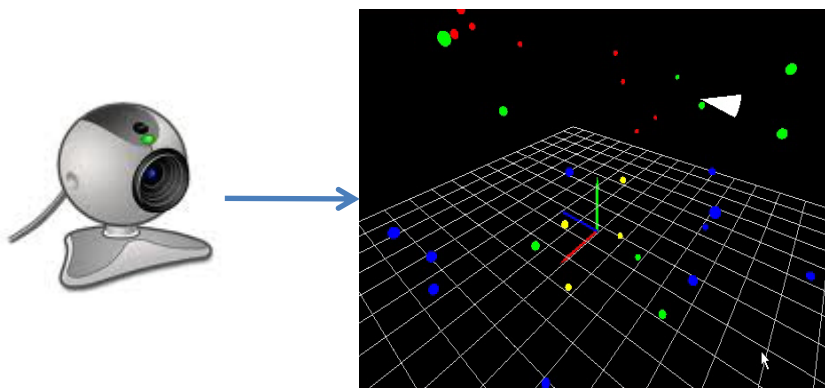


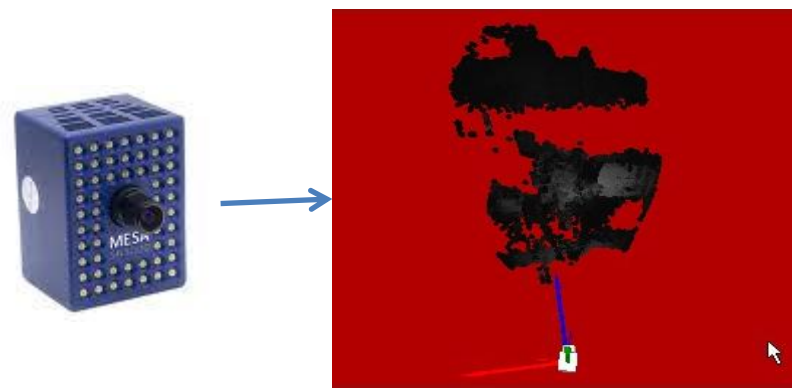
3D scene reconstruction using a Time-Of-Flight camera

- Introduction
- Camera Time-Of-Flight
- Data filtering
- Camera motion estimation and 3D map
- Meshes generation from points cloud
- Conclusion & perspectives

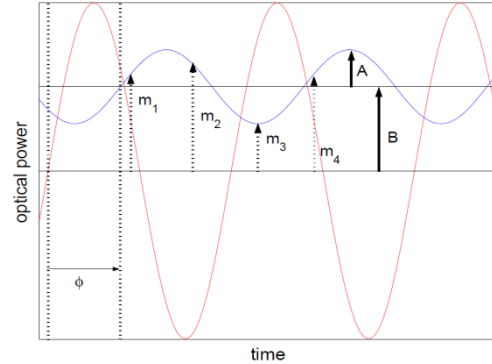
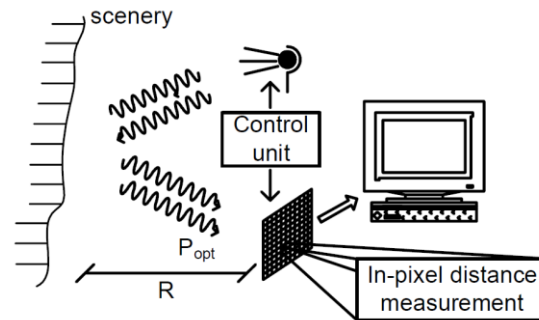
- Goal: perform a 3D reconstruction of the surrounding environment using a Time-of-flight (TOF) camera
- Context: static scene and moving camera (Simultaneous Localization And Mapping problem: SLAM)
- Previous works:
 - SLAM using monocular camera
 - Limitations: sparse reconstructed map
- New method: based on Time-Of-Flight sensor (dense reconstruction)



