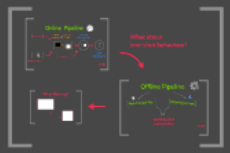
Researcher of the Week: superquadric-model module

Goal

Theory

Code

Pipeline



Dependencies

Connections

Giulia Vezzani - 13/05/2016

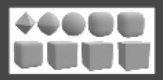
Goal

Object detection and modeling

iCub **looks** at the object

the object is **detected**

a **3D model** is computed



the model is **visualized**

Requests:

1. Detection ✔
2. Modeling ?
3. Visualization ?
4. ... and everything robust and in real-time !

Object detection and modeling

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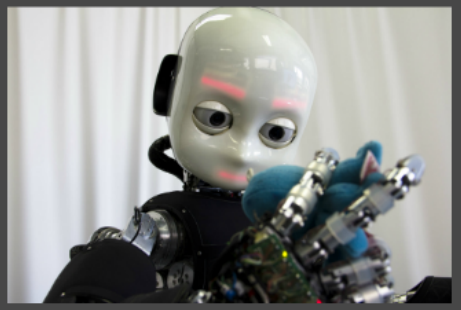

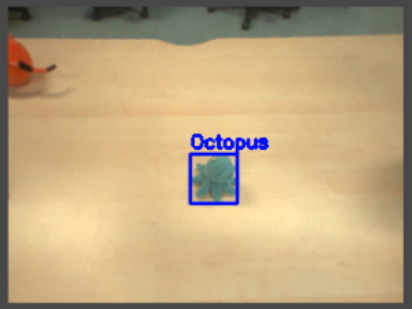

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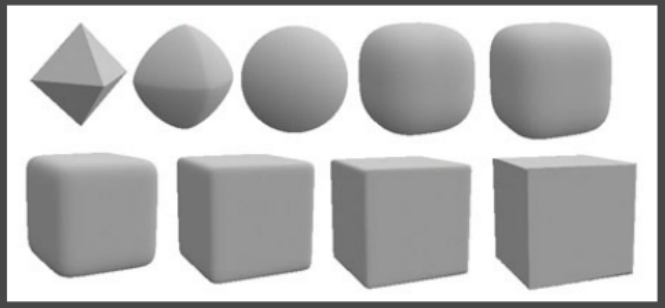

the model is **visualized**

Requests:

1. Detection ✓
2. Modeling ?
3. Visualization ?
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Theory

Superquadric fuctions

inside-outside function:

- F > 1: point outside
- F < 1: point inside
- F = 1: point on surface

$$F(x, y, z, \lambda) = \left(\left(\frac{x}{\lambda_1} \right)^{\frac{2}{\lambda_5}} + \left(\frac{y}{\lambda_2} \right)^{\frac{2}{\lambda_5}} \right)^{\frac{\lambda_5}{\lambda_4}} + \left(\frac{z}{\lambda_3} \right)^{\frac{2}{\lambda_4}}$$

[5 parameters for shape] + [6 parameters for pose]

Optimization problem

$$\min_{\lambda} \sum_{i=0}^{N} \left(\sqrt{\lambda_1 \lambda_2 \lambda_3} F^{\lambda_4}(x_i, y_i, z_i, \lambda) - 1 \right)^2$$

- shape indipendence
- minimum volume

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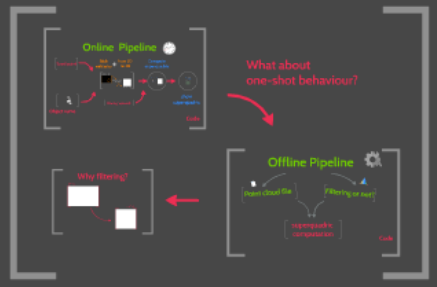


Dependencies

Dependencies:

