

# Advanced Techniques for Mobile Robotics

## ROS

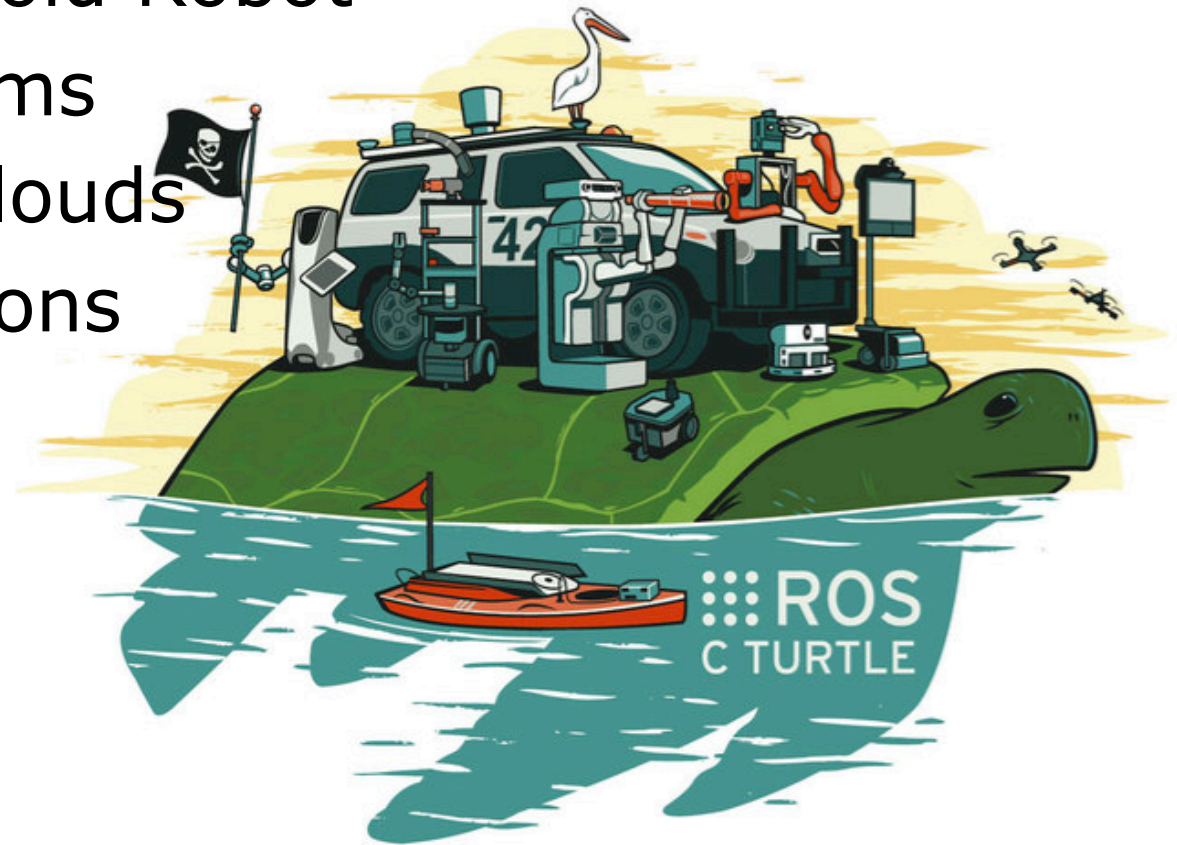
Wolfram Burgard, Cyrill Stachniss,  
Kai Arras, Maren Bennewitz

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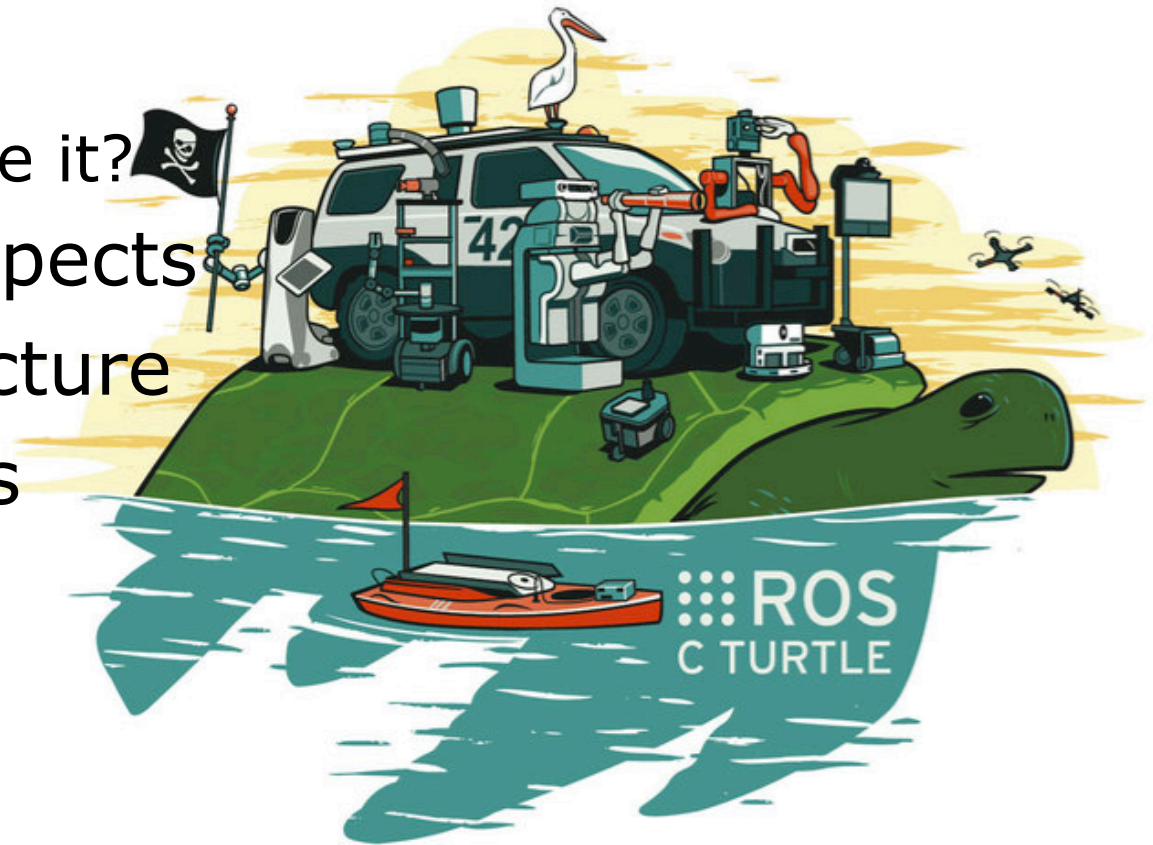
# Today's Lecture

- **ROS – Robot Operating System**
- PR2 – Humanoid Robot
- TF – Transforms
- PCL – Point Clouds
- ROS Applications



# ROS Overview Talk

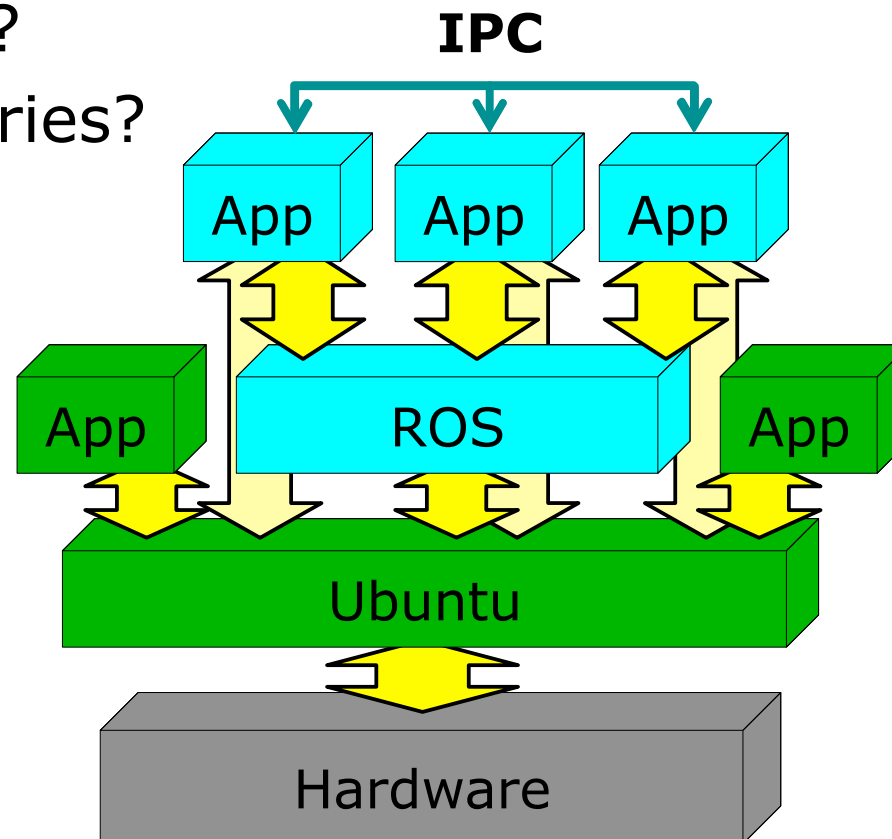
- High Level Overview
  - What is ROS?
  - Who made it?
  - Why do we use it?
- Middleware Aspects
- Software Structure
- Client Libraries



# High Level Overview

## What is ROS?

- An Operating System like Windows, GNU/Linux?
- A Software Distribution?
- A Middleware?
- A Set of Libraries?





# High Level Overview

## Who made ROS?



- Privately Owned Company
- Based in Menlo Park, California
- Hardware: PR2, Texai
- Software: ROS (OpenCV, Player, PCL)
- Strong Open Source Commitment

# High Level Overview

## Why do we use ROS?

- Great functionality
  - Middleware (this session)
  - Development tools (2<sup>nd</sup> Session)
  - Advanced libraries (3<sup>rd</sup> Session)
  - Hardware drivers
- Large scientific community
  - A lot of state-of-the-art software available
  - Easy to exchange/integrate/build-upon existing projects
  - Open source (mostly BSD)
  - Actively developed by full-time staff
- To stop reinventing the wheel...

How Robotics  
Research Keeps...

# Re-Inventing the Wheel

First, someone  
publishes...



...and they write  
code that barely  
works but lets  
them publish...



...a paper with  
a proof-of-  
concept robot.



This prompts  
another lab to  
try to build on  
this result...



But inevitably,  
time runs out...



...but they can't  
get any details  
on the software  
used to make it  
work...



...and countless  
sleepless nights  
are spent  
writing code  
from scratch.

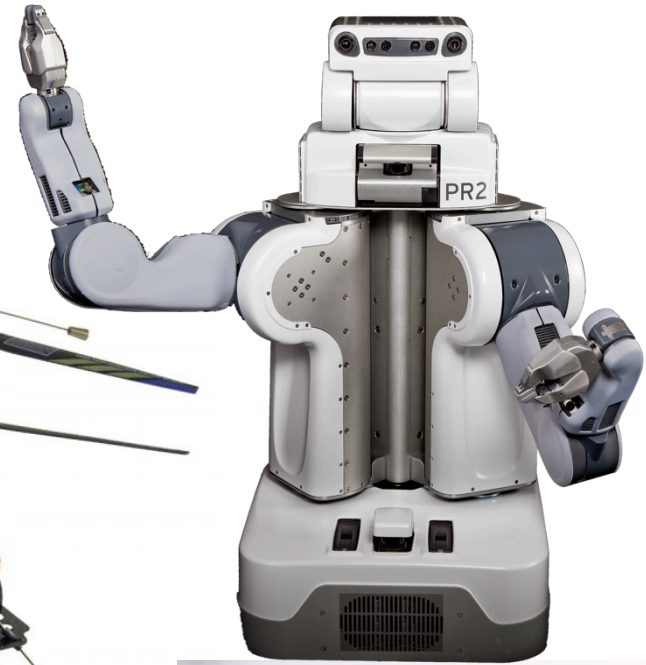
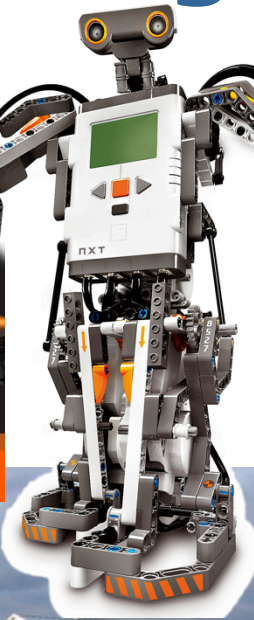
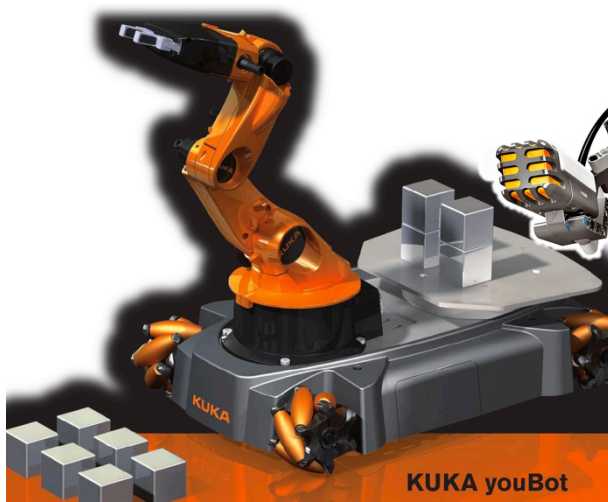


So, a grandiose  
plan is formed  
to write a new  
software API...



...and all the  
code used by  
previous lab  
members is a mess.

# Robots Using ROS





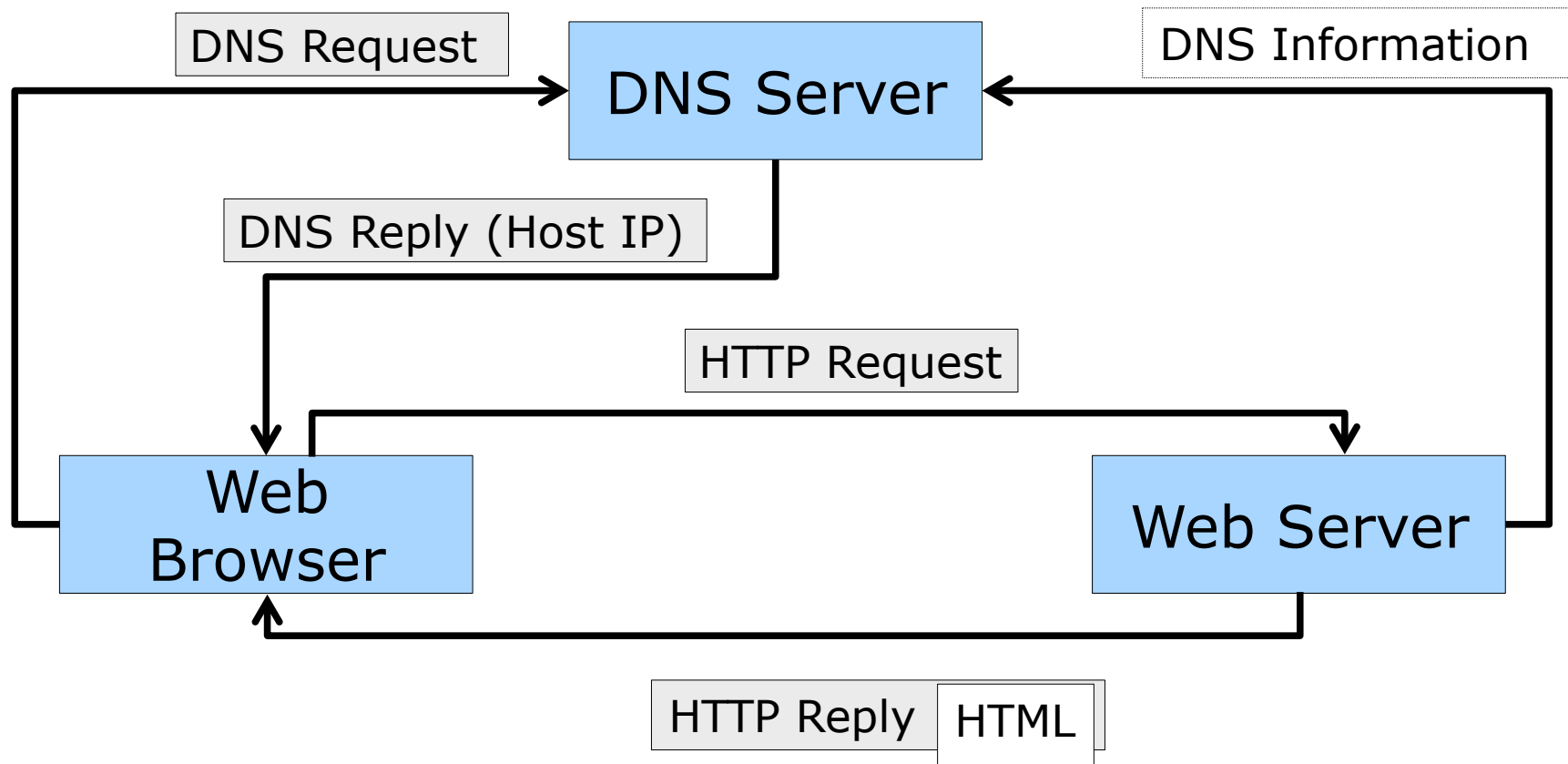
# ROS - a Middleware for Robots

“Middleware is a software that connects software components or applications.”

