

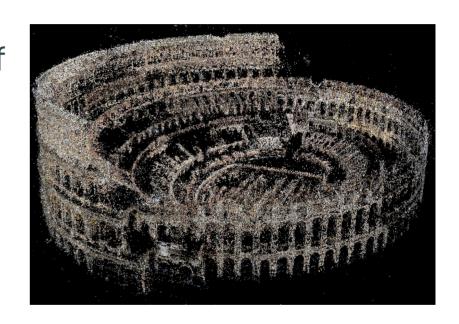
Università degli Studi di Padova

# 3D Data Processing Introduction to PCL and Open3D Basic Topics

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#### Point Cloud: a Definition

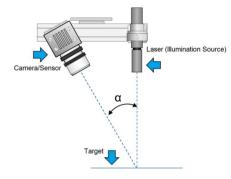
- A point cloud is a data structure used to represent a collection of multi-dimensional points and is commonly used to represent three-dimensional data.
- The points usually represent the X, Y, and Z geometric coordinates of a sampled surface.
- Each point can hold additional information: RGB colors, intensity values, etc...

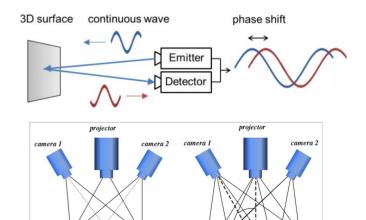


# Where Do They Come From?

- 2/3D Laser scans
- Laser triangulation
- Stereo cameras
- RGB-D cameras
- Structured light cameras
- Time of flight cameras







# Point Cloud Library

#### pointclouds.org

- The Point Cloud Library (PCL) is a standalone, large scale, open source (C++) library for 2D/3D image and point cloud processing.
- PCL is released under the terms of the BSD license, and thus free for commercial and research use.
- Among others, PCL depends on Boost, Eigen, OpenMP,...



#### PCL PointCloud

A PointCloud is a templated C++ class which basically contains the following data fields:

#### width (int)

- The total number of points in the cloud (equal with the number of elements in points) for unorganized datasets
- The width (total number of points in a row) of an organized point cloud dataset

#### height (int)

- Set to 1 for unorganized point clouds
- The height (total number of rows) of an organized point cloud dataset
- points (std::vector <PointT>): Contains the data array where all the points of type PointT are stored.

# Point Types

- PointXYZ float x, y, z
- PointXYZI float x, y, z, intensity
- PointXYZRGB float x, y, z, rgb
- PointXYZRGBA float x, y, z, uint32 t rgba
- Normal float normal(3), curvature
- PointNormal float x, y, z, normal(3), curvature
   → See pcl/include/pcl/point\_types.h for more examples.

#### PCL Structure

#### PCL is a collection of smaller, modular C++ libraries:

- **libpcl\_features**: many 3D features (e.g., normals and curvatures, boundary points, moment invariants, principal curvatures, Point Feature Histograms (PFH), Fast PFH, ...)
- **libpcl\_surface**: surface reconstruction techniques (e.g., meshing, convex hulls, Moving Least Squares, ...)
- **libpcl\_filters**: point cloud data filters (e.g., downsampling, outlier removal, indices extraction, projections, ...)
- libpcl\_io: I/O operations (e.g., writing to/reading from PCD (Point Cloud Data) and BAG files)
- libpcl\_segmentation: segmentation operations (e.g.,cluster extraction, Sample Consensus model fitting, polygonal prism extraction, ...)
- libpcl\_registration: point cloud registration methods (e.g., Iterative Closest Point (ICP), non linear optimizations, ...)
- libpcl\_range\_image: range image class with specialized methods
- It provides unit tests, examples, tutorials, ...

#### Point Cloud File Format

 Point clouds can be stored to disk as files, into the PCD (Point Cloud Data) format:

```
# Point Cloud Data ( PCD ) file format v .5
FIELDS x y z rgba
SIZE 4 4 4 4
TYPE F F F U
WIDTH 307200
HEIGHT 1
POINTS 307200
DATA binary
...<data>...
```

- Functions to load/save point clouds:
  - pcl::io::loadPCDFile
  - pcl::io::savePCDFile