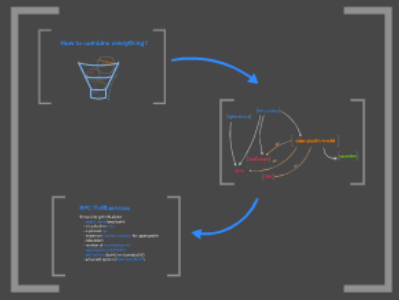
- YARP
- iCub
- OpenCV
- IPOPT: Interior Point OPTimizer

External application

- IOL: (Interactive Object Learning)
  - lbpExtract (Local Binary Pattern Extractor)
  - SFM (Structure from Motion)
  - OPC (Object Property Collector)

Code

Connections



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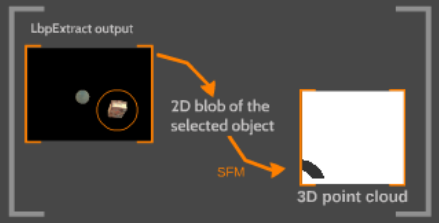
Code

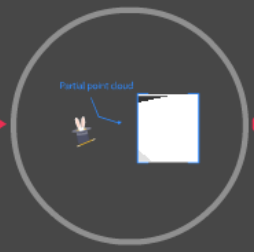
