

Maggio 2020

UNIVERSITÀ DEGLI STUDI DELLA BASILICATA







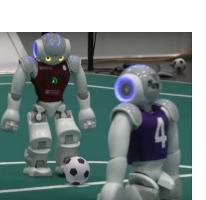
Corso di Visione e Percezione A.A. 2019/2020

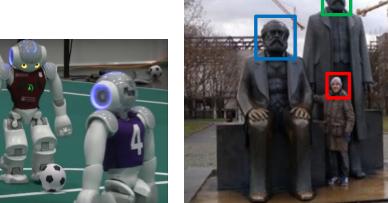
Docente

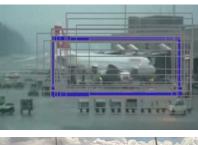
Domenico Daniele Bloisi

Visualizzazione

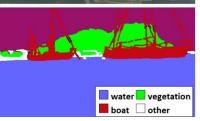
dati 3D



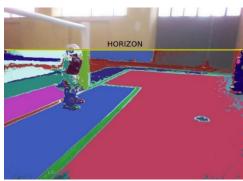














References and Credits

Queste slide sono adattate da:

Alberto Pretto – Sapienza Università di Roma Introduction to PCL: The Point Cloud Library Basic topics

http://www.dis.uniroma1.it/~pretto/download/pcl_intro.pdf

Gestione dati 2D

OpenCV (Open Source Computer Vision) is a library of programming functions for real-time computer vision

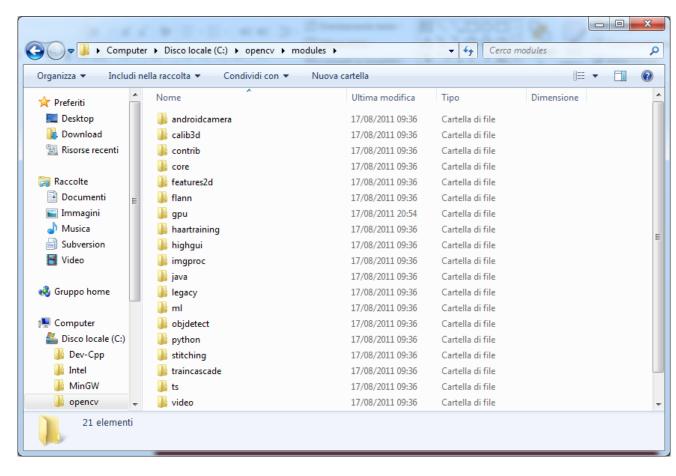


- BSD Licensed free for commercial use
- C++, C, Python and Java (Android) interfaces
- Supports Windows, Linux, Android, iOS and Mac OS
- More than 2500 optimized algorithms

Moduli OpenCV

OpenCV has a modular structure

- core
- imgproc
- video
- calib3d
- features2d
- objdetect
- highgui
- gpu





• ...

Processamento delle immagini

core - a compact module defining basic data structures, including the dense multidimensional array Mat and basic functions used by all other modules.

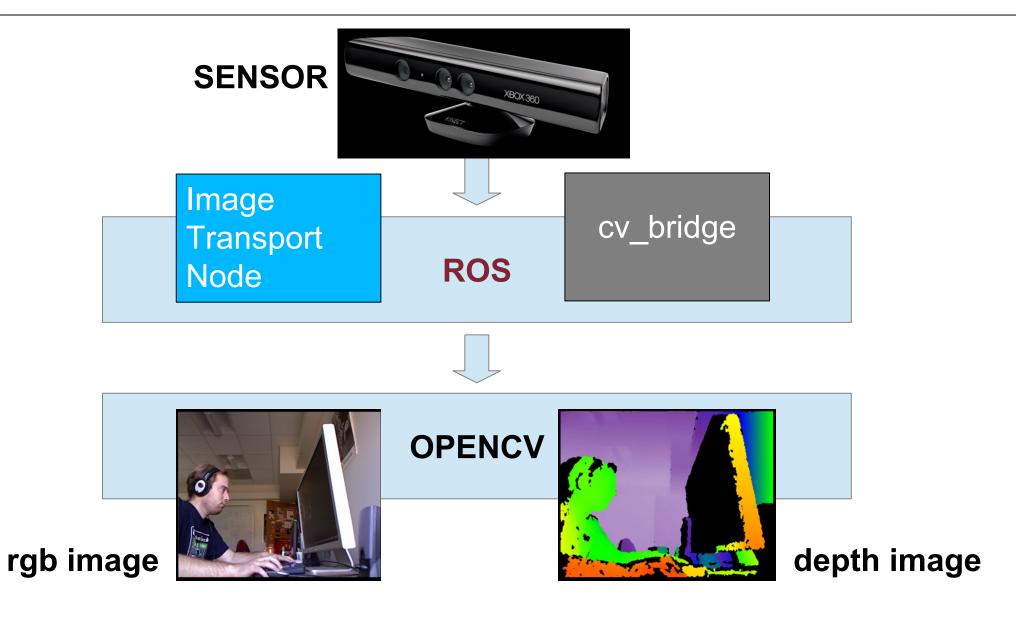
imgproc - an image processing module that includes linear and non-linear image filtering, geometrical image transformations (resize, affine and perspective warping, generic table-based remapping), color space conversion, histograms, and so on.

features2d - salient feature detectors, descriptors, and descriptor matchers.

highgui - an easy-to-use interface to video capturing, image and video codecs, as well as simple UI capabilities.

objdetect - detection of objects and instances of the predefined classes (for example, faces, eyes, mugs, people, cars, and so on).