#### Example: Create and Save a PC

```
#include<pcl/io/pcd io.h>
#include<pcl/point types.h>
//....
pcl::PointCloud<pcl::PointXYZ>:: Ptr cloud ptr (new
pcl::PointCloud<pcl::PointXYZ>);
cloud->width=50;
cloud->height=1;
cloud->isdense=false:
cloud->points.resize(cloud.width*cloud..height);
for(size t i=0; i<cloud.points.size(); i++)</pre>
    cloud \rightarrow points[i].x = 1024 * rand()/(RANDMAX + 1.0f);
    cloud \rightarrow points[i].y = 1024 * rand()/(RANDMAX + 1.0f);
    cloud \rightarrow points[i].z = 1024 * rand()/(RANDMAX + 1.0f);
pcl::io::savePCDFileASCII("testpcd.pcd",*cloud);
```

# Example: Visualize a PC

```
boost::shared_ptr<pcl::visualization::PCLVisualizer> viewer (new pcl::visualization::PCLVisualizer ("3D Viewer")); 
viewer->setBackgroundColor (0, 0, 0); 
viewer->addPointCloud<pcl::PointXYZ> ( in_cloud, cloud_color, "Input cloud" ); 
viewer->initCameraParameters (); 
viewer->addCoordinateSystem (1.0); 
while (!viewer->wasStopped ()) 
    viewer->spinOnce ( 1 );
```

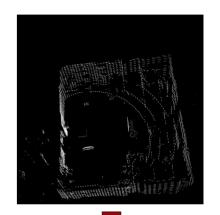
#### Basic Module Interface

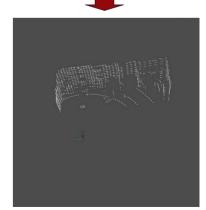
- Filters, Features, Segmentation all use the same basic usage interface:
- use setInputCloud() to give the input
- set some parameters
- call compute() or filter() or align() or ... to get the
  - output

# PassThrough Filter

Filter out points outside a specified range in one dimension

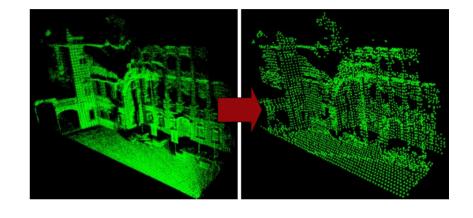
```
pcl::PassThrough<T> pass_through;
pass_through.setInputCloud (in_cloud);
pass_through.setFilterLimits (0.0,
0.5);
pass_through.setFilterFieldName ("z");
pass_through.filter( *cutted_cloud );
```





# Downsampling

Voxelize the cloud to a 3D grid. Each occupied voxel is approximated by the centroid of the points inside it.



```
pcl::VoxelGrid<T> voxel_grid;
voxel_grid.setInputCloud (input_cloud);
voxel_grid.setLeafSize (0.01, 0.01, 0.01);
voxel_grid.filter ( *subsamp_cloud ) ;
```

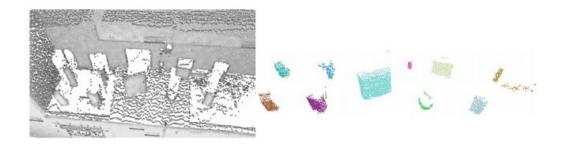
# Compute Normals



```
pcl::NormalEstimation<T, pcl::Normal> ne;
ne.setInputCloud (in_cloud);
pcl::search::KdTree<pcl::PointXYZ>::Ptr tree (new pcl::search::KdTree<pcl::PointXYZ> ());
ne.setSearchMethod (tree);
ne.setRadiusSearch (0.03);
ne.compute (*cloud_normals);
```

# Segmentation

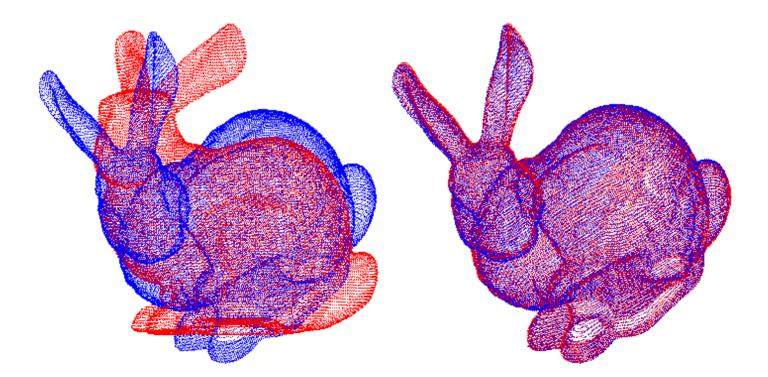
- A clustering method divides an unorganized point cloud into smaller, correlated, parts.
- EuclideanClusterExtraction uses a distance threshold to the nearest neighbors of each point to decide if the two points belong to the same cluster.



```
pcl::EuclideanClusterExtraction<T> ec;
ec.setInputCloud (in_cloud);
ec.setMinClusterSize (100);
ec.setClusterTolerance (0.05); // distance threshold
ec.extract (cluster_indices);
```

# Point Cloud Registration

We want to find the translation and the rotation that maximize the overlap between two point clouds



#### Iterative Closest Point

ICP iteratively revises the transformation (translation, rotation) needed to minimize the distance between the points of two raw scans.

