



pointcloudlibrary

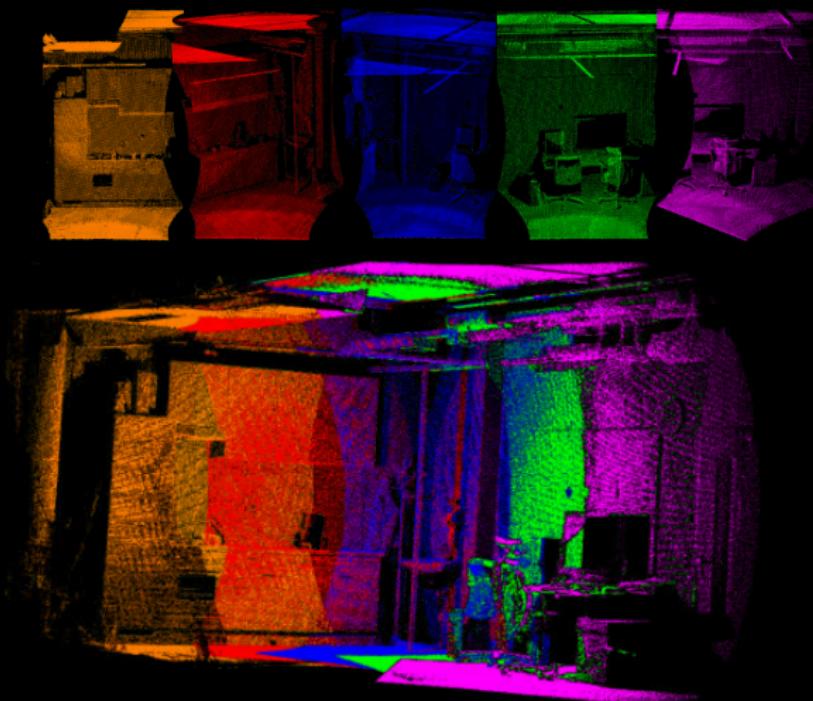


PCL :: Registration

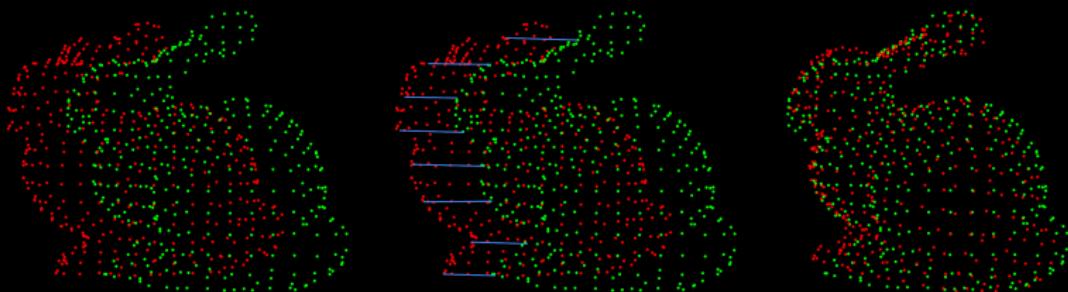
Jochen Sprickerhof

May 18, 2012

Registering **3D Point Clouds** for building 3D models of objects, table scenes, and whole rooms/buildings/outdoor environments.



Registering 3D Point Clouds



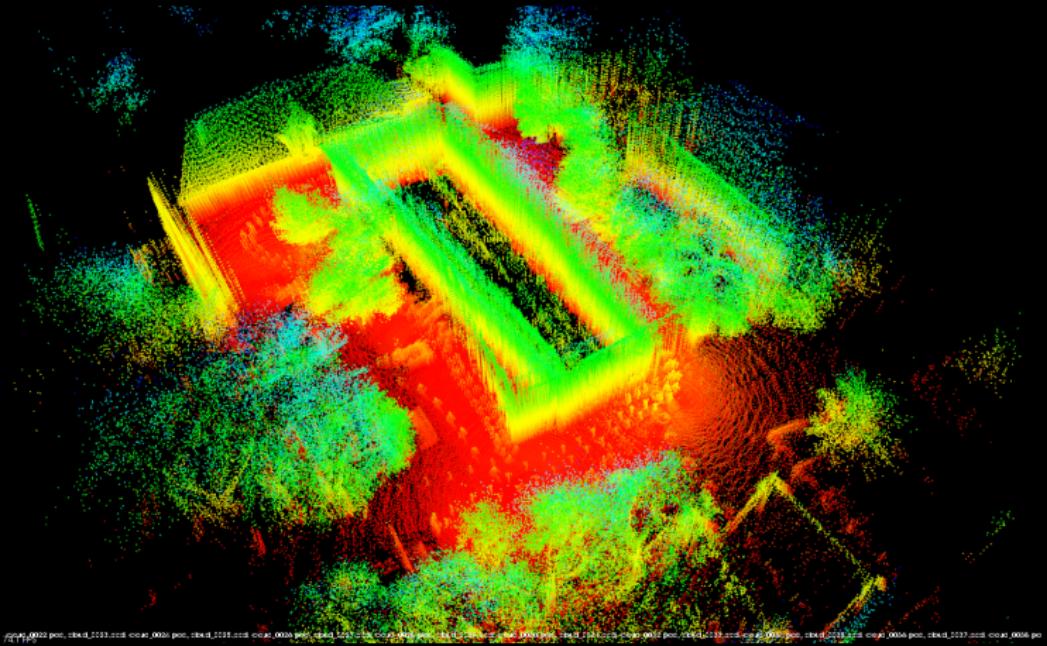
- ▶ Given a **source** point cloud and a **target** point cloud
 1. determine **correspondence pairs**,
 2. estimate a transformation that aligns the **correspondences**,
 3. apply the transformation to align **source** and **target**.