- Framework for interprocess communication
- Boosts modularization
- Enables transparent distribution of software in a network

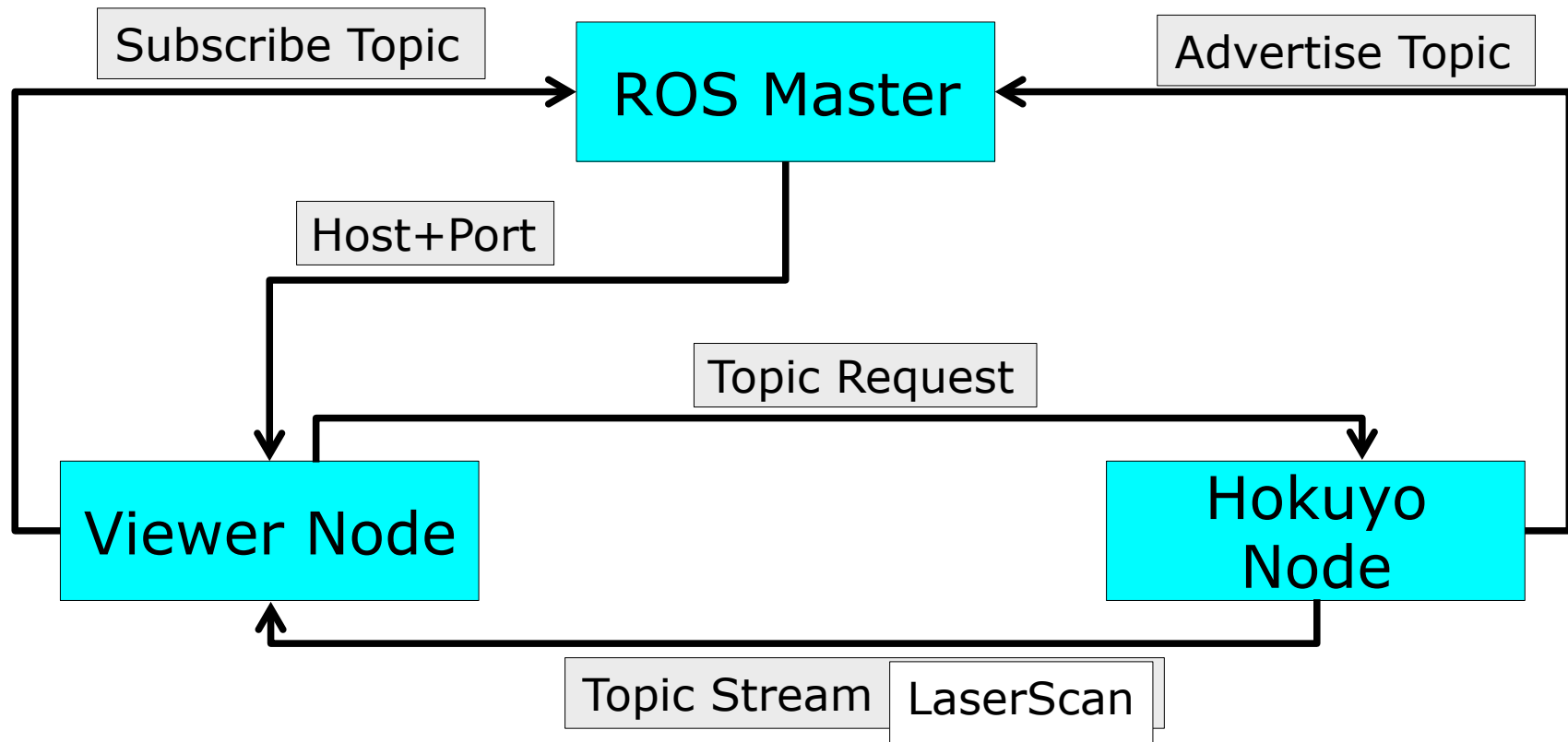
# ROS - a Middleware for Robots

Example: Communication on the Internet



# ROS - a Middleware for Robots

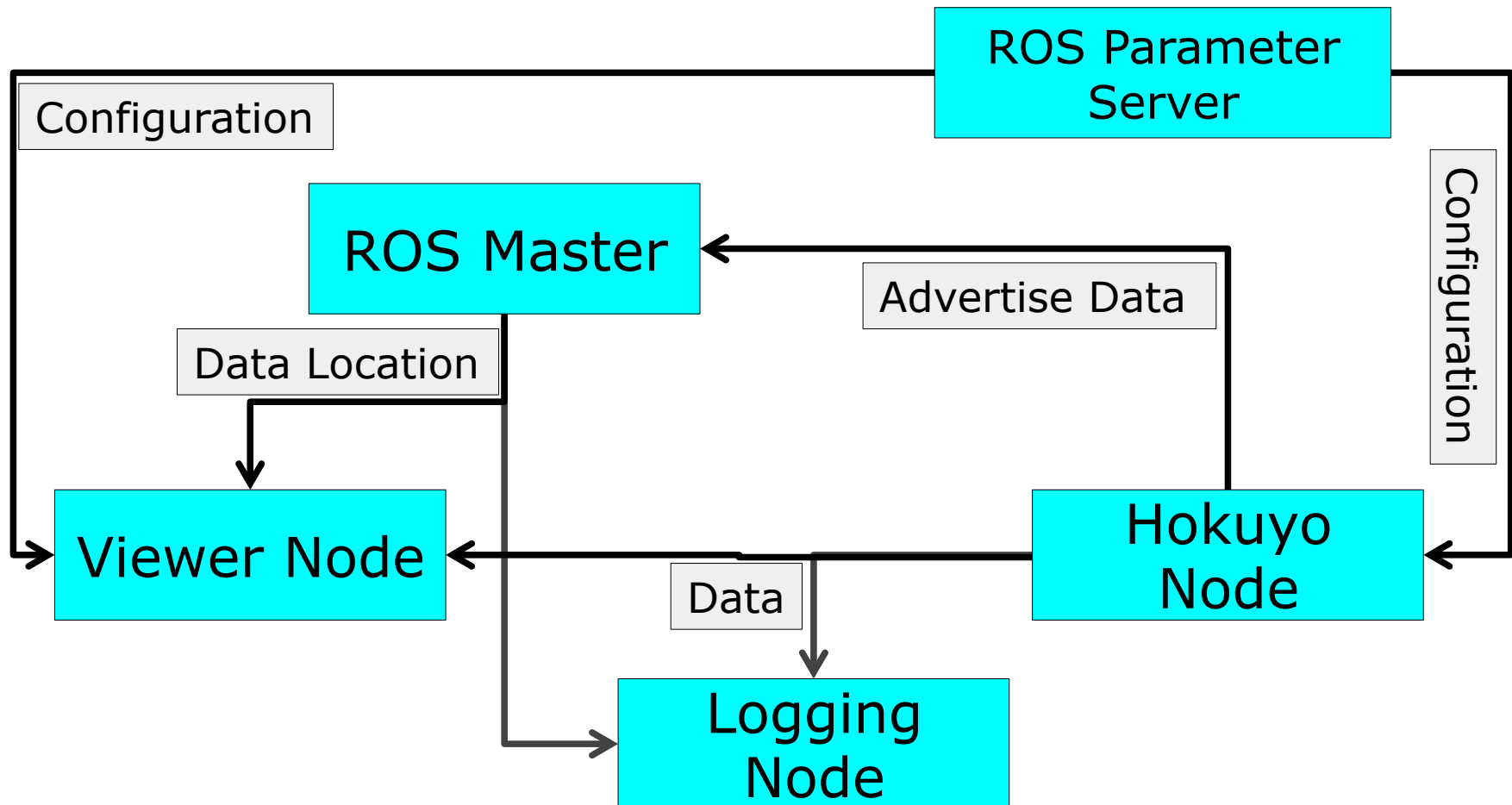
## Interprocess Communication Using ROS



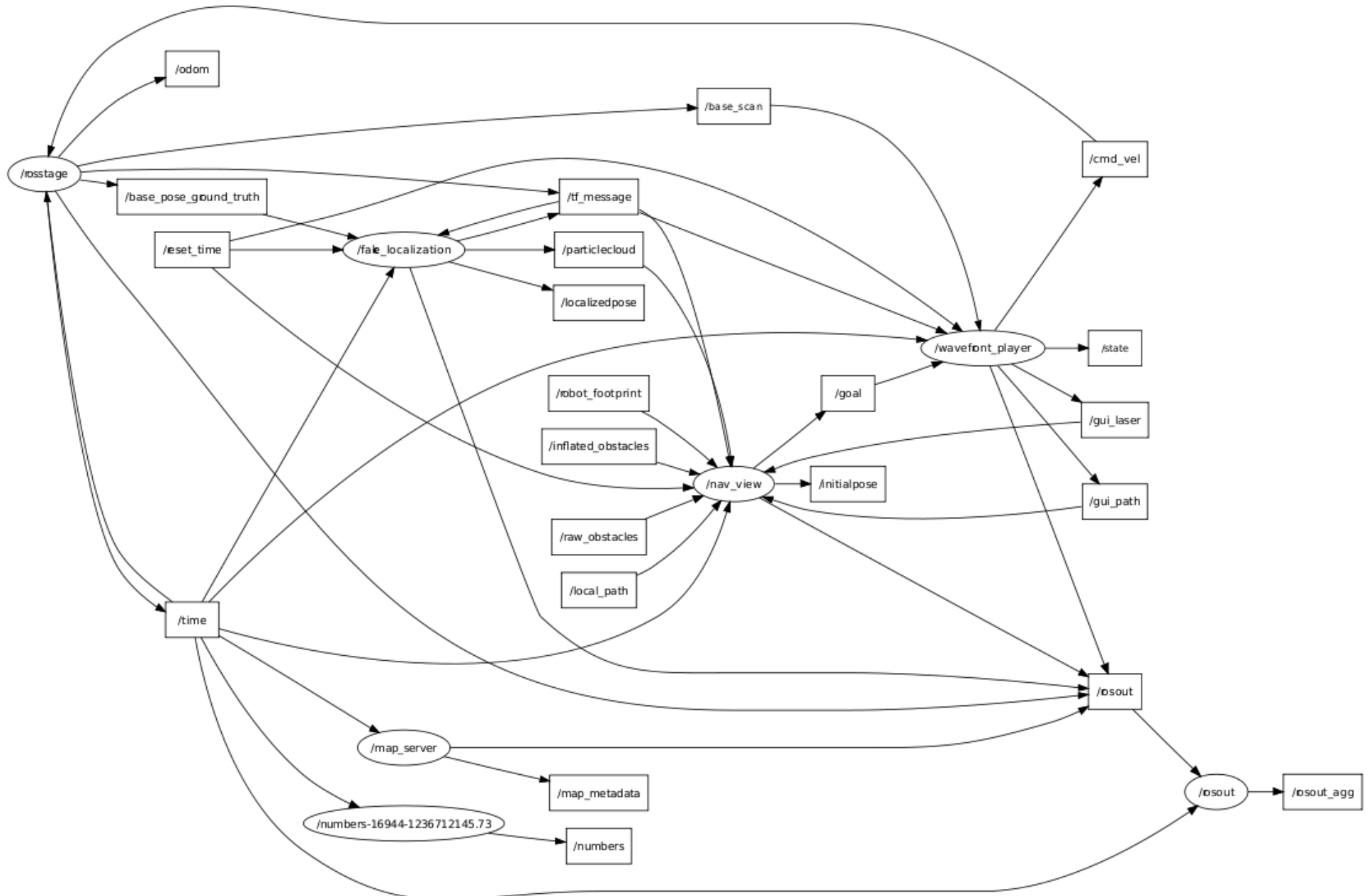


# ROS - a Middleware for Robots

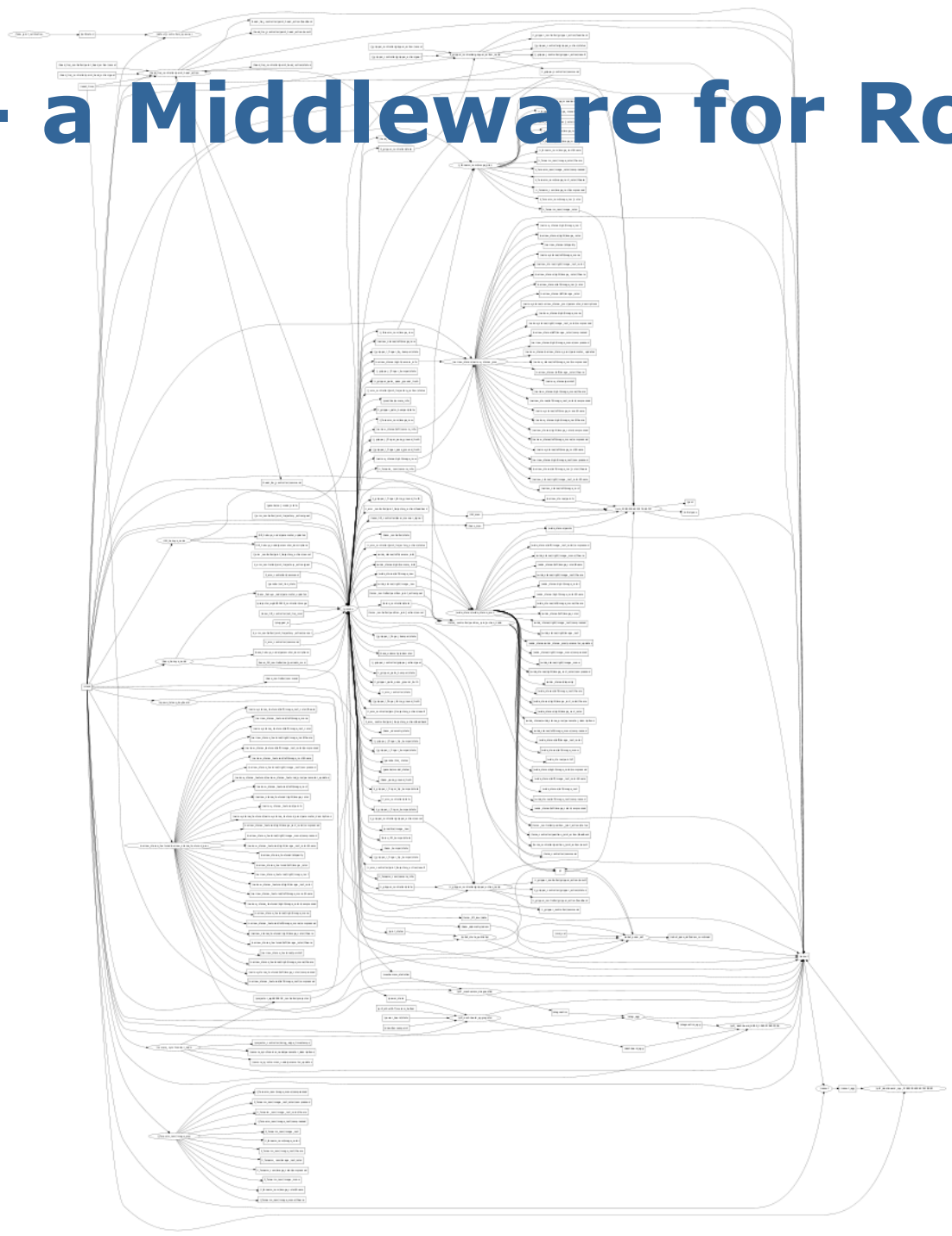
## Interprocess Communication Using ROS



# ROS - a Middleware for Robots



# ROS - a Middleware for Robots



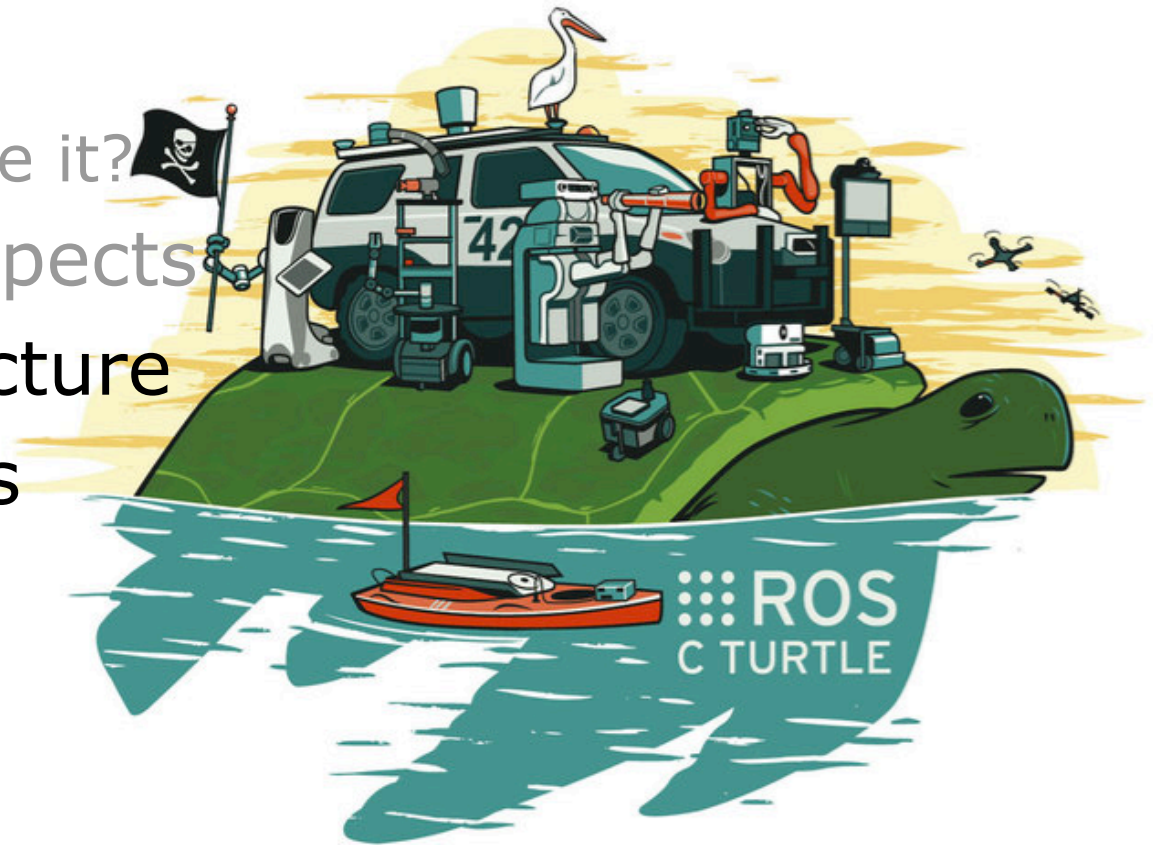
# ROS - a Middleware for Robots

There is more to it...

