

# Processing Laser Data

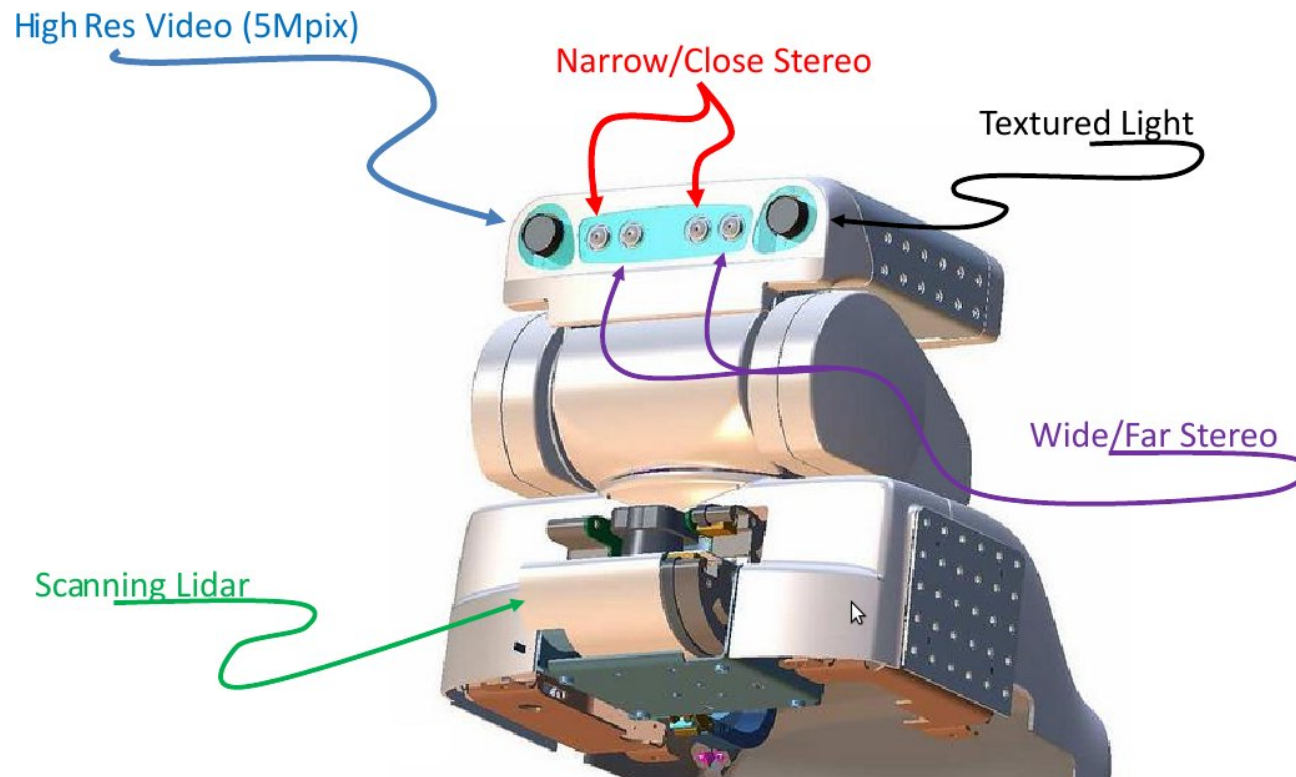
ROS + PR2 Training Workshop

# Outline

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- Lasers and 3D sensing
  - Visualizing Laser Scans
- 
- From LaserScan to PointCloud
  - What are Point Clouds?
  - Data representation
  - Visualizing PointCloud messages
  - Preview :: ROS C-Turtle (latest)

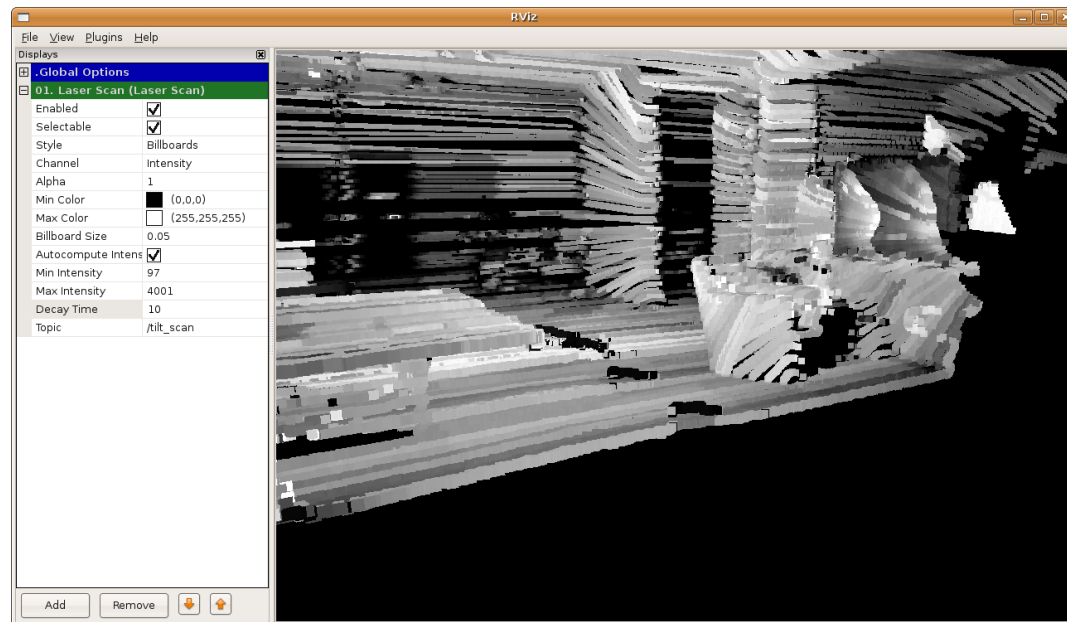
# Lasers and 3D sensing



- Stereo cameras in the head
- Tilting laser range finder
- Base laser range finder

# Visualizing Laser Scans

- `rviz` (<http://www.ros.org/wiki/rviz/DisplayTypes/LaserScan>)
  - `$ rosrun rviz rviz`
  - Add a Laser Scan display
  - Set the topic and the TF frames (Fixed/Target)