OpenCV e ROS



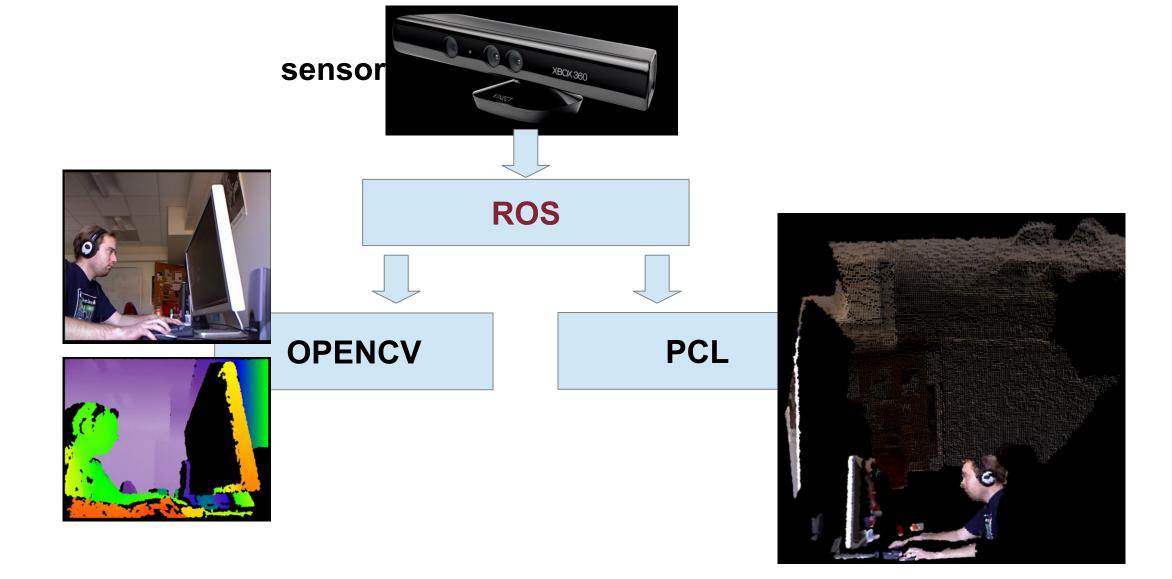
Gestione dati 3D

The Point Cloud Library (PCL) is a standalone, large scale, open project for 2D/3D image and point cloud processing



- Collection of Libraries focused on Point Cloud processing
- More than 450 developers/contributors
- Over 60 Tutorials and many examples
- BSD Licensed free for commercial use

PCL e ROS



Point cloud: a definition

- A point cloud is a data structure used to represent a collection of multi-dimensional points
- It is commonly used to represent three-dimensional data

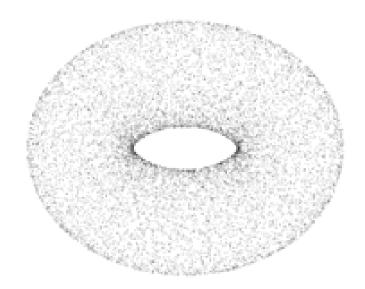


image from https://en.wikipedia.org/wiki/Point_cloud

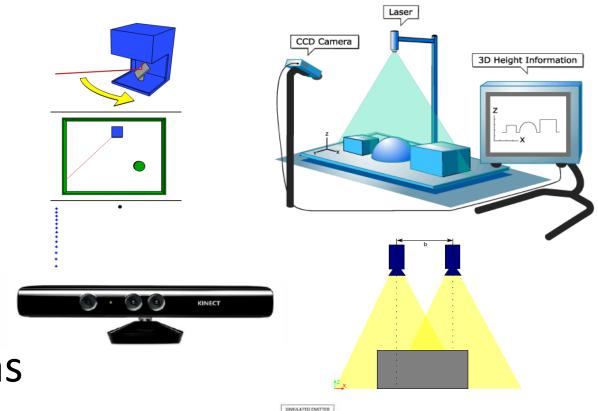
Point cloud: a definition

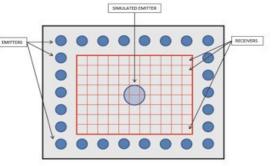
- The points in the point cloud usually represent the X, Y, and Z geometric coordinates of a sampled surface
- Each point can hold additional information: RGB colors, intensity values, etc...