Code: ICP

```
pcl::IterativeClosestPoint<pcl::PointXYZ, pcl::PointXYZ> icp;  
icp.setInputCloud(cloud_in);  
icp.setInputTarget(cloud_out);  
pcl::PointCloud<pcl::PointXYZ> Final;  
icp.align(Final);
```

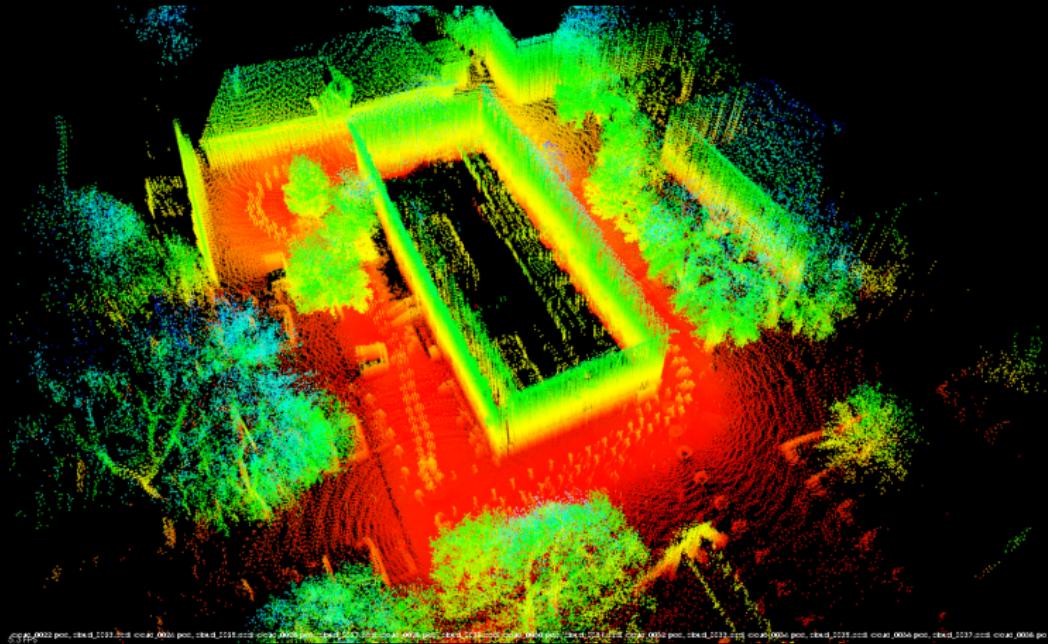
The ICP API: Termination Criteria

- ▶ Max. number of iteration steps
→ set via `setMaximumIterations(nr_iterations)`
- ▶ Convergence: Estimated transformation doesn't change
(the sum of differences between current and last
transformation is smaller than a user-defined threshold)
→ set via `setTransformationEpsilon(epsilon)`
- ▶ A solution was found (the sum of squared errors is smaller
than a user-defined threshold)
→ set via `setEuclideanFitnessEpsilon(distance)`



figures_0902.pcd, chkd_2039.pcd, chkd_0020.pcd, chkd_2038.pcd, chkd_0021.pcd, chkd_2037.pcd, chkd_0019.pcd, chkd_2036.pcd

Input Data



cloud_0022.pcd cloud_0023.pcd cloud_0024.pcd cloud_0025.pcd cloud_0026.pcd cloud_0027.pcd cloud_0028.pcd

```
pcl_icp -d 0.75 -r 0.75 -i 50 ../../data/cloud_0{022..130}.pcd
```