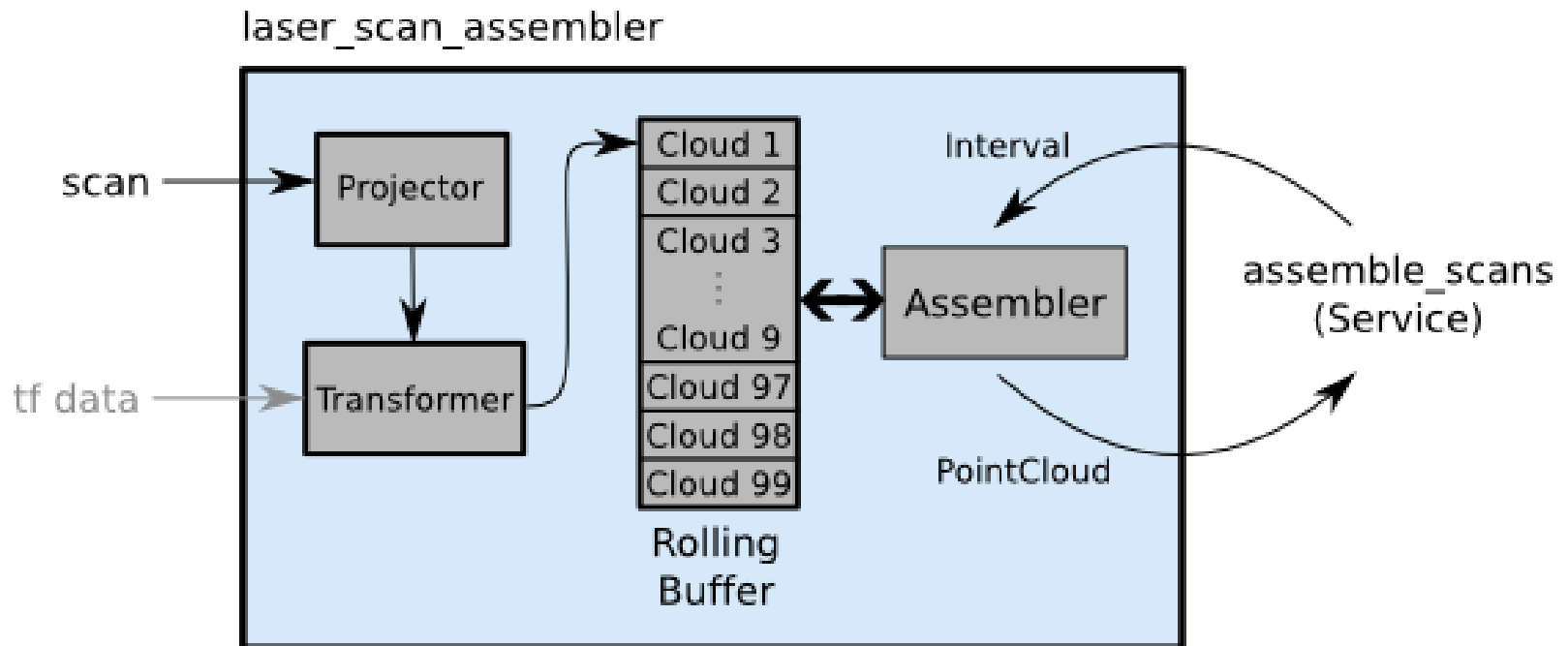
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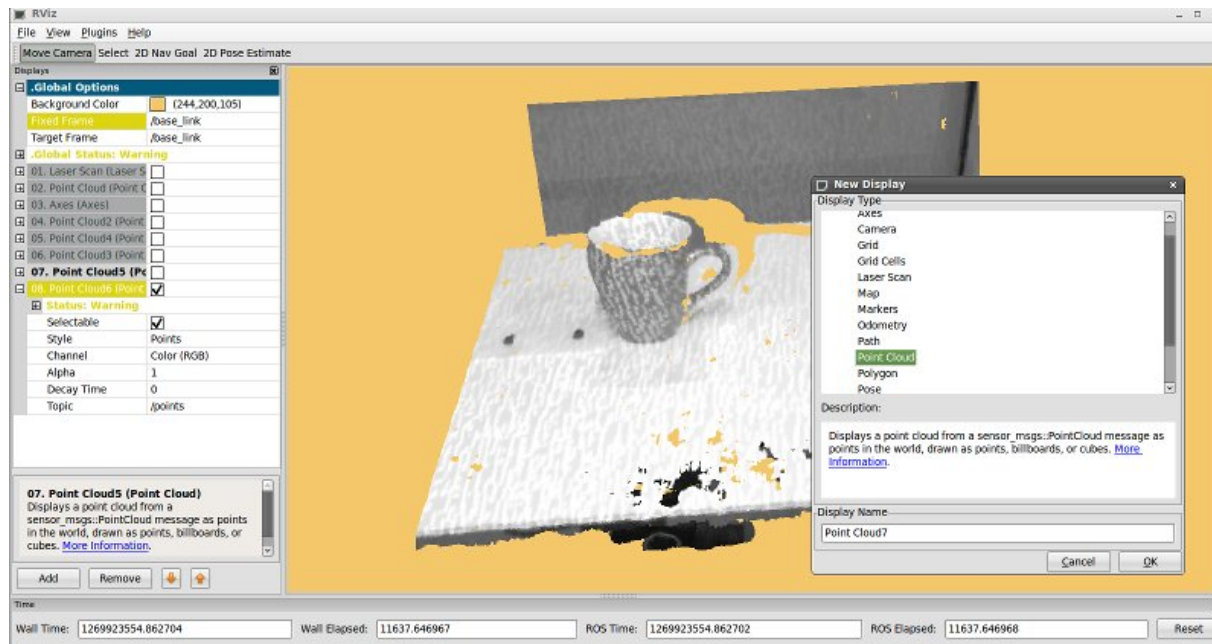
# From LaserScan to PointCloud

- Check if the tilt laser is being actuated ([http://www.ros.org/wiki/pr2\\_mechanism\\_controllers/LaserScannerTrajController](http://www.ros.org/wiki/pr2_mechanism_controllers/LaserScannerTrajController))
- laser\_assembler ([http://www.ros.org/wiki/laser\\_assembler](http://www.ros.org/wiki/laser_assembler))
  - create PointCloud from LaserScan messages



# Visualizing Point Clouds

- `rviz` (<http://www.ros.org/wiki/rviz/DisplayTypes/PointCloud>)
  - `$ rosrun rviz rviz`
  - Add a Point Cloud display
  - Set the topic and the TF frames (Fixed/Target)



# What are Point Clouds?



- Point Cloud = a “cloud” (i.e., collection) of  $nD$  points (usually  $n = 3$ )
- $p = \{x, y, z\} \rightarrow P = \{p_1, p_2, \dots, p_n\}$
- represent 3D information about the world