- (De-)Serialization under the hood
- Central multi-level logging facilities
- Service calls, preemptible actions
- Central time
- Dynamic reconfiguration

# ROS Overview

- High Level Overview
  - What is ROS?
  - Who made it?
  - Why do we use it?
- Middleware Aspects
- Software Structure
- Client Libraries



# ROS – Software Distribution

## Software Hierarchy

- **Release** – collection of stacks and packages
- **Stacks** – a full application suite
- **Package** – software (and interface definition) to solve a specific task
- **Node** – An executable with some useful functionality
- **Message/Service/Action** – Interface definitions

# ROS – Software Distribution



Boxturtle  
March 2010

C-Turtle  
August 2010



Diamondback  
March 2011

## Releases

Electric Emys  
November 2011





# ROS – Software Distribution

ROS.org

[About](#) | [Mailing Lists](#) | [code.ros.org](#)

S

Documentation

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News

[packages](#)

~3150

[repositories](#)

~270

[stacks](#)

~500

search

Name	Packages	Description
<a href="#">2dmapping_pr2_app</a>	1	A 2D mapping application for the PR2 robot platform.
<a href="#">2dnav_pr2_app</a>	1	A 2D navigation application for the PR2 robot platform.
<a href="#">arm_navigation</a>	4	arm_navigation
<a href="#">articulation</a>	5	Kinematic models for articulated objects (cabinet doors fitting, model selection and visualization).
<a href="#">art_vehicle</a>	10	ART autonomous vehicle support
<a href="#">asctec_drivers</a>	6	asctec_drivers
<a href="#">au_automow_common</a>	5	Auburn University Autonomous Lawnmower - Common
<a href="#">au_automow_drivers</a>	3	au_automow_drivers
<a href="#">au_automow_simulation</a>	0	au_automow_simulation
<a href="#">bag_experimental</a>	0	bag_experimental
<a href="#">billiards</a>	17	billiards

# ROS – Software Distribution

- Binary distribution via debian package management
- Source distribution via version control systems
- Tools for dependency resolution
  - Fetch binaries from apt-repository
  - Fetch source code from version control
  - Recursively build dependencies
- Central documentation wiki: [ros.org/wiki](http://ros.org/wiki)

Installation instructions & download mirror:

<http://ros.informatik.uni-freiburg.de>

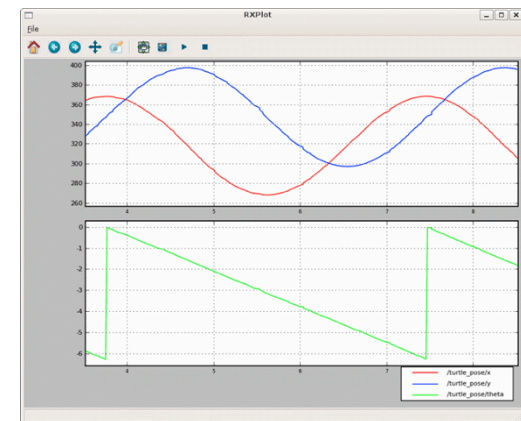
# ROS – Client Libraries

- What makes a program a ROS node?
- → Usage of a client library
  - C++ and Python
  - Octave/Matlab, Lisp, Java and Lua (experimental)
  - You can implement your own
- A client library embeds a program in the ROS interprocess communication network (i.e. attaches it to the middleware)

# ROS – Tool Ecosystem

There is more to it...

- Tools for analysis, debugging
- and visualization of IPC
- Live message view
- Recording and playback
- Many more...

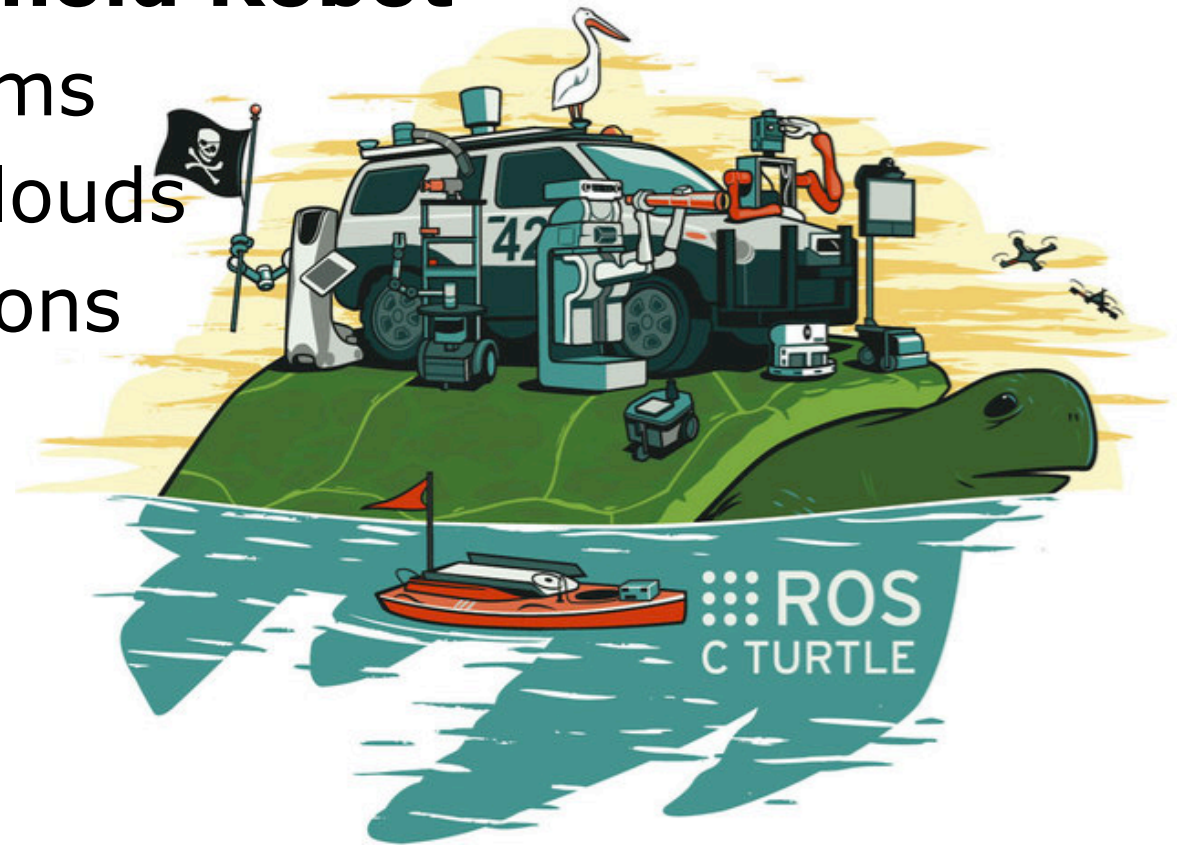


rxconsole window showing a list of messages. The messages are all warnings from the /turtlesim node, indicating that the turtle hit the wall. The messages are: "Oh no! I hit the wall! (Clamping from [x=11.143110, y=5.555555])". The severity is Warn and the node is /turtlesim. The window includes fields for Include, Exclude, and a checkbox for Regex, along with buttons for Clear Messages, Pause, and Setup.

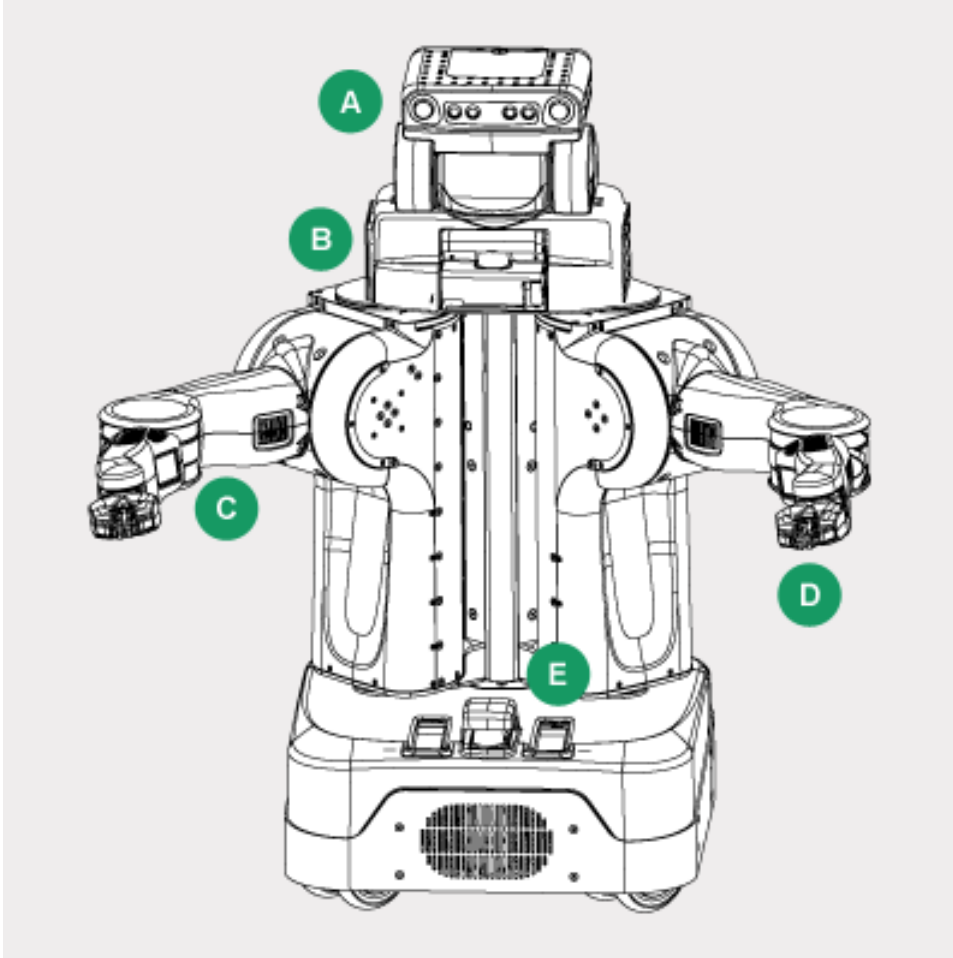
rxbag window showing a timeline of messages and corresponding image raw data. The timeline is labeled "Nov 06 2009 11:57:57.23" and shows a time axis from 0:00 to 7:00. The messages listed are: base\_scan, laser\_tilt\_controller/laser\_scanner\_signal, narrow\_stereo/left/camera\_info, narrow\_stereo/left/image\_raw, narrow\_stereo/right/camera\_info, narrow\_stereo/right/image\_raw, prosilica/cam\_info, prosilica/image\_throttled, tf, tilt\_scan, wide\_stereo/left/camera\_info, wide\_stereo/left/image\_raw, wide\_stereo/right/camera\_info, and wide\_stereo/right/image\_raw. The image raw data is visualized as a series of frames along the timeline.

# Today's Lecture

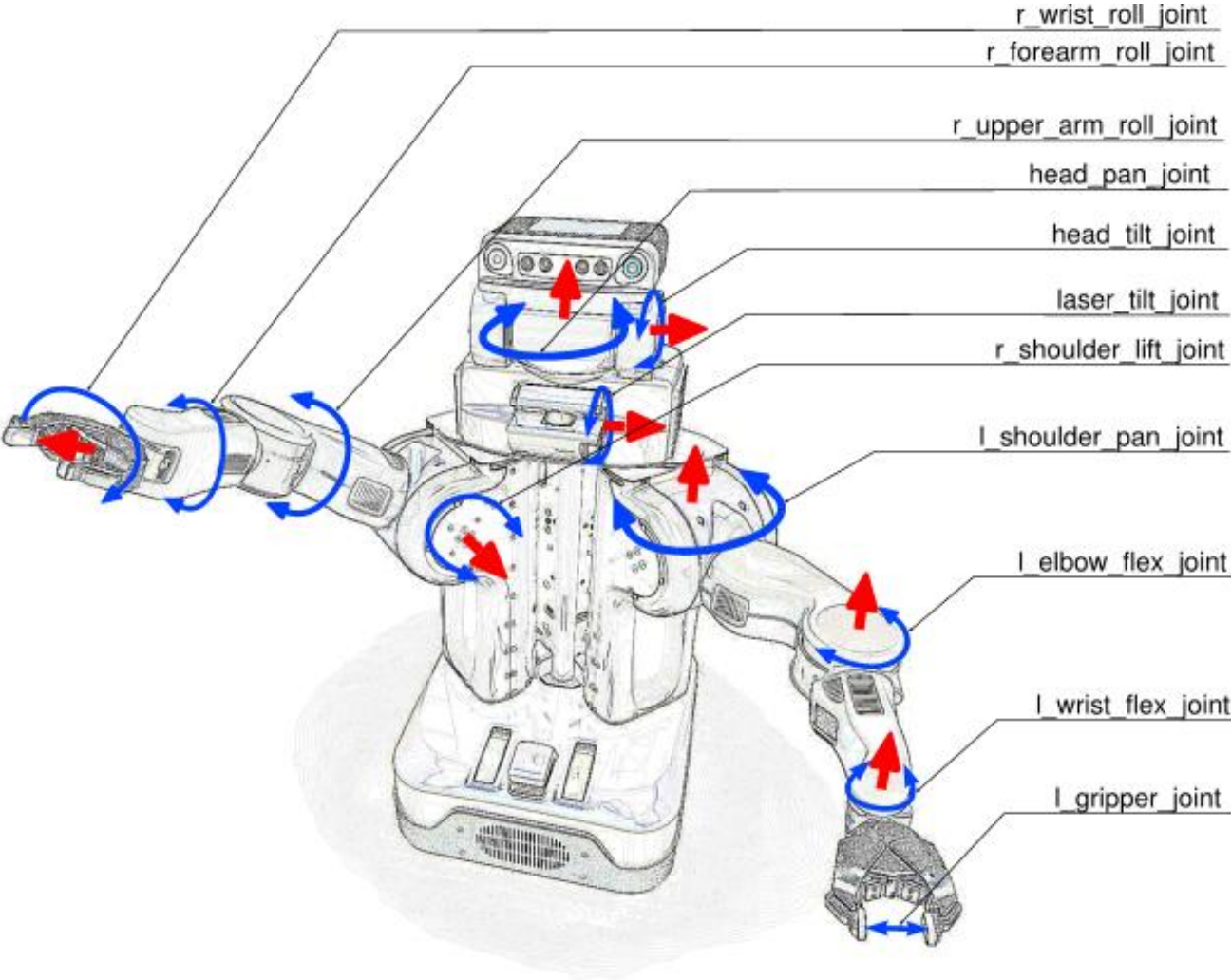
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- **PR2 – Humanoid Robot**
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- ROS Applications



# Sensors



# Joints

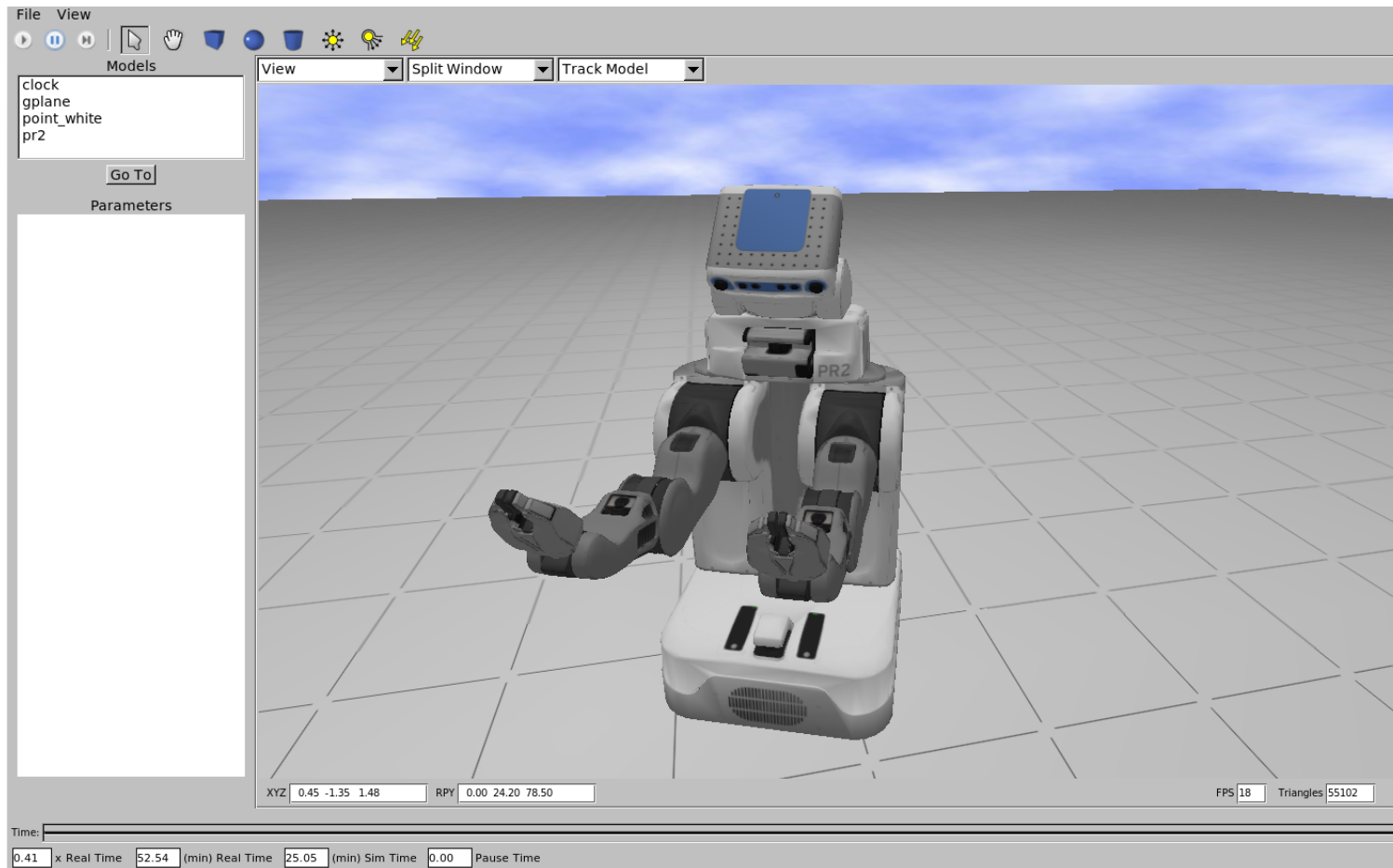




# PC Hardware

- 2x Onboard servers
  - Processors : Two Quad-Core i7 Xeon
  - Memory : 24 GB
- Internal hard drive: 500 GB
- Removable hard drive:: 1.5 TB

# PR2 Simulation



# PR2 Simulation

- Motivation:
  - Regression testing
  - Visualization
  - Design, optimization
  - Multiple developers, single robot
- Based on Open Source project Gazebo
- Uses Open Dynamics and Physics Engine
- Uses Ogre for rendering

