

Simultaneous Localization and Mapping (SLAM)

RSS Technical Lecture 16

April 9, 2012

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Text: Siegwart and Nourbakhsh S. 5.8

Navigation Overview

- Where am I? Where am I going?
 - Localization
 - Assumed perfect map, but imperfect sensing
- How can I get there from here?
 - Planning
 - Assumed perfect map, sensing, and actuation
- Exploration (Mapping and Localization)
 - Mapping
 - SLAM (Today)

SLAM Problem Statement

- Inputs:
 - No external coordinate reference
 - Time sequence of proprioceptive and exteroceptive measurements* made as robot moves through an initially unknown environment
- Outputs:
 - A *map** of the environment
 - A robot *pose estimate* associated with each measurement, in the coordinate system in which the map is defined

*Not yet fully defined

SLAM Problem -- Incremental

- State/Output:
 - Map of env't observed "so far"
 - Robot pose estimate w.r.t. map
- Action/Input:
 - Move to a new position/orientation
 - Acquire additional observation(s)
- Update State:
 - Re-estimate the robot's pose
 - Revise the map appropriately

SLAM Aspects

- What is a measurement?
 - What is a map?
 - How are map, pose coupled?
 - How should robot move?
 - What is hard about SLAM?
-
- But first: some intuition

Illustration of SLAM without Landmarks

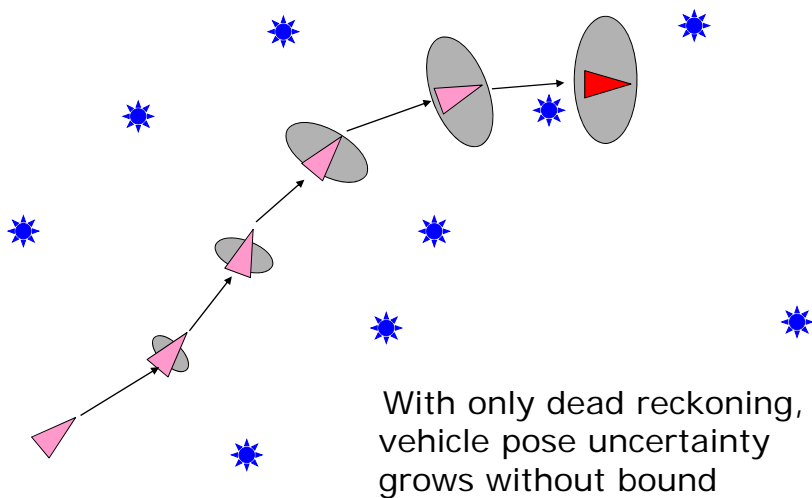
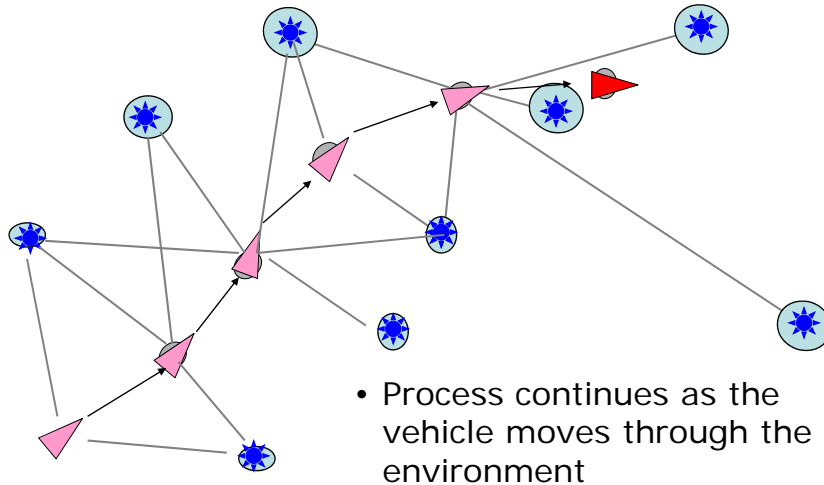