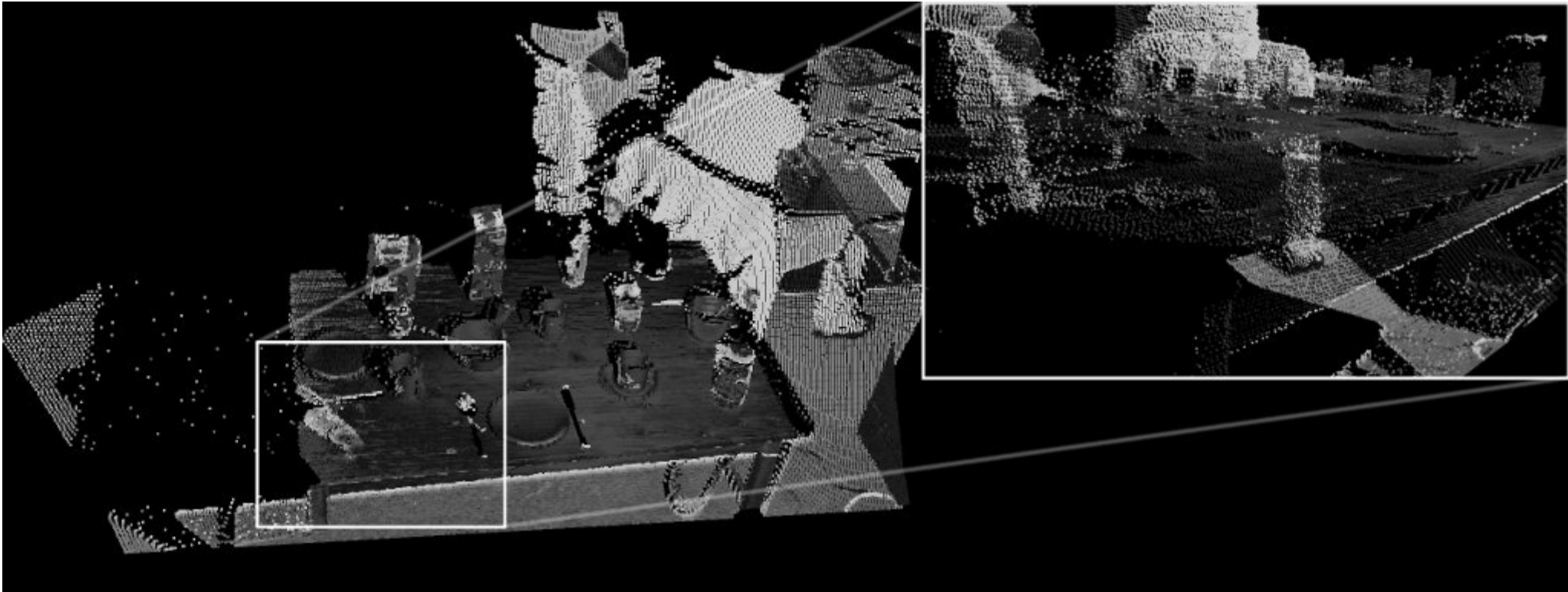


# What are Point Clouds?



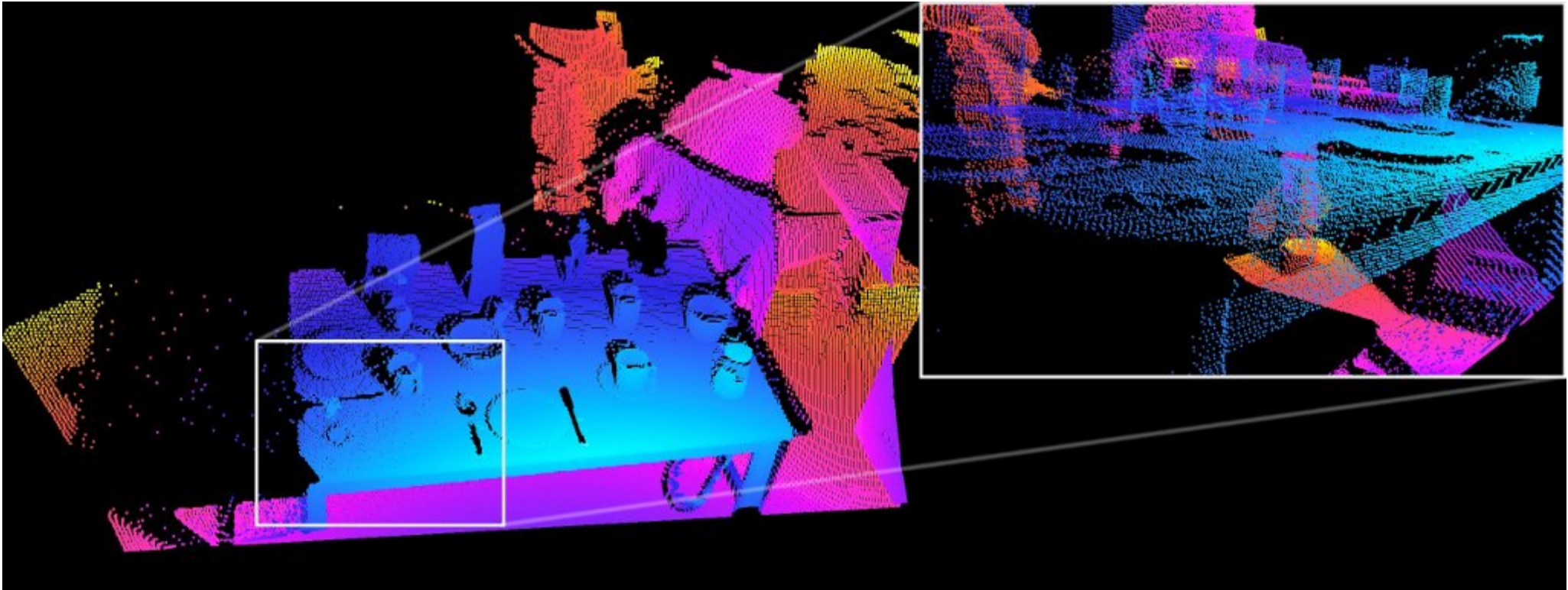
- besides XYZ data, each point  $p$  can hold additional information
- examples include: RGB colors, intensity values, distances, segmentation results, etc