Webcam SLAM



TOF SLAM



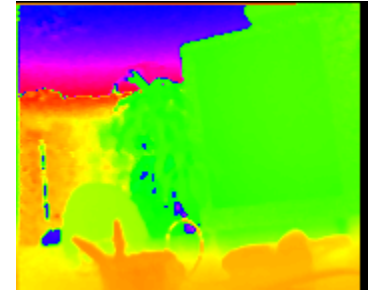
- Camera Time-of-flight (TOF) is an active range finder:
 - Emits light (LEDs) and Measures TOF (CCD sensor)
- Reconstructed signal (4 measurements each period) characterized by
 - Offset B: Mean of the total light incident on the sensor: background plus modulated signal
 - Amplitude: Object reflectivity and object distance
 - Phase shift: Distance
- The distance accuracy of the measurement depends on the acquired amplitude A and offset B

TOF camera (pros and cons)

- Swissranger SR3000
 - Number of LEDs: 55
 - Resolution: 176x144
 - Output: amplitude + distance
 - Frame rate: 30 fps



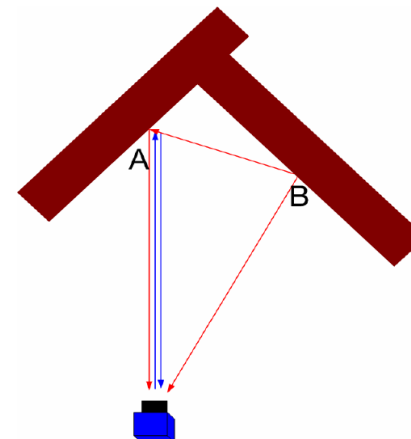
Amplitude



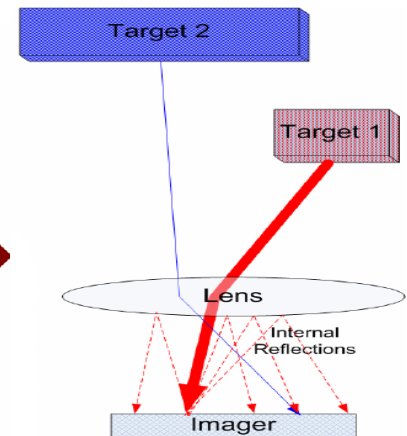
Depth

- Advantages:
 - Easy to extract range information
 - High frame rate
 - Not suffer from missing texture in the scene or bad lighting conditions
 - Compactness of the sensor

- Disadvantages:
 - Multiple reflections
 - Light scattering
 - Low resolution (176x144)
 - Camera network



