

Indoor EKF SLAM

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Including work by:

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Outline



- Line based indoor SLAM
- Using typical indoor geometric constraints
 - Right angles
 - Line = wall from floor to ceiling
- Indoor sub-maps
- Error recovery with point based SLAM



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Feature: Line

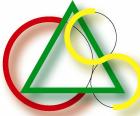


- Lines are extracted using “growing” followed by least square.
- Lines are represented as (ρ, α) , i.e. distance and angle.
- End-points maintained outside the filter
- 0.25m and 10 points used lower limit for the line extraction



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SLAM w. lines from laser

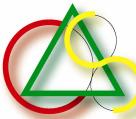


- EKF SLAM algorithm
- SICK laser data as input
- Features: lines



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More on implementation



- OO-framework
- Each feature-type must supply basic functions like; calc of innovation given measurement, specMatch(Measurement*), etc.
- Have tried lines, doors and points

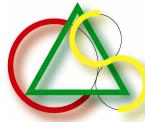


Data association



- If ($f.\text{match}(m) \&\& f.\text{specMatch}(m)$)
 - $f.\text{match}(m)$: Mahalanobis distance between feature f and measurement m .
 - $f.\text{specMatch}(m)$: Feature specific matching. For lines it requires an overlap.
- Accept measurement-feature match iff measurement matches only one feature in the map.





Platform 1: Asterix



- Nomad200
- 16 sonars
- SICK PLS
- 16 infrareds
- Synchro-drive
- Camera on pan-tilt
- Linux
- Single processor
PIII 450MHz



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Environment 1: CVAP/CAS building

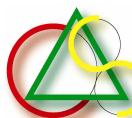


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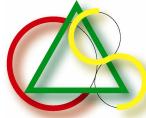
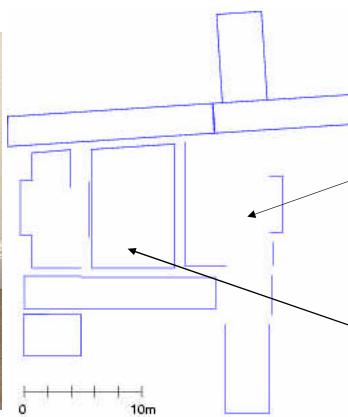
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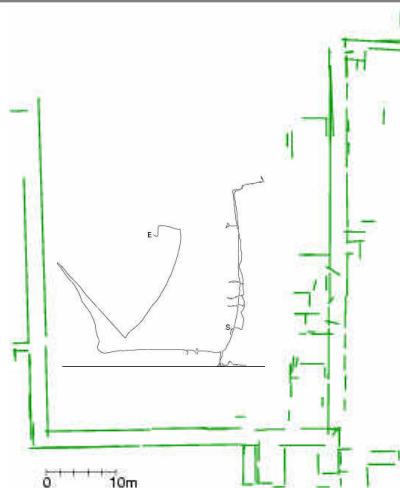
Env. and platform 2: Atrium in main building



Result at CVAP/CAS



- 388 m total distance
- 44 min (2640s)
- Robot controlled with joystick
- 127 lines



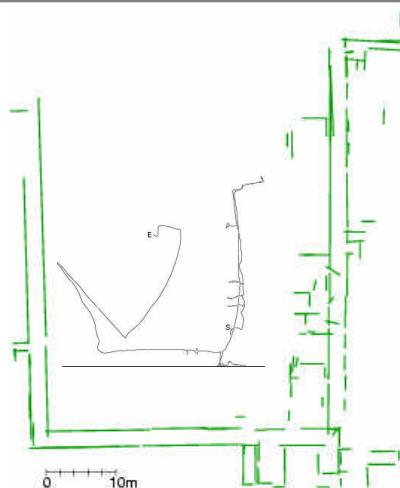
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Result at CVAP/CAS

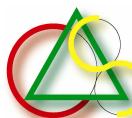
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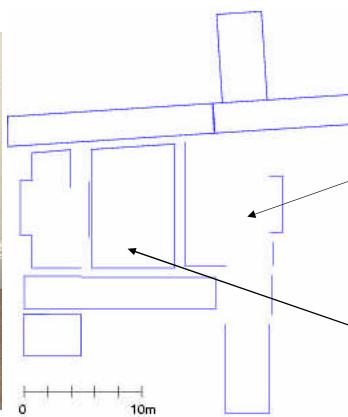
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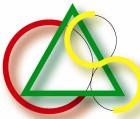
Env. and platform 2: Atrium in main building