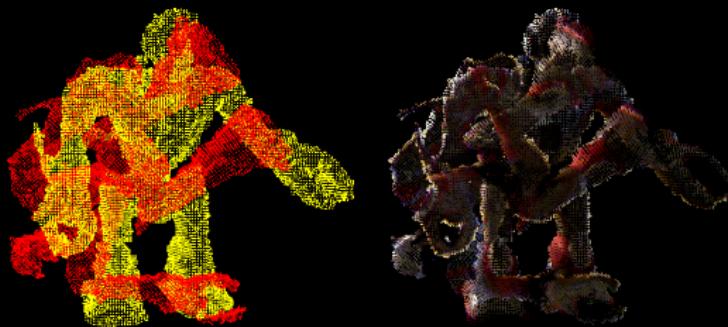
- ▶ Several methods for computing a transformation $T = (R, t)$ given correspondence pairs (d_i , m_i):
 - ▶ Point-to-point
 - ▶ Point-to-plane
 - ▶ Plane-to-plane
 - ▶ ... and many others
- ▶ Simple solution (based on SVD) for minimizing point-to-point distance (least squares error E):

$$E(T) = \sum_i (\mathbf{m}_i - (\mathbf{R}\mathbf{d}_i + \mathbf{t}))^2$$

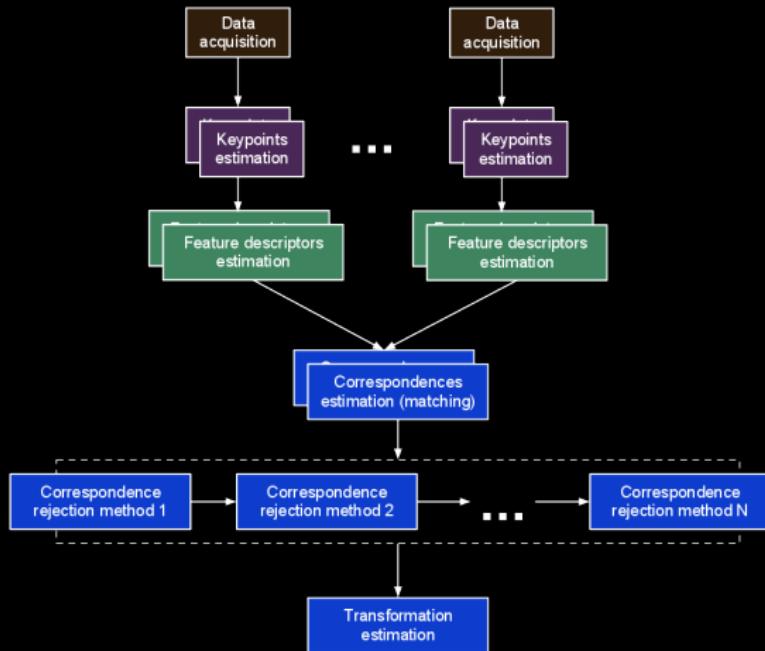
Code: Transformation estimation

```
Eigen::Matrix4f transformation;  
TransformationEstimationSVD<PointT, PointT> svd;  
svd.estimateRigidTransformation(src, trgt, corres, trans);
```

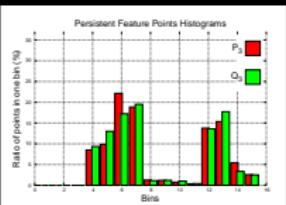
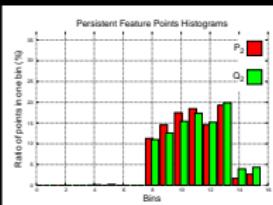
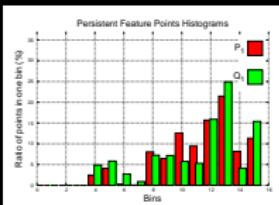
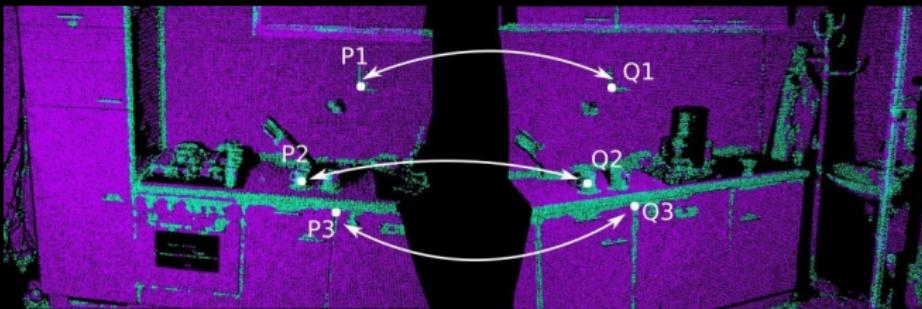
Example: Simple initial alignment



Problem: False correspondences!