Illustration of SLAM with Landmarks



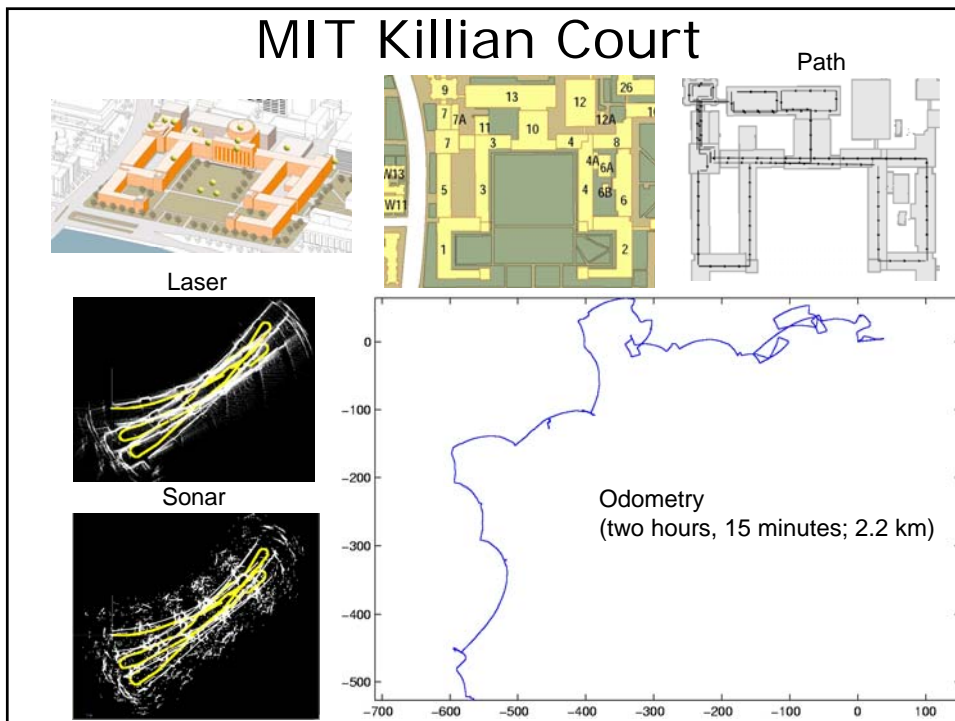
Why is SLAM Hard?

- “Grand challenge”-level robotics problem
 - Autonomous, persistent, collaborative robots mapping multi-scale, generic environments
- Map-making = learning
 - Difficult even for humans
 - Even skilled humans make mapping mistakes
- Scaling issues
 - Space: Large extent (combinatorial growth)
 - Time: Persistent autonomous operation
- “Chicken and Egg” nature of problem
 - If robot had a map, localization would be easier
 - If robot could localize, mapping would be easier
 - ... But robot has neither; starts from blank slate
 - Must also execute an *exploration strategy*
- **Uncertainty** at every level of problem

Uncertainty in Robotic Mapping

Uncertainty:	Continuous	Discrete
Scale:		
Local	Sensor noise	Data association
Global	Navigation drift	Loop closing

MIT Killian Court

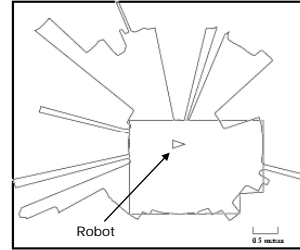


Common range-and-bearing sensors

Polaroid sonar ring
12 range returns,
one per 30
degrees, at ~4 Hz



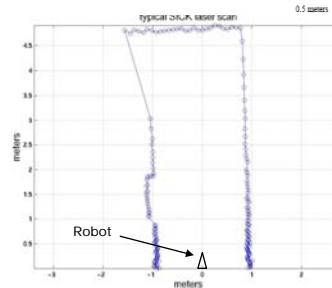
(+ servoed
rotation)



SICK laser scanner
180 range returns,
one per degree,
at 5-75 Hz

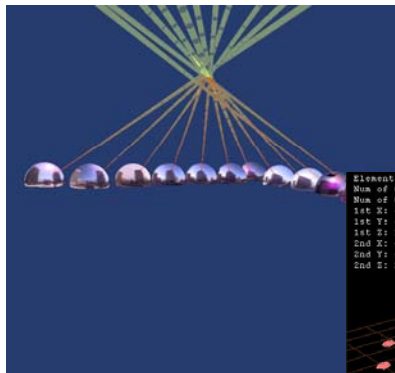


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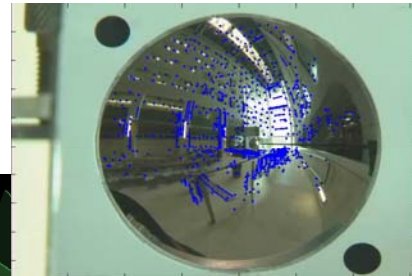