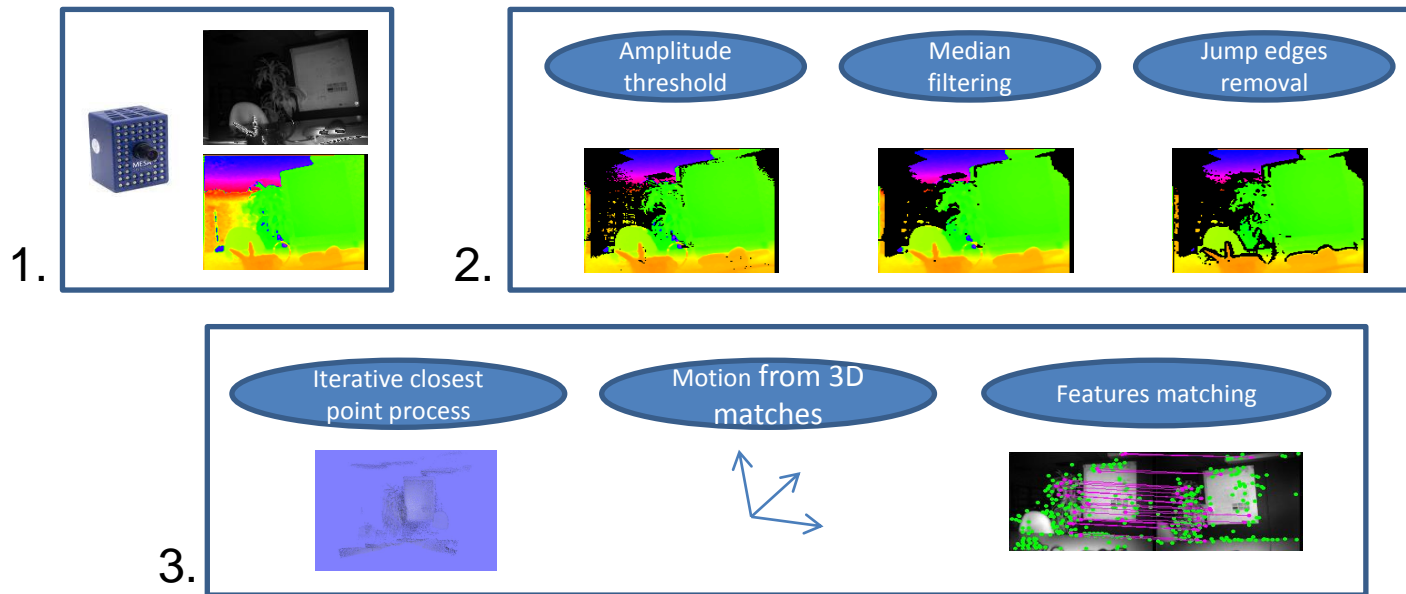
Multiple reflections



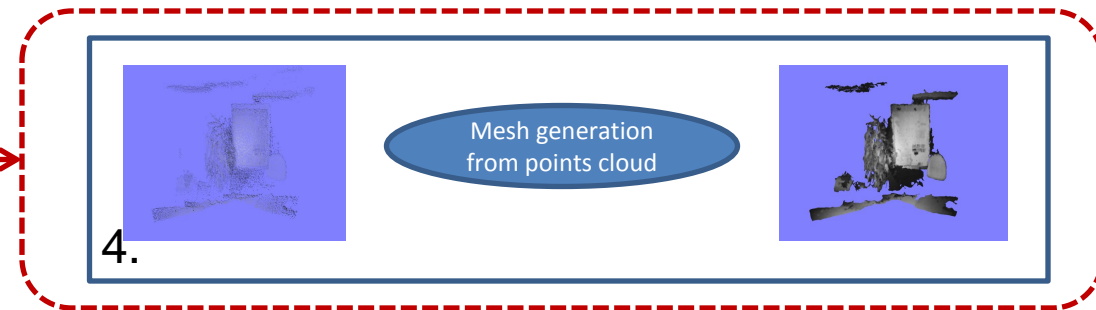
Light scattering

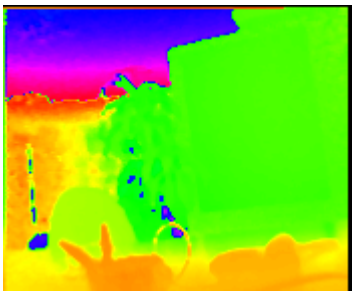
Whole process overview

ONLINE
OFFLINE



1. Acquisition (TOF)
2. Pre processing (Data filtering)
3. Processing (Map generation)
4. Post processing (Mesh building)



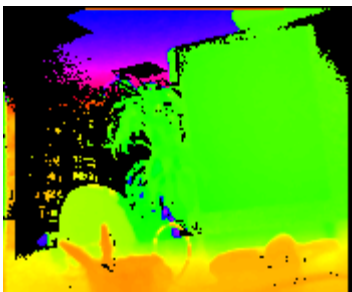


Raw distance



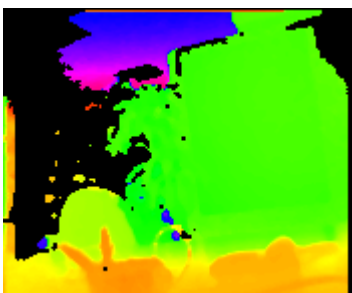
Raw amplitude

- Raw data provided by TOF camera are very noisy
- Necessary to clean it



Amplitude threshold

- Depth accuracy depend on amplitude value
 - Lower is the amplitude, lower is the accuracy
 - Far objects
 - Low reflectivity
- Threshold based on amplitude

