

# Final Presentation Master Thesis

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Frederike Duembgen

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# Outline

## Introduction

- Motivation

- Project Goals

- Theoretical Concepts

- Results Mapping

- Mapping

- Localization

## Conclusion

# Dense SLAM with B-Splines

## Motivation

Picture of Sparse SLAM    Picture of Dense SLAM

# Project Goal and Methodology

## Project Goals

Create framework for surface reconstruction and localization using moving stereo camera based on moving monocular camera case.

## Methodology

### Added functionalities

- Create spline surface representation from static stereo camera.
- Localize new stereo camera position using obtained map.

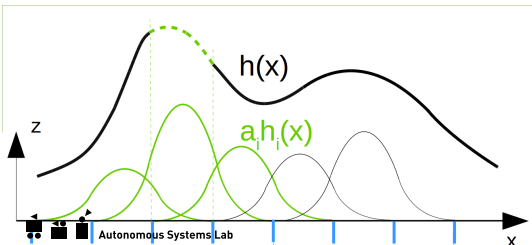
- Simulation environment ROS with `rviz` for pointcloud and `opencv` for image handling.
- Implementation of optimization algorithm using Eigen's sparse matrix solvers.
- Localization method using photometric errors only and generic rotation representation

# B-Splines for surface representation

## Theoretical Concepts

*Splines*: piecewise polynomial function of degree  $< d$ .

*B(asis)-Splines*: Specific choice of finite-support splines calculated by Cox de Boor recursion formula (support  $s = d + 1$ ).

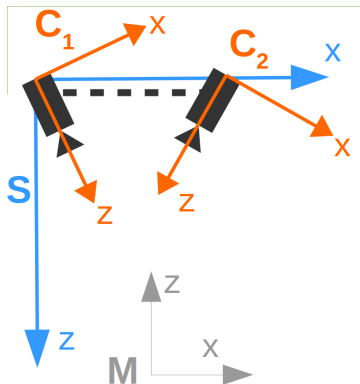


$$h(x_j) = \sum_{i=0}^M a_i h_i(x_j)$$

$$= \sum_{i=k}^{k+s} a_i h_i(x_j), \text{ for } j = 1 \dots N$$

# Camera setup

## Theoretical Concepts



Camera poses described by  ${}^M\mathbf{r}_{MC_k}$  and  $\mathbf{C}_{C_kM}$  for  $k = 1, 2$  or

$$\xi_{C_k} := \begin{bmatrix} {}^M\mathbf{r}_{MC_k}, \Phi_{CM} \end{bmatrix}^T$$

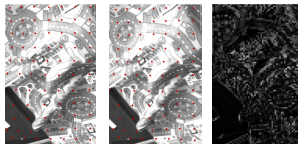
$$\xi_S := \begin{bmatrix} {}^M\mathbf{r}_{MS}, \Phi_{SM} \end{bmatrix}^T$$

with  $\Phi_{CM}, \Phi_{SM} \in \mathbb{R}^3$  [?]

If stereo rotation matrices  $\mathbf{R}_k := \mathbf{C}_{SC}$  and baseline  $T_x := s\mathbf{r}_{C_1C_2}$  are known, one pose of  $\{\xi_{C_1}, \xi_{C_2}, \xi_S\}$  is sufficient for all poses to be defined.

# Photometric errors for mapping

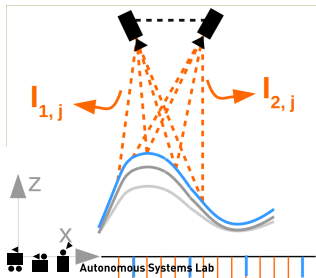
## Theoretical Concepts



Photometric error of grid point  $x_j, y_j$ :

$$r_j = I_1(\mathbf{u}_{j,1}) - I_2(\mathbf{u}_{j,2}) ,$$

with  $I_1, I_2$  interpolated intensities at the locations  $\mathbf{u}_{j,k}$  in camera  $k = 1$  and  $k = 2$ .



