

Point Cloud Data

ENV Analysis Type A:

Memory used:	74812 kbytes
Points in cloud:	459701
Polygons generated:	60759
Objects:	251 (697)

Wireframe

Adaptive polygonal optimisations: ON
 Adaptive curvature analysis: ON

SD Tree statistics:

Number of faces stored in cache:	9521
Maximum tree depth of ENV analysis:	32.5
Average tree depth detected:	17.54
Tree nodes:	2578
Tree faces:	21049
Average leaf:	22,9074
Raycasts required:	17226

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

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(your friendly 3D processing library. 0.1)



Why PCL? Why PointCloud2?

I. sensor_msgs/PointCloud => inefficient

- not aligned (e.g., SSE)
- no way to represent organized data
- large overhead
- the world is “float” ☺

Header header

geometry_msgs/Point32[] points

ChannelFloat32[] channels

float32 x string name

float32 y float32[] values

float32 z

Why PCL? Why PointCloud2?

II. point_cloud_mapping => C library

- easy to use for simple stuff

```
void computeCentroid (const sensor_msgs::PointCloud &points, sensor_msgs::PointCloud &centroid);
```

- hard for more “complicated” things

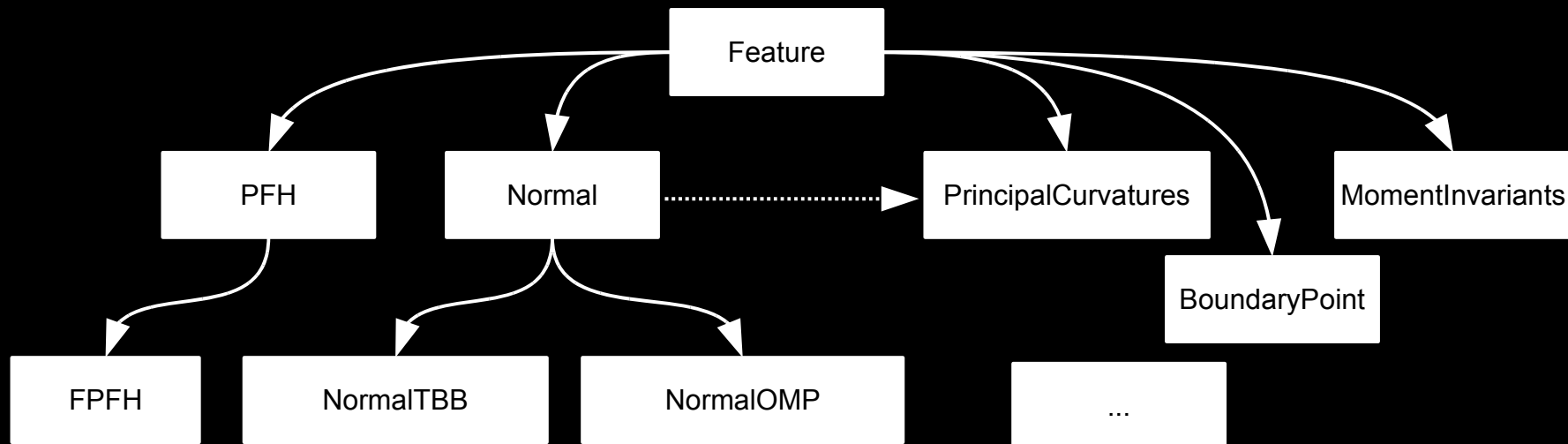
```
void computeOrganizedPointCloudNormalsWithFiltering (sensor_msgs::PointCloud &points, const sensor_msgs::PointCloud &surface, int k, int downsample_factor, int width, int height, double max_z, double min_angle, double max_angle, const geometry_msgs::Point32 &viewpoint);
```

- no way to store state, hard to reuse code as “building blocks”, too many nodes
- worst possible example: N versions of the same thing (e.g., same planar detector copied in 6 packages!) => maintainability is an issue! If 1 copy is updated, the rest are left behind!