Where do they come from?

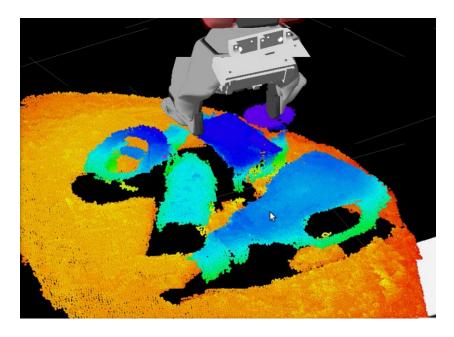
- 2/3D Laser scans
- Laser triangulation
- Stereo cameras
- RGB-D cameras
- Structured light cameras
- Time of flight cameras



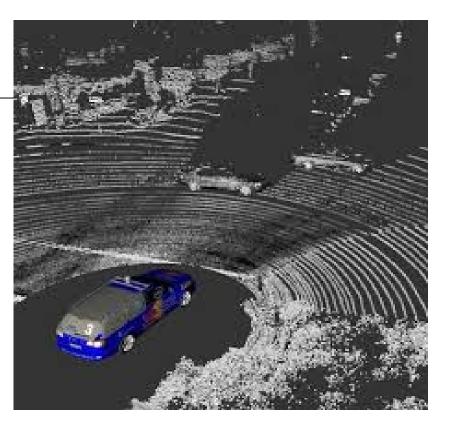


Point clouds in robotics

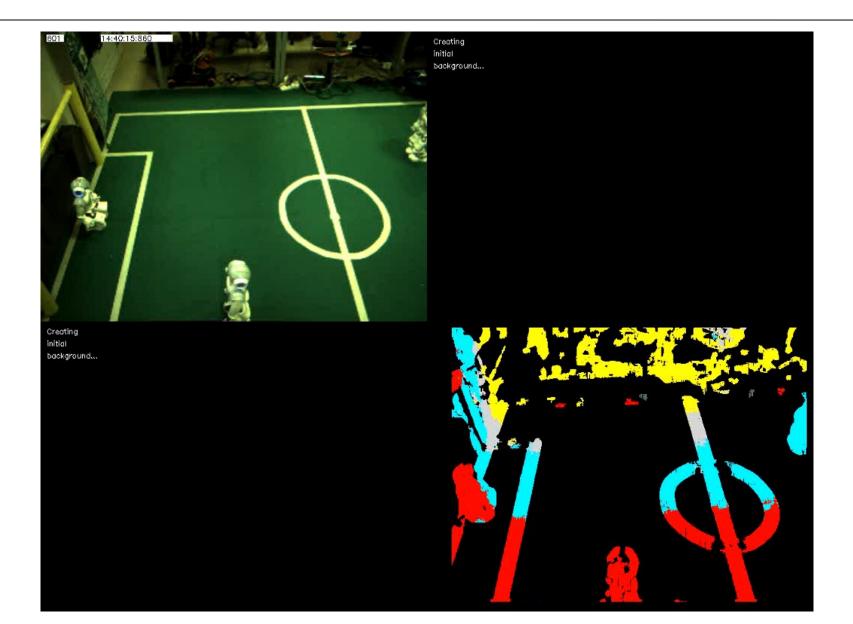
- Navigation/Obstacle avoidance
- Object recognition and registration
- Grasping and manipulation







Offside detection



Grasping



https://youtu.be/HIMIEOdsttU

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

PCL + ROS

 PCL provides the 3D processing pipeline for ROS, so you can also get the perception pcl stack and still use PCL standalone

- Among others, PCL depends on:
 - ✓ Boost
 - ✓ Eigen
 - ✓ OpenMP

PCL Basic Structures: PointCloud

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

- width (int) specifies the width of the point cloud dataset in the number of points.
 - → 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) Specifies the height of the point cloud dataset in the number of points
 - → 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.

PointCloud vs PointCloud2

We distinguish between two data formats for the point clouds:

- PointCloud<PointType> with a specific data type (for actual usage in the code)
- PointCloud2 as a general representation containing a header defining the point cloud structure (e.g., for loading, saving or sending as a ROS message)
- Conversion between the two frameworks is easy:
 - →pcl::fromROSMsg and pcl::toROSMsg
- Important: clouds are often handled using smart pointers, e.g.:
 - →PointCloud<PointType>::Ptr cloud_ptr;

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

CMakeLists.txt

```
project(pcl test)
cmake minimum required (VERSION 2.8)
cmake policy(SET CMP0015 NEW)
find package (PCL 1.7 REQUIRED)
add definitions(${PCL DEFINITIONS})
include directories (... ${PCL INCLUDE DIRS})
link directories(... ${PCL LIBRARY DIRS})
add executable (pcl test pcl test.cpp ...)
target link libraries(pcl test ${PCL LIBRARIES})
```