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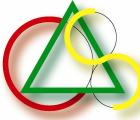
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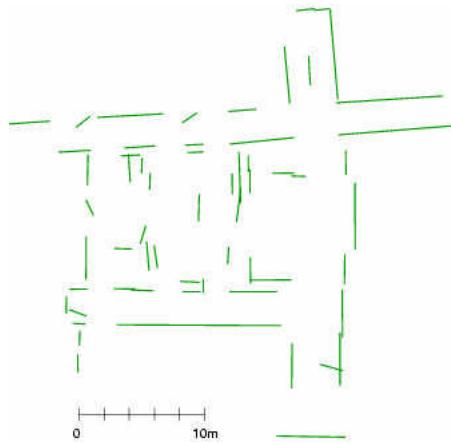
Conclusions

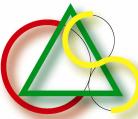
- Computation time only small fraction of total time for experiment. BUT full SLAM is $O(N^2)$ and at the end scans had to be skipped.
- Not so good data association
- Large odometric perturbations not handled well (thresholds!)



Result in Atrium

- 165 m total distance, twice around a loop
- 22 min (1320s)
- Data collected while being a tour guide for students
- 67 lines

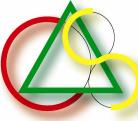




Conclusions III



- Doors are tricky to extract consistently when the walls are thick.
- Different features require different motion strategies. Lines often extend along the path, whereas for example points don't
- Lines are rather simple in this aspect

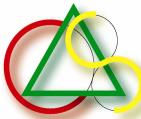


Conclusions II



- Bad registration between laser and odometry (not a real-time OS)
- Fast rotations (+ bad registration) are problematic
- The SISO-rule applies (Shit In) Shit Out)
- Poor end-point management of lines
- Door leafs are in the map (must be treated with care, DYNAMIC!!)
- A closed door splits the map! (end of corridor)

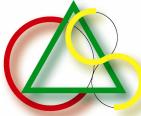




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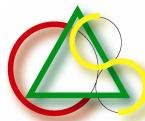


Lessons learned



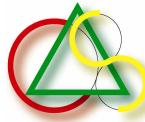
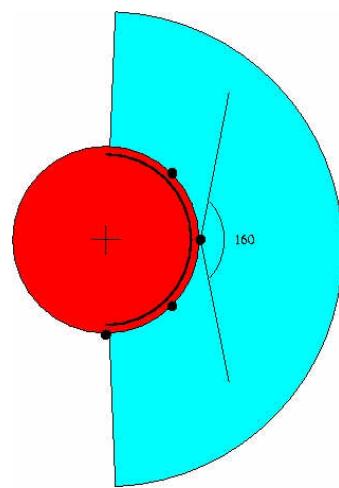
- Make sure that your input, is good. Remember SISO.
- Make sure your environment is large enough, some problems are not visible otherwise.
- Take extra care when the robot is rotating fast or better, put SLAM in control of motion!





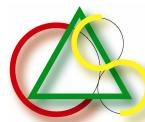
Platform continued

- 1 sonar transmitter and 4 microphones in horizontal plane.
- Range: approx 2.5 m

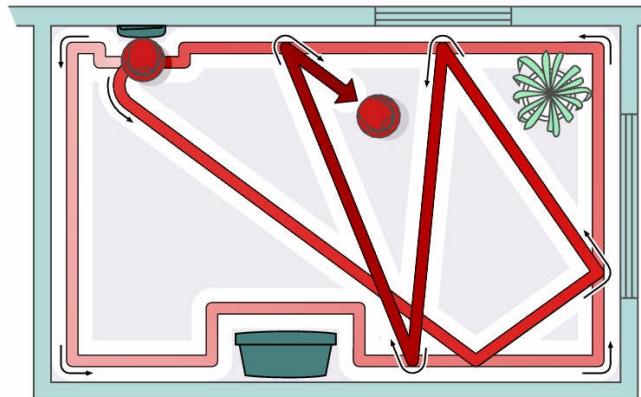


Platform 3: Trilobite vacuum cleaner





“Mapping” strategy



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Utilizing architectural constraints on Trilobite



- Problem:
 - Processor from “refrigerator”! (integers only)
 - Often stuff on floor along walls) slips easily
- Idea:
 - Use architectural constraint; Walls are straight and many things indoor tend to be positioned with right angles between them.