Why PCL? Why PointCloud2?

III. more on reusability, architecture

- Algorithmically:
 - door detection = table detection = wall detection = ...
 - the only thing that changes is: parameters (constraints)!
- Inheritance simplifies development and testing (C++ rocks!)



Why PCL? Why PointCloud2?

III. more on reusability, architecture

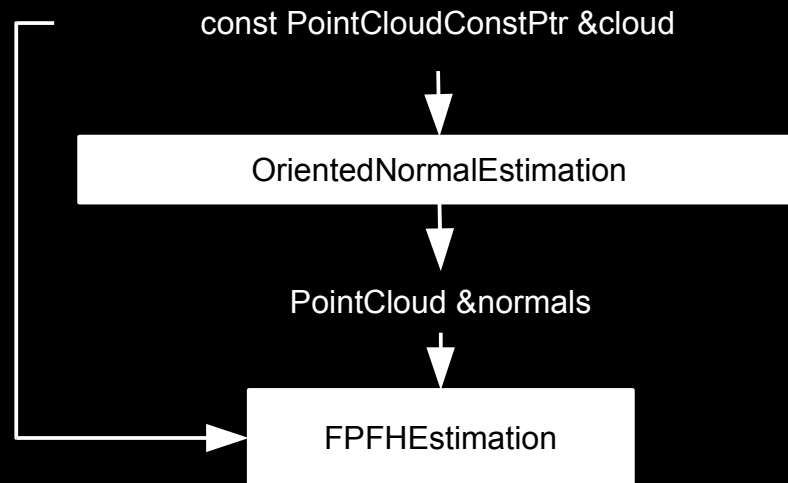
- Algorithmically:
 - door detection = table detection = wall detection = ...
 - the only thing that changes is: parameters (constraints)!
- Inheritance simplifies development and testing (C++ rocks!)

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

Why PCL? Why PointCloud2?

IV. PointCloud is the largest data type in ROS. Need ways to pass data without sacrificing reusability, performance, etc...

- Ideally, all in *shared memory*! N processing building blocks, joint access.



Solutions!

I. sensor_msgs/PointCloud2

- binary blob + channel/field description

II. PCL

- fully templated modern C++ library, SSE optimizations (Eigen backend + custom), OpenMP and TBB enabled
- data passing through Boost shared pointers

III. Nodelets

- “nodes within nodes”
- same ROS API as nodes, dynamically loadable
- optimizations for zero-copy Boost shared_ptr passing

sensor_msgs/PointCloud2

```
#This message holds a collection of nD points, as a binary blob.
Header header

# 2D structure of the point cloud. If the cloud is unordered,
# height is 1 and width is the length of the point cloud.
uint32 height
uint32 width

# Describes the channels and their layout in the binary data blob.
PointFiel[] fields

bool    is_bigendian # Is this data bigendian?
uint32  point_step   # Length of a point in bytes
uint32  row_step     # Length of a row in bytes
uint8[] data         # Actual point data, size is (row_step*height)
bool is_dense        # True if there are no invalid points
```


sensor_msgs/PointCloud2

```
#This message holds the description of one point entry in the  
#PointCloud2 message format.
```

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

```
string name      # Name of field  
uint32 offset   # Offset from start of point struct  
uint8 datatype  # Datatype enumeration see above  
uint32 count    # How many elements in field
```

sensor_msgs/PointCloud2

- PointField 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
```

- 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
```

sensor_msgs/PointCloud2

- Binary blobs are hard to work with
- custom converter, Publisher/Subscriber, transport tools, filters, etc, similar to images (the whole nine yards)
- templated types: PointCloud2 \rightarrow PointCloud<T>
- examples of T:
 - ```
struct PointXYZ
{
 float x;
 float y;
 float z;
}
```
  - ```
struct Normal
{
    float normal[3];
    float curvature;
}
```