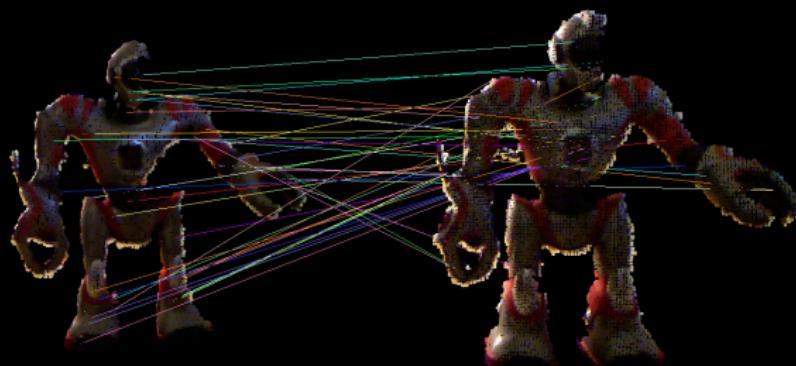
1. Compute sets of **keypoints**
2. Compute (**local**) feature descriptors
3. Match features to find correspondences
4. Estimate transformation from correspondences



Code: Matching Features

```
CorrespondenceEstimation<FeatureT, FeatureT> est;  
est.setInputCloud (source_features);  
est.setInputTarget (target_features);  
est.determineCorrespondences (correspondences);
```

Example: Found correspondences / matches



Rejecting false correspondences (outliers) using SAC

- ▶ Draw three correspondences pairs d_i, m_i
- ▶ Estimate transformation (R, t) for these samples
- ▶ Determine inlier pairs with $((Rd_i + t) - m_i)^2 < \epsilon$
- ▶ Repeat N times, and use (R, t) having most inliers

Code: SAC-based correspondence rejection

```
CorrespondenceRejectorSampleConsensus<PointT> sac;
sac.setInputCloud(source);
sac.setTargetCloud(target);
sac.setInlierThreshold(epsilon);
sac.setMaxIterations(N);
sac.setInputCorrespondences(correspondences);
sac.getCorrespondences(inliers);
Eigen::Matrix4f transformation = sac.getBestTransformation();
```

Example: SAC-based correspondence rejection

Initial correspondences:



Inliers:



Problem: In case of less descriptive features, the best match m_i may not be the true correspondence for d_i !

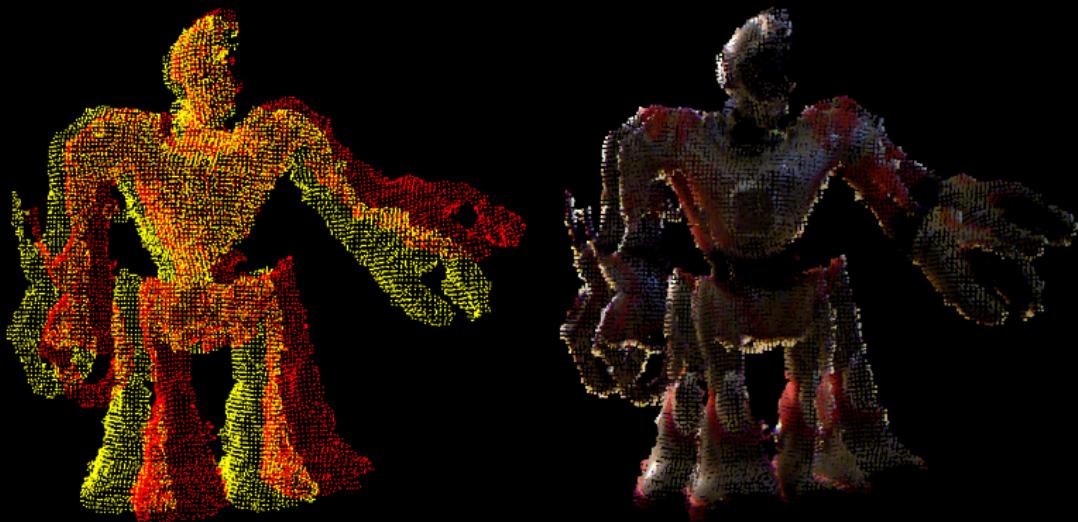
SAC-IA: Sampled Consensus-Initial Alignment