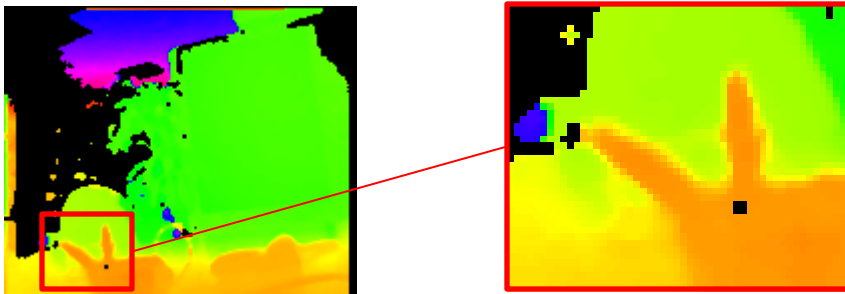


Median filtering

- Pepper and salt noise is very important due to
 - Sensor noise as light scattering
 - Amplitude thresholding
- Median filtering

Jump edge filtering

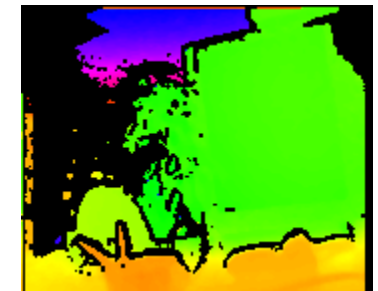
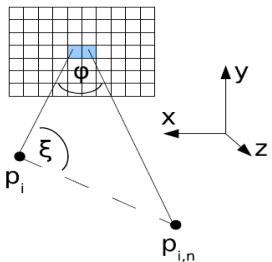
- Problems on abrupt distance transitions
 - TOF camera smooth abrupt depth transitions
 - A small measurement error provide an important error on distance value



- Green: chair
- Orange: pencils
- Yellow: Noise (smooth transition)

-> JUMP EDGE FILTERING

- Higher is the $\xi_{i,n}$ angle value, higher is the depth transition
- Knowing camera intrinsic parameters, compute $\xi_{i,n}$ (from ϕ and 3D points)
- Remove pixels with at least 2 neighbors with an angle $\xi_{i,n} > s$

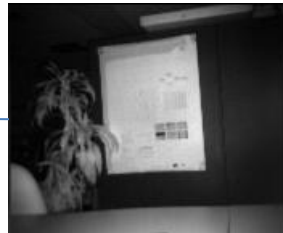


Jump edges filtering

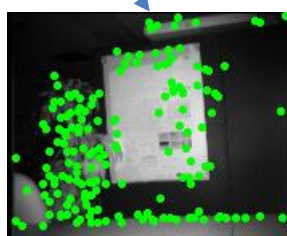
Amplitude image (t-n)



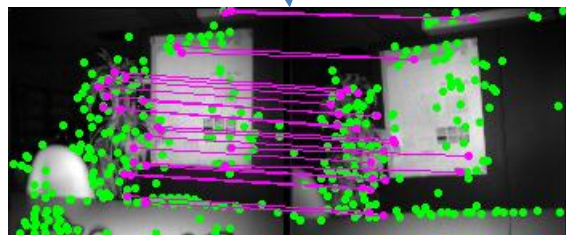
Amplitude image (t)