Other possibilities: Stereo/monocular vision; Robot itself (stall, bump sensing)

Tracking & long-baseline monocular vision

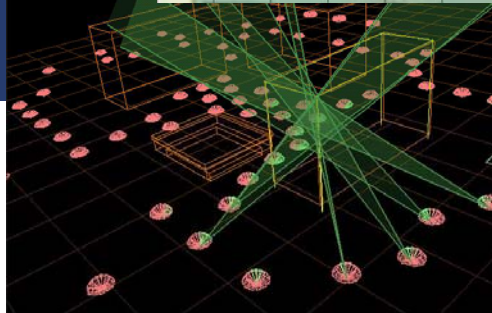


```
Element ID: 00-150
Num of Observations: 7
Num of Commits: 1
1st X: -1402.749
1st Y: -3007.098
1st Z: 1876.276
2nd X: -1421.749
2nd Y: -3007.098
2nd Z: 200.000
```



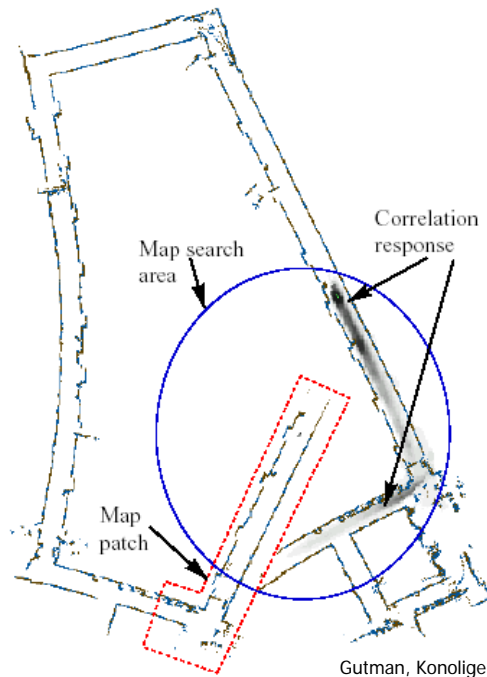
Bosse

Track points, edges, texture
patches from frame to frame;
triangulate to recover local 3D
structure. Also called "SFM,"
Structure From camera Motion,
or object motion in the image



Chou

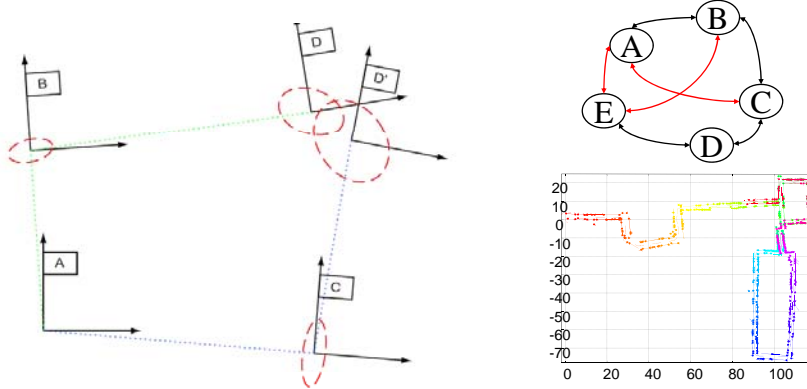
Loop Closing



What is a map?

- Collection of *features* with some *relationship* to one another
- What is a *feature*? ← **Uncertainty**
 - Occupancy grid cell
 - Line segment
 - Surface patch
- What is a feature *relationship*?
 - Rigid-body transform (metrical mapping)
 - Topological path (chain of co-visibility)
 - Hybrid representation (geom. + topo.)

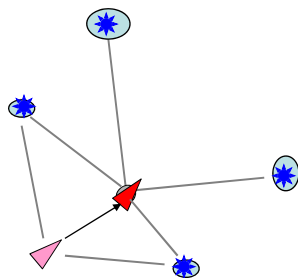
Atlas hybrid maps (Bosse et al.)



- Features: point, line, patch clouds
- Geometry: rigid frames, submaps
- Topology: map adjacencies
- Hybrid: uncertain map-to-map transformations

What is *pose* w.r.t. a map?