# What are Point Clouds?



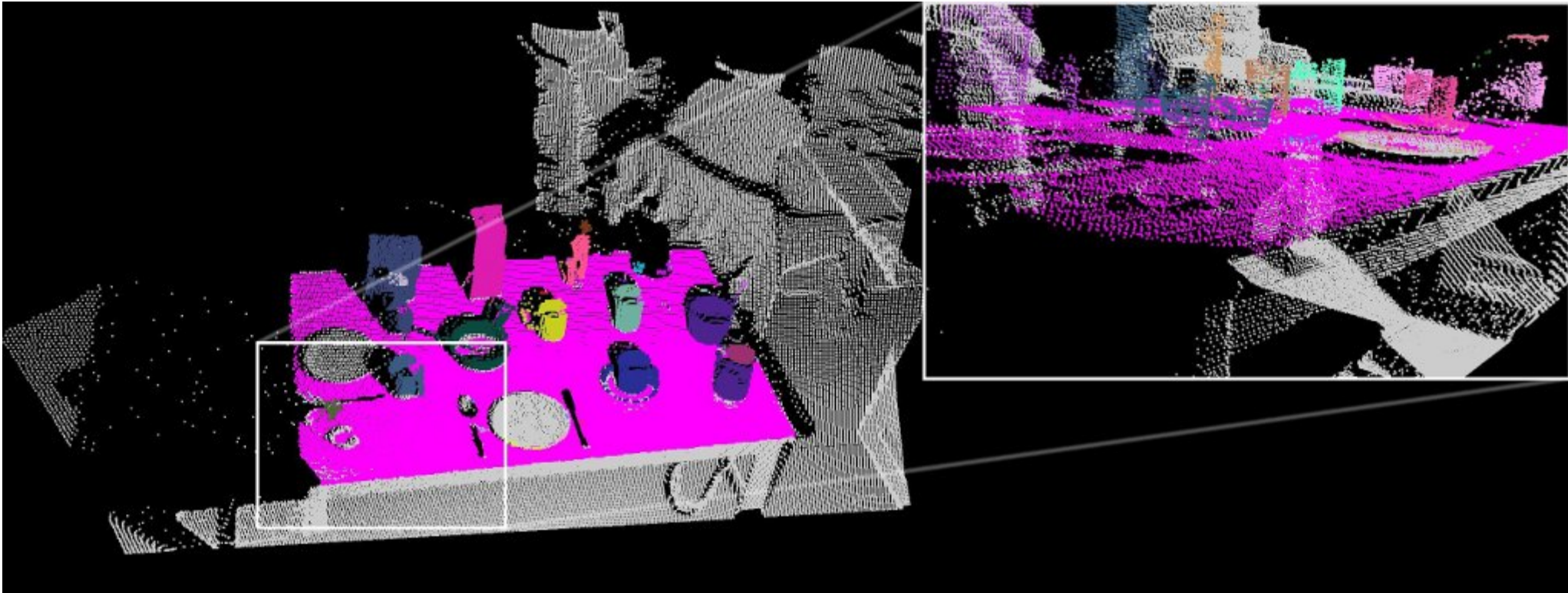
- intensity data

# What are Point Clouds?



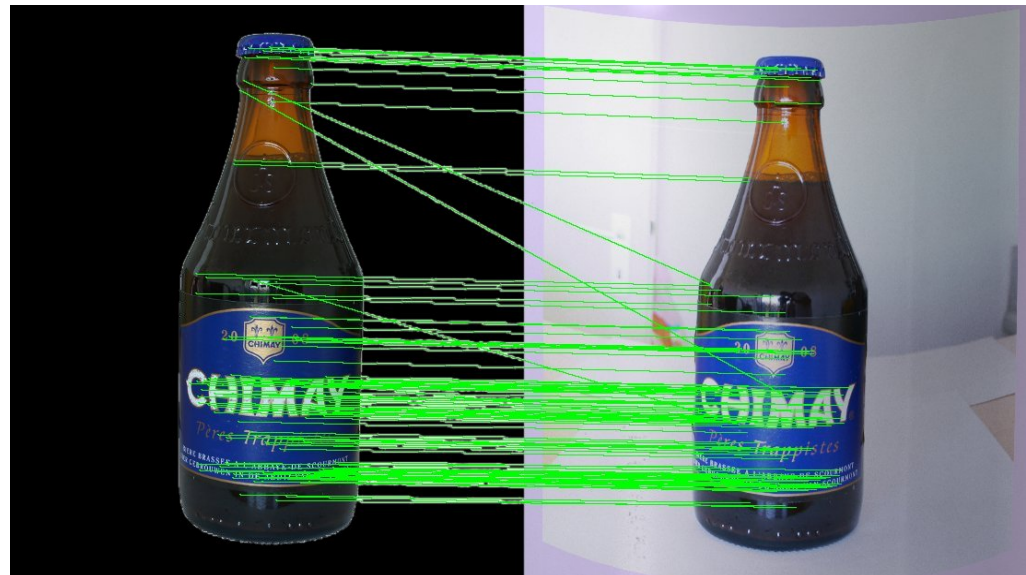
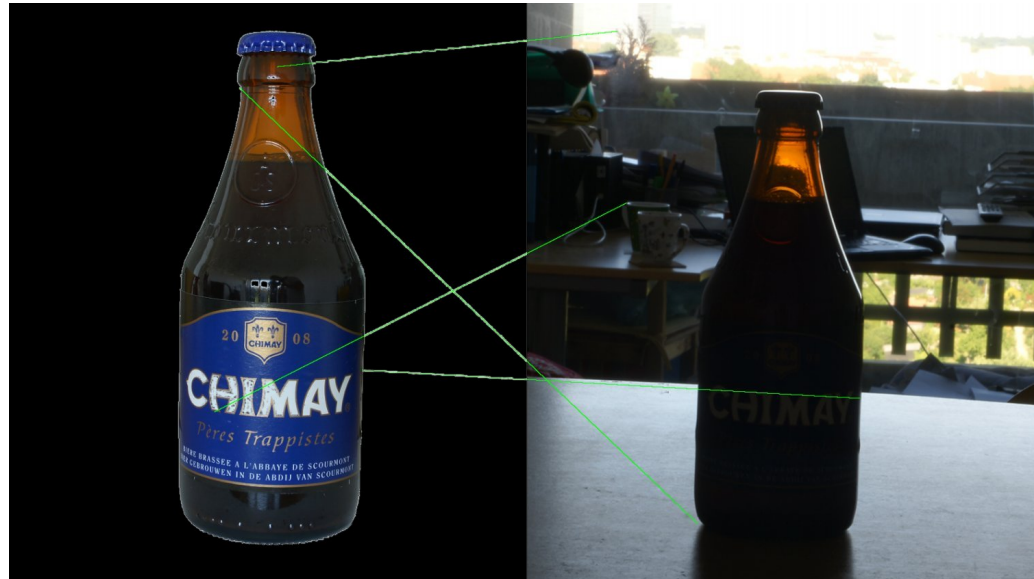
- distance data

# What are Point Clouds?

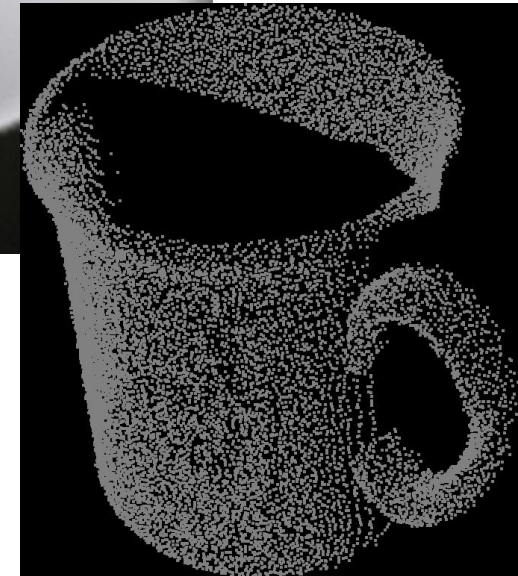
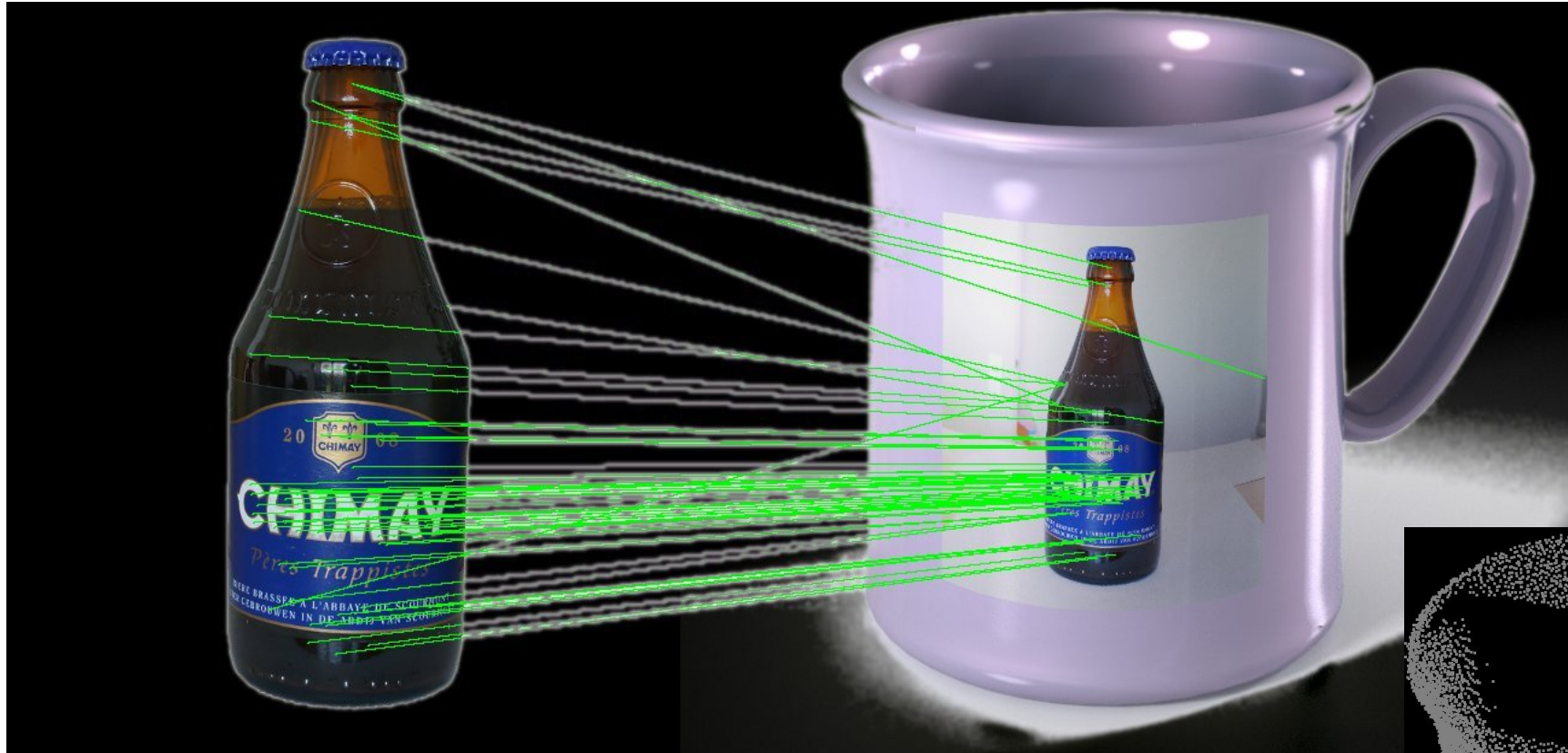


