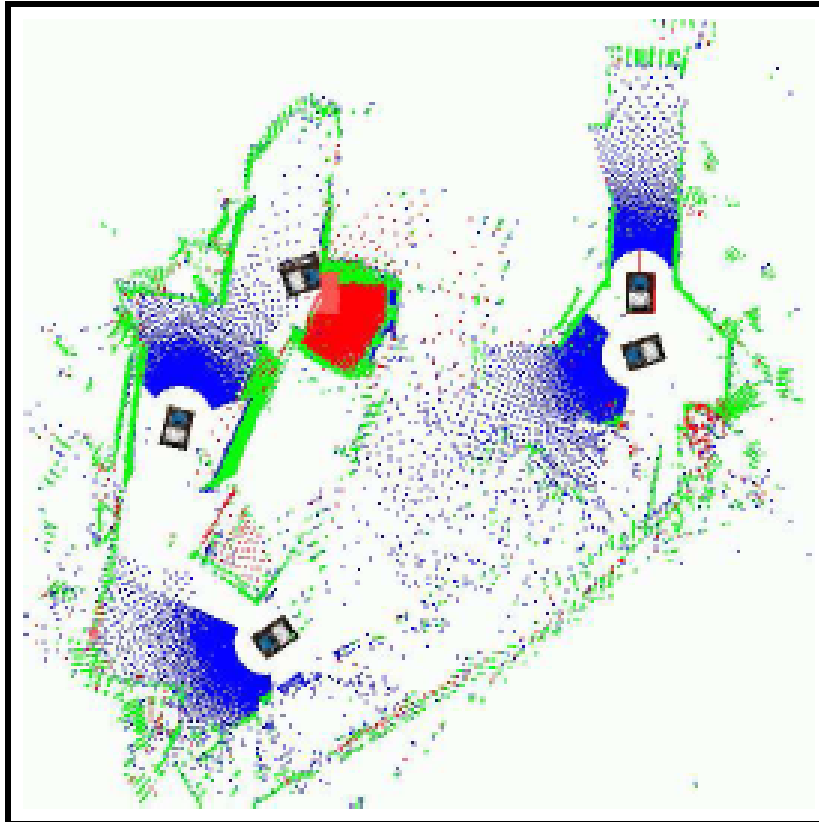


# Kurt3D RoboCup Rescue (Osaka 2005)

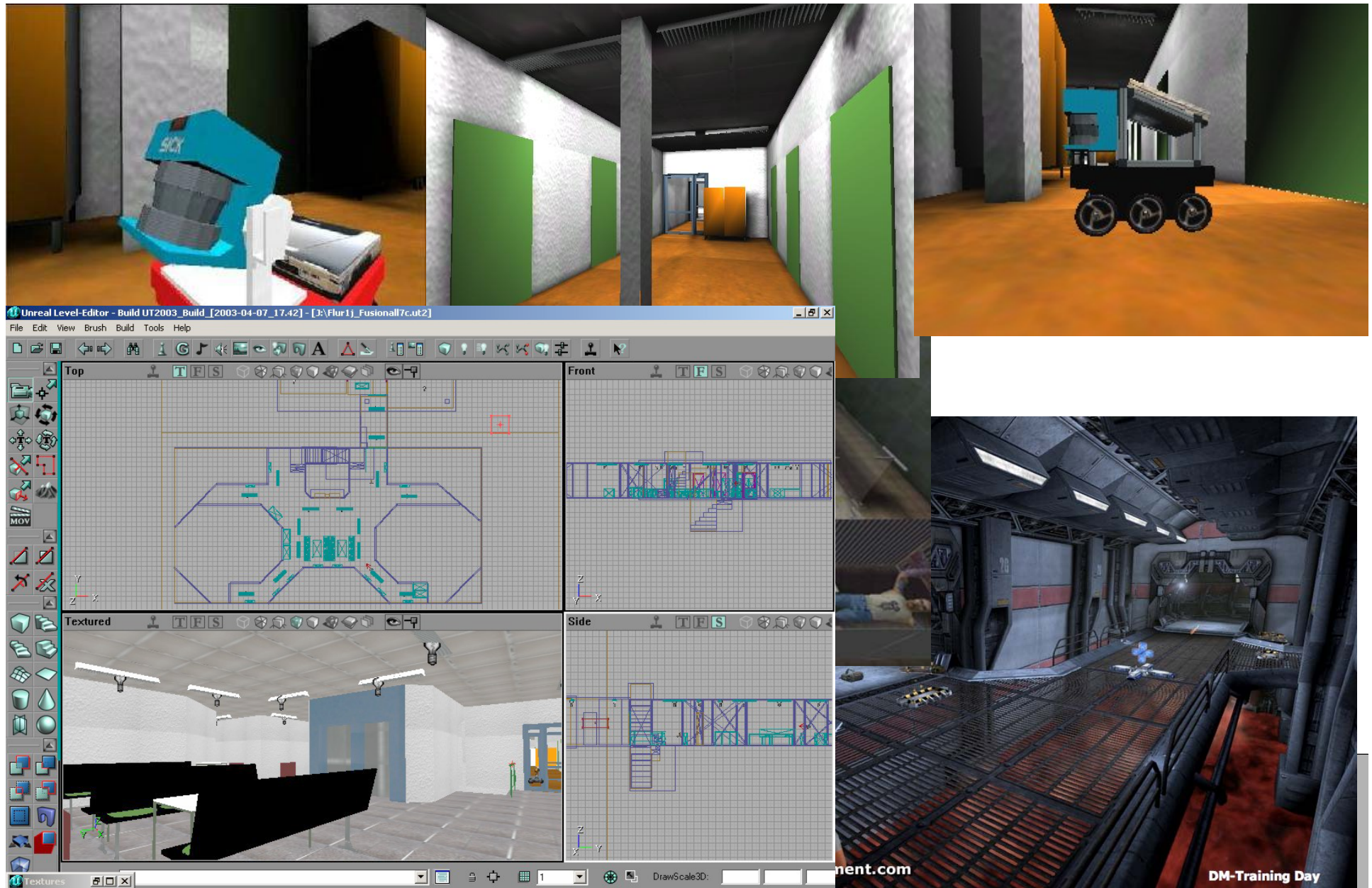


# Kurt3D RoboCup Rescue (Osaka 2005)

3D maps:

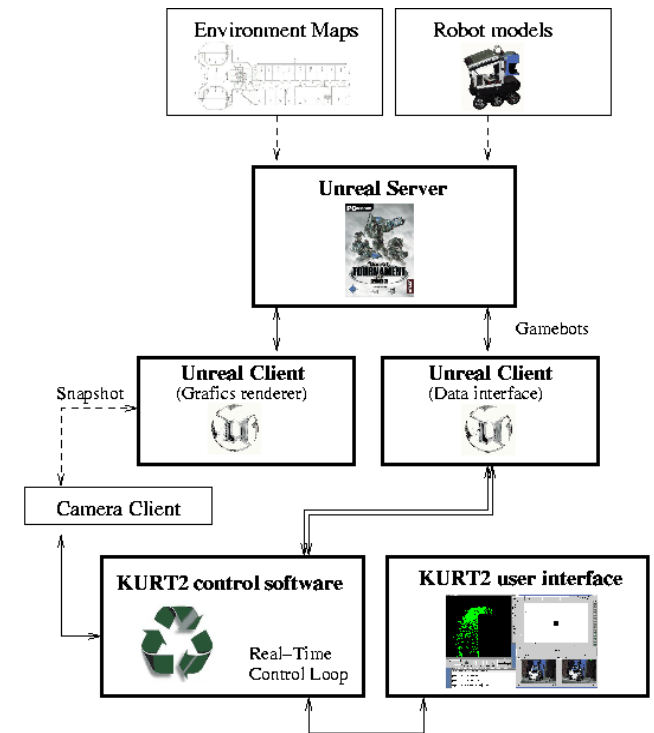


# 3D scans from simulation (1)



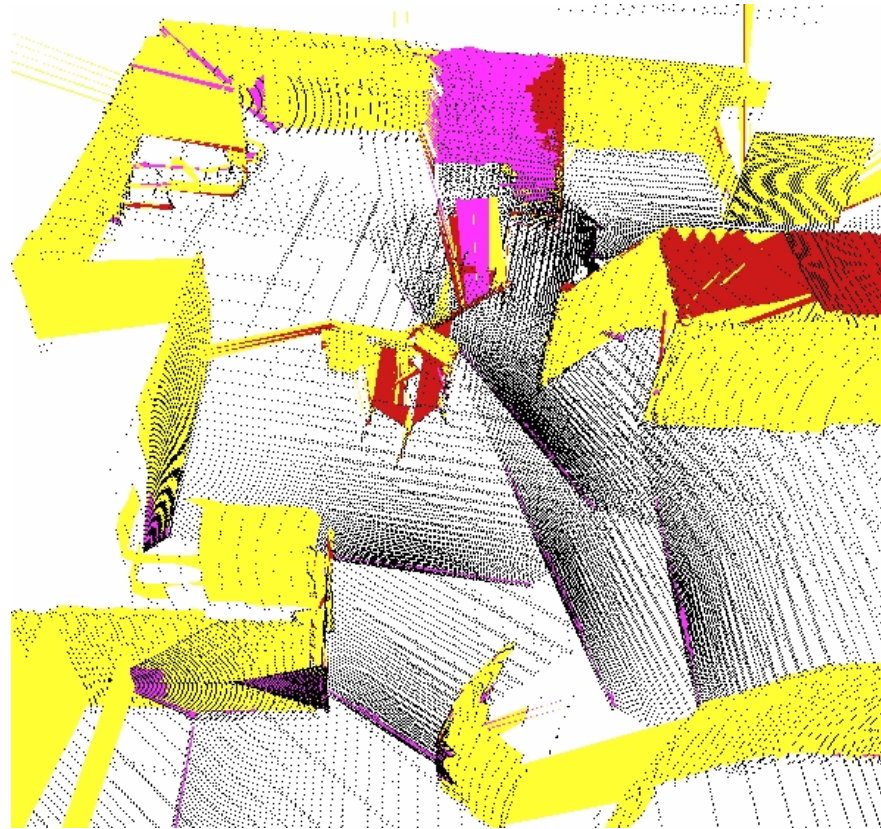
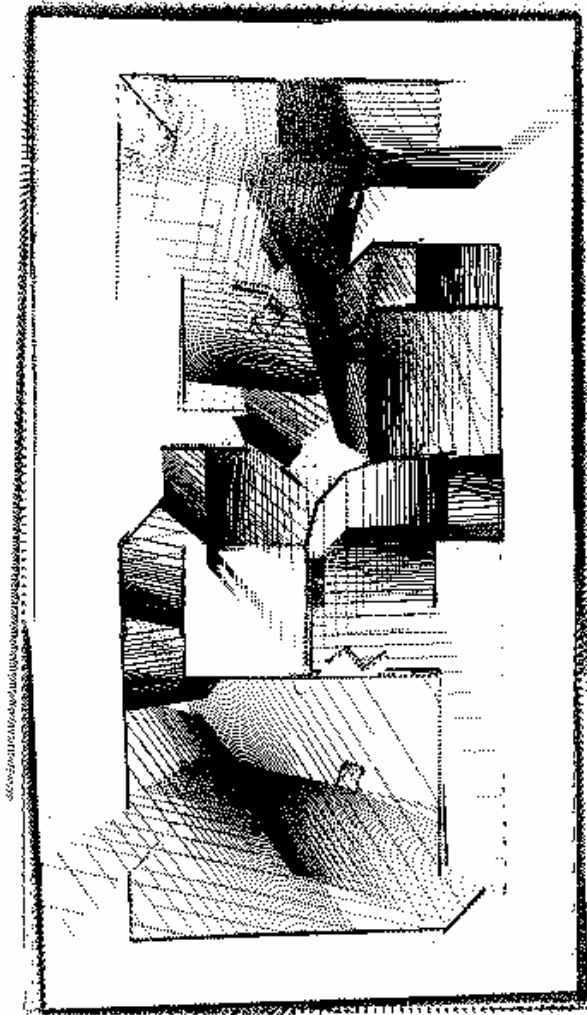


# 3D scans from simulation (2)



**Identical interfaces !!!**

# 3D scans from simulation (3)



- Everyone can test algorithms for robotics in 3D

# Contributions

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- Practical (on-line, on-board) variant of ICP for high-resolution point sets due to
  - point reduction and
  - efficient representation (Cached-KD-trees)
  - starting guess based on octree representation
- Generating overall consistent 2D maps with global error minimization
- Tested on various datasets (including borrowed ones, e.g., CMU mine mapping)
- Capable of and demonstrated for full 6D SLAM
- Integrated into robot controller for 3D environment mapping
- RoboCup Rescue as evaluation for our mapping approach

2004 second place, SSRR 2005 best paper award, 2005 6th place.