- Pose estimate that is (maximally) *consistent* with the estimated features observed from vicinity
- Consistency can be evaluated locally, semi-locally, or globally

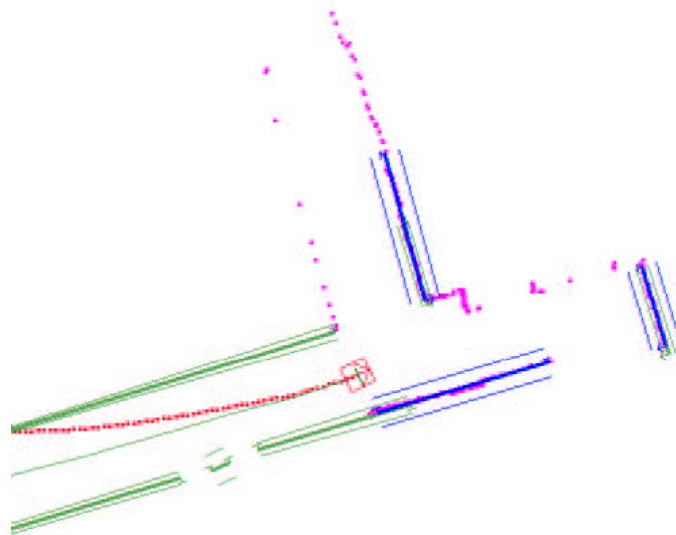


- Note tension between estimation *precision* and solution *consistency*

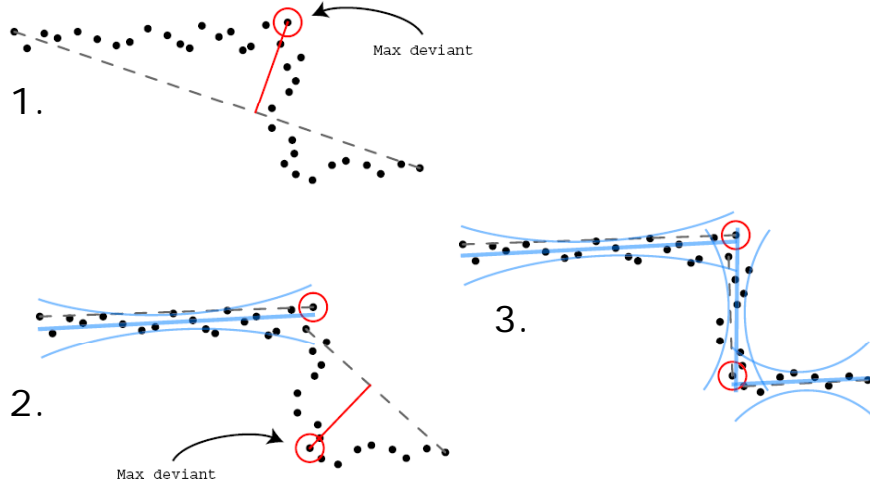
Example

- SLAM with laser scanning
- Observations
- Local mapping
 - Iterated closest point
- Loop closing
 - Scan matching
 - Deferred validation
 - Search strategies

Observations

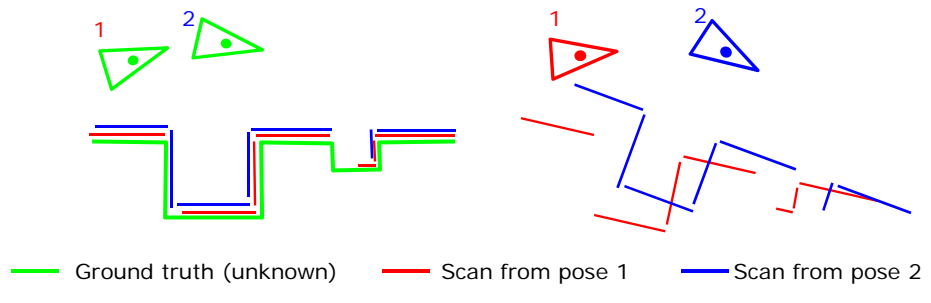


Observations



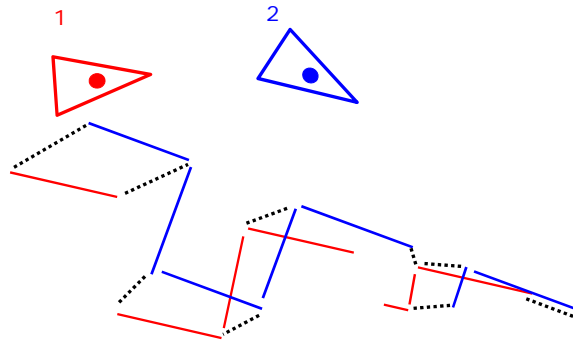
Scan Matching

- Robot scans, moves, scans again
- Short-term odometry/IMU error causes misregistration of scans
- *Scan matching* is the process of bringing scan data into alignment