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

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 U
WIDTH 307200
HETGHT 1
POINTS 307200
DATA binary
...<data>...
```

Funtions: pcl::io::loadPCDFile and pcl::io::savePCDFile

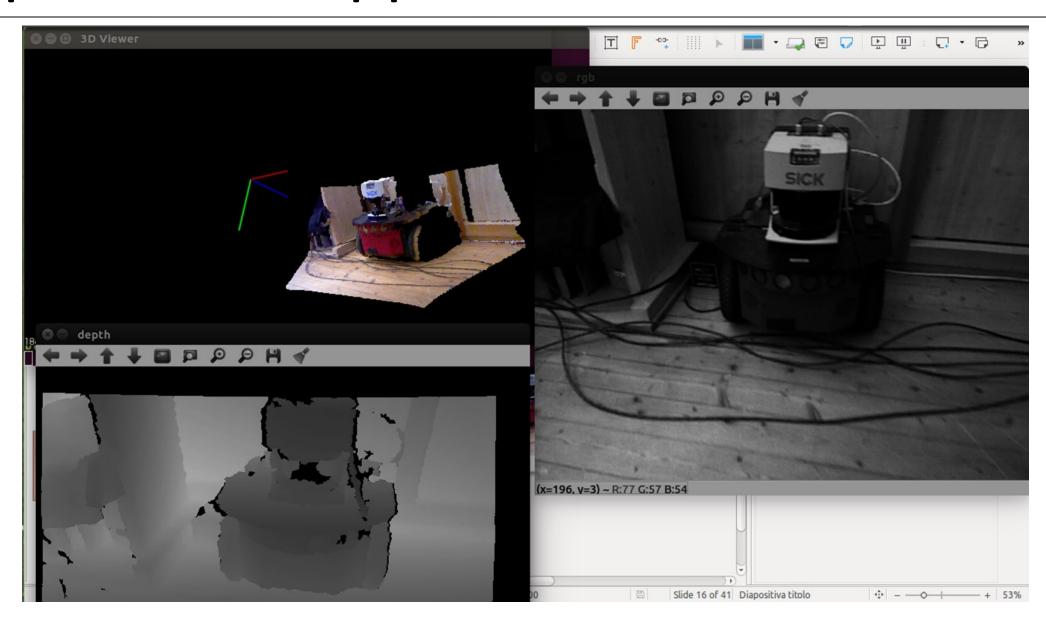
Example: create and save a PC