# PR2 Simulation

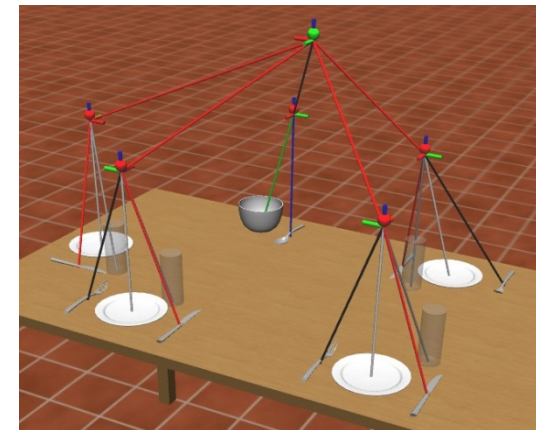
- Requirements:
  - Graphics card with 3D acceleration (Nvidia, ATI)
  - Linux supported driver
  - Core2Duo processor
  - > 2GB Ram

# Simulation Fidelity

- Manipulation:
  - Simulated real-time PR2 etherCAT node communicates actuator states (motor efforts and position)
- Perception:
  - Simulated sensor nodes for cameras, lasers and imu
  - Stream data as well as services provided
- Self collision checks are disabled
- Anti-gravity arms
- No laser scan duration

# Learning to Set a Table

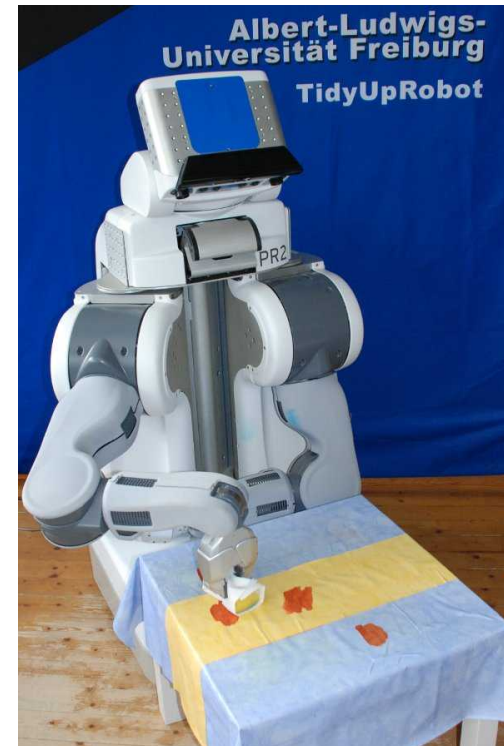
- Observe example scenes of breakfast tables
- Learn a **hierarchical scene model**
  - Level 1: physical objects (segmented point clouds)
  - Level 2: covers (constellations of physical objects)
  - Level 3: table scenes (constellations of covers)
- Sample from the model to **generate new scenes**



# Learning to Clean a Table

- Uncertainty about how the dirt looks like
- Idea: “Dirt is that what we can remove”

1. Cluster image into color classes
2. Observe table state
3. Try wiping each class
4. Updating belief about the expected
5. Clean the entire table
6. Observe again





# Learning to Clean a Table

Learning to Clean a Table

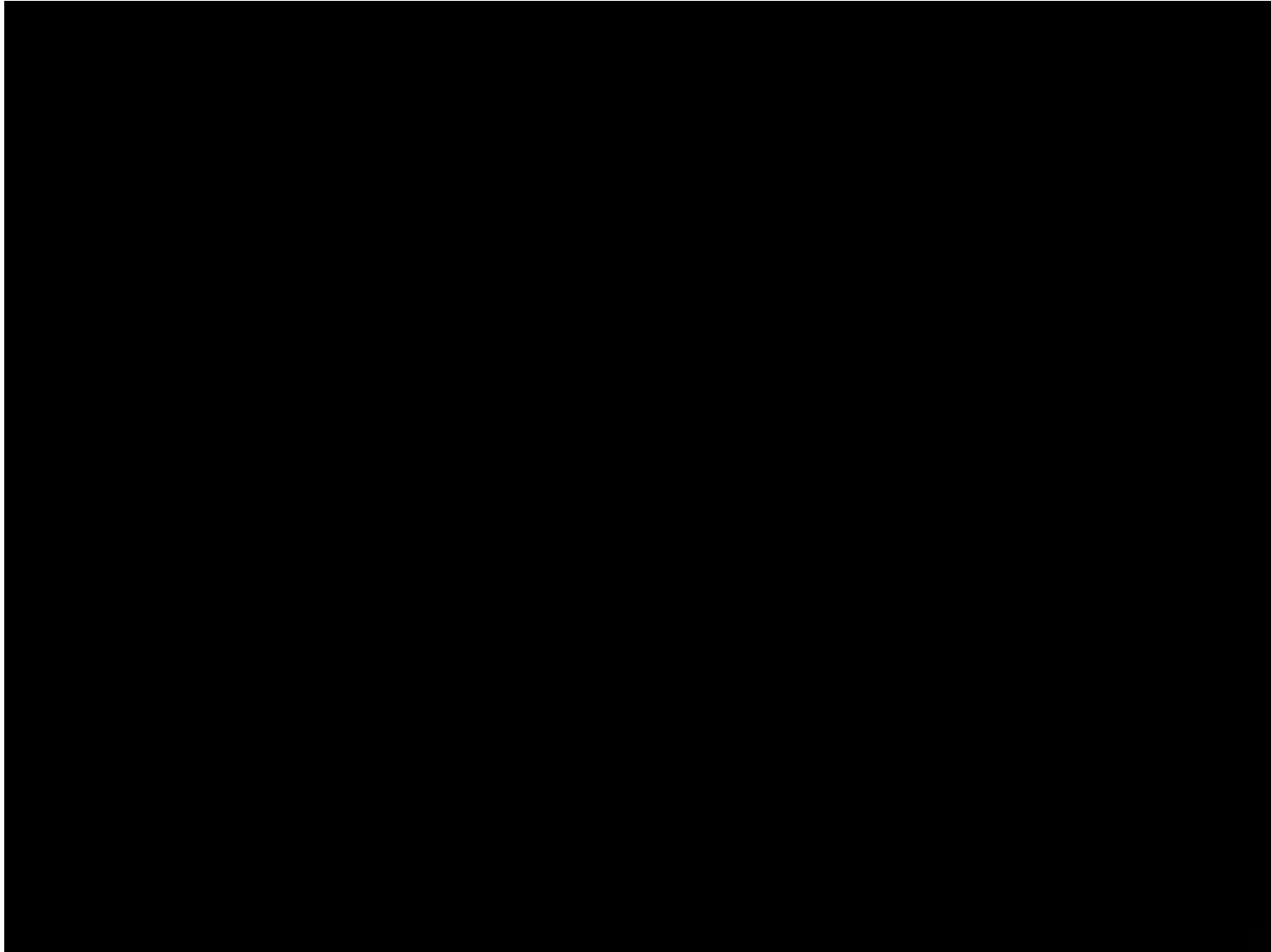
Jürgen Hess, Jürgen Sturm, Wolfram Burgard

Autonomous Intelligent System Lab,  
University of Freiburg, Germany

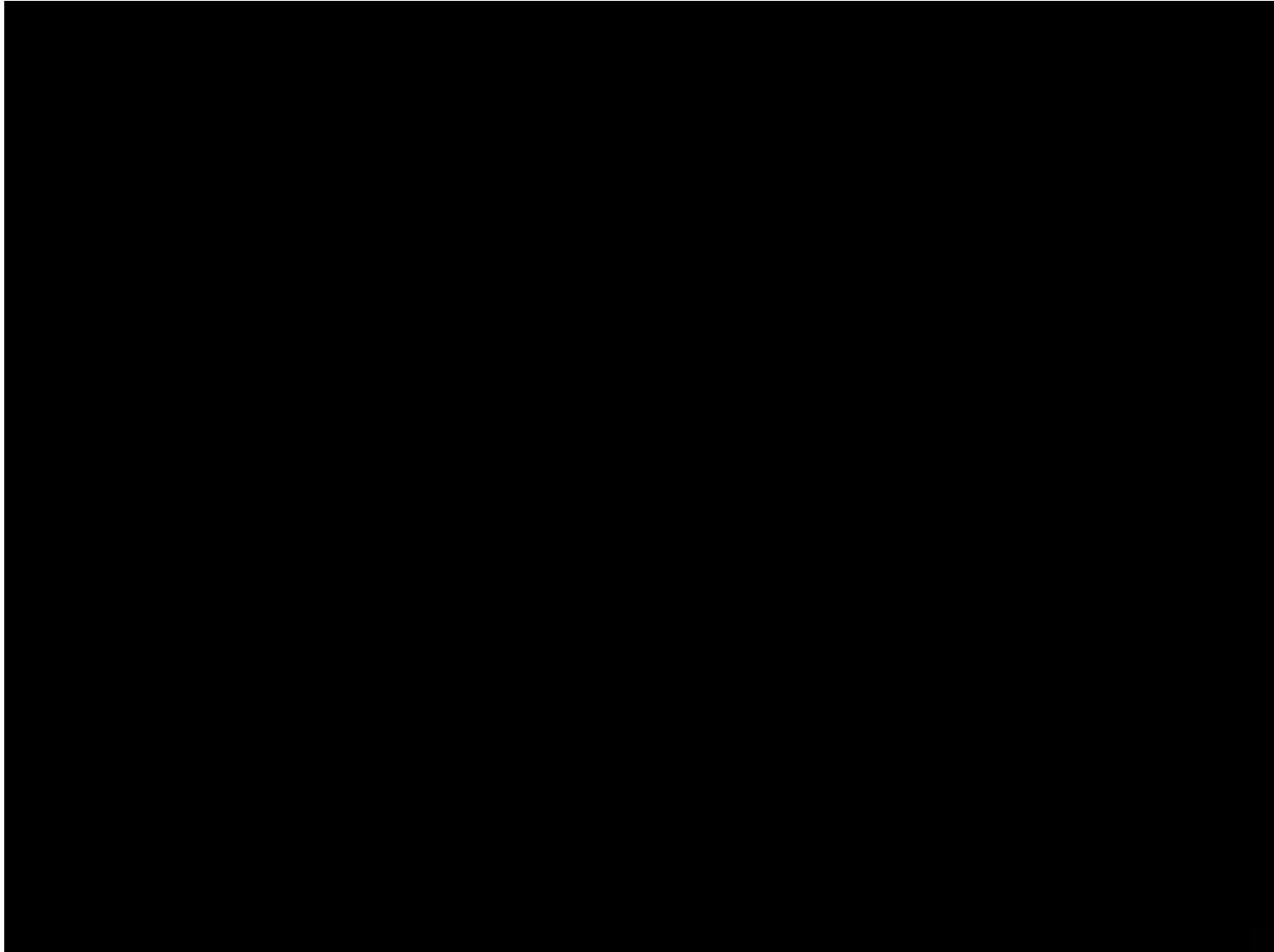
# Student Projects

- Portrait Bot:
  - Recognize faces in camera images
  - Extract edges
  - Draw edge image onto a whiteboard
  - Available in May
- Two-handed cleaning with a broom:
  - Objective: adapting coverage plan to using a broom (e.g. different poses/pattern for sweeping in corners or under a table)

# Student Project: Portrait Bot

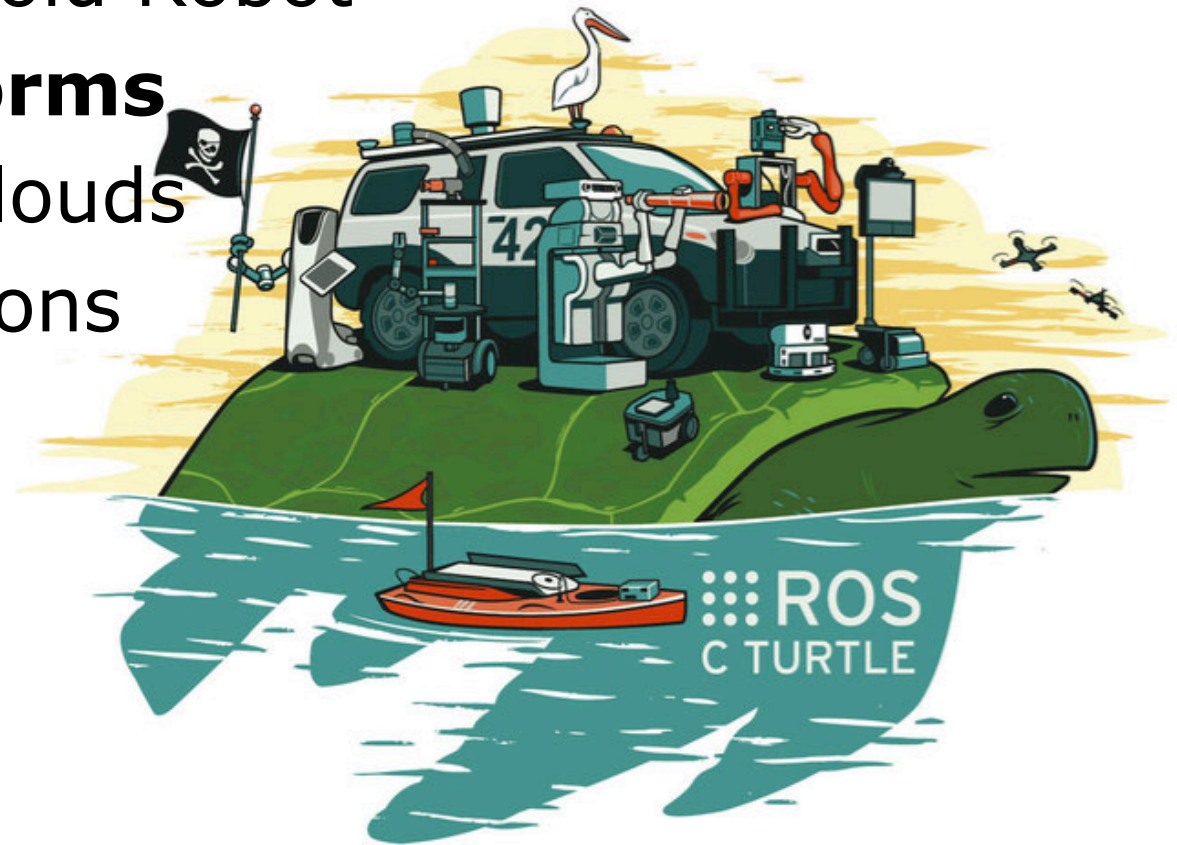


# Student Project: Sweeping



# Today's Lecture

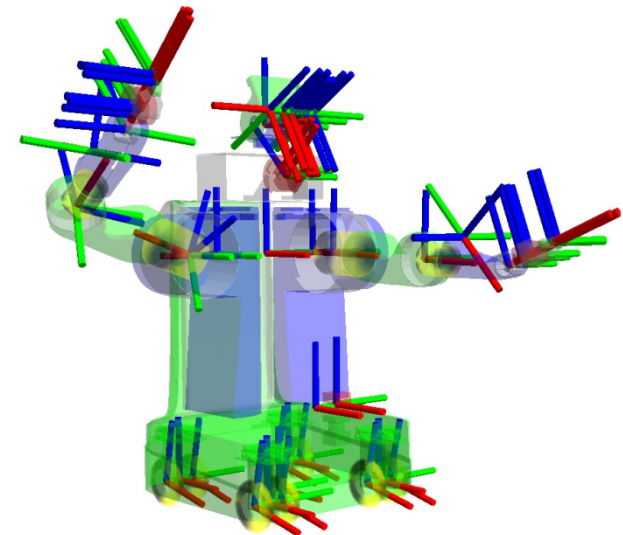
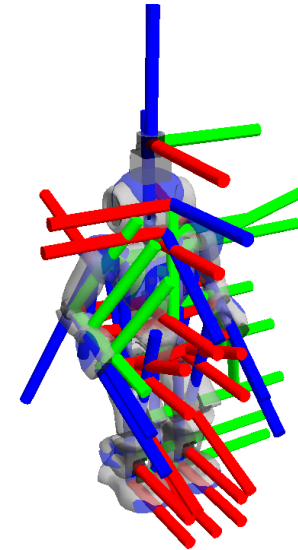
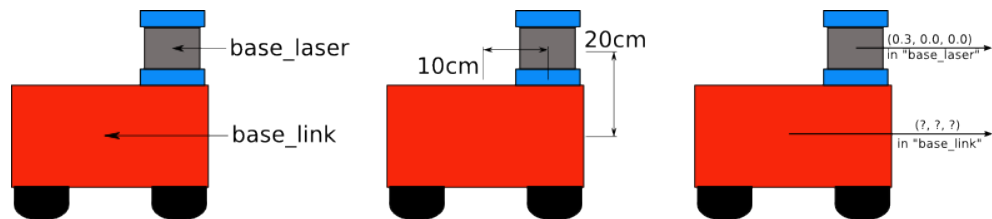
- ROS – Robot Operating System
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- **TF – Transforms**
- PCL – Point Clouds
- ROS Applications



# Motivation

- Multiple sensors and actuators on robots
- Robot and sensors not static
- Multi-robot cooperation

➔ **Where is this object relative to my gripper? Where is my arm in the world?**



# What is tf?

## A coordinate frame tracking system

- Standardized protocol for publishing transform data to a distributed system
- Helper classes and methods for:
  - Publishing coordinate frame data: *TransformBroadcaster*
  - Collecting transform data and using it to manipulate data: *Transformer, TransformListener, tf::MessageFilter, ...*
- Currently: Only tree(s) of transformations, but any robot(s) / sensor layout
- Conversion functions, mathematical operations on 3D poses



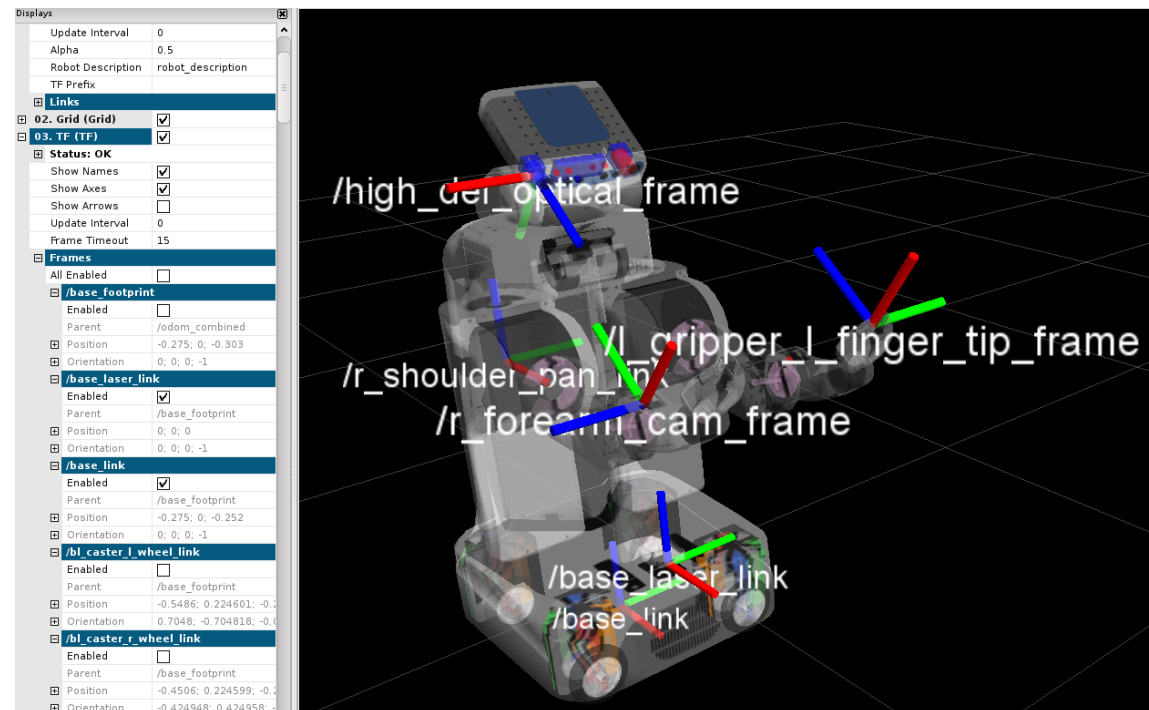
# tf is Distributed!