Nodelet

- dynamically loadable C++ ROS entity
- uses ROS API for communication
 - Publish/Subscribe
 - parameters
 - remappings
- walks like a node, talks like a node, except:
 - nodelets reside in the same process, while
 - each node is a separate process
 - **because it's single-process, communication can be efficient!!!** Instead of publishing data, we publish Boost shared pointers to data! (roscpp optimizations)

Nodelet Implementation Example

```
class Plus : public nodelet::Nodelet
{
    void init()
    {
        private_nh_.getParam("value", value_);
        pub=private_nh_.advertise<std_msgs::Float64>("out", 10);
        sub=private_nh_.subscribe("in", 10, &Plus::callback, this);
    };
    void callback(const std_msgs::Float64::ConstPtr& input)
    {
        boost::shared_ptr<std_msgs::Float64> output;
        output = boost::make_shared<std_msgs::Float64> ();
        output->data= input->data + value_;
        pub.publish(output);
    };
    ros::Publisher pub;
    ros::Subscriber sub;
    double value_;
};
```

Dynamic Loading with Pluginlib

- Library declaration

```
PLUGINLIB_DECLARE_CLASS
```

```
(nodelet_tutorial_math, Plus, nodelet_tutorial_math::Plus, nodelet::Nodelet);
```

- Manifest Export

```
<export>
```

```
  <nodelet plugin="${prefix}/nodelet_math.xml"/>
```

```
</export>
```

- Plugin description file

```
<library path="lib/libnodelet_math">
```

```
  <class name="nodelet_tutorial_math/Plus"
```

```
    type="nodelet_tutorial_math::Plus"
```

```
    base_class_type="nodelet::Nodelet">
```

```
  <description>Add a value and republish. </description>
```

```
  </class>
```

```
</library>
```


Running from the command line

```
$ rosrun nodelet standalone_nodelet  
$ rosrun nodelet nodelet foo_name  
nodelet_tutorial_math/Plus standalone_nodelet
```

- This will run a nodelet manager then call into it to dynamically load a nodelet.

Nodelets in Launch Files

- An example of launching a nodelet from a file

```
<launch>
  <node pkg="nodelet"
        type="standalone_nodelet"
        name="nodelet_manager"/>
  <node pkg="nodelet"
        type="nodelet"
        name="$(anon nodelet)"
        args="Plus3 nodelet_tutorial_math/Plus nodelet_manager">
    <param name="value" type="double" value="2.5"/>
    <remap from="Plus3/in" to="Plus2/out"/>
  </node>
</launch>
```

PCL

- Features:
 - complete Eigen backend (patches go upstream)
 - SSE, OpenMP, TBB optimizations (future CUDA)
 - fully templated on the point type