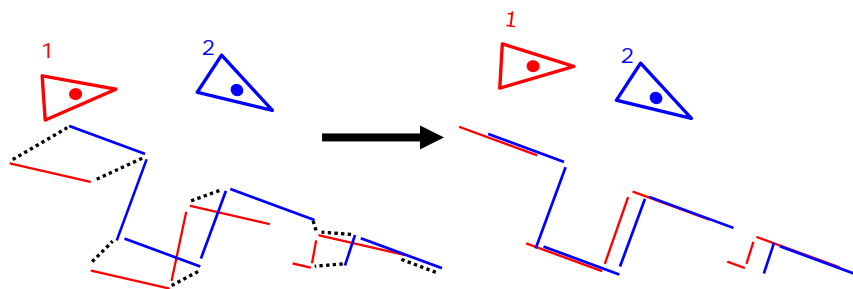
Iterated Closest Point

- For each point in scan 1
 - Find closest point in scan 2



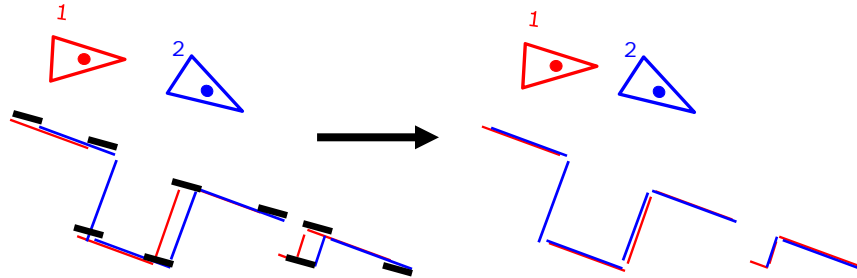
Iterated Closest Point

- Find the transformation that best aligns the matching sets of points



Iterated Closest Point

- ... Repeat until convergence



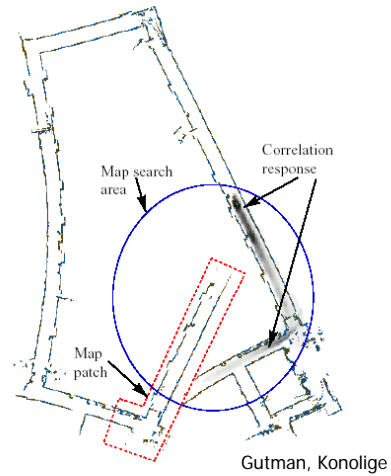
- Can ICP across scans, across a scan and a (sub)map, or even across submaps!

Limitations / failure modes

- Computational cost (two scans of size n)
 - Naively, $O(n^2)$ plus cost of alignment step
- False minima
 - If ICP starts far from true alignment
 - If scans exhibit repeated local structure
- Bias
 - Anisotropic point sampling
 - Differing sensor fields of view (occlusion)
- Lots of research on improved ICP methods (see, e.g., Rusinkiewicz)

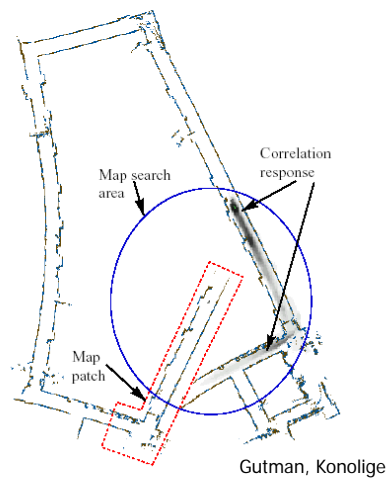
Loop Closing

- ICP solves small-scale, short-duration SLAM fairly well
- But now, consider:
 - Large scale
 - High uncertainty