- Two types of tf nodes: Publishers & Listeners
- Listeners: Listen to /tf and cache all data heard up to cache limit (10 sec default)
- Publishers: Publish transforms between coordinate frames on /tf
- No central source of tf information, or history before a node was started

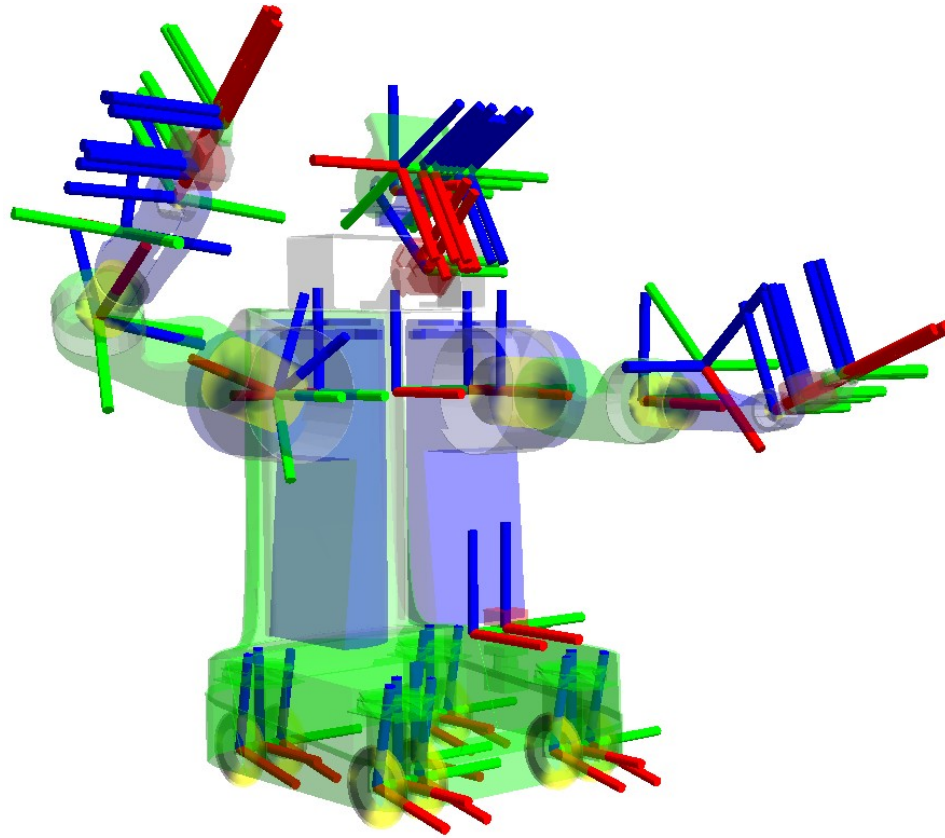
# Debugging Tools

- Command Line Tools
  - `tf_echo`: Print a specific transform to the screen
  - `tf_monitor`: Display statistics about transforms
  - `roswtf`: Debug common tf configuration errors

- Visualizations
  - Rviz tf plugin
  - `view_frames`



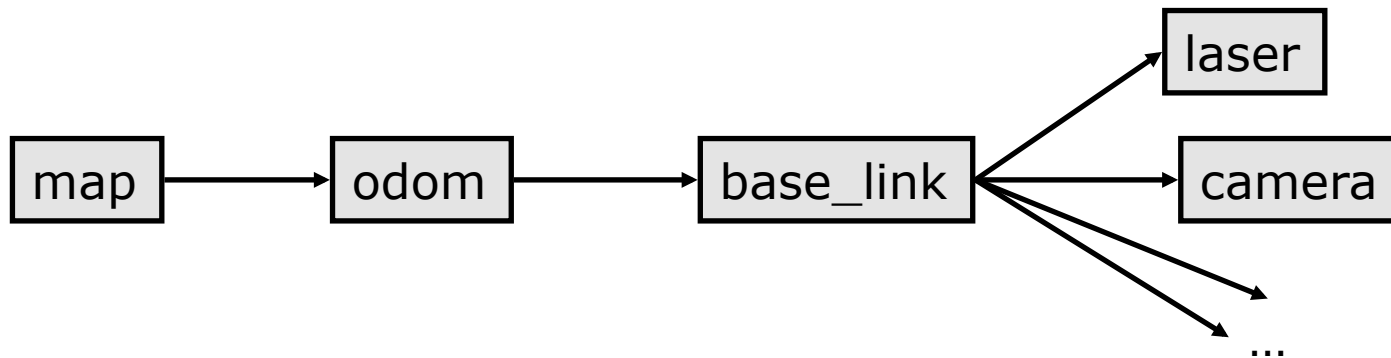
# Coordinate Frames on the PR2



- Coordinate frames for every link and sensor of the robot
- Each sensor publishes data in its own coordinate frame

# Common Setup for Mobile Robots

- odom -> base\_link provided by robot odometry
- map -> odom provided by localization method (e.g. *amcl* in ROS)
  - How to transform odom link in map frame so that base\_link is "correct"?

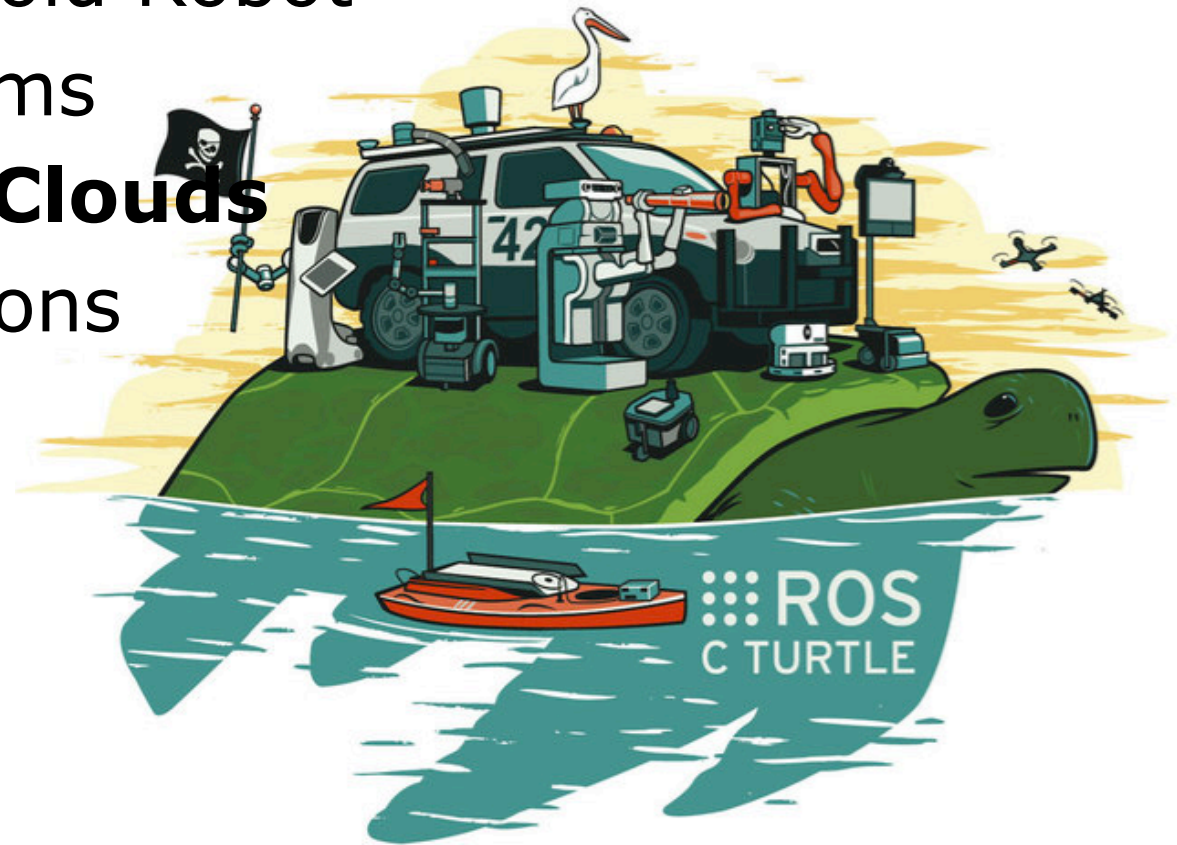


# Sources of Transformations

- URDF file defines robot “body layout”
  - Joints, links, sensor positions, visualization
- tf for all robot links automatically published
- Static transforms published from command line
- In your own node (e.g. localization)

# Today's Lecture

- ROS – Robot Operating System
- PR2 – Humanoid Robot
- TF – Transforms
- **PCL – Point Clouds**
- ROS Applications



# What are Point Clouds



- Point Cloud = a “cloud” (i.e., collection) of  $nD$  points (usually  $n = 3$ )
- $\mathbf{p}_i = \{x_i, y_i, z_i\} \longrightarrow \mathcal{P} = \{\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_i, \dots, \mathbf{p}_n\}$
- used to represent 3D information about the world

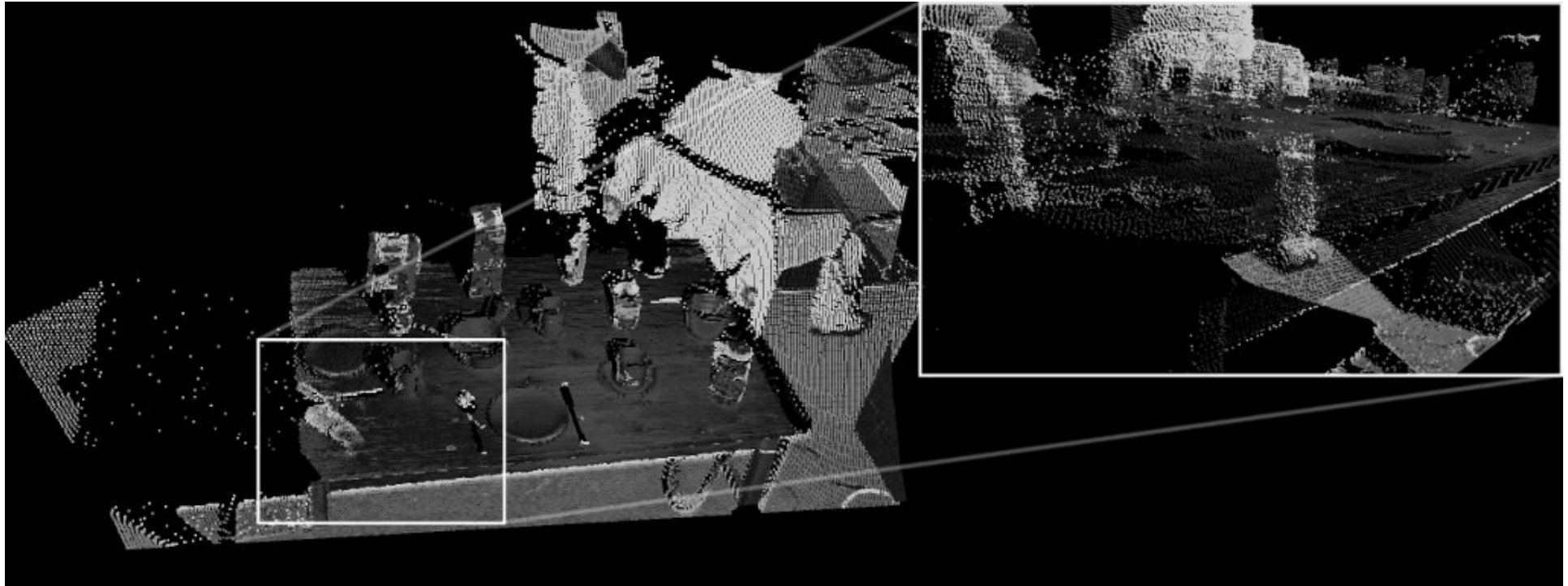


# What are Point Clouds

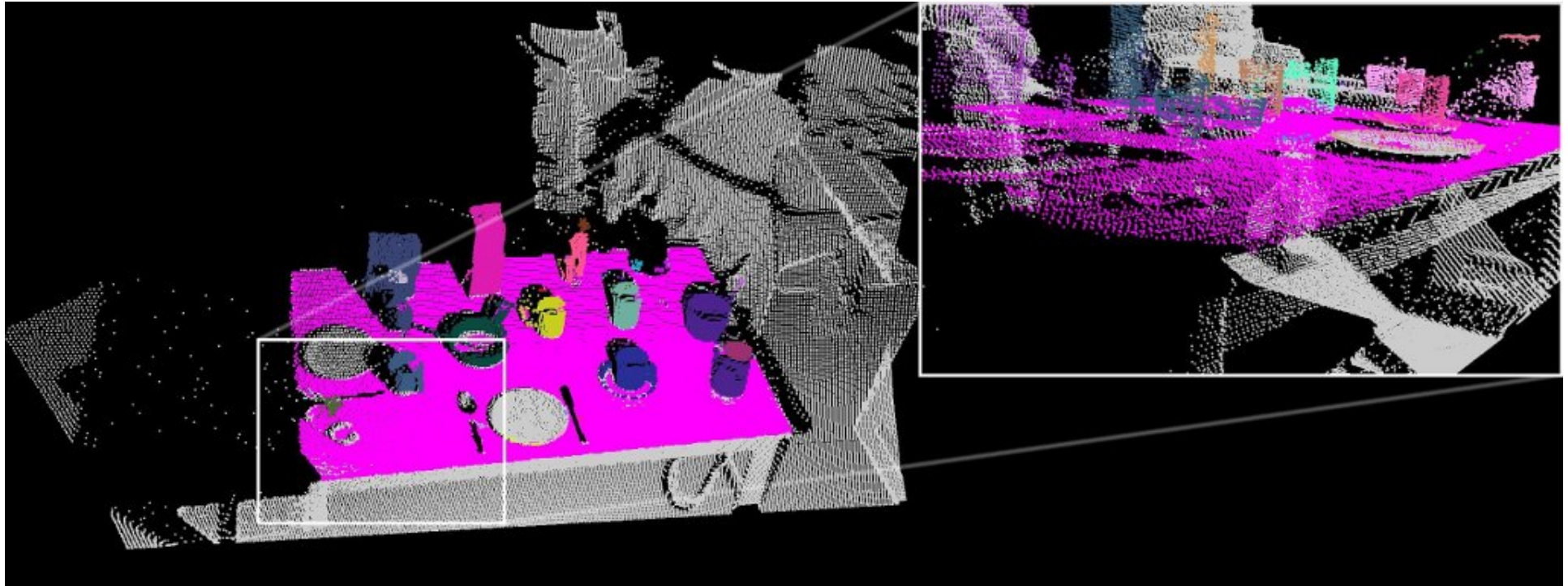


- besides XYZ data, each point  $p$  can hold additional information
- examples include: RGB colors, intensity values, distances, segmentation results, etc...

# What are Point Clouds



# What are Point Clouds



# Where do they come from?

- Laser scans (high quality)
- Stereo cameras (passive & fast but dependent on texture)
- Time of flight cameras (fast but not as accurate/robust)
- Kinect-Style Sensors
- Simulation

■ ...