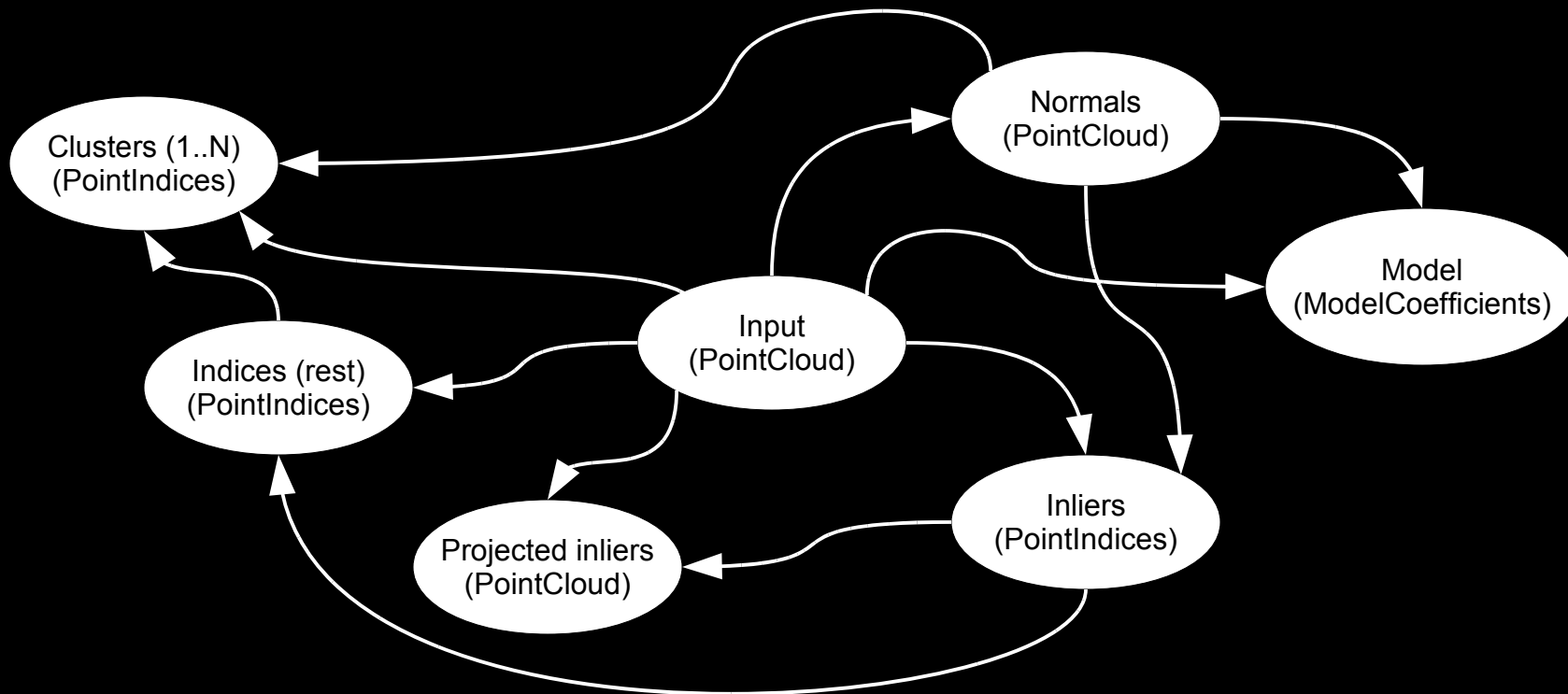
```
template <typename PointT>
class Foo: public boost::enable_shared_from_this <Foo<PointT> >, public nodelet::Nodelet
{
    typedef point_cloud::PointCloud<PointT> PointCloud;
    ...
}
PointCloud<pcl::PointXYZ> cloud;
cloud.points[i].x = ...
Foo<pcl::PointXYZ> bar;
```

PCL

- Structure:
 - C++ libraries, with a few simple C-style methods: 6 in progress + more soon
 - `libPCL_features`, `libPCL_surface`, `libPCL_filters`, `libPCL_segmentation`, `libPCL_io` – tested/testing and documented
 - `libPCL_registration` – design in progress
 - Features, segmentation, io → unit tested
 - C++ classes are templated building blocks (nodelets!) via inheritance and `init()`
 - “standard” node examples still exist

PCL

- Philosophy: *write once, parameterize everywhere*
- PPG: **P**erception **P**rocessing **G**raphs



PCL

- Misc, stats:
 - Approx 77 classes, over 25k lines of code (10-15% done by my estimation in terms of the functionality that I want ☺ - oh well, need more coders!)
 - External dependencies (for now) on [eigen](#), [cminpack](#), [ANN](#), [FLANN](#), [TBB](#)
 - Internal dependencies (excluding the obvious) on [dynamic_reconfigure](#), [message_filters](#)
 - ...