```
#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->points[i].x = 1024*rand()/(RANDMAX+1.0f);
    cloud->points[i].y = 1024*rand()/(RANDMAX+1.0f);
    cloud \rightarrow points[i].z = 1024*rand()/(RANDMAX+1.0f);
  pcl::io::savePCDFileASCII("testpcd.pcd", *cloud);
  //....
```

Visualize a cloud

viewer->spinOnce(1);

depth2cloud.cpp



Basic Module Interface

Filters, Features, Segmentation all use the same basic usage interface:

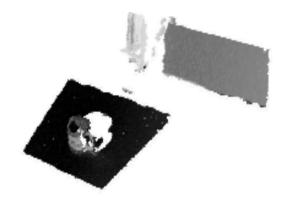
use setInputCloud() to give the input

- set some parameters
- call compute() or filter() or align() or ... to get the output

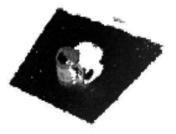
PassThrough Filter

Filter out points outside a specified range in one dimension.

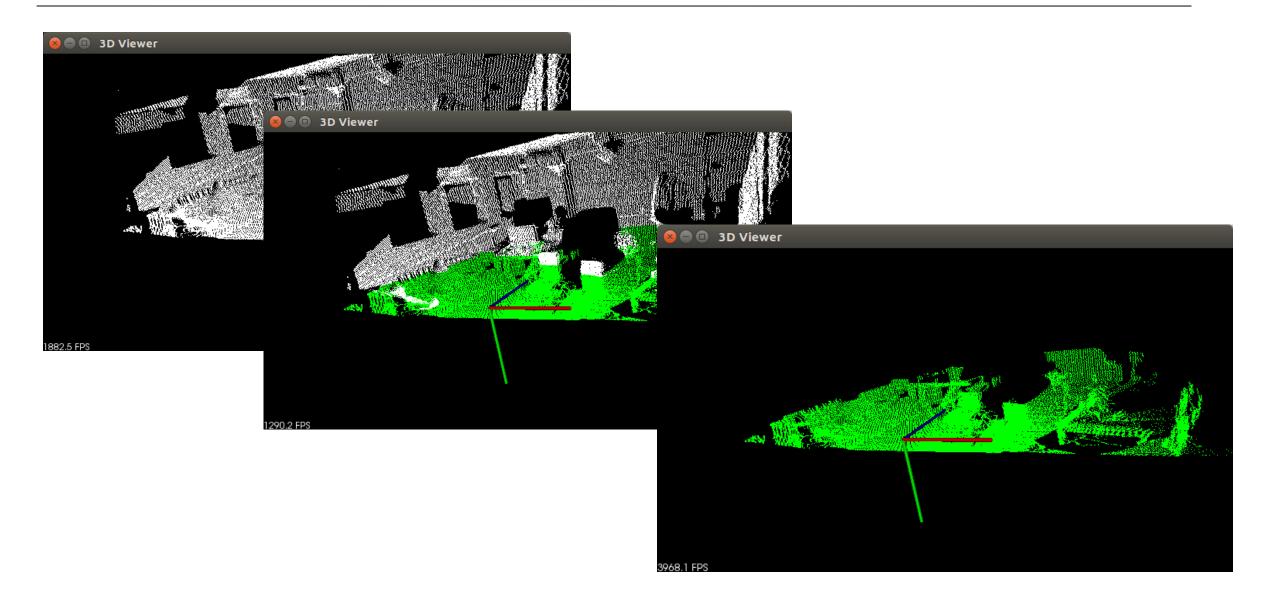
```
pcl::PassThrough<T> pass_through;
pass_through.setInputCloud(in_cloud);
pass_through.setFilterLimits (0.0, 0.5);
pass_through.setFilterFieldName("z");
pass through.filter(*cutted cloud);
```







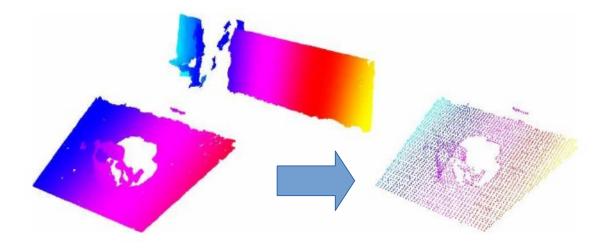
cloud_filters.cpp



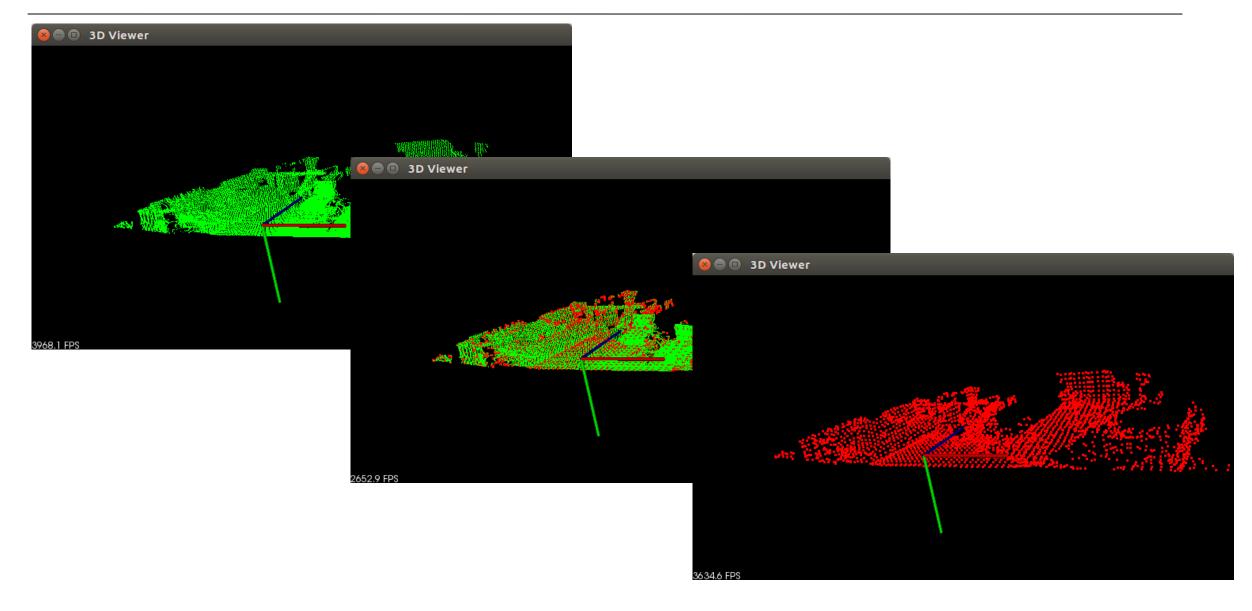
Downsampling

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);
```