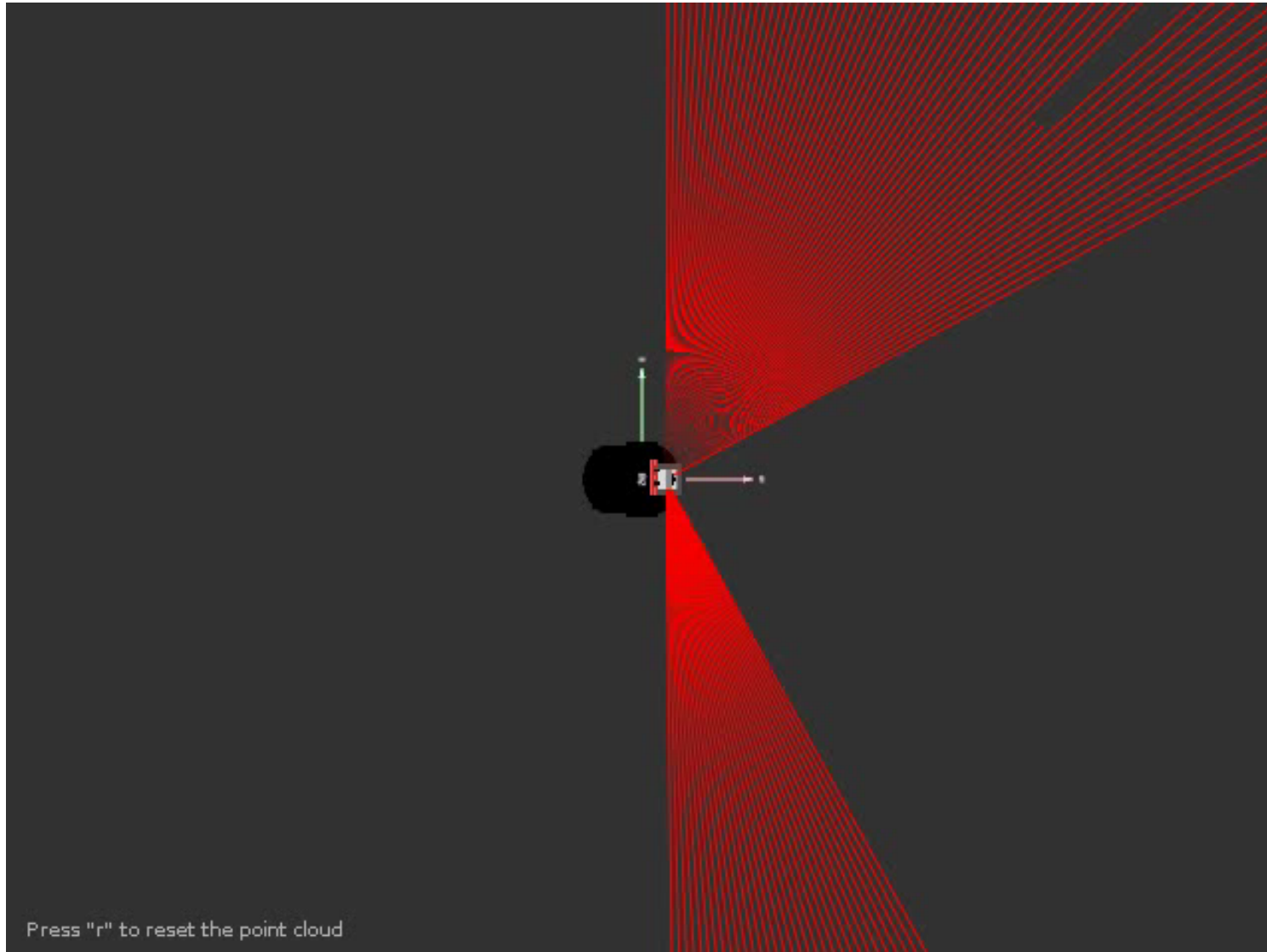
# Tilting Laser Scanner



Current laser beams and resulting point cloud



# Tilting Scanner on Moving Robot

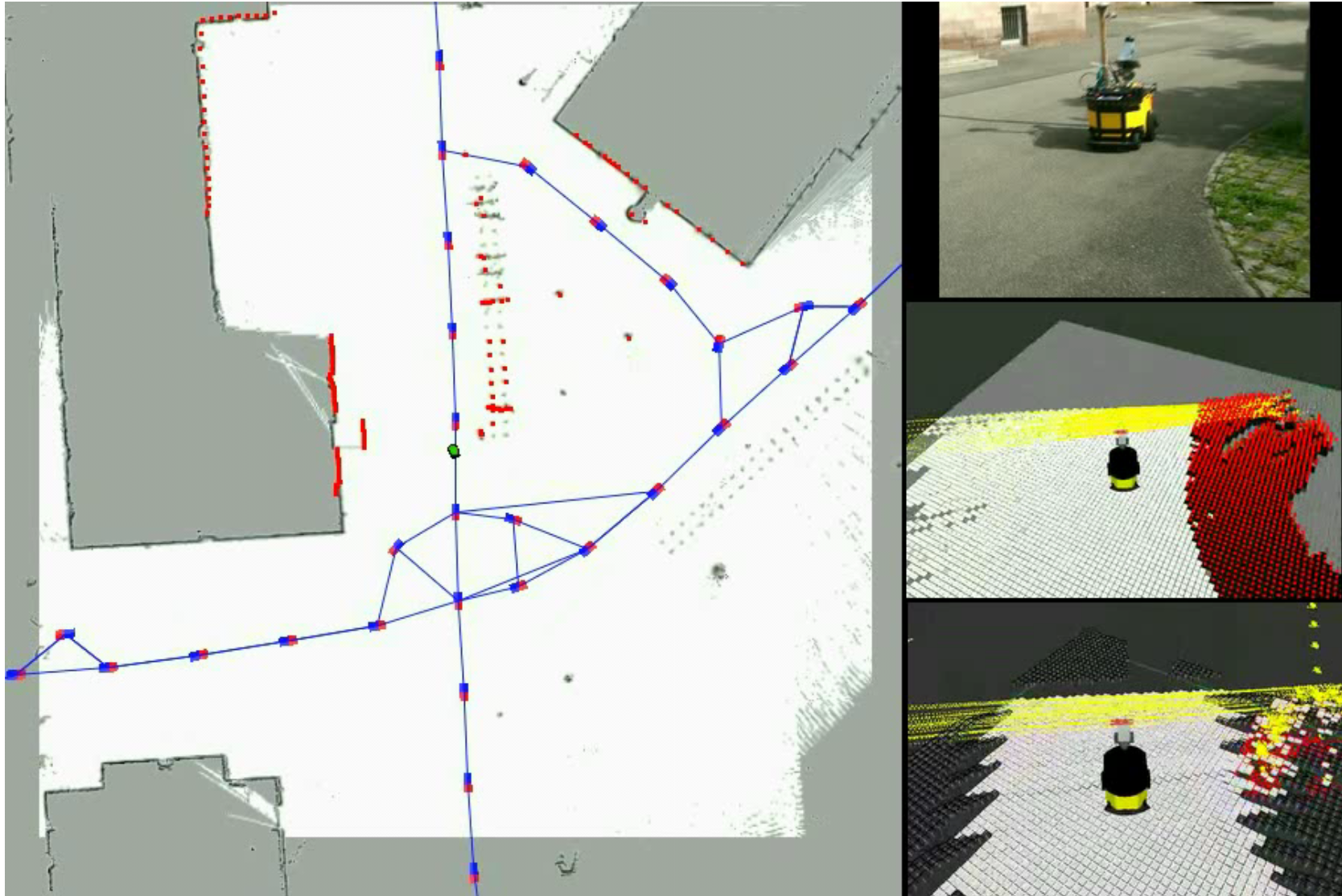


# For what!?

- Spatial information of the environment has many important applications
  - Navigation / Obstacle avoidance
  - Object recognition
  - Grasping
  - ...

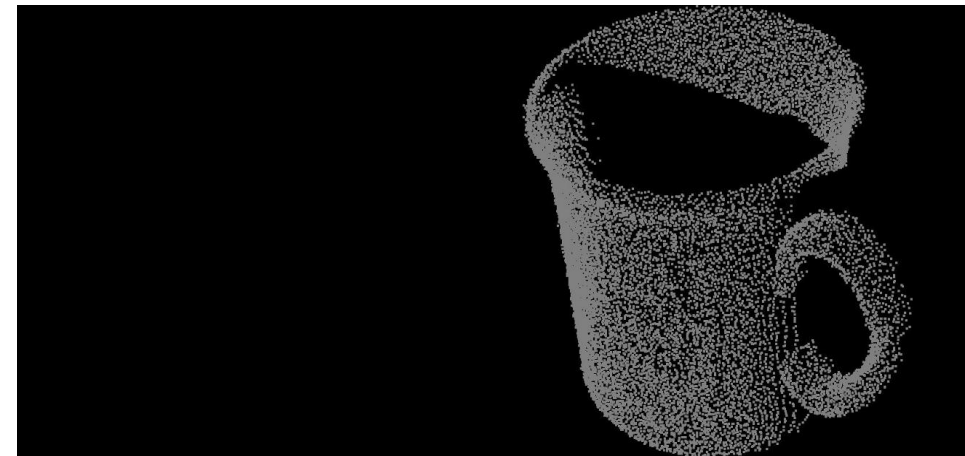
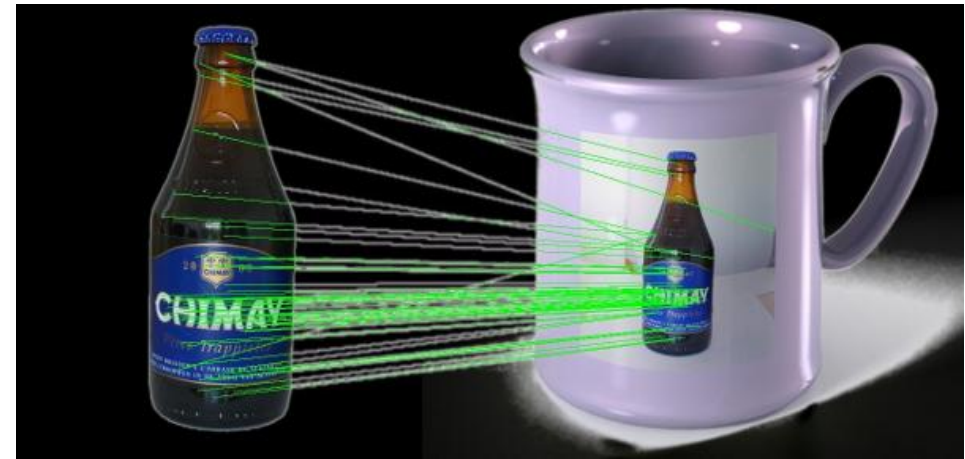
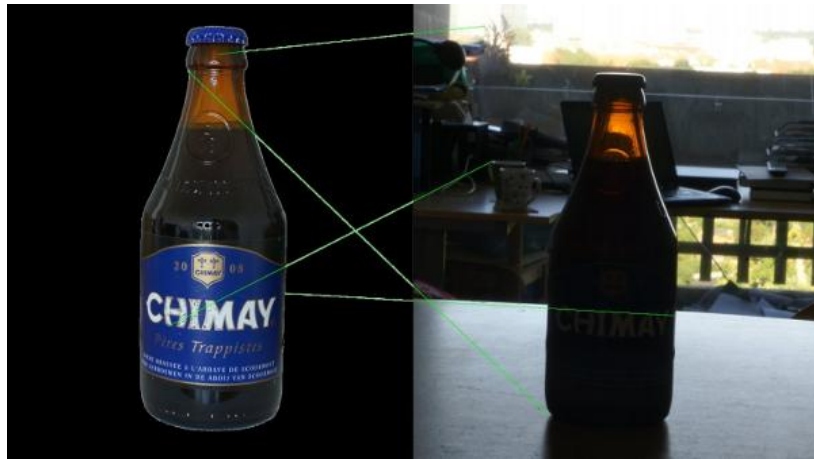


# Obstacle & Terrain Detection



# Object Recognition

Point Clouds can complement and supersede images when they are ambiguous.



# What is PCL?

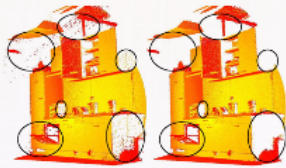
## PCL

- is a 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
- will be made independent from ROS in one of the next releases

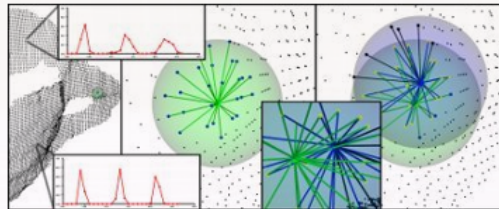


# PCL Modules

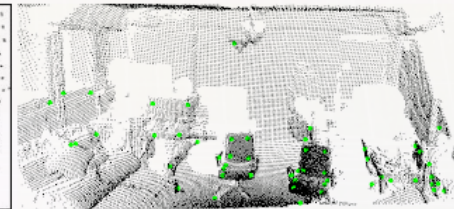
**filters**



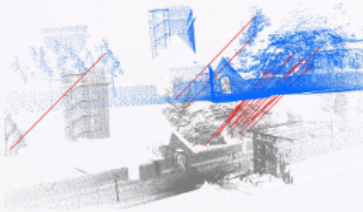
**features**



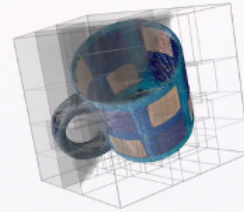
**keypoints**



**registration**



**kdtree**



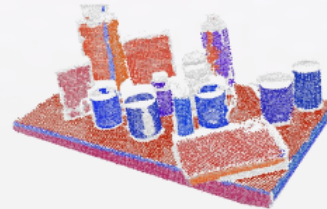
**octree**



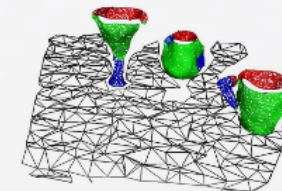
**segmentation**



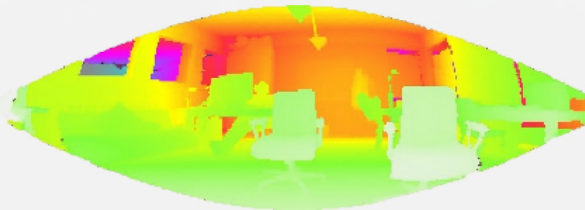
**sample\_consensus**



**surface**



**range\_image**



**io**



**visualization**



# Data Representation

- PointCloud class (templated over the point type):

```
template <typename PointT>  
class PointCloud;
```

- Important members:

```
std::vector<PointT> points; // the data  
uint32_t width, height; // scan structure?
```

# Point Types

- Examples of PointT:

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

or

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

See `pcl/include/pcl/point_types.h` for more examples.

# PointCloud2 Message

- 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 (for loading, saving or sending as a ROS message)

Conversion between the two is easy:

`pcl::fromROSMsg` and `pcl::toROSMsg`



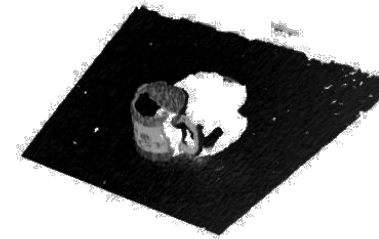
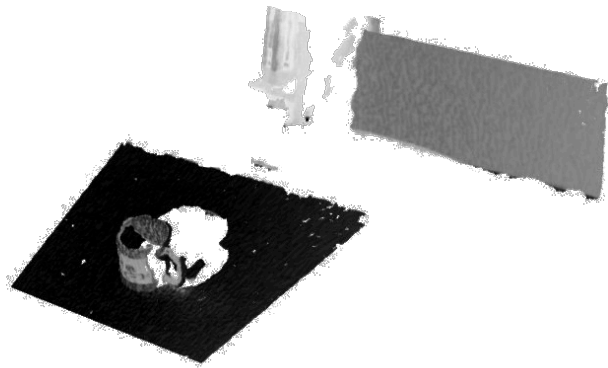
# Basic Interface

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

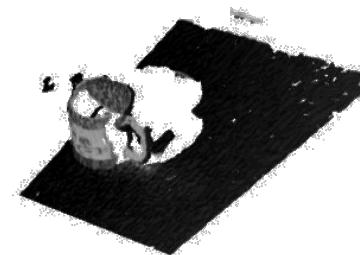
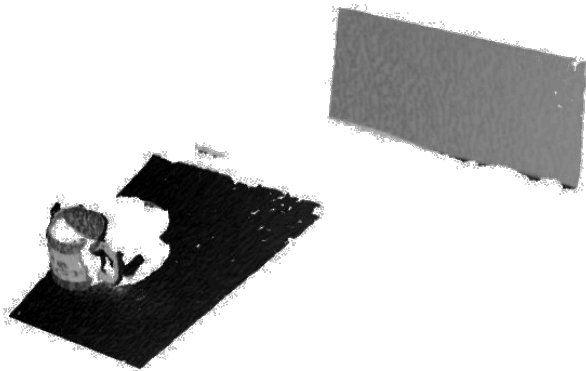
- Create the object
- use `setInputCloud` to give the input
- set some parameters
- call `compute` to get the output

# Filter Example 1

```
pcl::PassThrough<T> p;  
p.setInputCloud (data);  
p.FilterLimits (0.0, 0.5);  
p.SetFilterFieldName ("z");
```

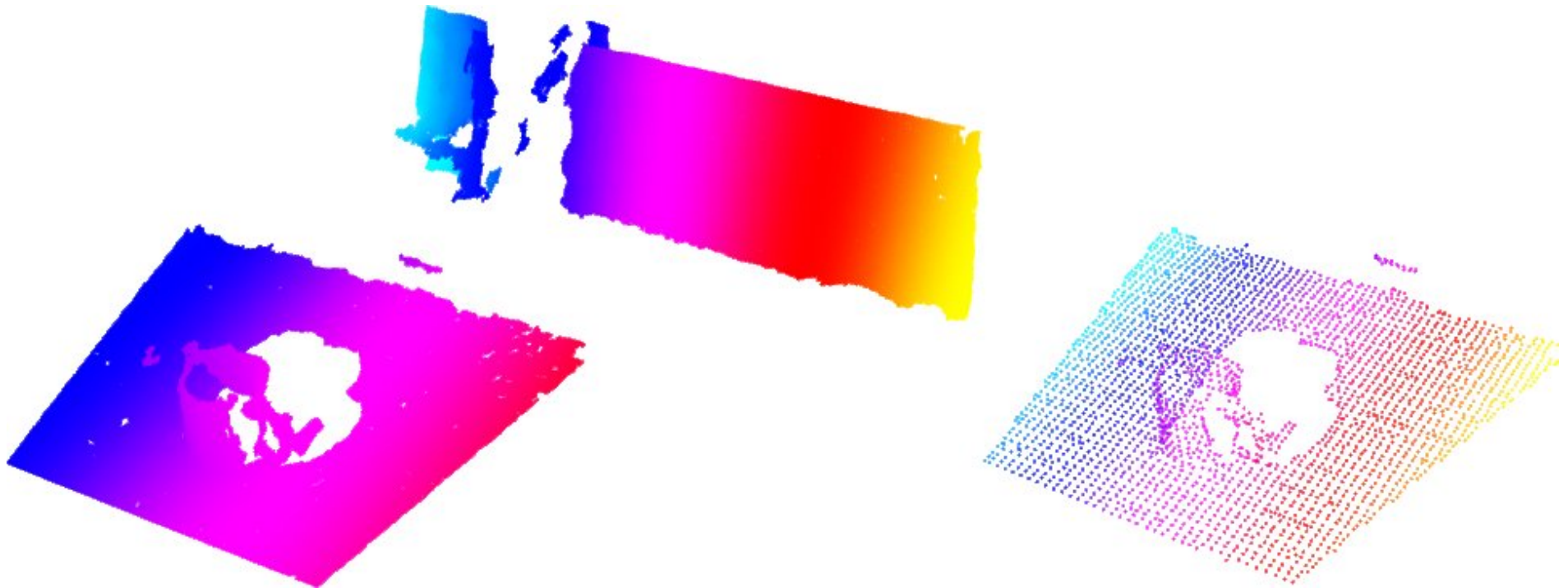


```
filter_field_name = "x"; | filter_field_name = "xz";
```