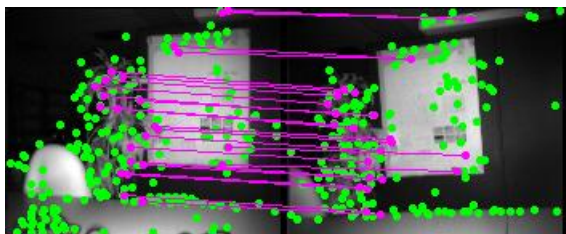
2D features
extraction



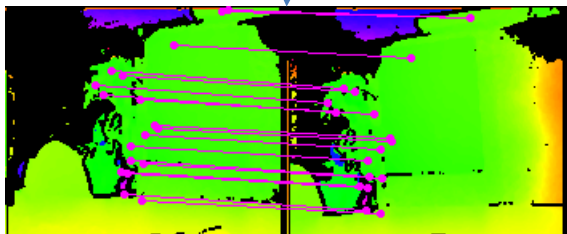
Features
matching



- Use Amplitude image (2D information) to make matches between 2 images
 - Detect SIFT features into the current frame
 - Match features using SIFT descriptors between the last reference frame and the current one



Extract 3D matches



Resolve equation system



Camera motion

- For each match $p_1 = Rp_2 + t$
- Rotation matrix is represented as a quaternion
 - Reduce number of parameters to estimate
 - Maintain constraints (matrix normalization)

$$\mathbf{q} = (w, x, y, z) \mapsto \mathbf{R} = \begin{bmatrix} w^2 + x^2 - y^2 - z^2 & 2xy - 2wz & 2xz + 2wy \\ 2xy + 2wz & w^2 - x^2 + y^2 - z^2 & 2yz - 2wx \\ 2xz - 2wy & 2yz + 2wx & w^2 - x^2 - y^2 + z^2 \end{bmatrix}$$

- Estimate q from 3D matches (rotation)
 - Linear equation system (SVD)
- Apply rotation found to the first point set
- Compute centroids of each point set
- Extract t from point centroids (translation)

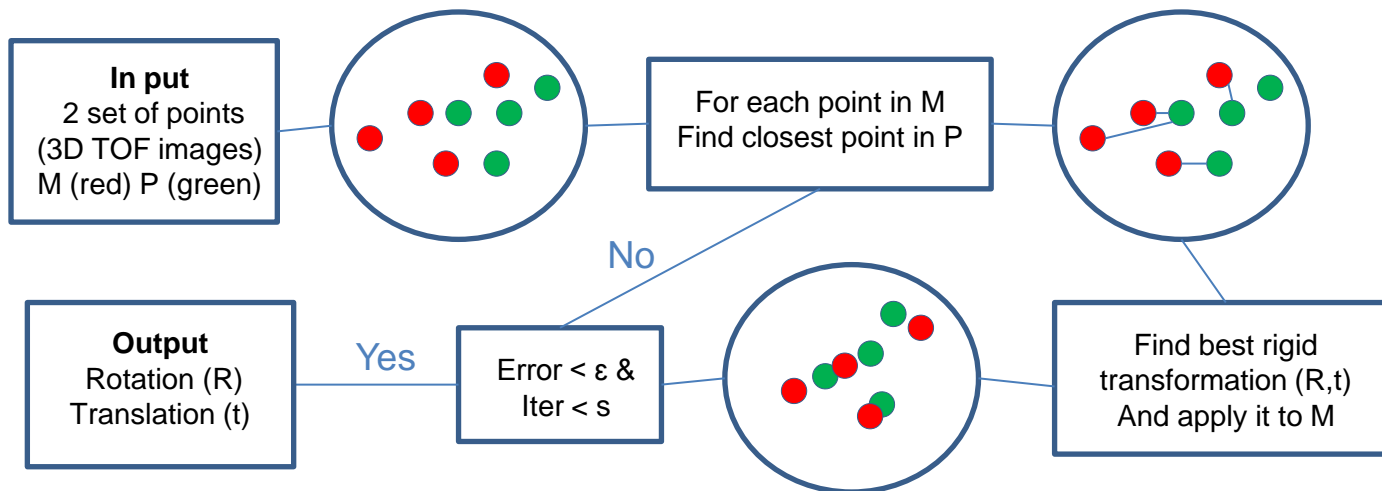
- Two kind of outliers
 - Bad matches between 2 SIFT points
 - Bad distance measurement by TOF camera