$$\mathbf{u}_{j,k} = \mathbf{K}_k D_k(\mathbf{T}_k(M \mathbf{r}_{MX_j}))$$

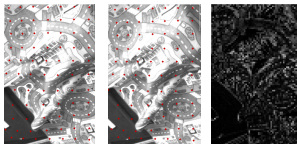
$$\begin{aligned} \mathbf{T}_k(M \mathbf{r}_{MX_j}) &= \pi(\mathbf{C}_k \mathbf{r}_{C_k X_j}) \\ &= \pi(\mathbf{C}_{C_k M}(M \mathbf{r}_{MX_j} - M \mathbf{r}_{MC_k})) \end{aligned}$$

3D point given by spline map:

# Photometric errors for mapping

## Theoretical Concepts

The analytical Jacobian



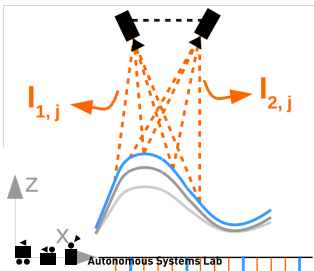
$$\mathbf{J}_r(\mathbf{a}) = \frac{\partial \mathbf{r}(\mathbf{a})}{\partial \mathbf{a}} \in \mathbb{R}^{N \times M}$$

is obtained by the chain rule:

$$\mathbf{J}_r(\mathbf{a}) = (\mathbf{J}_{pixel,1} \mathbf{J}_{camera,1} (M \mathbf{r}_{MX_j}) - \mathbf{J}_{pixel,2} \mathbf{J}_{camera,2} (M \mathbf{r}_{MX_j})) \mathbf{J}_{splines}$$

with

$$\mathbf{J}_{pixel,k} = \frac{\partial I_k(\mathbf{u}_k)}{\partial \tilde{\mathbf{u}}_k}, \quad \mathbf{J}_{camera,k}(M \mathbf{r}_{MX_j}) = \frac{\partial \tilde{\mathbf{u}}_k}{\partial M \mathbf{r}_{MX_j}}$$





# Optimization problem for mapping

## Theoretical Concepts

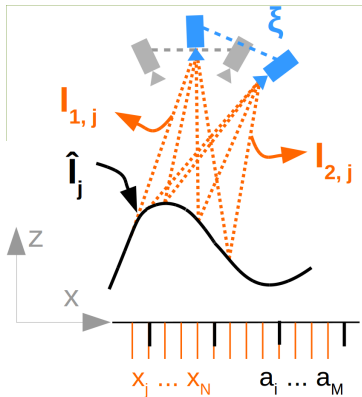
$$\begin{aligned}\hat{\mathbf{a}} &= \arg \min_{\mathbf{a} \in \mathbb{R}^M} f(\mathbf{a}) \\ &= \arg \min_{\mathbf{a} \in \mathbb{R}^M} \frac{1}{2} \left( \sum_{j=0}^N w_j r_j(\mathbf{a})^2 + \beta \mathbf{a}^T \mathbf{B} \mathbf{a} + \gamma \mathbf{a}^T \mathbf{G} \mathbf{a} \right),\end{aligned}$$

with

- **bending** and **gradient** energy regularization terms and
- **weight** representing the average visibility of point  $j$ .

# Photometric errors for localization

## Theoretical Concepts



Photometric error of grid point  $x_j, y_j$ :

$$r_{j,1} = l_1(\mathbf{u}_{j,1}) - \hat{l}(x_j, y_j)$$

$$r_{j,2} = l_2(\mathbf{u}_{j,1}) - \hat{l}(x_j, y_j) ,$$

with  $l_1, l_2$  interpolated intensities at pixels  $\mathbf{u}_{j,k}$  in camera  $k = 1$  and  $k = 2$  and  $\hat{l}(x_j, y_j)$  the estimated intensity from previous step.

# Photometric errors for localization

## Theoretical Concepts

The analytical Jacobian

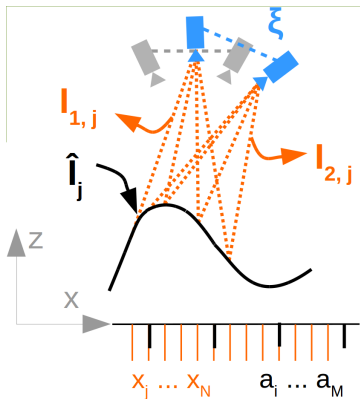
$$\mathbf{J}_r(\xi) = \frac{\partial \mathbf{r}(\xi)}{\partial \xi} \in \mathbb{R}^{N \times 6}$$

is obtained by the chain rule:

$$\mathbf{J}_r(\xi) = \mathbf{J}_{pixel} \mathbf{J}_{camera}(\xi)$$

with

$$\mathbf{J}_{pixel} = \frac{\partial l(\mathbf{u})}{\partial \tilde{\mathbf{u}}}, \quad \mathbf{J}_{camera}(\xi) = \frac{\partial \tilde{\mathbf{u}}}{\partial \xi}$$