Usage examples

- Simple downsampling and filtering

```

<launch>
  <node pkg="nodelet" type="standalone_nodelet" name="pcl_manager" output="screen" />

  <!-- Downsample the data -->
  <node pkg="nodelet" type="nodelet" name="foo" args="voxel_gridVoxelGrid pcl_manager">
    <remap from="/voxel_grid/input" to="/narrow_stereo_textured/points" />
    <rosparam>
      # -[ Mandatory parameters
      leaf_size: [0.015, 0.015, 0.015]
      # -[ Optional parameters
      filter_field_name: "z" # The field name that holds distance values (for filtering)
      filter_limit_min: 0.8 # points closer than 0.8m from the viewpoint will not be considered
      filter_limit_max: 5.0 # points closer than 5.0m from the viewpoint will not be considered
      use_indices: false # false by default
    </rosparam>
  </node>
  ...
</launch>

```

Usage examples

- Normal estimation

```

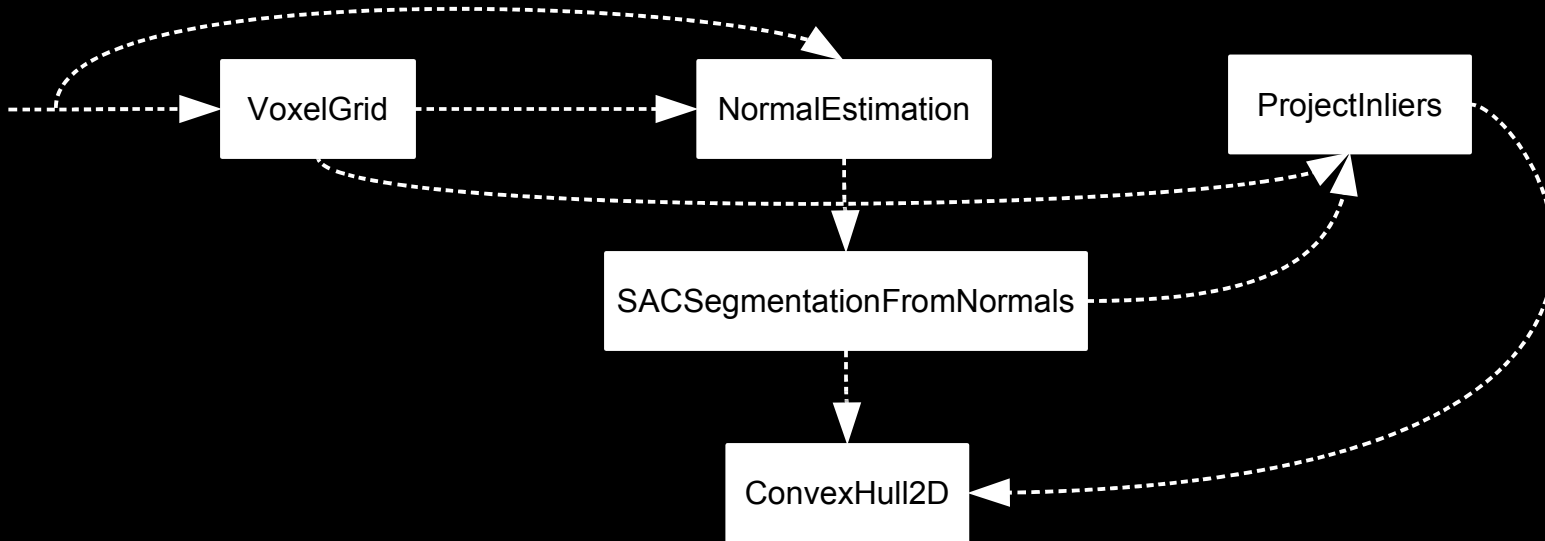
<launch>
  <node pkg="nodelet" type="standalone_nodelet" name="pcl_manager" output="screen" />