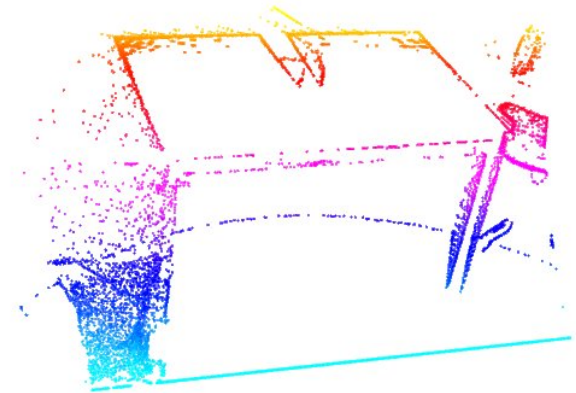
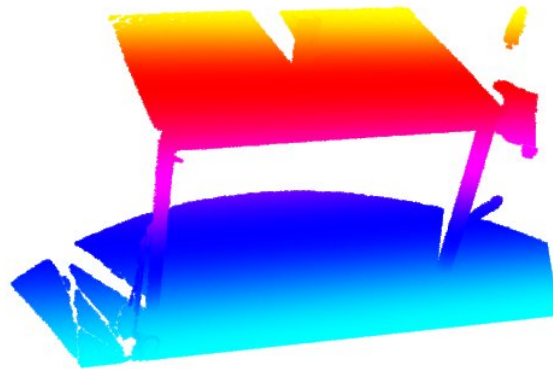
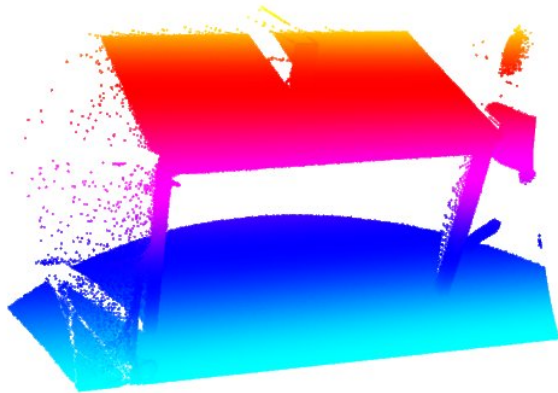
# Filter Example 2

```
pcl::VoxelGrid<T> p;  
p.setInputCloud (data);  
p.FilterLimits (0.0, 0.5);  
p.SetFilterFieldName ("z");  
p.setLeafSize (0.01, 0.01, 0.01);
```



# Filter Example 3

```
pcl::StatisticalOutlierRemoval<T> p;  
p.setInputCloud (data);  
p.setMeanK (50);  
p.setStddevMulThresh (1.0);
```



# Features Example 1

```
pcl::NormalEstimation<T> p;  
p.setInputCloud (data);  
p.SetRadiusSearch (0.01);
```





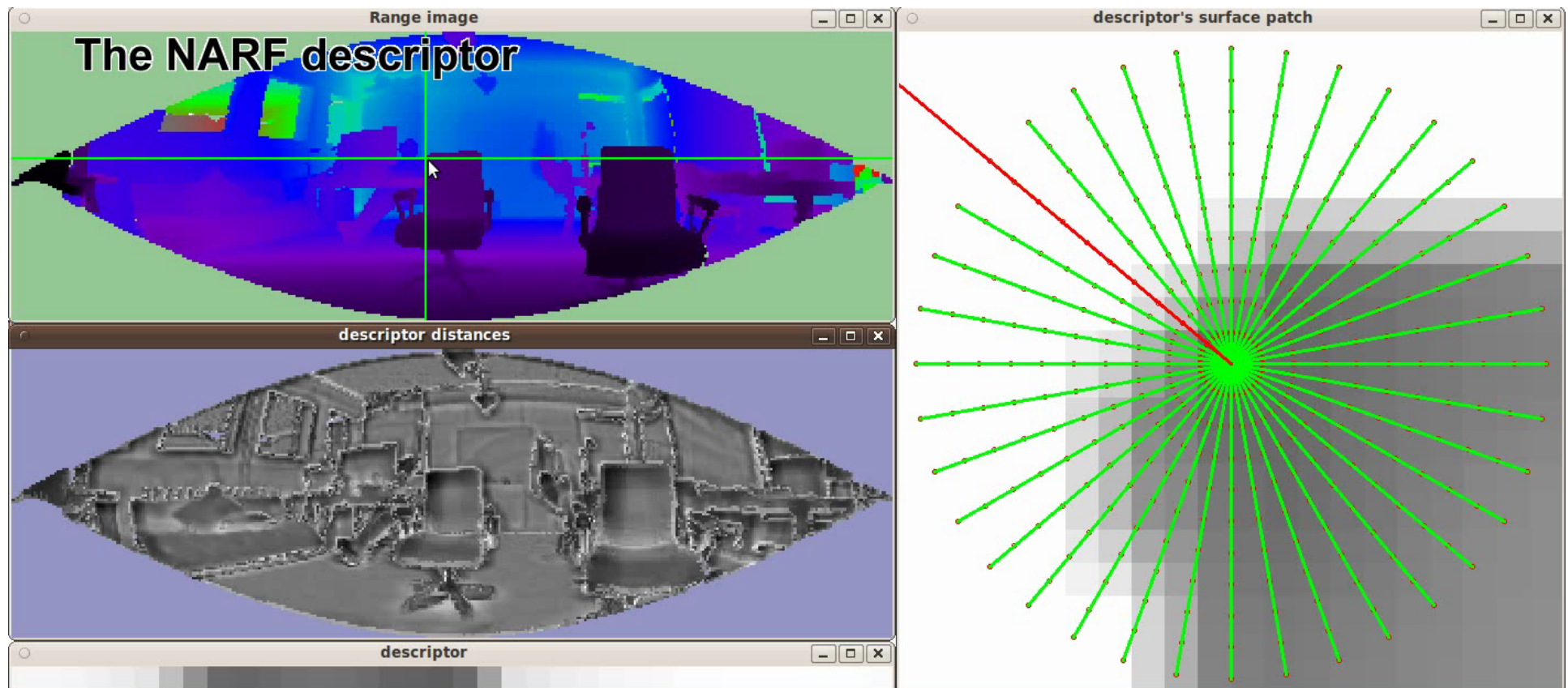
# Features Example 2

```
pcl::BoundaryEstimation<T,N> p;  
p.setInputCloud (data);  
p.setInputNormals (normals);  
p.SetRadiusSearch (0.01);
```



# Features Example 3

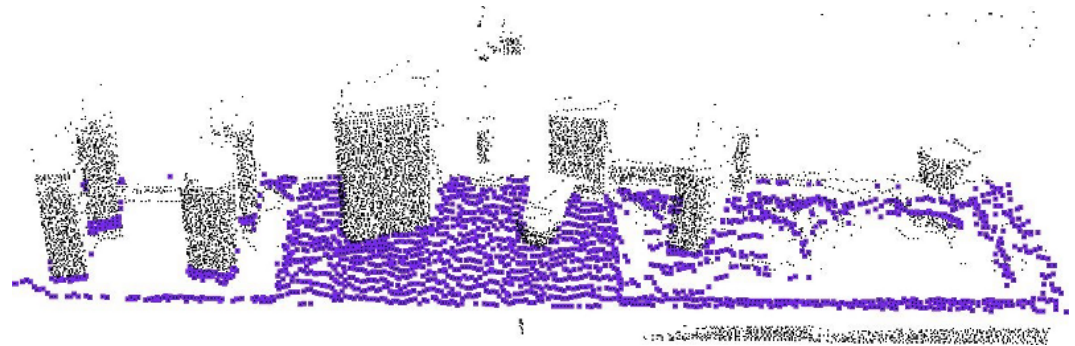
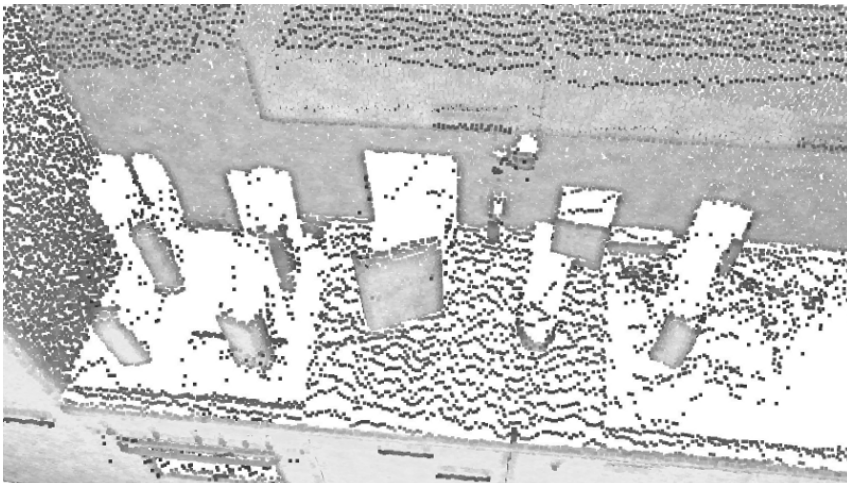
```
NarfDescriptor narf_descriptor(&range_image);  
narf_descriptor.getParameters().support_size = 0.3;  
narf_descriptor.getParameters().rotation_invariant = false;  
PointCloud<Narf36> narf_descriptors;  
narf_descriptor.compute(narf_descriptors);
```





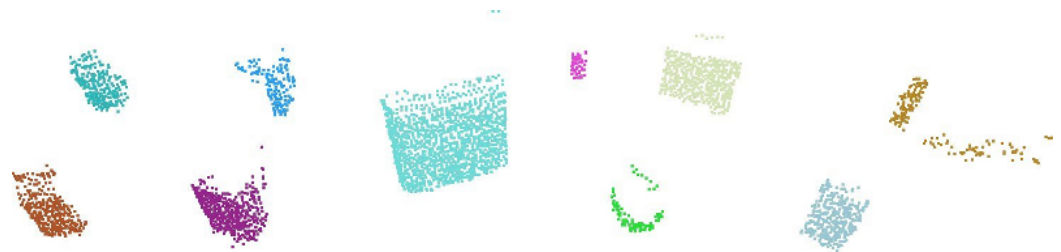
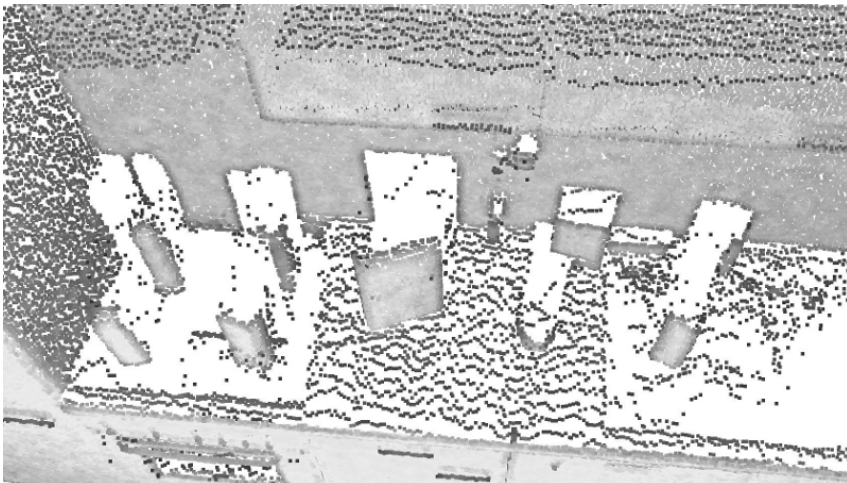
# Segmentation Example 1

```
pcl::SACSegmentation<T> p;  
p.setInputCloud (data);  
p.setModelType (pcl::SACMODEL_PLANE);  
p.setMethodType (pcl::SAC_RANSAC);  
p.setDistanceThreshold (0.01);
```



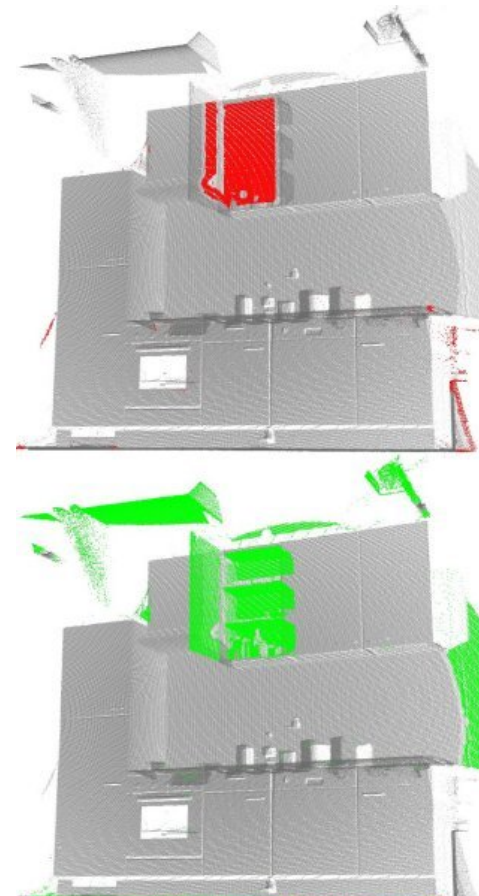
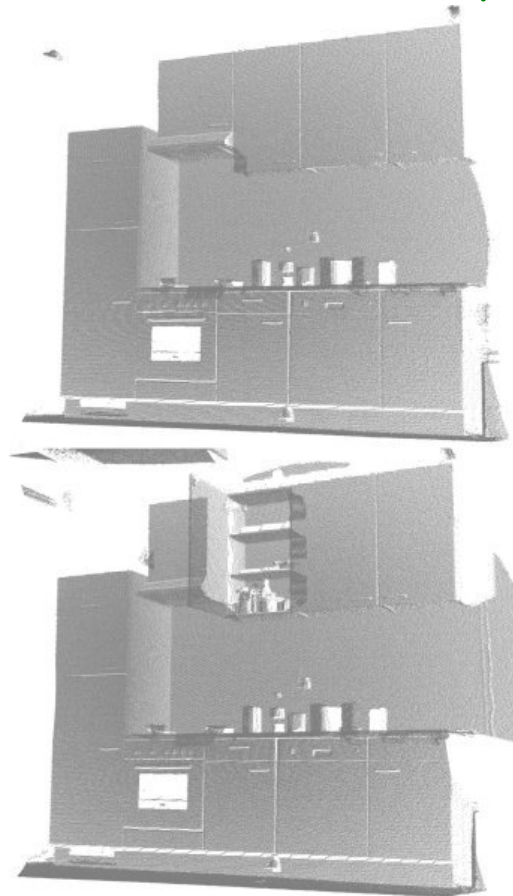
# Segmentation Example 2

```
pcl::EuclideanClusterExtraction<T> p;  
p.setInputCloud (data);  
p.setClusterTolerance (0.05);  
p.setMinClusterSize (1);
```



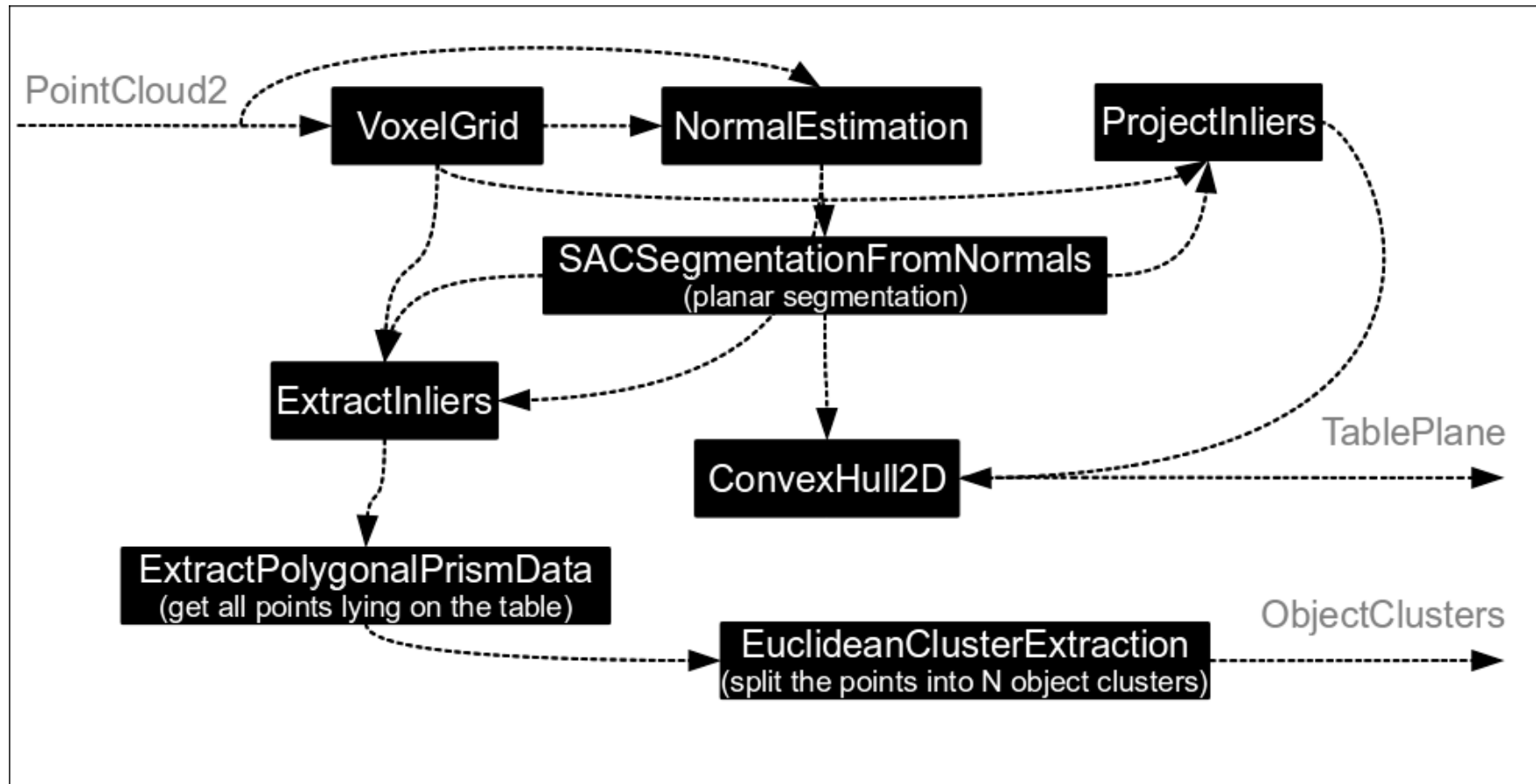
# Segmentation Example 3

```
pcl::SegmentDifferences<T> p;  
p.setInputCloud (source);  
p.setTargetCloud (target);  
p.setDistanceThreshold (0.001);
```