- segmentation data

# Why are Point Clouds important?



# Why are Point Clouds important?



# Data representation

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- a point  $\mathbf{p}$  is a n-tuple, e.g.

$$\mathbf{p}_i = \{x_i, y_i, z_i, r_i, g_i, b_i, \dots\}$$

- a Point Cloud  $\mathbf{P}$  is represented as a collection of points  $\mathbf{p}_i$ , e.g.  $\mathbf{P} = \{p_1, p_2, \dots, p_n\}$
- in terms of data structures, an XYZ point can be represented as: *float32 x, float32 y, float32 z*
- an n-dimensional point is then: *float32[] point*
- therefore a Point Cloud  $\mathbf{P}$  is *Point[] points*

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# Preview :: ROS C-Turtle (latest)

## Preview

- PointCloud2 message types: compact, aligned, efficient representations for point clouds
- PCL (Point Cloud Library): a full package for 3D processing

# PointCloud2

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- Point Clouds are big (!)
  - Operation on them are typically slower
  - They are expensive to store (float/double)
- Solutions:
  - Store each dimension data in different (the most appropriate) formats, e.g., *rgb* - 24bits, instead of 3x4 (sizeof float)
  - Group data together, and keep it aligned (SSE) to speed up computations
  - Support organized data - *nD images*

# PointCloud2

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- The ROS PointCloud2 data format:

**Header** header

uint32 height

uint32 width

**PointField[]** fields

bool is\_bigendian

uint32 point\_step

uint32 row\_step

uint8[] data

bool is\_dense

# PointField

---

- Where PointField is:

uint8 INT8=1, UINT8=2, INT16=3,  
UINT16=4, INT32=5, UINT32=6,  
FLOAT32=7, FLOAT64=8

string **name**  
uint32 **offset**  
uint8 **datatype**  
uint32 **count**

- Examples:

“x”, 0, 7, 1

“y”, 4, 7, 1

“z”, 8, 7, 1

“rgba”, 12, 6, 1

“normal\_x”, 16, 8, 1

“normal\_y”, 20, 8, 1

“normal\_z”, 24, 8, 1

“fpfh”, 32, 7, 33

# pcl::PointCloud<T>

- Binary blobs are hard(er) to work with
- We provide converters, Publishers / Subscribers, filters, etc, similar to images
- **PointCloud2** → **PointCloud<T>**
- Examples of T:

```
struct PointXYZ
{
    float x;
    float y;
    float z;
}
```

```
struct Normal
{
    float normal[3];
    float curvature;
}
```

# Point Cloud Data (PCD) format

- In addition, point clouds can be stored to disk as files, into the PCD format:

**FIELDS** x y z rgba

**SIZE** 4 4 4 4

**TYPE** F F F U

**WIDTH** 307200

**HEIGHT** 1

**POINTS** 307200

**DATA** binary

...

- **DATA** can be either **binary** or **ascii**:

**DATA** ascii

0.0054216 0.11349 0.040749

-0.0017447 0.11425 0.041273

# Point Cloud Library (PCL)

**Point Cloud Data**

ENV Analysis Type A:

Memory used:	7482 kbytes
Points in cloud:	459701
Polygons generated:	60759
Objects:	251 (697)

**Wireframe**

Relative polygonal optimisations:  ON

Relative curvature analysis:  ON

**SD Tree statistics:**

Number of faces stored in cache:	89091
Maximum tree depth of ENV analysis:	32.5
Average tree depth created:	6.24
Tree nodes:	2578
Tree faces:	21789
Average leaf:	22.9074
Raycasts required:	9726

**Surface**

Execution list:

- 3D point coordinates, intensity
- Sparse Outlier Removal
- Cloud Resampling (RMLS)
- Normal and Curvature Estimation
- Normal Consistency Propagation
- Feature Persistence Analysis
- Region Segmentation
- Model Fitting

Result: no errors

**POINT CLOUD LIBRARY**

# What is P(oint) C(loud) L(ibrary)

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- PCL is:
  - fully **templated** modern C++ library for 3D point cloud processing
  - uses **SSE** optimizations (Eigen backend) for fast computations on modern CPUs
  - uses **OpenMP** and Intel **TBB** for parallelization
  - passes data between modules (e.g., algorithms) using **Boost shared pointers**



# What is P(oint) C(loud) L(ibrary)

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- 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, ...)

