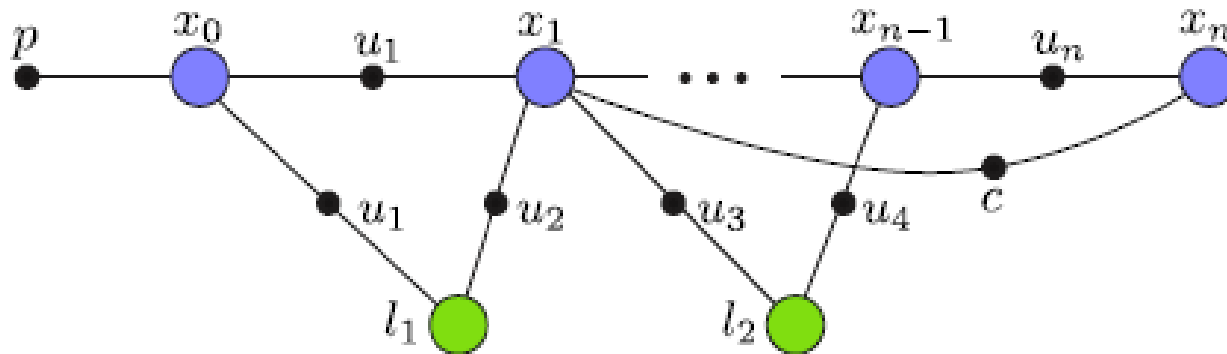# Localization using GTSAM and ROS

Bonus project for the lecture **Autonomous Driving**

Technical University of Munich

Robotics and Embedded Systems

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Source: MIT isam

Motivation: why localization?

- Autonomous cars need to **know** their position at **any** time

- GPS alone **error-prone** and **not always available**

- Car odometry **inaccurate** which leads to **accumulating error**

- Sensors such as Lidar or cameras can give us **information** about our **surroundings and the own position**

- In localization, a **map is available** which **represents the environment** (**landmarks, images, scans, pointclouds…**)

Motivation: why factor graphs?

- Localization can be solved using **Bayes/Kalman Filters, Monte Carlo Localization…**

- Factor graphs are the **most flexible solution**, as they

    - give us free control over the **type and amount of factors** we incorporate

    - allow the usage of **non-linear** motion and measurement models

    - can deal with **multi-modal** beliefs

    - break the task down into a graph problem that **scales well** with large environments and state representations

- More on the theory and GTSAM here and here

Task

1. Define and visualize a rectangular vehicle

2. Parametrize sensor position and field of view as well as accuracy

3. Define a few landmarks

4. Let the rectangular car drive a curve between the landmarks

5. Localize the vehicle with GTSAM based on the parametrized positions and sensors

Implementation

- Usage of **GTSAM 4.0**

  - one of the most complete **c++ libraries** for factor graphs and SLAM

  - comes with a **Matlab interface** for fast evaluation/plotting

- Usage of **ROS Kinetic LTS**

  - **Open Source** Framework

  - includes many **efficient libraries and tools** for robotic applications

  - solves **communication** between seperated packages in **real time**

System architecture

- Every **100 ms** the controller sends **odometry commands** (+ error estimates) that cause the car to drive in a circle

- After receiving the commands, the car
  - **simulates the driving** (to get the ground truth position)
  - **senses** the landmarks in a local circle around the ground truth and according to a prespecified detection accuracy
  - **localizes** itself given odometry and sensed landmarks

- Real time visualization (in **rviz**) of
  - the **path and position** of the ground truth, the localization result and the accumulated odometry
  - landmark **map** and **currently sensed local landmarks**

