

Robot Mapping

Graph-Based SLAM with Landmarks

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Graph-Based SLAM (Chap. 15)

- Use a **graph** to represent the problem
- Every **node** in the graph corresponds to a pose of the robot during mapping
- Every **edge** between two nodes corresponds to a spatial constraint between them
- **Graph-Based SLAM:** Build the graph and find a node configuration that minimize the error introduced by the constraints

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The Graph

So far:

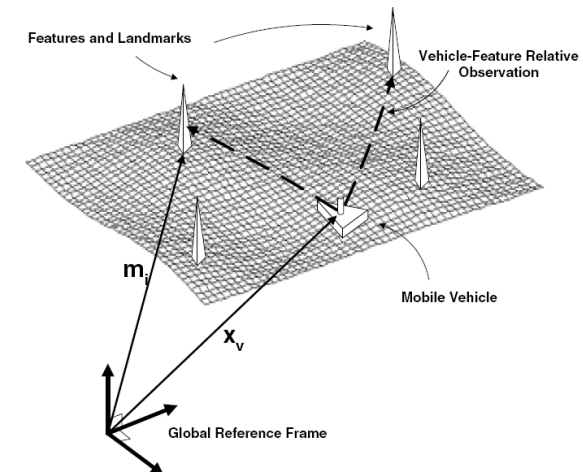
- Vertices for robot poses (x, y, θ)
- Edges for virtual observations (transformations) between robot poses

Topic today:

- How to deal with landmarks

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Landmark-Based SLAM



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Real Landmark Map Example

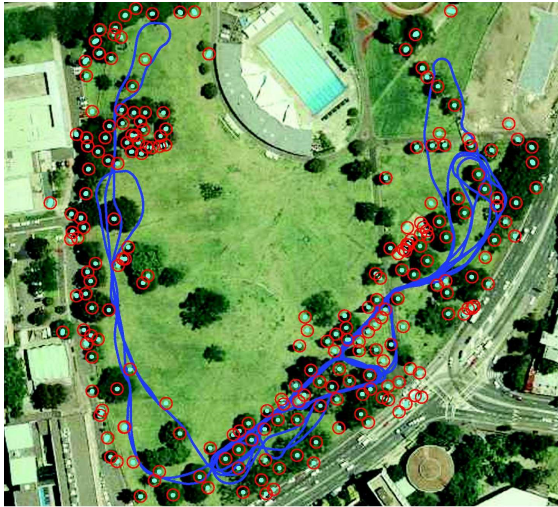
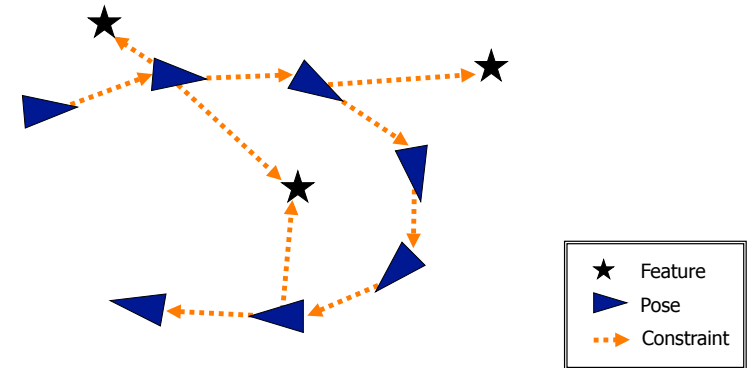


Image courtesy: E. Nebot

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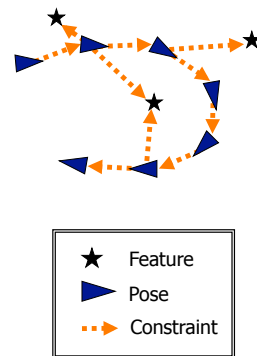
The Graph with Landmarks



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The Graph with Landmarks

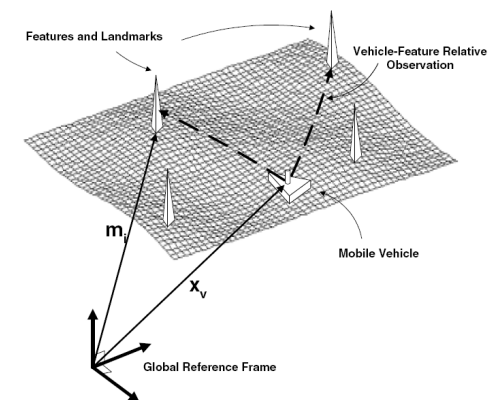
- **Nodes** can represent:
 - Robot poses
 - Landmark locations
- **Edges** can represent:
 - Landmark observations
 - Odometry measurements
- The minimization optimizes the landmark locations and robot poses



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2D Landmarks

- Landmark is a (x, y) -point in the world
- Relative observation in (x, y)



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Landmarks Observation

- Expected observation (x-y sensor)

$$\hat{\mathbf{z}}_{ij}(\mathbf{x}_i, \mathbf{x}_j) = \mathbf{R}_i^T(\mathbf{x}_j - \mathbf{t}_i)$$

↑ robot
↑ landmark
↑ robot translation

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Landmarks Observation

- Expected observation (x-y sensor)

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↑ robot
↑ landmark
↑ robot translation

- Error function

$$\begin{aligned} \mathbf{e}_{ij}(\mathbf{x}_i, \mathbf{x}_j) &= \hat{\mathbf{z}}_{ij} - \mathbf{z}_{ij} \\ &= \mathbf{R}_i^T(\mathbf{x}_j - \mathbf{t}_i) - \mathbf{z}_{ij} \end{aligned}$$