Online Pipeline

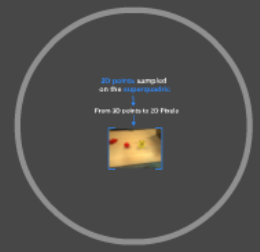
Seed point

blob extractor + from 2D to 3D



Compute superquadric



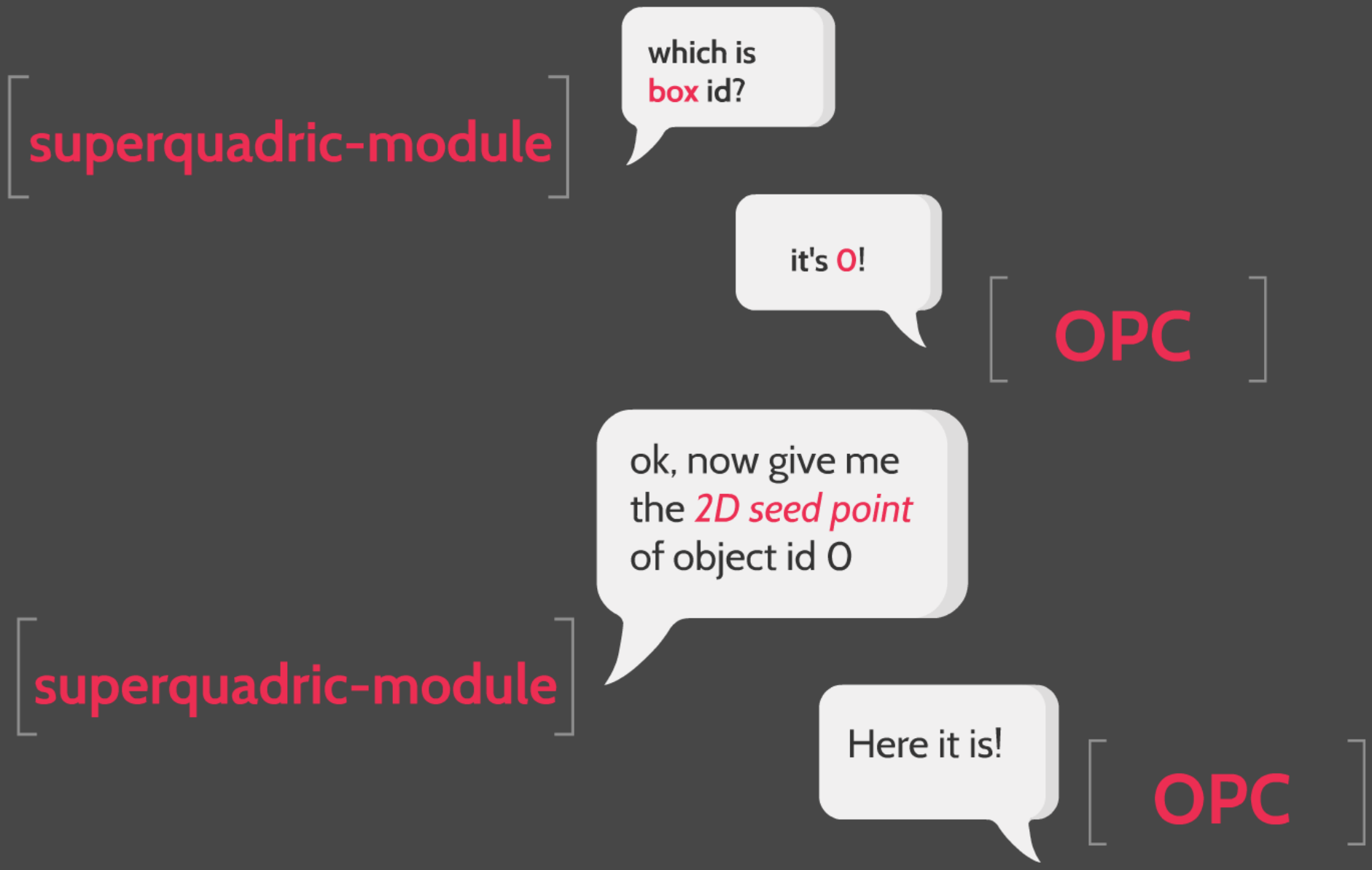


show superquadric

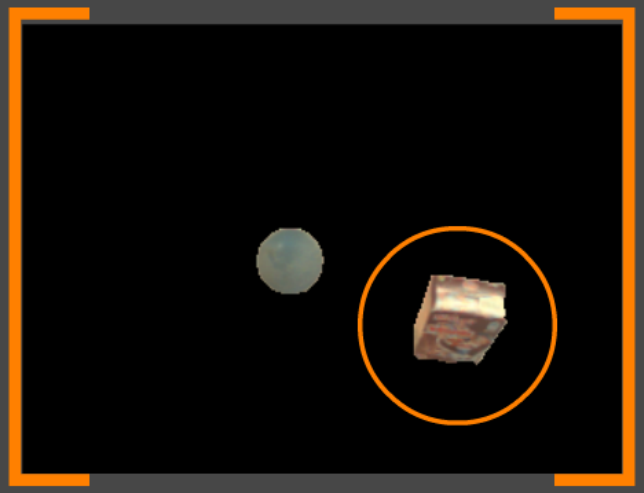
Object name

Filtering (optional)

Code

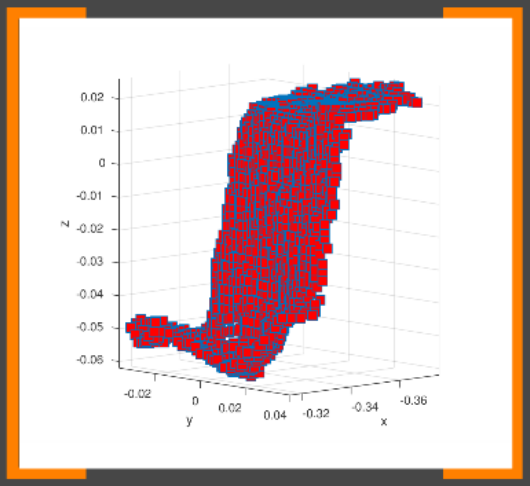

superquadric-module
which is box id?
it's 0!
OPC
ok, now give me the 2D seed point of object id 0
superquadric-module
Here it is!
OPC