# Higher level example

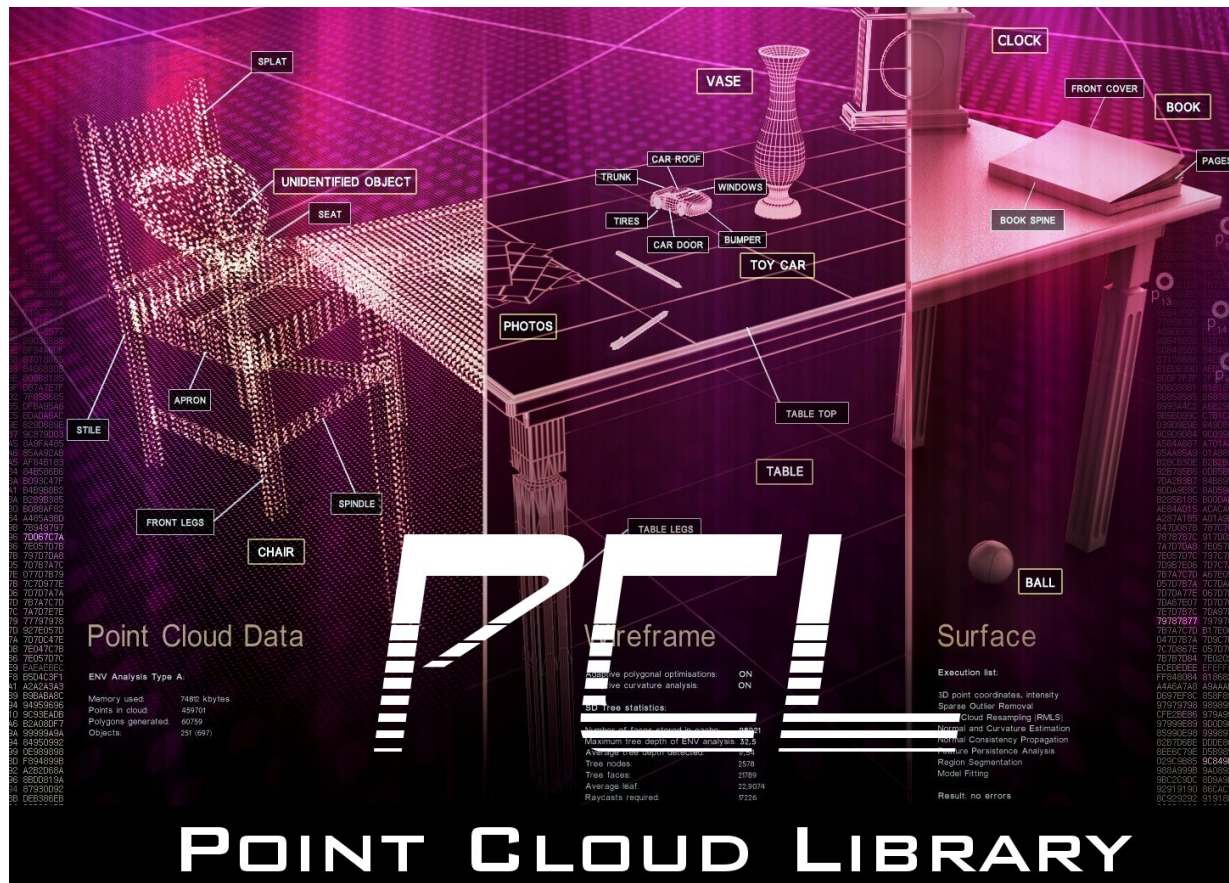
How to extract a table plane and the objects on it?





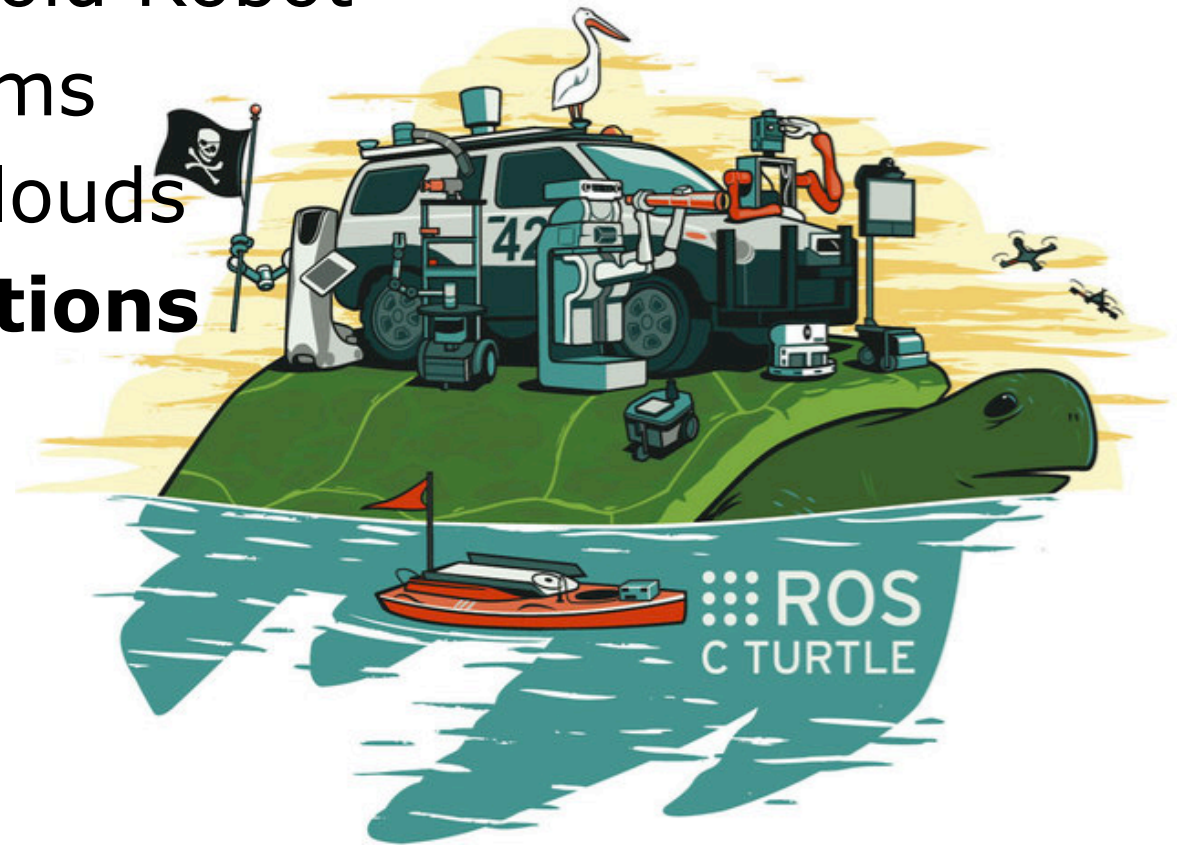
# More details

- See <http://www.ros.org/wiki/pcl> and <http://www.pointclouds.org>



# Today's Lecture

- ROS – Robot Operating System
- PR2 – Humanoid Robot
- TF – Transforms
- PCL – Point Clouds
- **ROS Applications**



# SLAM on Dense Colored Clouds

- Our goal: SLAM systems for RGB-D sensors
  - 3D environment representation
  - 6DOF trajectory estimation
  - Online Operation
  - No dependency on further sensors (e.g. Odometry, Laser)

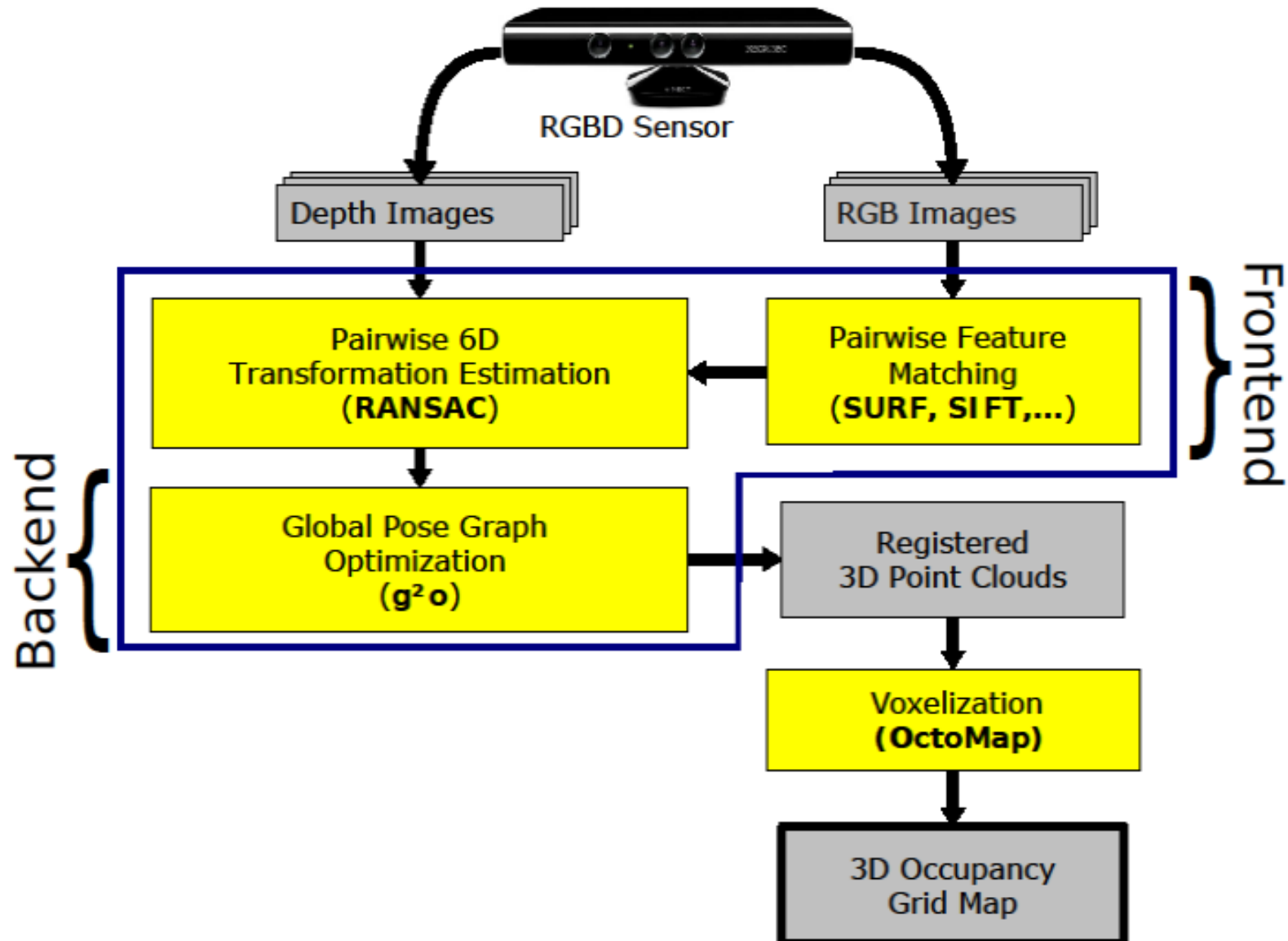


- New kinect-style sensors provide:
  - Dense RGB-D data
  - High framerate (30 Hz)
  - Low weight (440g)
  - Low-cost ( $\sim 100\text{€}$ )





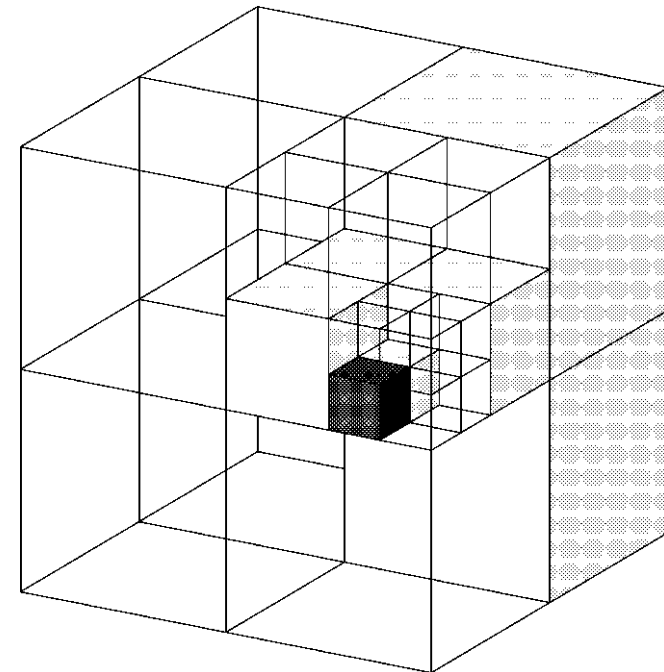
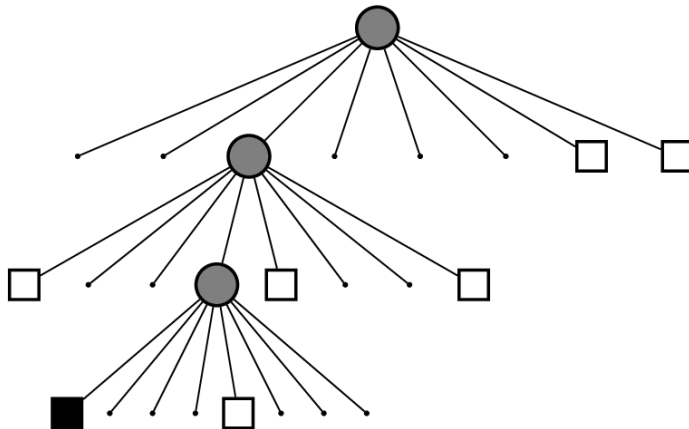
# Approach Overview



# Map Representation

- Point clouds are inefficient for applications such as collision detection and navigation
- We use the OctoMap framework
  - Octree-based data structure
  - Recursive subdivision of space into octants
  - Volumes allocated as needed

➔ **Smart 3D grid**



# OctoMap Framework

## Probabilistic Update of Voxels

- Full 3D model
- Probabilistic
- Multi-resolution
- Memory efficient



- Available from [octomap.sf.net](http://octomap.sf.net)

# Octomap with Per-Voxel Colors

Probabilistic 3D mapping using  
OctoMap and RGBDSLAM

Kai M. Wurm, Felix Endres  
Autonomous Intelligent Systems Lab  
University of Freiburg, Germany



# Kinematic Models



## A Probabilistic Framework for Learning Kinematic Models of Articulated Objects

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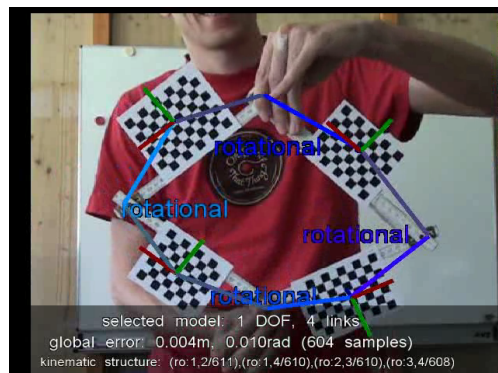
Jürgen Sturm, Cyrill Stachniss,

Wolfram Burgard

University of Freiburg, Germany

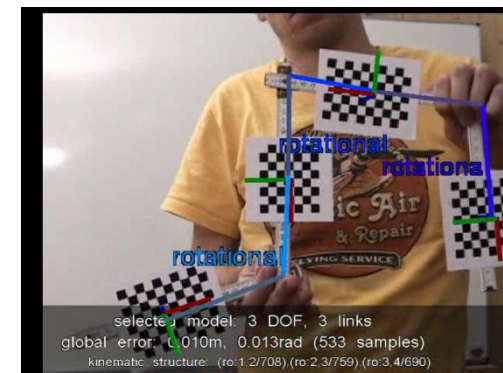
# Kinematic Models

- Fit and select appropriate models for observed motion trajectories of articulated objects
- Estimate structure and DOFs of articulated objects with  $n > 2$  object parts



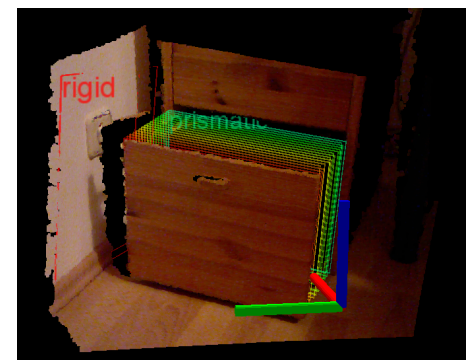
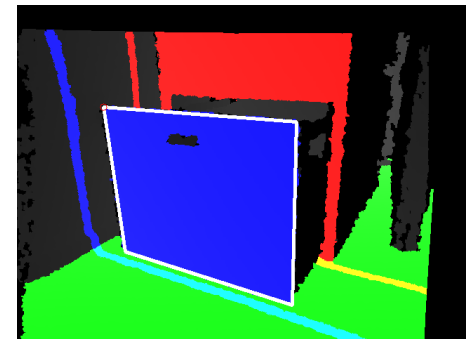
← 1-DOF closed chain

3-DOF open chain →



# Kinematic Models

- Marker-less perception:  
Detect and track articulated objects in depth images and learn their kinematic models
- Approach:
  - Segment planes
  - Fit pose candidates and filter
  - Learn kinematic models







# A Probabilistic Framework for Learning Kinematic Models of Articulated Objects

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Wolfram Burgard

University of Freiburg, Germany

# Thank You...

...for your attention