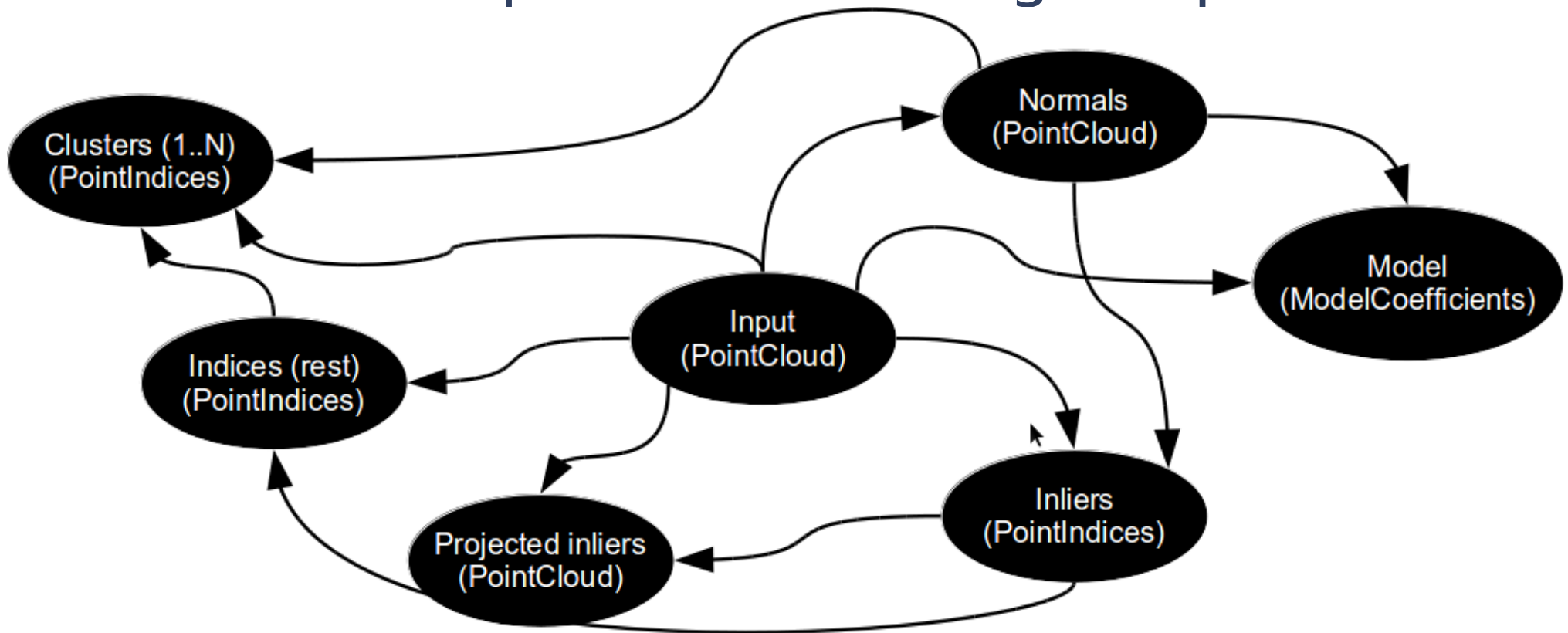
# What is P(oint) C(loud) L(ibrary)

- **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, ...)
- unit tests, examples, tutorials
- C++ classes are templated building blocks (**nodelets!**)

# PCL Philosophy

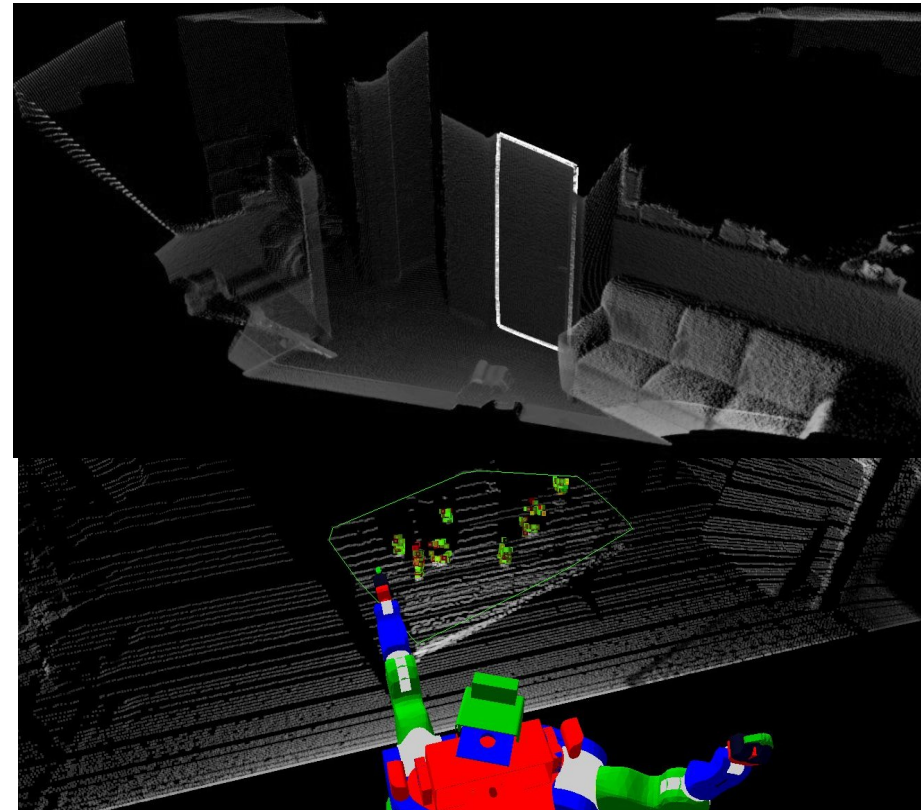
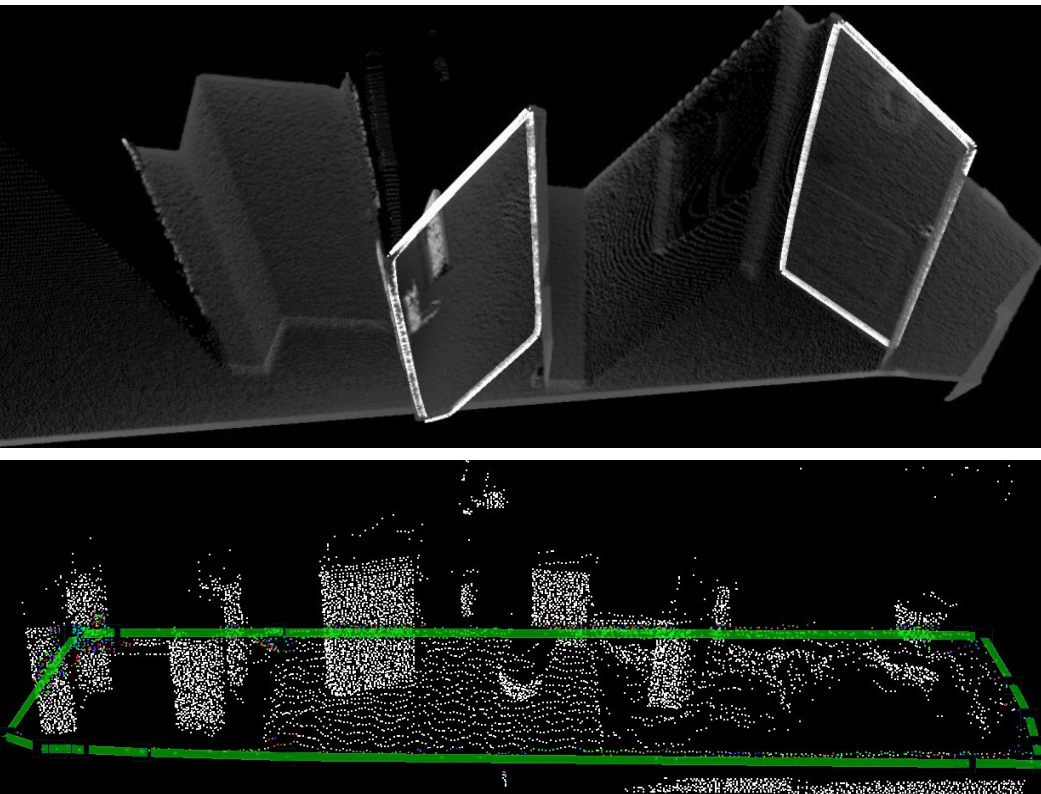
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- **Philosophy:** *write once, parameterize everywhere*
- **PPG: Perception Processing Graphs**