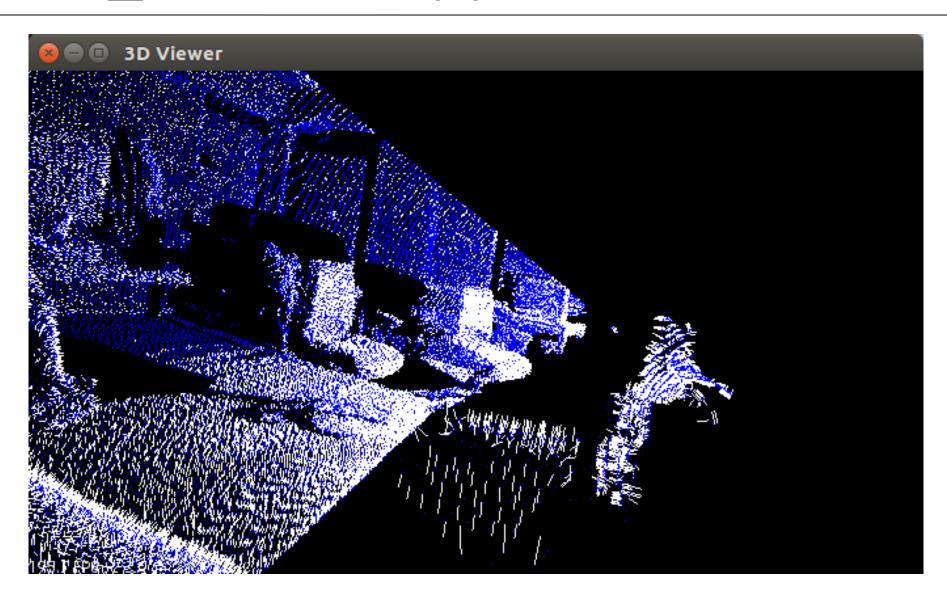
cloud_filters.cpp



Features example: 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)
```

cloud_normals.cpp



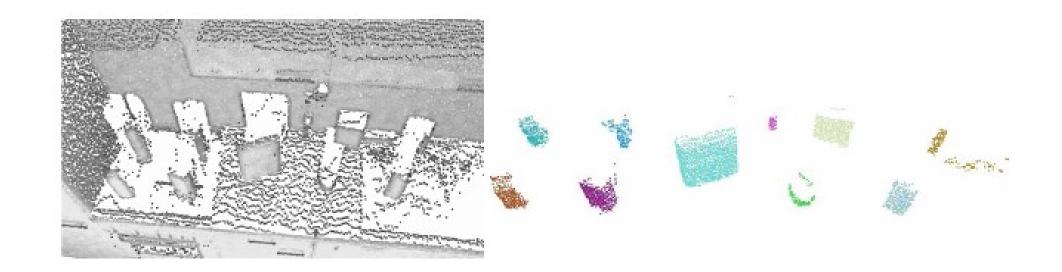
Segmentation

A clustering method divides an unorganized point cloud into smaller, correlated, parts

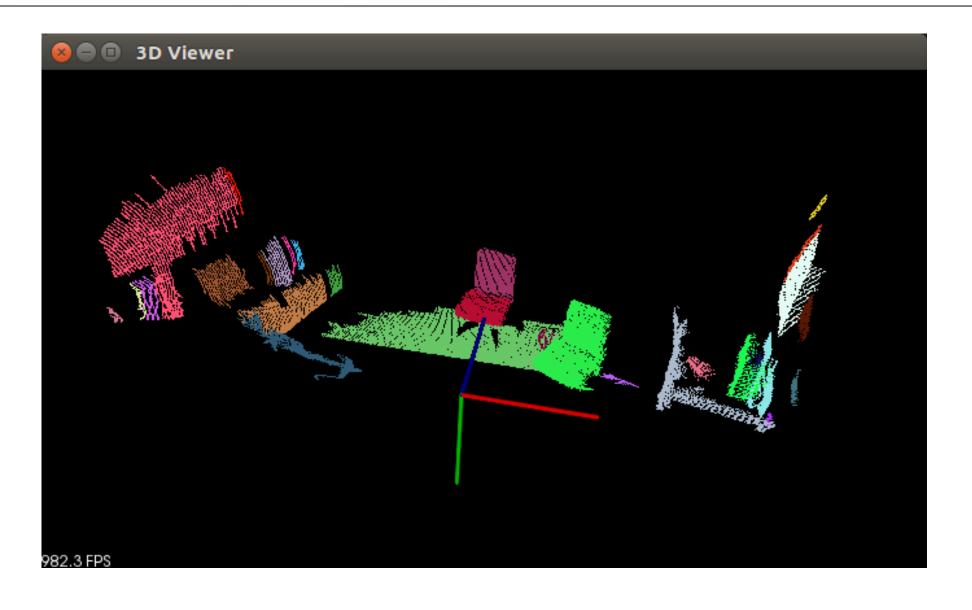
EuclideanClusterExtraction uses a distance threshold to the nearest neighbors of each point to decide if the two points belong to the same cluster.

Segmentation example

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

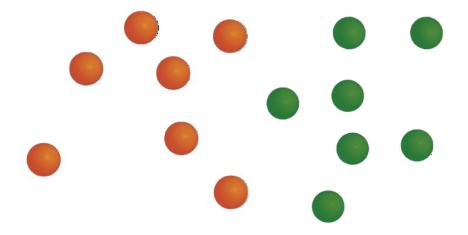


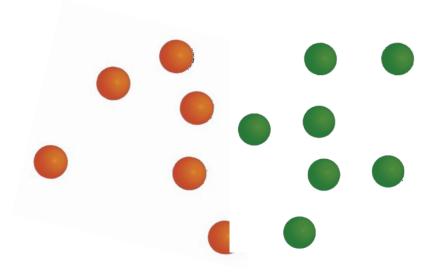
clustering.cpp



Point Cloud Registration

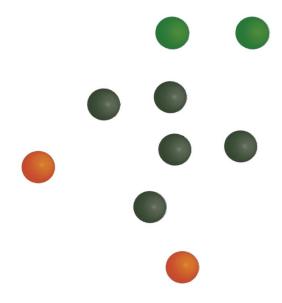