1. Draw n points d_i from the source cloud (with a minimum distance d in between).
2. For each drawn d_i :
 - 2.1 get k closest matches, and
 - 2.2 draw one of the k closest matches as m_i (instead of taking closest match)
3. Estimate transformation (R, t) for these samples
4. Determine inlier pairs with $((Rd_i + t) - m_i)^2 < \epsilon$
5. Repeat N times, and use (R, t) having most inliers

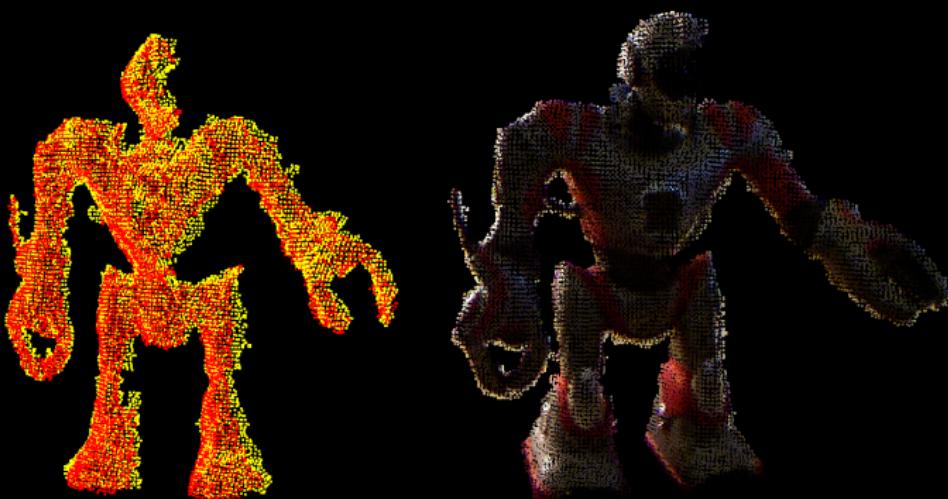
Code: Sampled Consensus-Inital Alignment

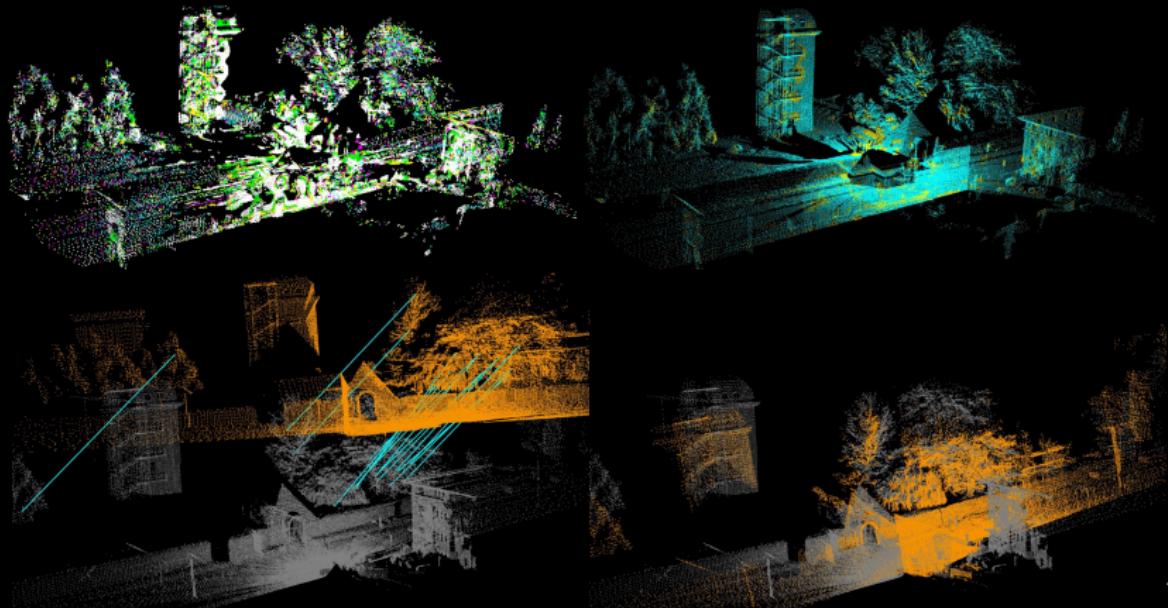
```
pcl::SampleConsensusInitialAlignment  
    <PointT, PointT, FeatureT> sac_ia;  
sac_ia.setNumberOfSamples (n)  
sac_ia.setMinSampleDistance (d);  
sac_ia.setCorrespondenceRandomness (k);  
sac_ia.setMaximumIterations (N);  
sac_ia.setInputCloud (source);  
sac_ia.setInputTarget (target);  
sac_ia.setSourceFeatures (source_features);  
sac_ia.setTargetFeatures (target_features);  
sac_ia.align (aligned_source);  
Eigen::Matrix4f = sac_ia.getFinalTransformation ();
```