LbpExtract output



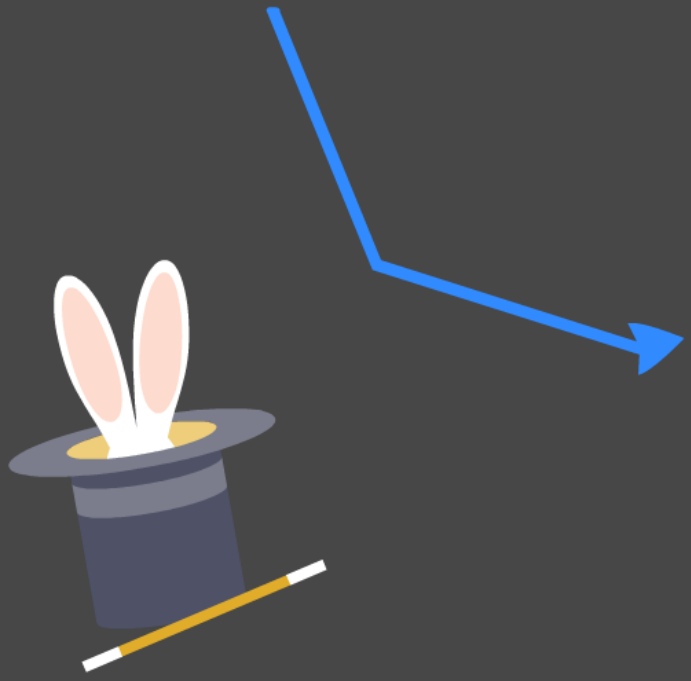
2D blob of the selected object

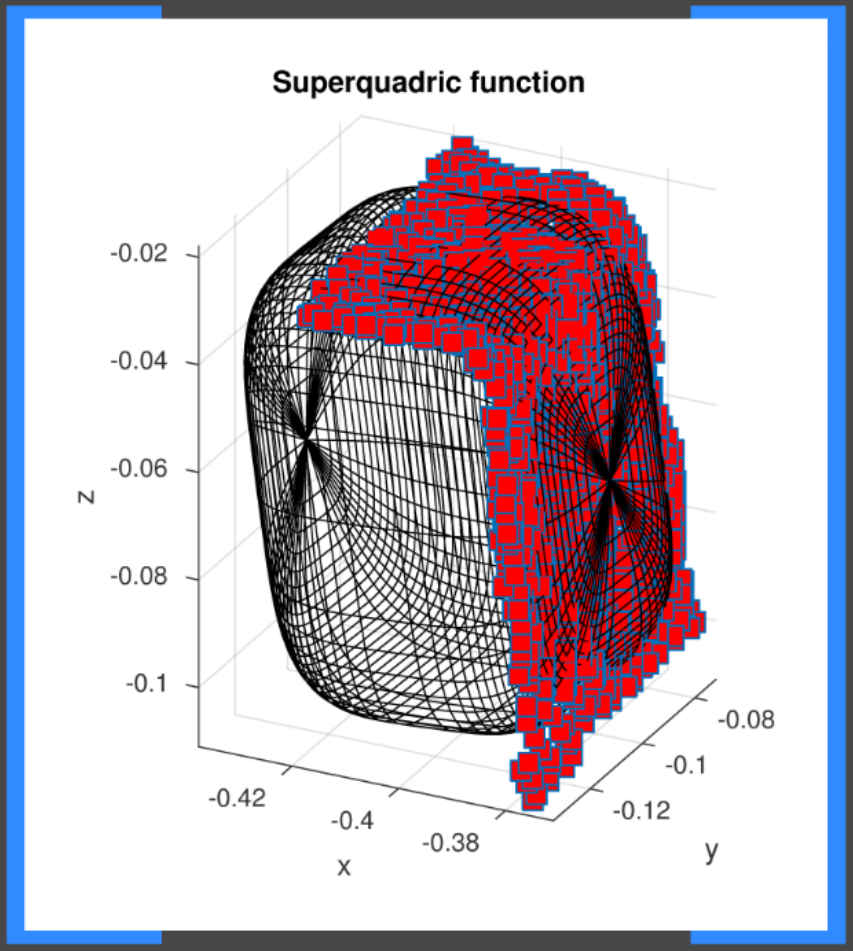
SFM



3D point cloud

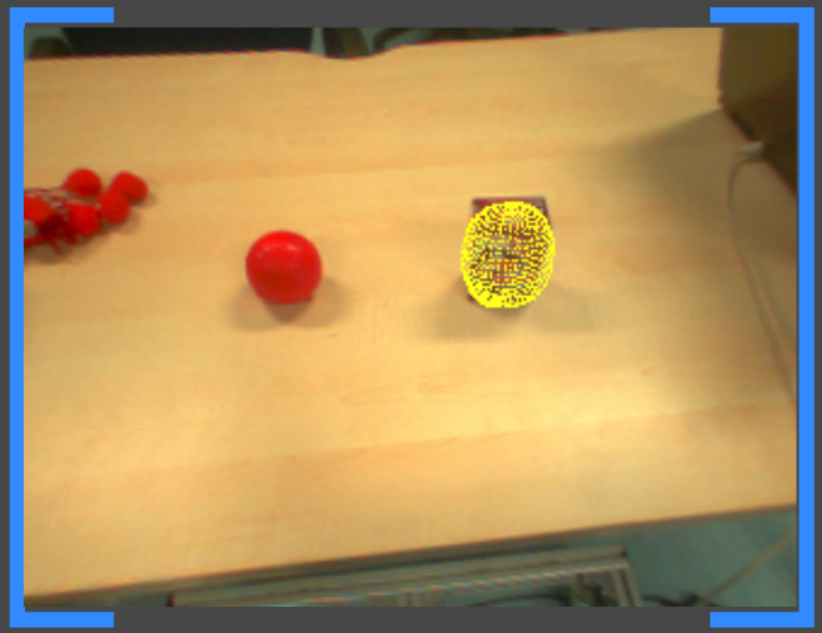
Partial point cloud





3D points sampled
on the superquadric