→ RANSAC

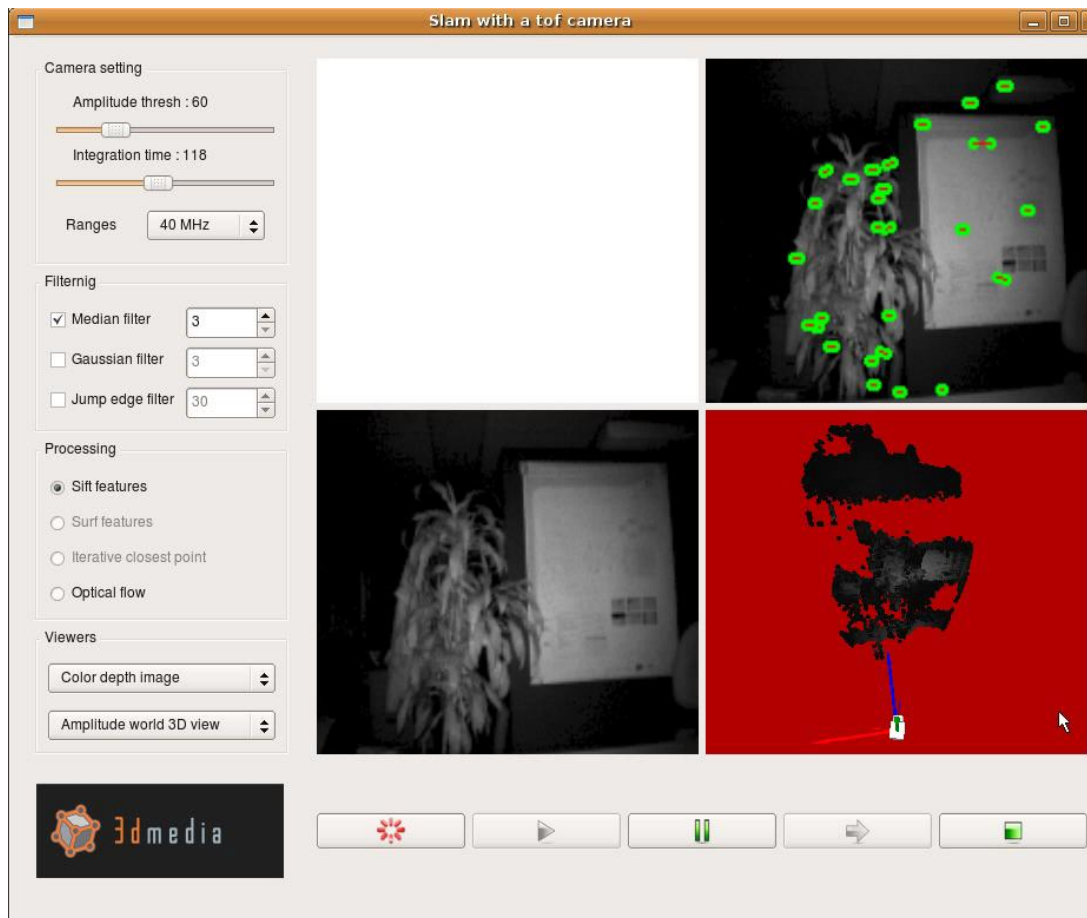
- while iterations < k
 - Select randomly 3 points into the two 3D point sets $S1$ and $S2$
 - Estimate q and t
 - Apply (q,t) to not selected points in $S1$ ($S'1$)
 - Compute error between $S'1$ and corresponding points in $S2$
 - Extract list of P points with an error < e ($S''1$)
 - If ($P > d$)
 - evaluate (q,t) with all points into ($S''1$)
 - If (q,t) is the best model, update the best model (q,t)
- Return the best model (q,t)

- If RANSAC finished correctly and estimate motion is higher than a threshold (rotation or translation)
 - Update camera motion and 3D map
 - Current image is the next reference image