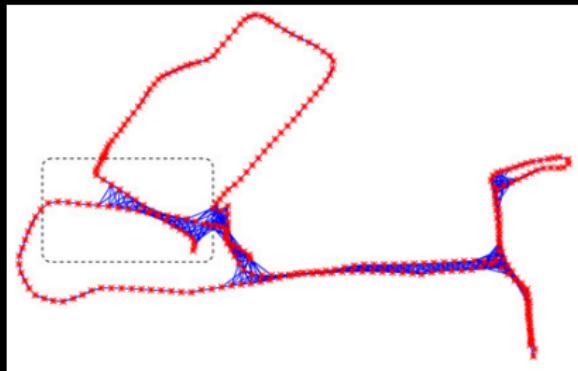
Example: Sampled Consensus-Initial Alignment



Sampled Consensus-Initial Alignment + refinement

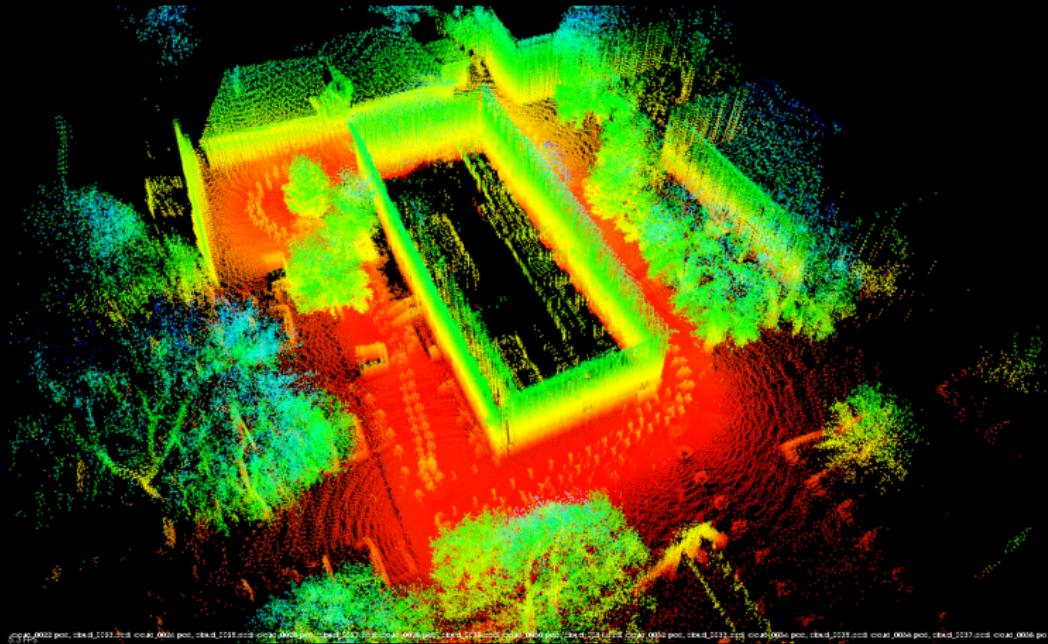






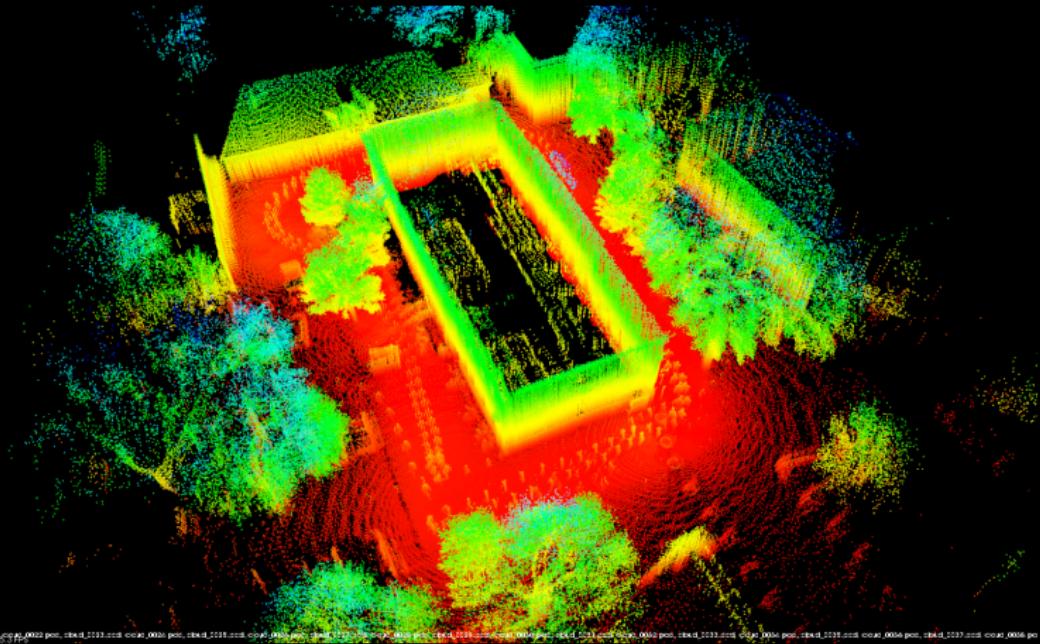
[Borrmann 2007]

```
pcl::registration::LUM<PointType> lum;
lum.setMaxIterations (lumIter);
lum.setConvergenceThreshold (0.001f);
for (int i = 1; i < n; i++)
{
    lum.addPointCloud (cloud[i]);
}
//for all point pairs (i, j)
lum.setCorrespondences (i, j, correspondences);