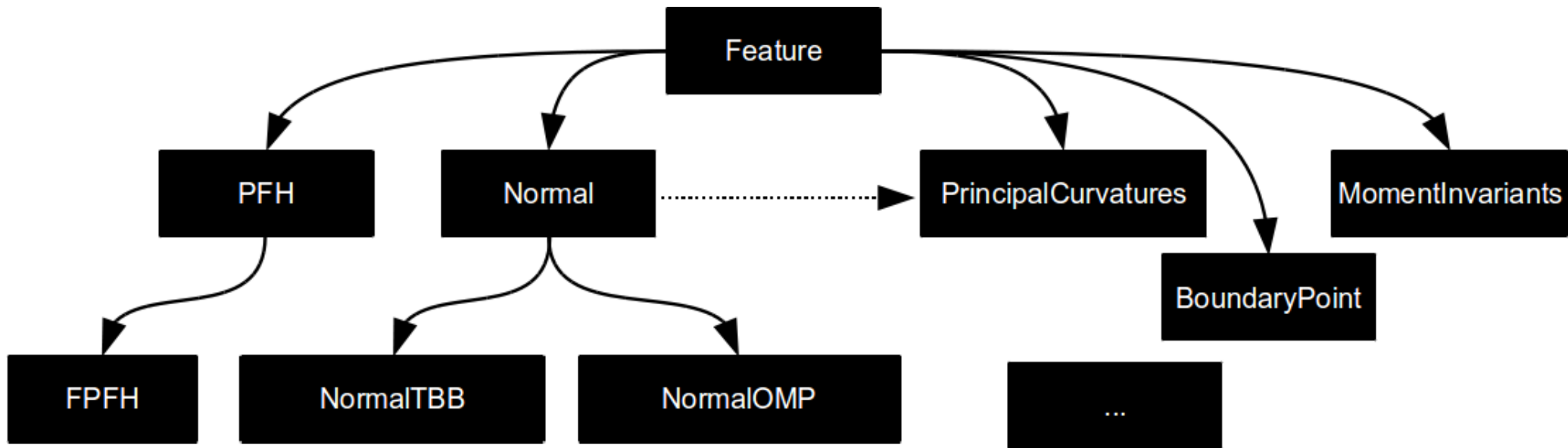
# Why PPG? Example

- Algorithmically:
  - door = table = wall detection = ...
  - the only thing that changes is: **parameters** (constraints)!



# More on PCL Architecture

- Inheritance simplifies development:



```
pcl :: Feature < PointT > feat ;  
feat = pcl :: Normal < PointT > ( input );  
feat = pcl :: FPFH < PointT > ( input );  
feat = pcl :: BoundaryPoint < PointT > ( input );  
feat . compute ( & output );
```

# PCL Statistics

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- Misc, stats:
  - 9 releases so far (*latest: 0.1.8*)
  - over 100 classes
  - over 25k lines of code
  - external dependencies (for now) on eigen, cminpack, ANN, FLANN, TBB
  - internal dependencies (excluding the obvious) on dynamic\_reconfigure, message\_filters
- tf\_pcl package for TF integration

# PCL :: Exercise

**Point Cloud Data**

ENV Analysis Type A:

Memory used: 7482 kbytes  
 Points in cloud: 459701  
 Polygons generated: 60759  
 Objects: 251 (697)

**Wireframe**

Aggressive polygonal optimizations: ON  
 Aggressive curvature analysis: ON

**SD Tree statistics:**

Number of faces stored in cache: 85021  
 Maximum tree depth of ENV analysis: 32.5  
 Average tree depth created: 7.24  
 Tree nodes: 2578  
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 Raycasts required: 9726

**Surface**

Execution list:

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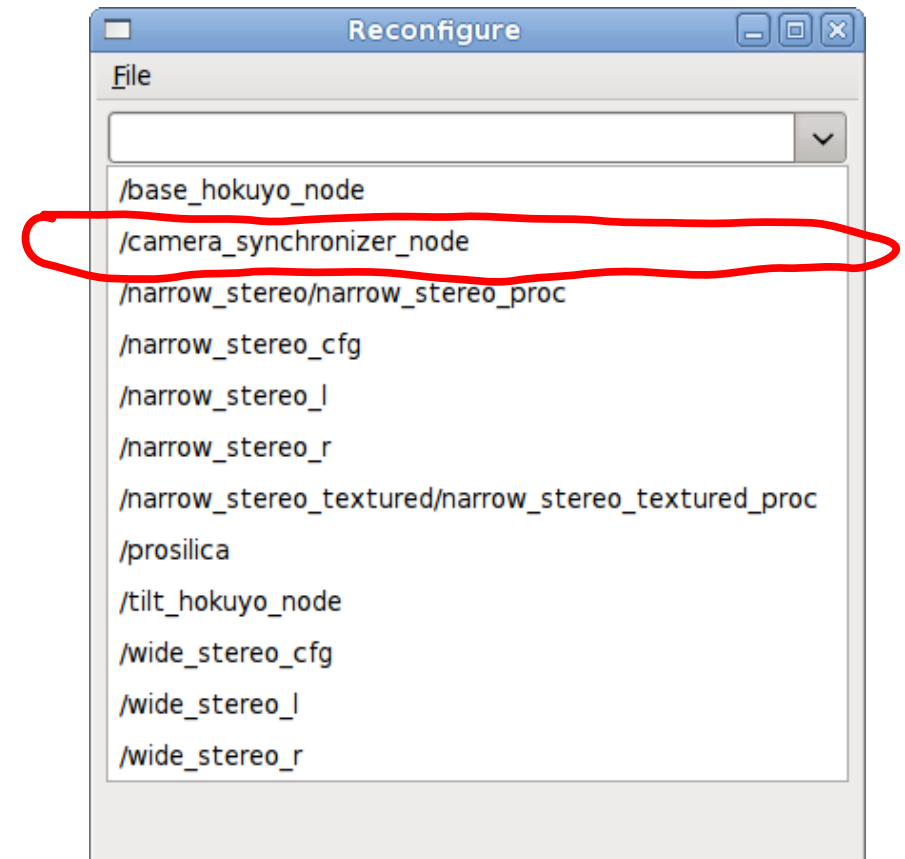
Result: no errors

