

# 3D Camera Localization - Probabilistic Robotics

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## 1 Introduction

This project aims to reconstruct the poses of a vehicle, using a Least-Squares approach, using only the odometry and anonymized observations. Using G2O simulator, we obtain a landmarks based map, the robot's perceptions and the transition of the robot between two poses (the odometry).

## 2 Implementation

### 2.1 Data Structure

All the project's data structures are parsed from the G2O file by the *g2o\_parser* class.

- **Landmark:** This object store the position of a landmark and his ID.
- **Pose:** Used to represent a robot's pose and his ID.
- **Transition:** This structure represent the transition of the robot from pose A to B. In it are stored the two poses and their relative ID.
- **Observation:** This kind of data structure is used to represent the robot's observations. It contains the pose's ID, a vector of 2d landmarks and two vector with the relative ID and depth of each one.
- **DictPoints:** This dictionary is used to store match of 2D and 3D projection of a landmark.

### 2.2 Camera

This class is used to represent a camera, initialized with the resolution (rows and columns), the camera matrix and the initial position in the space. In it we find the two principal methods *projectPoints* and *unprojectPoint*, used to transform a point from 2D to 3D and vice versa.

### 2.3 Distance Map

A distance map approach is used in order to compute the data associations. After each transition, a distance map is computed using the G2O's observations as reference and are computed the correspondences with the projections of the landmarks observed as reference.

## 2.4 Main

The main class is used to parse the data coming out from the G2O simulation into the main data structures of the project and, for each transition of the robot, the observation in a range are stored and used to compute the data association, in order to use the least square to correct the predicted position of the robot.

## 3 Least-Squares Approach

### 3.1 Prediction

The prediction of the robot's pose at time  $t$  is obtained applying the odometry transition at time  $t$  to the pose  $t-1$  (*current\_camera\_pose* \* *cameraToRobot* \* *motion* \* *robotToCamera*).  
 $h^{[n]} = proj(KX^{-1}p^{[n]}).$

### 3.2 Error

The error is computed as the difference between the 2D projection of the landmarks, using the predicted pose, and the observation of the same:  
 $e^{[n,m]} = h^{[n]} - z^{[m]}$

### 3.3 Jacobian

For each landmarks, the column  $i$  of the  $2 \times 6$  jacobian is computed using the formulae of the numerical differentiation:  $\frac{f(i+\epsilon) - f(i-\epsilon)}{2\epsilon}$ , where  $i$  is the  $i$ -th element of the state  $\mathbf{X}$ .

### 3.4 Correction

$$\begin{aligned} H &\leftarrow H + J_i^T \Omega_i J_i \\ b &\leftarrow b + J_i^T \Omega_i e_i \\ \Delta x &\leftarrow solve(H\Delta x = -b) \\ X^* &\leftarrow X^* \boxplus \Delta x \end{aligned}$$

## 4 Experimental Results

## 5 Front-End