We want to find the translation and the rotation that maximize the overlap between two point clouds











Iterative Closest Point

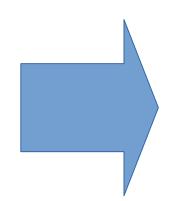
ICP iteratively revises the transformation (translation, rotation) needed to minimize the distance between the points of two raw scans

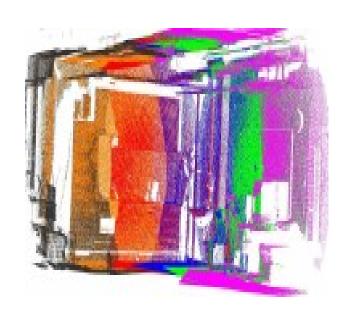
Input: points from two raw scans, initial estimation of the transformation, criteria for stopping the iteration

Output: refined transformation

Iterative Closest Point: Example







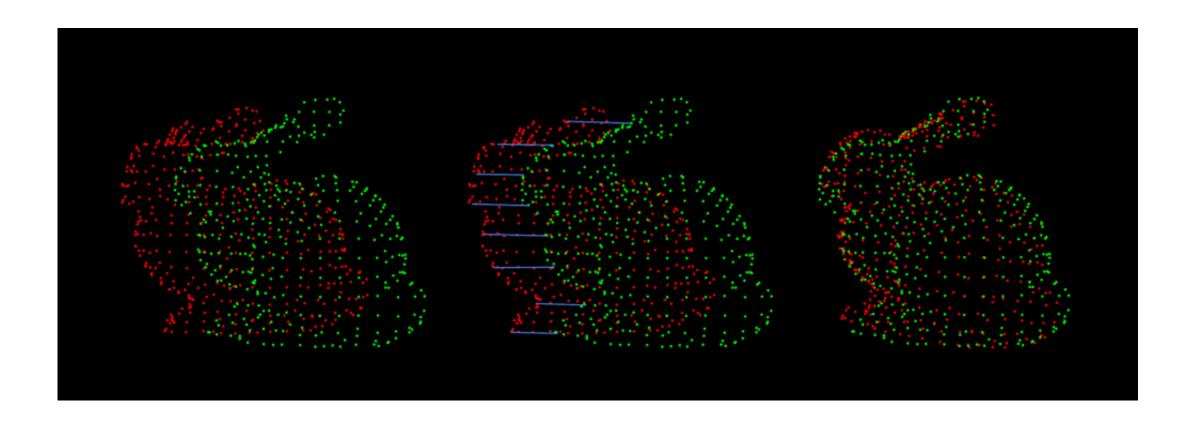
Iterative Closest Point: Algorithm

- 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)

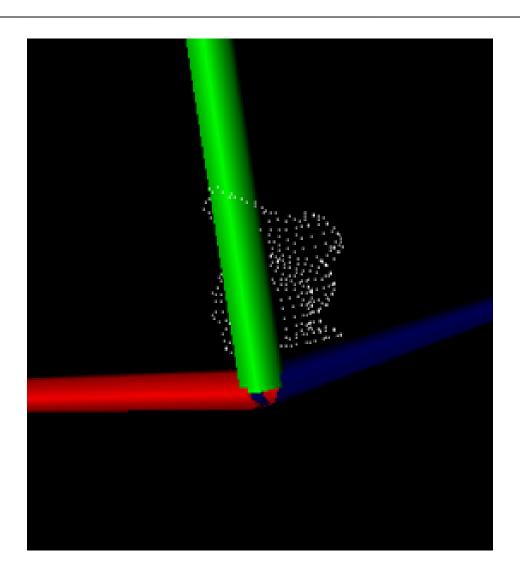
Iterative Closest Point: Code