From 3D points to 2D Pixels



What about one-shot behaviour?

Offline Pipeline

Point cloud file

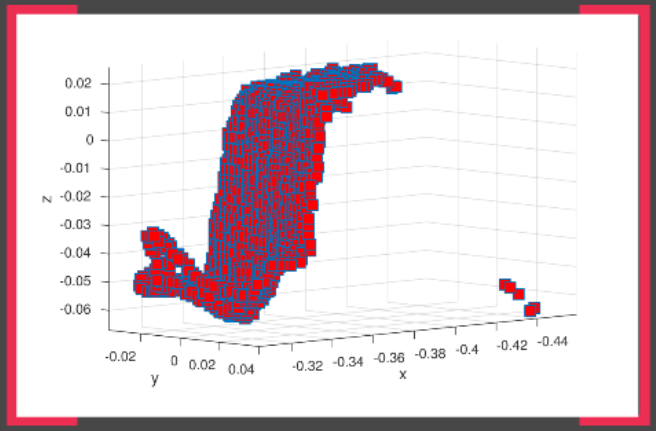
Filtering or not?

superquadric computation

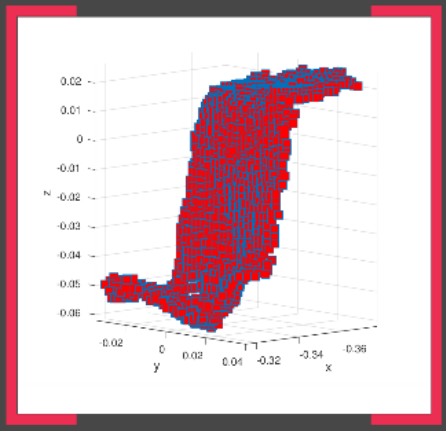
Code

Why filtering?