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Algorithm



- Extract lines using Hough transform from sonar
 - First line becomes reference line
 - Lines must be long enough
- Validate extracted line against reference line
 - Must be close enough to parallel or orthogonal (with uncertainty taken into account)
- Correct the pose as if all angular error was produced in the last corner (an approximation)



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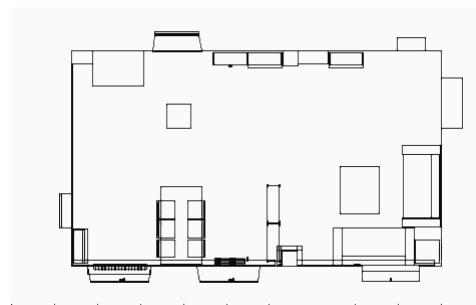


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Results

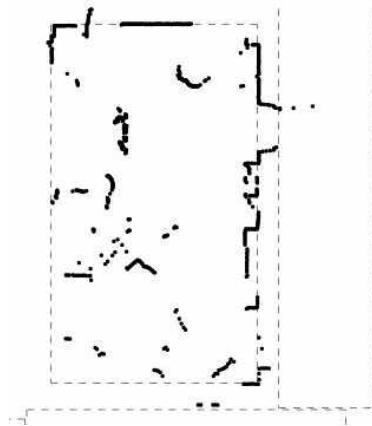


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Clutter in sensor height!



Outline

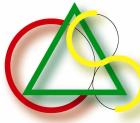
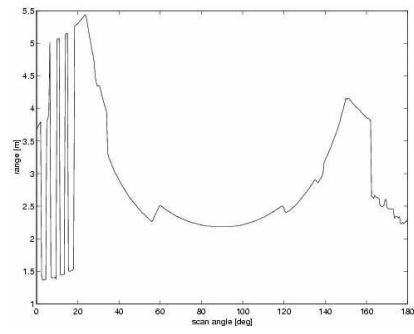
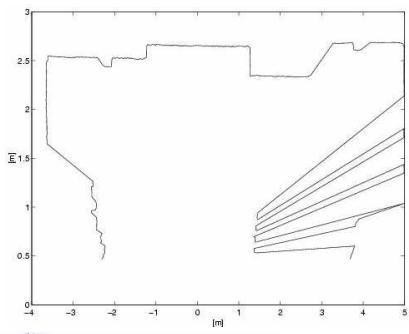


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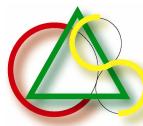
Less clutter higher up

- Wall often easy to find at its intersection with the ceiling.



SICK on pan-tilt





Pan the SICK and extract intersection points

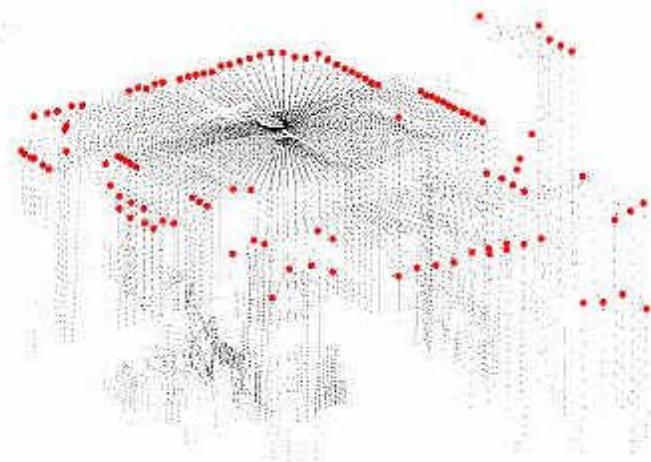


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Compare the result!