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Bearing Only Observations

- A landmark is still a 2D point
- The robot observe only the bearing towards the landmark
- Observation function

$$\hat{\mathbf{z}}_{ij}(\mathbf{x}_i, \mathbf{x}_j) = \text{atan} \frac{(\mathbf{x}_j - \mathbf{t}_i).y}{(\mathbf{x}_j - \mathbf{t}_i).x} - \theta_i$$

↑ robot
↑ landmark
↑ robot-landmark angle
↑ robot orientation

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Bearing Only Observations

- Observation function

$$\hat{\mathbf{z}}_{ij}(\mathbf{x}_i, \mathbf{x}_j) = \text{atan} \frac{(\mathbf{x}_j - \mathbf{t}_i).y}{(\mathbf{x}_j - \mathbf{t}_i).x} - \theta_i$$

↑ robot
↑ landmark
↑ robot-landmark angle
↑ robot orientation

- Error function

$$\mathbf{e}_{ij}(\mathbf{x}_i, \mathbf{x}_j) = \text{atan} \frac{(\mathbf{x}_j - \mathbf{t}_i).y}{(\mathbf{x}_j - \mathbf{t}_i).x} - \theta_i - \mathbf{z}_{ij}$$

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The Rank of the Matrix H

- What is the rank of \mathbf{H}_{ij} for a 2D landmark-pose constraint?

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The Rank of the Matrix H

- What is the rank of \mathbf{H}_{ij} for a 2D landmark-pose constraint?
 - The blocks of \mathbf{J}_{ij} are a 2x3 matrices
 - \mathbf{H}_{ij} cannot have more than rank 2
 $\text{rank}(A^T A) = \text{rank}(A^T) = \text{rank}(A)$

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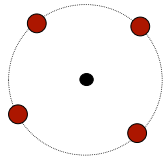
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- What is the rank of \mathbf{H}_{ij} for a bearing-only constraint?
 - The blocks of \mathbf{J}_{ij} are a 1x3 matrices
 - \mathbf{H}_{ij} has rank 1

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Where is the Robot?

- Robot observes one landmark (x,y)
- Where can the robot be relative to the landmark?



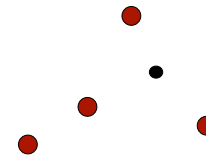
The robot can be somewhere on a circle around the landmark

It is a 1D solution space (constrained by the distance and the robot's orientation)

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Where is the Robot?

- Robot observes one landmark (bearing-only)
- Where can the robot be relative to the landmark?



The robot can be anywhere in the x-y plane

It is a 2D solution space (constrained by the robot's orientation)

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Rank

- In landmark-based SLAM, the system can be under-determined
- The rank of \mathbf{H} is **less or equal** to the sum of the ranks of the constraints
- To determine a **unique solution**, the system must have **full rank**

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Questions

- The rank of \mathbf{H} is **less or equal** to the sum of the ranks of the constraints
- To determine a **unique solution**, the system must have **full rank**
- **Questions:**
 - How many 2D landmark observations are needed to resolve for a robot pose?
 - How many bearing-only observations are needed to resolve for a robot pose?

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Under-Determined Systems

- No guarantee for a full rank system
 - Landmarks may be observed only once
 - Robot might have no odometry
- We can still deal with these situations by adding a “damping” factor to \mathbf{H}
- Instead of solving $\mathbf{H}\Delta\mathbf{x} = -\mathbf{b}$, we solve

$$(\mathbf{H} + \lambda\mathbf{I})\Delta\mathbf{x} = -\mathbf{b}$$

What is the effect of that?

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$$(\mathbf{H} + \lambda\mathbf{I})\Delta\mathbf{x} = -\mathbf{b}$$

- Damping factor for \mathbf{H}
- $(\mathbf{H} + \lambda\mathbf{I})\Delta\mathbf{x} = -\mathbf{b}$
- The damping factor $\lambda\mathbf{I}$ makes the system positive definite
- Weighted sum of Gauss Newton and Steepest Descent

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Simplified Levenberg Marquardt

- Damping to regulate the convergence using backup/restore actions

```
x: the initial guess
while (! converged)
   $\lambda = \lambda_{init}$ 
   $\langle \mathbf{H}, \mathbf{b} \rangle = \text{buildLinearSystem}(\mathbf{x});$ 
   $E = \text{error}(\mathbf{x})$ 
   $\mathbf{x}_{old} = \mathbf{x};$ 
   $\Delta\mathbf{x} = \text{solveSparse}(\mathbf{H} + \lambda\mathbf{I}) \Delta\mathbf{x} = -\mathbf{b};$ 
   $\mathbf{x} += \Delta\mathbf{x};$ 
  If ( $E < \text{error}(\mathbf{x})$ ) {
     $\mathbf{x} = \mathbf{x}_{old};$ 
     $\lambda *= 2;$ 
  } else {  $\lambda /= 2;$  }
```

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Bundle Adjustment

- 3D reconstruction based on images taken at different viewpoints
- Minimizes the reprojection error
- Often Levenberg Marquardt
- Developed in photogrammetry during the 1950ies

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Summary

- Graph-Based SLAM for landmarks
- The rank of \mathbf{H} matters
- Levenberg Marquardt for optimization

Literature

Bundle Adjustment:

- Triggs et al. "Bundle Adjustment — A Modern Synthesis"