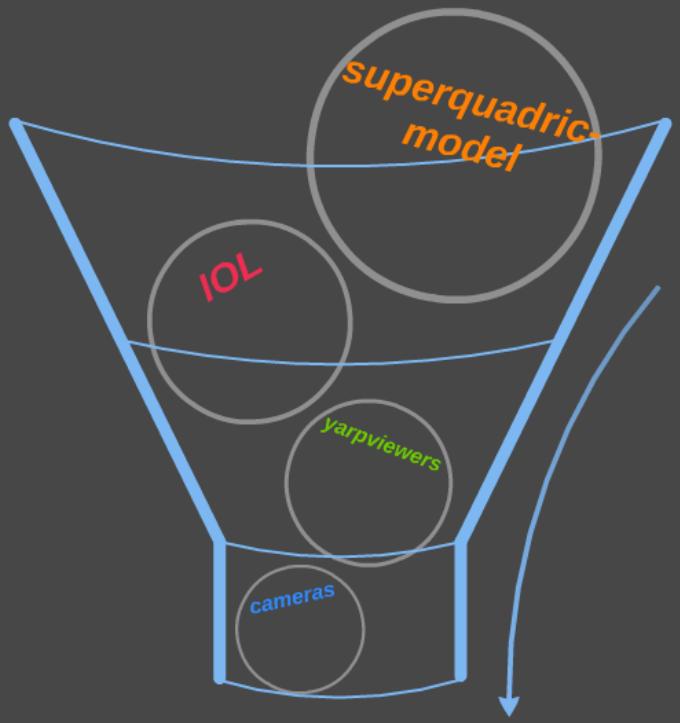
Before ...