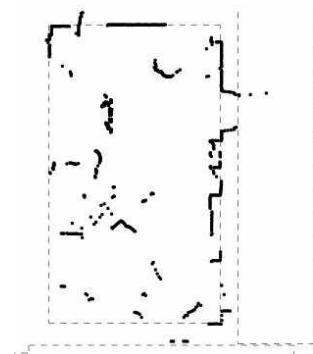
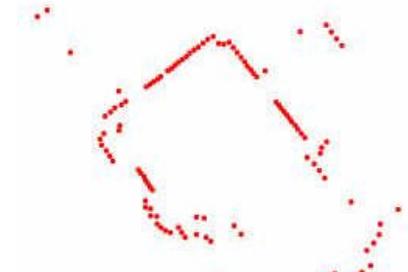


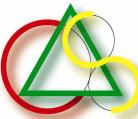
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Sub-maps



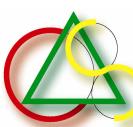
- At some point it might be necessary to split the map into several sub-maps. This typically gives:
 - Requires less memory and processing time
 - Higher local accuracy
 - Sub-optimal if not all information is used



Outline



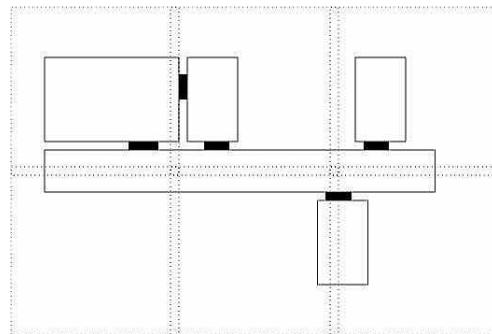
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Typical sub-map division



- Square grids
- Equal in size
- Placement not dependant on the environment

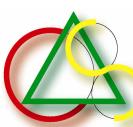


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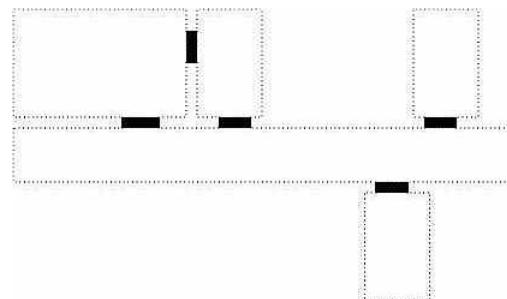
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Indoor sub-maps



- Rooms give natural sub-map division
- Visibility constraints between sub-maps almost given for free



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Error recovery by Guido Zunino



- It is not a question of IF errors will occur, rather WHEN they will occur.
- Must be able to recover from them
- Examples:
 - Large un-modeled odometric perturbations, common when passing e.g. thresholds
 - Erroneous data associations
 - Linearization errors (map slippage)



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Recovery scheme



- Localization:

- Match local map consisting of last detected features against last verified map) pose estimate x_L and covariance P_L

- Restoration:

- Use (x_L, P_L) as measurement in de-correlation and EKF update as in (Leonard & Feder 1999)
- Continue mapping
- Old feature matches) map assumed verified



Simple error recovery strategy



- Detection:

- No matching measurement when they where expected) error might have occurred
- Allow only M iterations with matches that where expected but did not occur

- Recovery:

- Store copies of old verified maps
- Re-localize in last verified map





Example with sonar point landmarks



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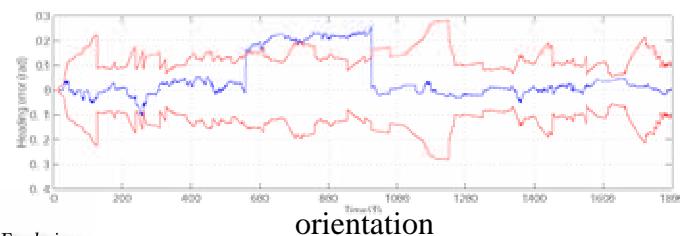
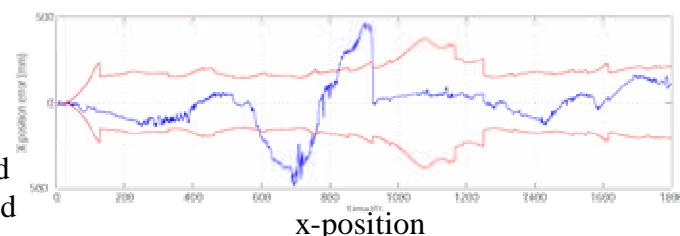
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Error vs. 2σ bounds



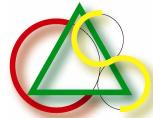
Errors estimated
using laser based
tracker running
in parallel



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Conclusions



- Robust although very simple
- Examples of failure situations:
 - Erroneous matching delaying detection
 - Un-modeled occlusions causing failure to detect feature) unnecessary recovery and loss of accuracy



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