# Optimization problem for localization

## Theoretical Concepts

$$\hat{\xi} = \arg \min_{\xi \in \mathbb{R}^6} \frac{1}{2} \sum_{j=0}^N r_j(\xi)^2$$

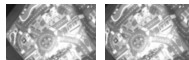
# Datasets and parameters

## Results Mapping

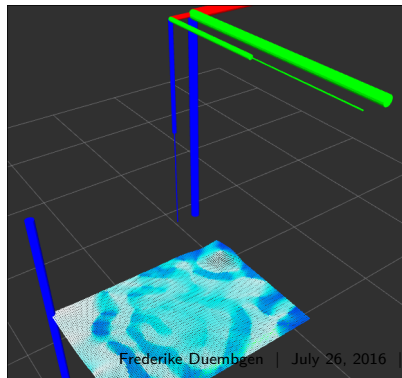
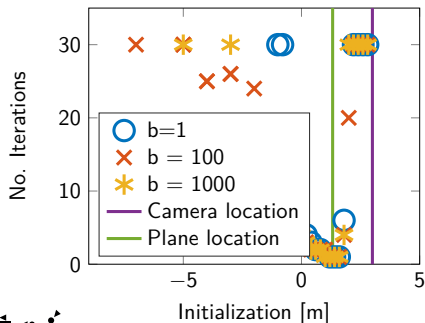
Dataset	Plane test	Middlebury [?]	Inhouse
Ground truth	analytical	structured light	pattern matching
Images	rectified	rectified	non rectified
Calibration	+++	++	+
Mapping	yes	yes	yes
Localization	yes	no	no
Spline resolution	20 × 20	75 × 100	
Map dimensions	0.9 × 1.2	1.5 × 2.0	
Map resolution	90 × 120		

# Plane test case

## Mapping

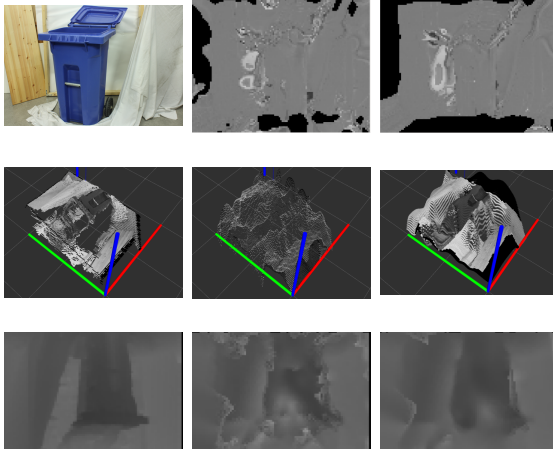


Convergence Study Plane Test



# Middlebury dataset

## Mapping



# Inhouse dataset

## Mapping

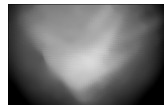
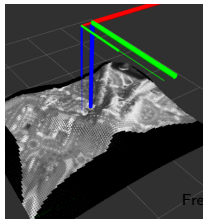
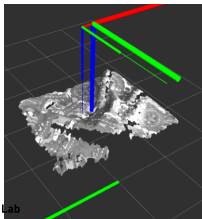
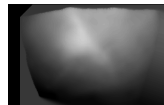
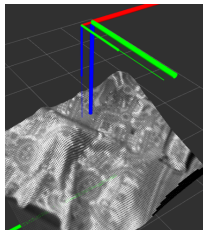
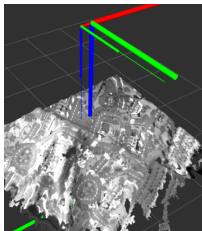
Parameters:  $\beta = 10$   $\gamma = 100e3$   $d = 100$

Parameters:

$\beta = 10$

$\gamma = 1e6$

$d = 100$





# Plane test case

## Localization

# Achievements

## Localization

# Suggestions For Future Work

## Localization