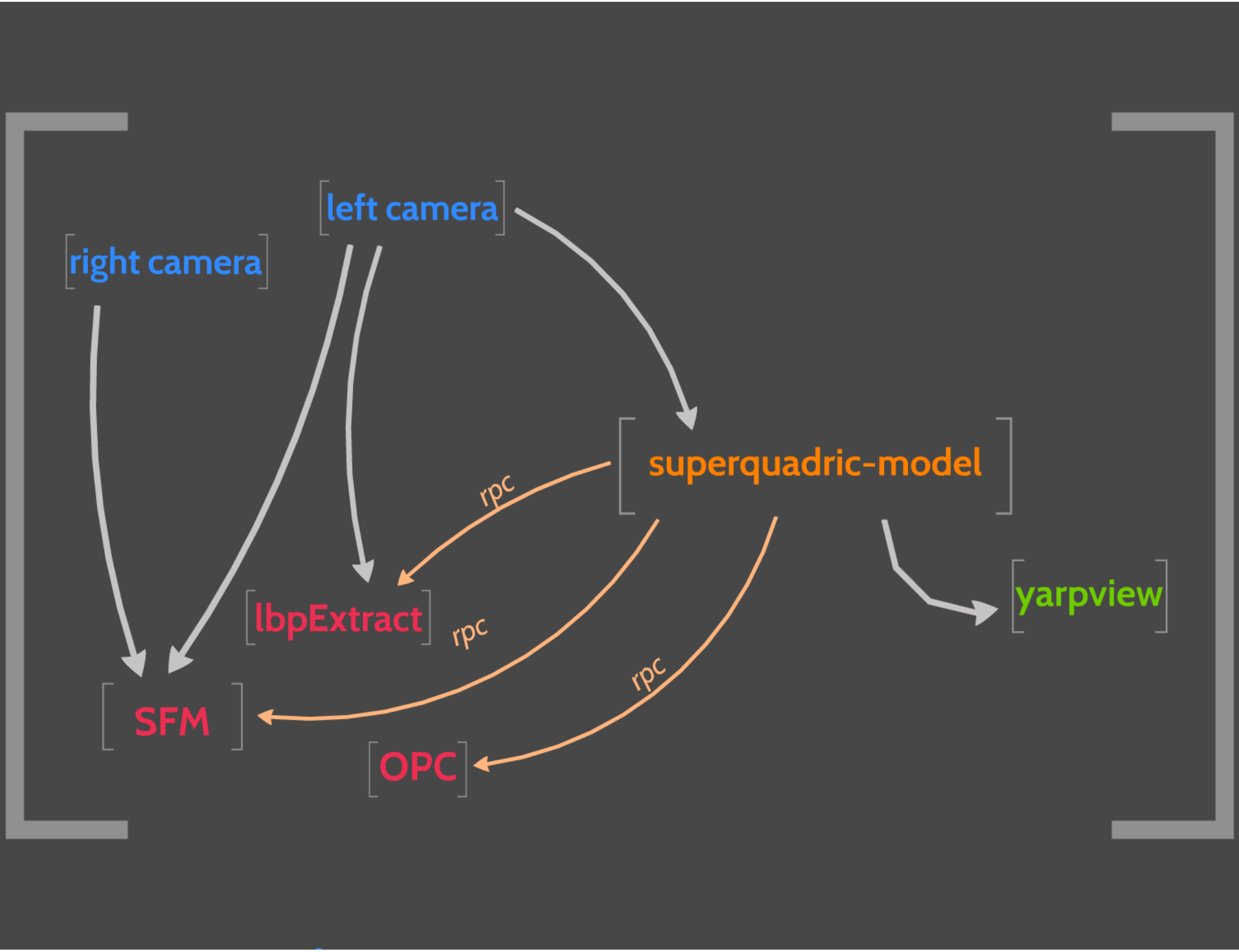


... after!



How to combine everything?




left camera
right camera
superquadric-model
rpc
lbpExtract
rpc
yarpview
rpc
SFM
OPC

RPC Thrift services

Set and/or get info about:

- object name/seed point
- visualization color
- exploited eye
- maximum number of points for superquadric calculation
- number of visualized points
- superquadric parameters
- plot options (points or superquadric)
- advanced options (filtering & IPOPT)

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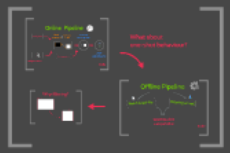


Dependencies

Connections

Giulia Vezzani - 13/05/2016

You can find
all the information about superquadric-model
module on the github repo:

https://github.com/·robotology/superquadric-model

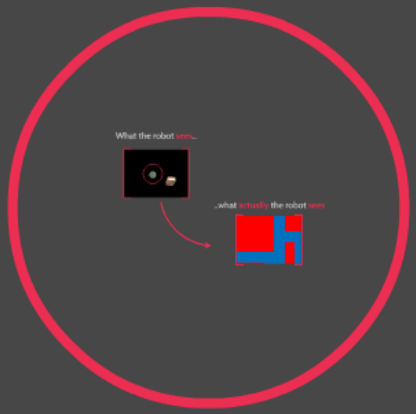
… What happens next?

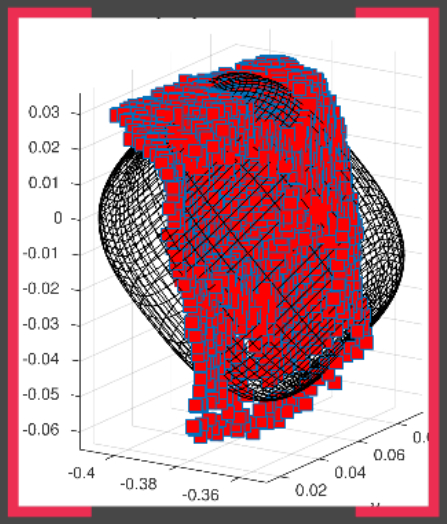
Noteworthy ...

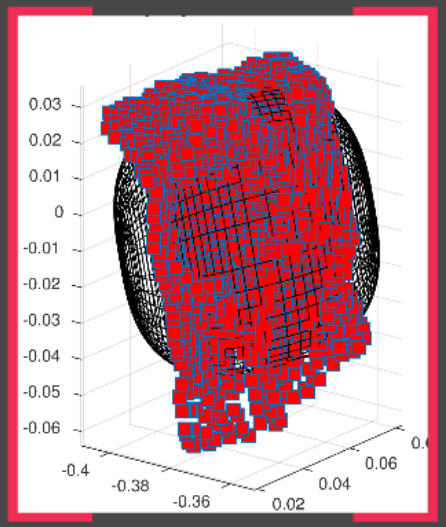
Wrong solution (at first glance)