lum.compute ();
```



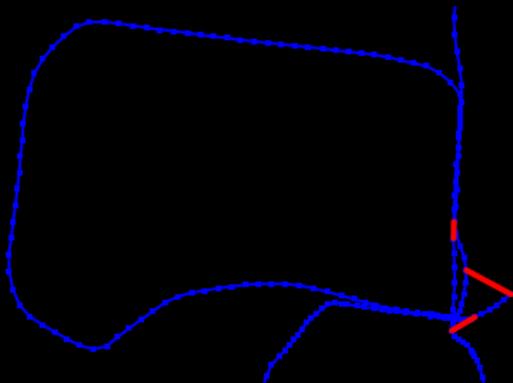
cloud_0022.pcd cloud_0023.pcd cloud_0024.pcd cloud_0025.pcd cloud_0026.pcd cloud_0027.pcd cloud_0028.pcd

```
pcl_icp -d 0.75 -r 0.75 -i 50 ../../data/cloud_0{022..130}.pcd
```

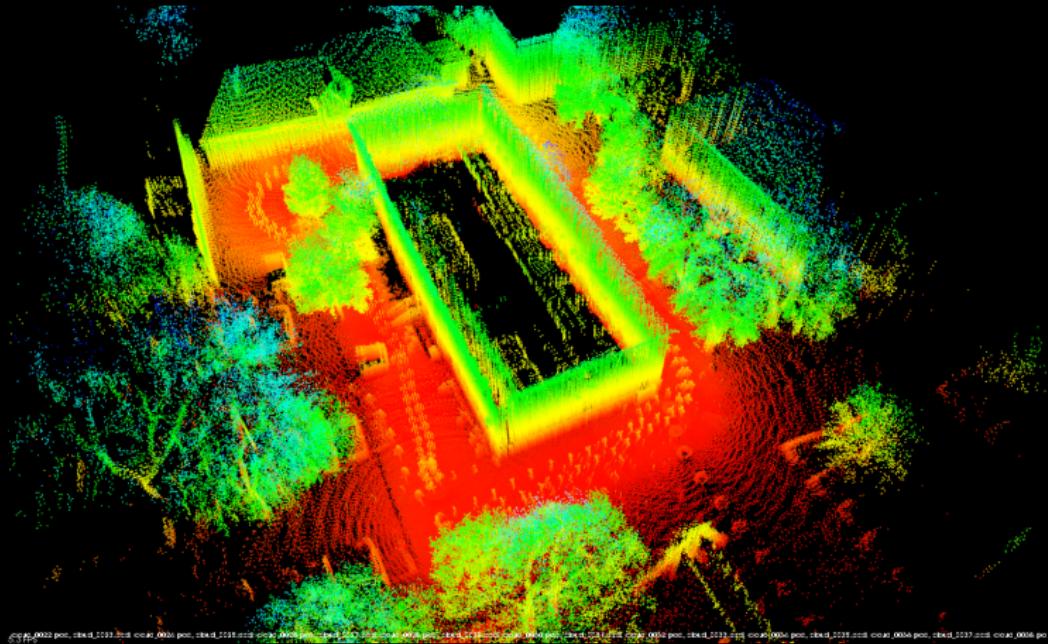


022.pcd 023.pcd 024.pcd 025.pcd 026.pcd 027.pcd 028.pcd 029.pcd 030.pcd 031.pcd 032.pcd 033.pcd 034.pcd 035.pcd 036.pcd 037.pcd 038.pcd

```
pcl_lum ..../icp/cloud_0{022..130}.pcd
```