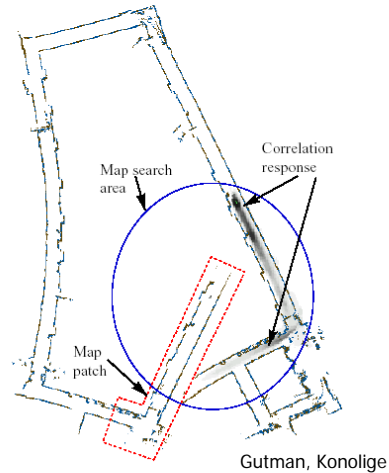
Loop Closing

- Naive ICP ruled out:
 - Too CPU-intensive
- Assume we have a *pose uncertainty bound*
- This limits portion of existing map that must be searched
- Still have to face the problem of matching two partial scans that are far from aligned



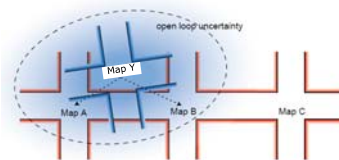
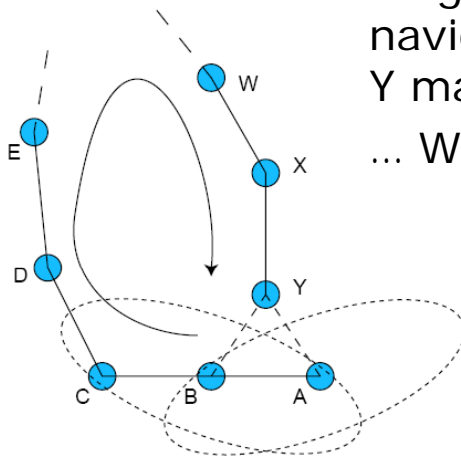
Scan Matching Strategies

- Exhaustive search
 - Discretize robot poses
 - Find implied alignments
 - Assign score to each
 - Choose highest score
 - Pros, Cons?
- Randomized search
 - Choose minimal sufficient match, at random
 - Align and score
 - Choose highest score
 - RANSAC (1981)
 - Pros, Cons?



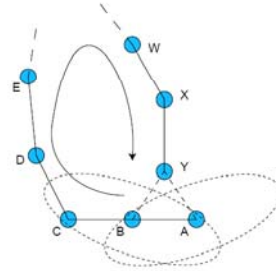
Loop Closing Ambiguity

- Consider SLAM state after ABC ... XY
Large open-loop navigation uncertainty
Y matches *both* A & B
... What to do?



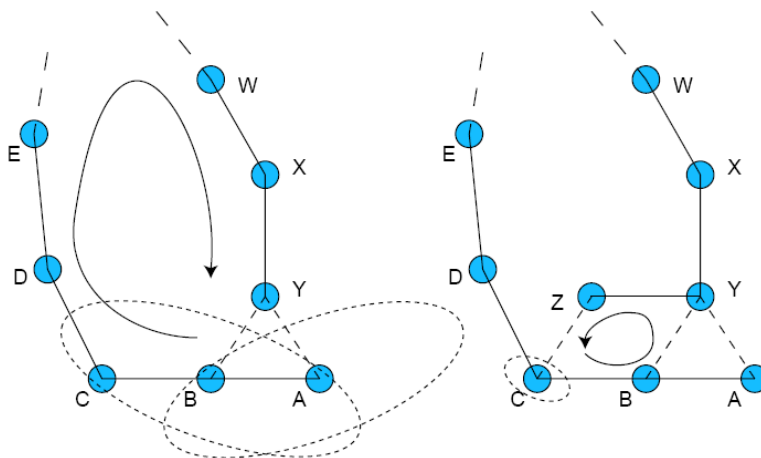
Loop Closing Choices

- Choose neither match
 - Pros, cons?
- Choose one match
 - Pros, cons?
- Choose both matches
 - Pros, cons?



Deferred Loop Validation

- Continue SLAM until Z matches C
- Examine graph for \sim identity cycle



Some SLAM results

- See rvsn.csail.mit.edu group page

Summary

- SLAM is a hard robotics problem:
 - Requires sensor fusion over large areas
 - Scaling issues arise quickly with real data
- Key issue is managing *uncertainty*
 - Low-level to high-level
 - Continuous and discrete
- Saw several SLAM strategies
 - Local and global alignment
 - Randomization
 - Deferred validation
- SLAM is only part of the solution for most applications (need naming, semantics)