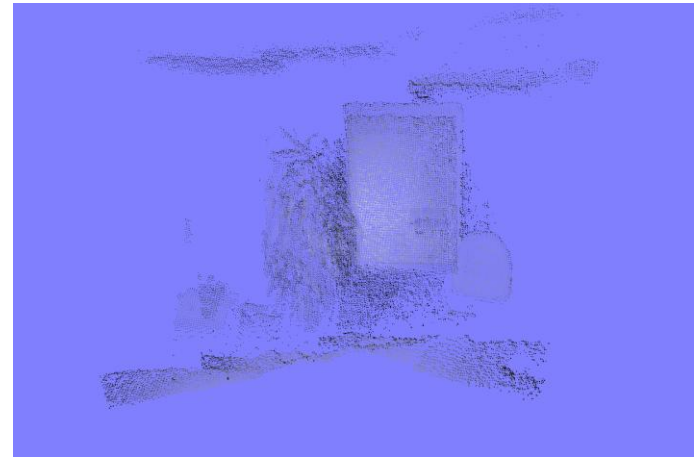
- Estimate the motion using only few matches is not enough
- A global optimization is required to improve preciseness of the motion estimation → Iterative Closest Point (ICP)



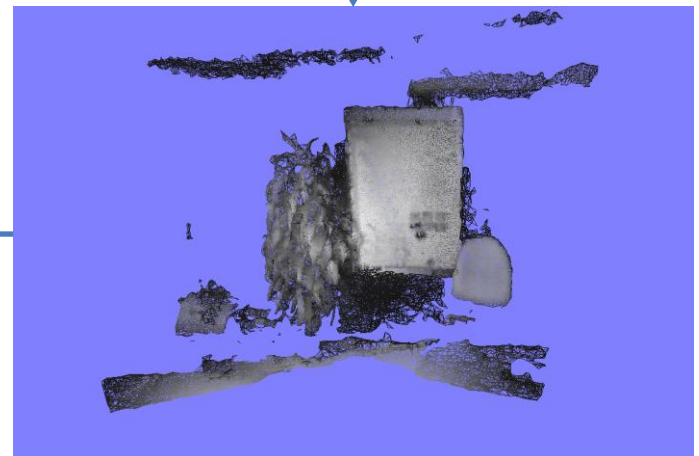
- Whole distance maps are used to estimate transformation
- The output motion is validated if error is small (not based on iter) and motion is lower than a threshold



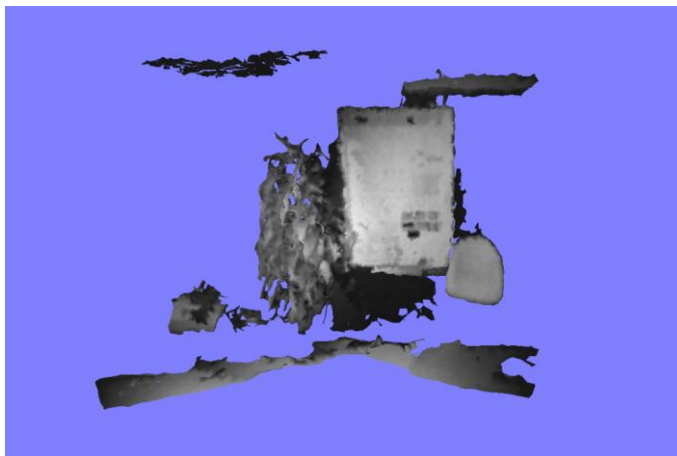
- Generate a mesh from 3D point cloud (meshlab library)
 - **Ball pivoting**, marching cubes
Poisson's reconstruction
- Filter mesh
 - **Based on objects size**
 - Subdivision algorithms
 - **Hole filling**
 - **Laplacian smoothing**, ...



Surface reconstruction



Filtering



- Conclusion
 - SLAM using TOF camera
 - Dense 3D map reconstruction of the scene
 - Map represented by meshes

- Perspectives
 - Investigate different versions of ICP algorithm (Trimmed ICP, ...)
 - Calibration of the depth map (relative to amplitude image, ...)
 - Use a temporal motion estimation model (Extended Kalman Filtering, Particle Filtering)

Thank you for your attention

Questions?