**POINT CLOUD LIBRARY**

# Re: Using the Texture Projector

```
$ rosrun dynamic_reconfigure  
reconfigure_gui
```

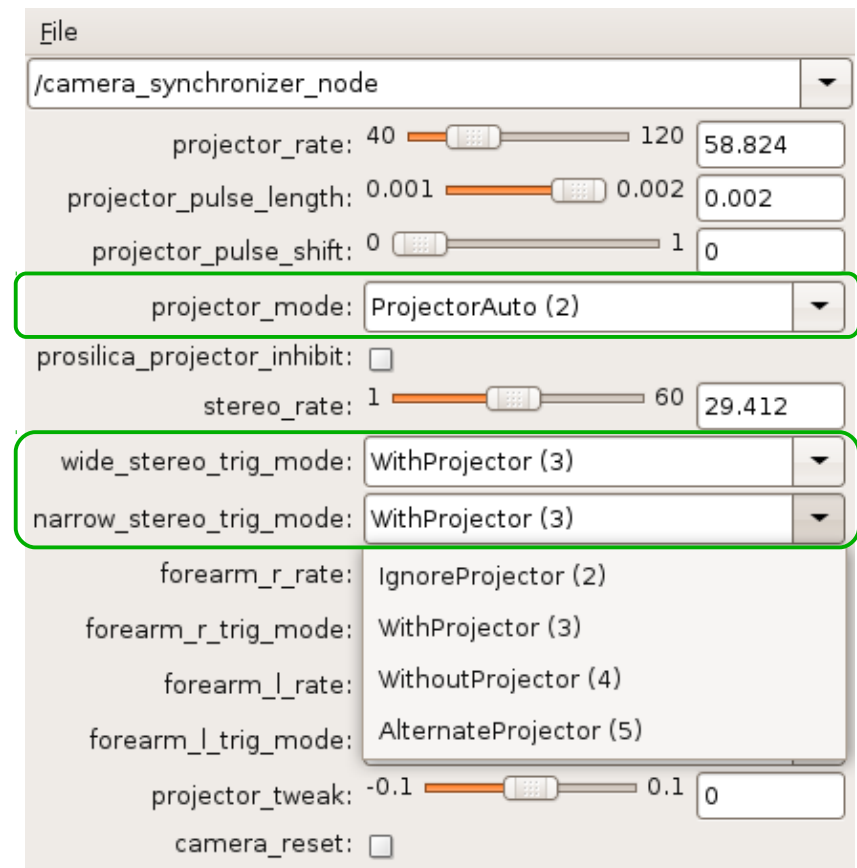
- Turn texture projector on/off





# Re: Using the Texture Projector

- projector\_mode – whether projector is turned on
- \*\_trig\_mode – whether the camera syncs with the projector on all, no, or some frames



# PCL Tutorials

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- <http://www.ros.org/wiki/pcl/Tutorials>
- Downsampling data (*VoxelGrid*)
- Planar model segmentation (*SACSegmentation*)
- Exercise: build a node that segments a table in front of the robot
- **PROBLEM:** Not all tools support PointCloud2 yet (!)
- **SOLUTION:**
  - `$ rosrun point_cloud_converter point_cloud_converter points2_in:=MYTOPIC`

# PCL VoxelGrid - downsampling

```
1 #include <pcl/filters/voxel_grid.h>
2 #include <pcl/point_types.h>
3 int main (int argc, char **argv) {
4     ros::init (argc, argv, "pcl_demo");
5     ros::NodeHandle nh;
6     point_cloud::Publisher<pcl::PointXYZ> pub_downsampled;
7     pub_downsampled.advertise (nh, "downsampled", 1);
8
9     pcl::PointCloud<pcl::PointXYZ> cloud, cloud_downsampled;
10    pcl::VoxelGrid<pcl::PointXYZ> grid;
11    grid.setFilterFieldName ("z");
12    grid.setLeafSize (0.01, 0.01, 0.01);
13    grid.setFilterLimits (0.4, 1.6);
14    while (nh.ok ()) {
15        sensor_msgs::PointCloud2ConstPtr cloud2_blob_ptr =
16        ros::topic::waitForMessage<sensor_msgs::PointCloud2>("/narrow_stereo_
17        extured/points2");
18        point_cloud::fromMsg (*cloud2_blob_ptr, cloud);
19        grid.setInputCloud
20        (boost::make_shared<pcl::PointCloud<pcl::PointXYZ> > (cloud));
21        grid.filter (cloud_downsampled);
22        pub_downsampled.publish (cloud_downsampled);
23    }
```

# PCL Segmentation - planar

```
#include <pcl/filters/project_inliers.h>
#include <pcl/segmentation/sac_segmentation.h>
#include <pcl/PointIndices.h>
#include <pcl/ModelCoefficients.h>
...
point_cloud::Publisher<pcl::PointXYZ> pub_plane;
pub_plane.advertise (nh, "plane", 1);
...
pcl::PointIndices plane_inliers;
pcl::ModelCoefficients plane_coefficients;
...
pcl::SACSegmentation<pcl::PointXYZ> seg;
seg.setDistanceThreshold (0.05);
seg.setMaxIterations (1000);
seg.setModelType (pcl::SACMODEL_PLANE);
seg.setMethodType (pcl::SAC_RANSAC);

pcl::ProjectInliers<pcl::PointXYZ> proj;
proj.setModelType (pcl::SACMODEL_PLANE);
...
seg.setInputCloud (boost::make_shared<pcl::PointCloud<pcl::PointXYZ> > (cloud_downsampled));
seg.segment (plane_inliers, plane_coefficients);

proj.setInputCloud (boost::make_shared<pcl::PointCloud<pcl::PointXYZ> > (cloud_downsampled));
proj.setIndices (boost::make_shared<pcl::PointIndices> (plane_inliers));
proj.setModelCoefficients (boost::make_shared<pcl::ModelCoefficients> (plane_coefficients));
...
pub_plane.publish (cloud_plane);
```

# If time allows: PCL **nodelets**!

---

- Goal:
  - *write once, parameterize everywhere* ⇒ **modular code**
  - ideally, each algorithm is a “*building block*” that consumes input(s) and produces some output(s)
  - in ROS, this is what we call a **node**. inter-process data passing however is inefficient. ideally we need shared memory.
- Solution:
  - **nodelets** = “nodes in nodes” = single-process, multi-threading

# If time allows: PCL **nodelets**!

---

- Nodelets:
  - same ROS API as nodes (subscribe, advertise, publish)
  - dynamically (un)loadable
  - optimizations for zero-copy Boost shared\_ptr passing
  - PCL nodelets use dynamic\_reconfigure for on-the-fly parameter setting

# PCL VoxelGrid nodelet example

```
<launch>
  <node pkg="nodelet" type="nodelet" name="foo"
    args="load pcl/VoxelGrid pcl_manager">
    <remap from="/foo/input"
      to="/narrow_stereo_textured/points"/>
    <rosparam>
      leaf_size: [0.015 , 0.015 , 0.015]
      filter_field_name: "z"
      filter_limit_min: 0.8
      filter_limit_max: 5.0
    </rosparam>
  </node>
</launch>
```

# Questions?

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<http://www.ros.org/>

<http://www.ros.org/wiki/pcl>

[ros-users@code.ros.org](mailto:ros-users@code.ros.org)