  <!-- Estimate point normals -->
  <node pkg="nodelet" type="nodelet" name="foo" args="normal_estimation NormalEstimation pcl_manager">
    <remap from="/normal_estimation/input" to="/voxel_grid/output" />
    <remap from="/normal_estimation/surface" to="/narrow_stereo_textured/points" />
    <rosparam>
      # -[ Mandatory parameters
      # Set either 'k_search' or 'radius_search'
      k_search: 0
      radius_search: 0.1
      # Set the spatial locator. Possible values are: 0 (ANN), 1 (FLANN), 2 (organized)
      spatial_locator: 0

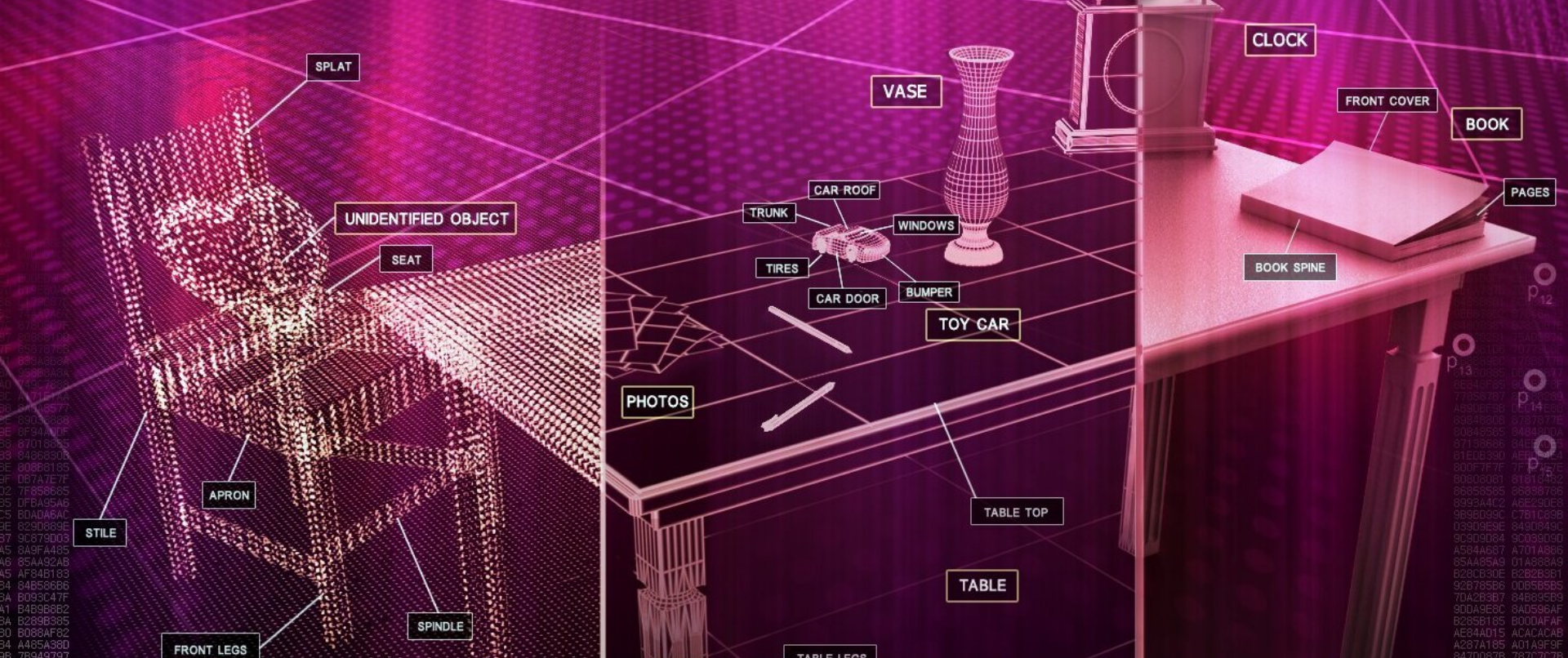
      # -[ Optional parameters
      use_indices: false # false by default
      use_surface: false # false by default
    </rosparam>
  </node>
  ...
</launch>

```

Usage examples

- (improved) table segmentation (constrained!)





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Thanks to Tully Foote, Patrick Mihelich, Josh Faust, Kurt Konolige from Willow Garage, and our colleagues at TUM!