Global LM csv Map

Local landmarks

Controller

Odometry commands

Car
- drive
- sense
- localize
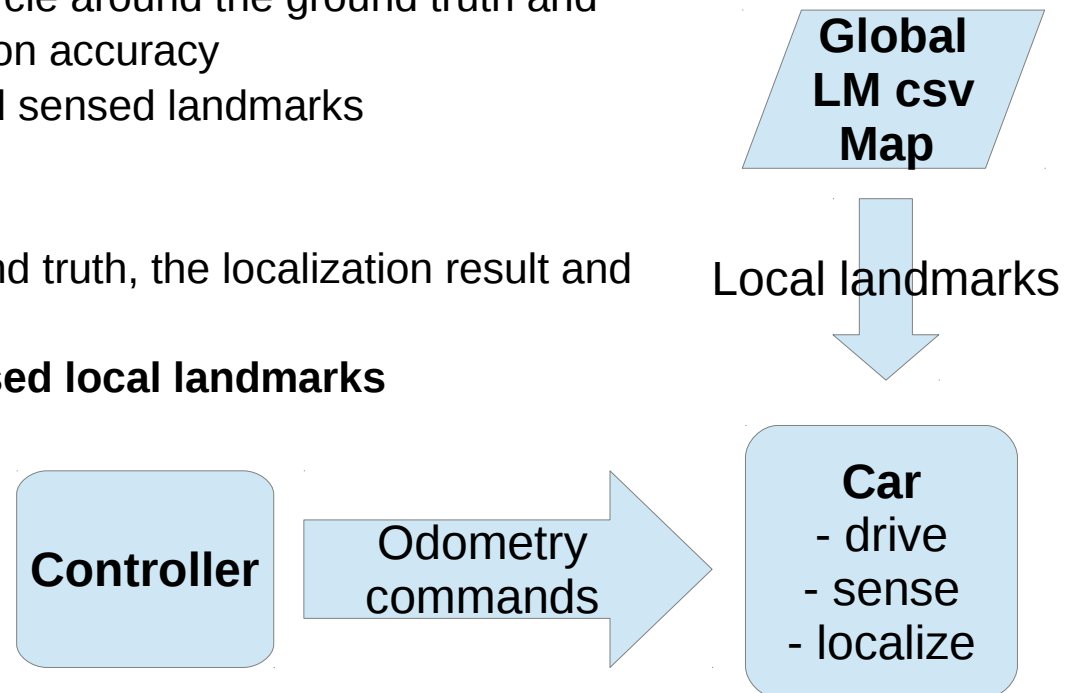
Localization

- Factors used are:
  - **Initial** factor for initial pose estimation
  - **Between** factor for relative (odometry) commands
 - **Unary** factors from the measurements

- Motion model:

$$\begin{pmatrix} x_{i+1} \\ y_{i+1} \\ \Theta_{i+1} \end{pmatrix} = \begin{pmatrix} x_i \\ y_i \\ \Theta_i \end{pmatrix} + \begin{pmatrix} v_x * \cos(\Theta_i) - v_y * \sin(\Theta_i) \\ v_x * \sin(\Theta_i) + v_y * \cos(\Theta_i) \\ v_{\Theta_i} * \Delta t \end{pmatrix}$$

- Measurement model: $h(z) = \begin{pmatrix} x \\ y \\ \Theta \end{pmatrix}$

Try it yourself!

- 1. Requirements
  - ROS
  - GTSAM

- 2. Download from this repository

- 3. Create a catkin workspace and
  copy the files into it

- 4. Invoke the following commands
  - catkin_make
  - source devel/setup.bash
  - roscore
  - rosrun car car_node
  - rosrun controller controller_node

- 5. For visualization, invoke rosrun rviz rviz and make sure that rviz
  subscribes to the topics specified in car/src/car_node.cpp
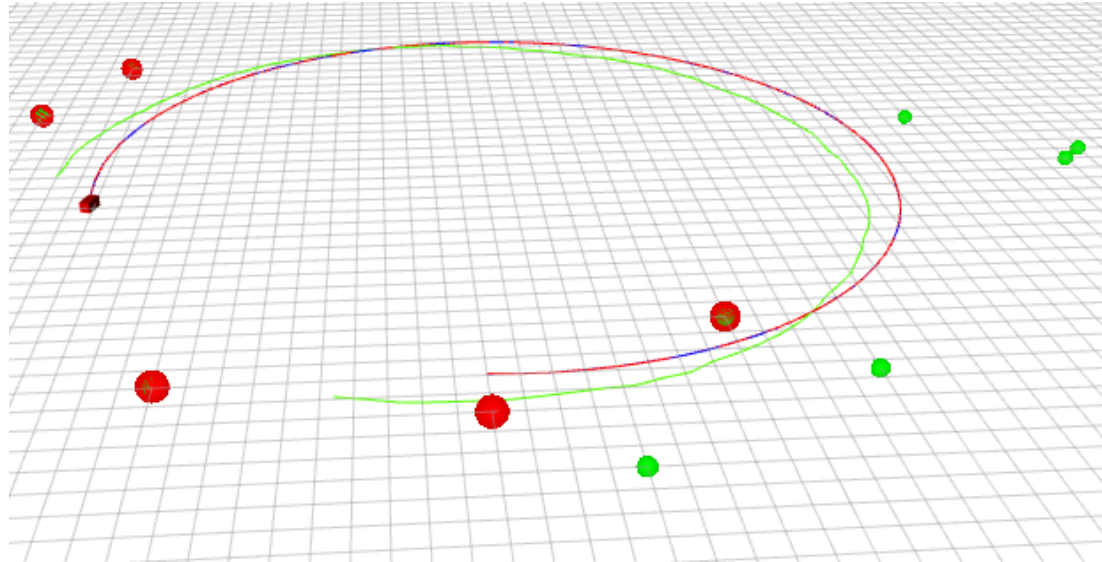
Try it yourself!

- 6. Experiment with:

  - Odometry (commands and noise)

  - Landmarks (change map and measurement noise)

  - Maximum size of graph

- 7. Build your own projects!