- **Inputs**: points from two raw scans, initial estimation of the transformation, criteria for stopping the iteration.
- Output: refined transformation.



#### Iterative Closest Point

The algorithm steps are:

- 1) Associate points of the two cloud using the nearest neighbor criteria.
- 2) Estimate transformation parameters using a mean square cost function.
- 3) Transform the points using the estimated parameters.
- 4) Iterate (re-associate the points and so on).

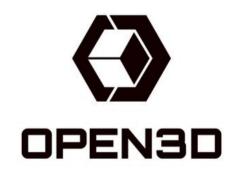
#### Iterative Closest Point

```
IterativeClosestPoint<PointXYZ, PointXYZ> icp;
// Set the input source and target
icp.setInputCloud (cloud_source);
icp.setInputTarget (cloud_target);
// Set the max correspondence distance to 5cm
icp.setMaxCorrespondenceDistance (0.05);
// Set the maximum number of iterations (criterion 1)
icp.setMaximumIterations (50);
// Set the transformation epsilon (criterion 2)
icp.setTransformationEpsilon (1e-8);
// Set the euclidean distance difference epsilon (criterion 3)
icp.setEuclideanFitnessEpsilon (1);
// Perform the alignment
icp.align (cloud_source_registered);
// Obtain the transformation that aligned cloud_source to
//cloud_source_registered
Eigen::Matrix4f transformation = icp.getFinalTransformation ();
```

# Open3D

#### open3d.org

- Open3D is an open-source library that supports rapid development of software that deals with 3D data. The Open3D frontend exposes a set of carefully selected data structures and algorithms in both C++ and Python.
- Open3D is released under the MIT license. Its use is encouraged for both research and commercial purposes, as long as proper attribution is given



#### Open3D PointCloud

A Open3D PointCloud contains the following data fields:

- points\_ (std::vector< Eigen::Vector3d >) -Contains the 3D coordinates of the points
- colors\_ (std::vector< Eigen::Vector3d >) Contains the RGB color for each point in points\_
- normals\_ (std::vector< Eigen::Vector3d > ) Contains the surface normal for each point in points\_

#### Open3D Modules

Open3D functions are gathered in different modules, some of them are:

- geometry: contains the class definition of point cloud and mesh structures
- camera: to store camera intrinsics and extrinsics parameters
- visualization: functions useful to display 3D structures
- io: I/O operations (e.g., writing to/reading from PLY, PCD,...)

# Load and Visualize Multiple PC

```
open3d::geometry::PointCloud pc0;
open3d::geometry::PointCloud pc1;
open3d::io::ReadPointCloud("path_to_pc0", pc0);
open3d::io::ReadPointCloud("path_to_pc1", pc1);
auto pc0 pointer =
std::make_shared<open3d::geometry::PointCloud>(pc0);
auto pc1_pointer =
std::make_shared<open3d::geometry::PointCloud>(pc1);
open3d::visualization::DrawGeometries({pc0 pointer, pc1 pointer});
```

# Downsample, Normals Estimation

```
//Downsample
double voxel_size = 0.007
std::shared_ptr<open3d::geometry::PointCloud>
&pcd_down_ptr = pcd.VoxelDownSample(voxel_size);
//Estimate point cloud normals
pcd_down_ptr->EstimateNormals(
 open3d::geometry::KDTreeSearchParamHybrid(
                        voxel_size*2, 30));
```

# ICP Registration

```
Eigen::Matrix4d transformation = Eigen::Matrix4d::Identity();
double threshold = 0.02;
double relative_fitness=1e-6;
double relative rmse1e-6;
int max iteration=1000;
auto result =
 open3d::pipelines::registration::RegistrationICP(
     source, target, threshold, transformation,
     open3d::pipelines::registration::
       TransformationEstimationPointToPoint(),
     open3d::pipelines::registration::
       ICPConvergenceCriteria(relative_fitness,
                              relative_rmse, max_iteration)
```