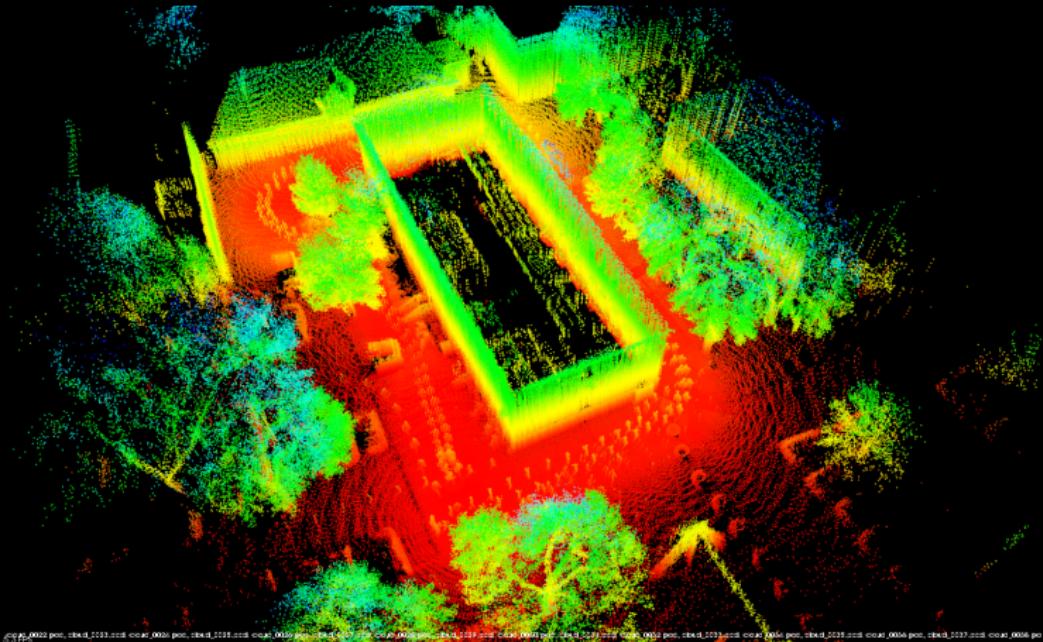


```
pcl::registration::ELCH<PointType> elch;
elch.setReg (icp);
for (int i = 1; i < n; i++)
{
    elch.addPointCloud (cloud[i]);
}
elch.setLoopStart (first);
elch.setLoopEnd (last);
elch.compute ();
```



cloud_0022.pcd cloud_0023.pcd cloud_0024.pcd cloud_0025.pcd cloud_0026.pcd cloud_0027.pcd cloud_0028.pcd

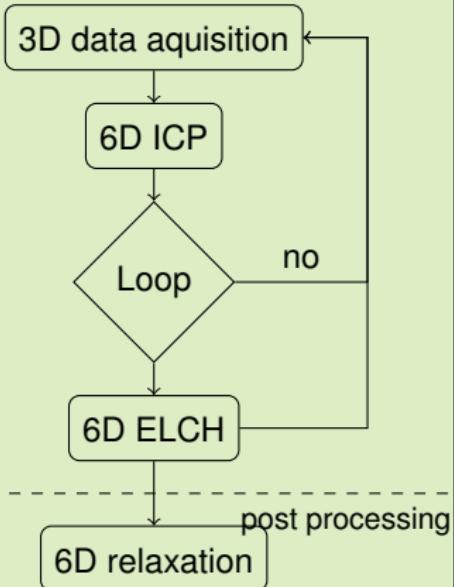
```
pcl_icp -d 0.75 -r 0.75 -i 50 ../../data/cloud_0{022..130}.pcd
```



cloud_0022.pcd, cloud_0023.pcd, cloud_0024.pcd, cloud_0025.pcd, cloud_0026.pcd, cloud_0027.pcd, cloud_0028.pcd, cloud_0029.pcd, cloud_0030.pcd, cloud_0031.pcd, cloud_0032.pcd, cloud_0033.pcd, cloud_0034.pcd, cloud_0035.pcd, cloud_0036.pcd

```
pcl_elch -d 1.0 -r 1.0 -i 100 ..\icp\cloud_0{022..130}.pcd
```

Pipeline



6D relaxiation:

- ▶ 6D Lu & Milios (LUM)
- ▶ G2O
- ▶ TORO

thanks to Frits Florentinus