Possible solutions

Noisy point clouds

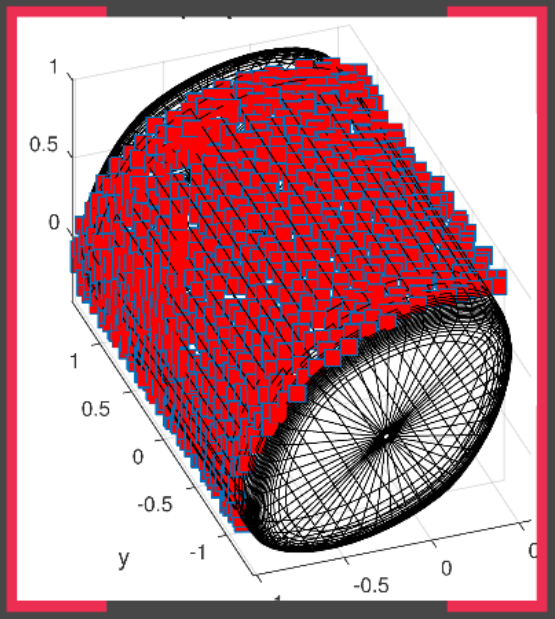




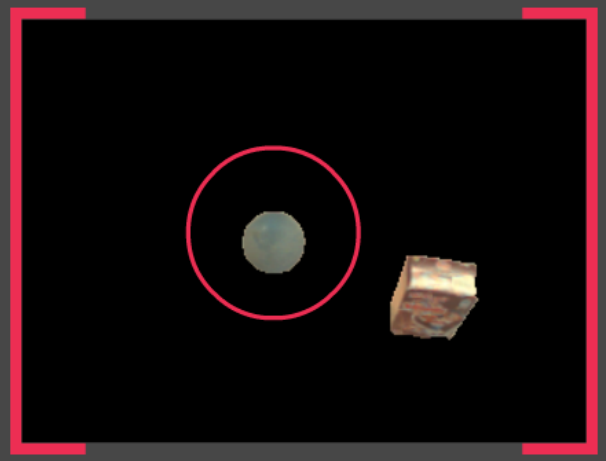


Same cost function value!

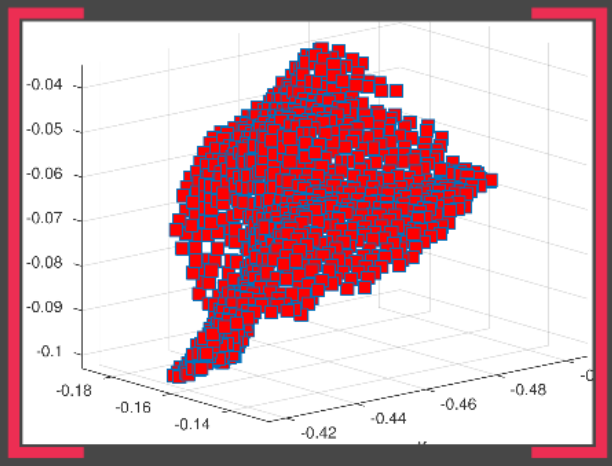
Partial non-noisy clouds:



What the robot sees..



..what actually the robot sees



Ideas

1. Merged point clouds (ICP)
2. Point cloud time refinement and filtering (Voxelization)
3. 2D surface overlapping
4. Different optimization problem formulation

$$\min_{\lambda} \forall(\lambda)$$

$$\sum_{i=0}^{N} (F(x_i, y_i, z_i, \lambda) - 1)^2 = 0$$

$$\min_{\lambda} \mathbb{V}(\lambda)$$

$$\sum_{i=0}^{N} \left(F(x_i, y_i, z_i, \lambda) - 1\right)^2 = 0$$

Future works

[coding & applications]

- Superquadric visualization ?

VTK/OpenGL

- Grasping application 

Good grasping pose 

Trajectory planning

Obstacle avoidance 

Thank you for your attention!

..any questions or comments?

Then, let's try our code!