```
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);
// Align cloud source to cloud source registered
Eigen::Matrix4f transformation = icp.getFinalTransformation();
```

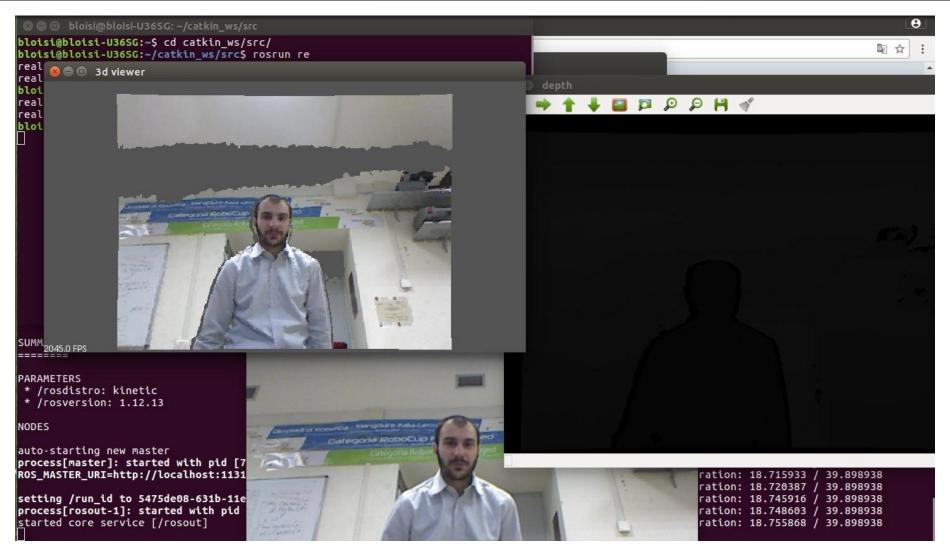
Iterative Closest Point: Example



icp.cpp



basic 3D visualizer



https://github.com/dbloisi/basic 3d visualizer

basic 3D visualizer: CMakeLists.txt

```
🔊 🖨 📵 CMakeLists.txt (~/catkin_ws/src/basic_3d_visualizer) - gedit
         I₽
Open ▼
                                                                                                      Save
 1 cmake minimum required(VERSION 2.8.3)
 2 project(basic 3d visualizer)
 4 ## Compile as C++11, supported in ROS Kinetic and newer
 5 # add compile options(-std=c++11)
 7 ## Find catkin macros and libraries
 ## if COMPONENTS list like find_package(catkin REQUIRED COMPONENTS xyz)
9 ## is used, also find other catkin packages
10 find package(catkin REQUIRED COMPONENTS
11 cv bridge
12 pcl conversions
13 pcl ros
14 гозсрр
15 sensor msgs
16 std msgs
17)
19 find package(OpenCV REOUIRED
20 COMPONENTS
21 opency highqui
22 opencv objdetect
23 CONFIG
24)
25 ## System dependencies are found with CMake's conventions
27 # Tind package(Boost REQUIRED COMPONENTS system)
28
30 ## Uncomment this if the package has a setup.py. This macro ensures
31 ## modules and global scripts declared therein get installed
                                                          CMake ▼ Tab Width: 8 ▼
                                                                                     Ln 27. Col 49
                                                                                                      INS
```

basic 3D visualizer: CMakeLists.txt

```
CMakeLists.txt (~/catkin_ws/src/basic_3d_visualizer) - gedit
 151 ## The above recommended prefix causes long target names, the following renames the
152 ## target back to the shorter version for ease of user use
153 ## e.g. "rosrun someones pkg node" instead of "rosrun someones pkg someones pkg node"
154 # set target properties(${PROJECT NAME} node PROPERTIES OUTPUT NAME node PREFIX "")
156 ## Add cmake target dependencies of the executable
157 ## same as for the library above
158 # add_dependencies(${PROJECT_NAME} node ${${PROJECT_NAME} EXPORTED_TARGETS} ${catkin_EXPORTED_TARGETS})
160 ## Specify libraries to link a library or executable target against
161 # target link libraries(${PROJECT NAME} node
162 # ${catkin LIBRARIES}
163 # )
164 target link libraries(basic 3d visualizer
165 ${catkin LIBRARIES}
166 S{OpenCV LIBRARIES}
167)
169 ############
170 ## Install ##
171 #############
172
173 # all install targets should use catkin DESTINATION variables
174 # See http://ros.ora/doc/api/catkin/html/adv user auide/variables.html
175
176 ## Mark executable scripts (Python etc.) for installation
177 ## in contrast to setup.py, you can choose the destination
178 # install(PROGRAMS
179 # scripts/my python script
180 # DESTINATION ${CATKIN PACKAGE BIN DESTINATION}
181 # )
                                                              CMake ▼ Tab Width: 8 ▼
                                                                                        Ln 27. Col 49
```



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Corso di Visione e Percezione A.A. 2019/2020 Docente

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Visualizzazione

dati 3D



