

# Visual odometry for Corobots

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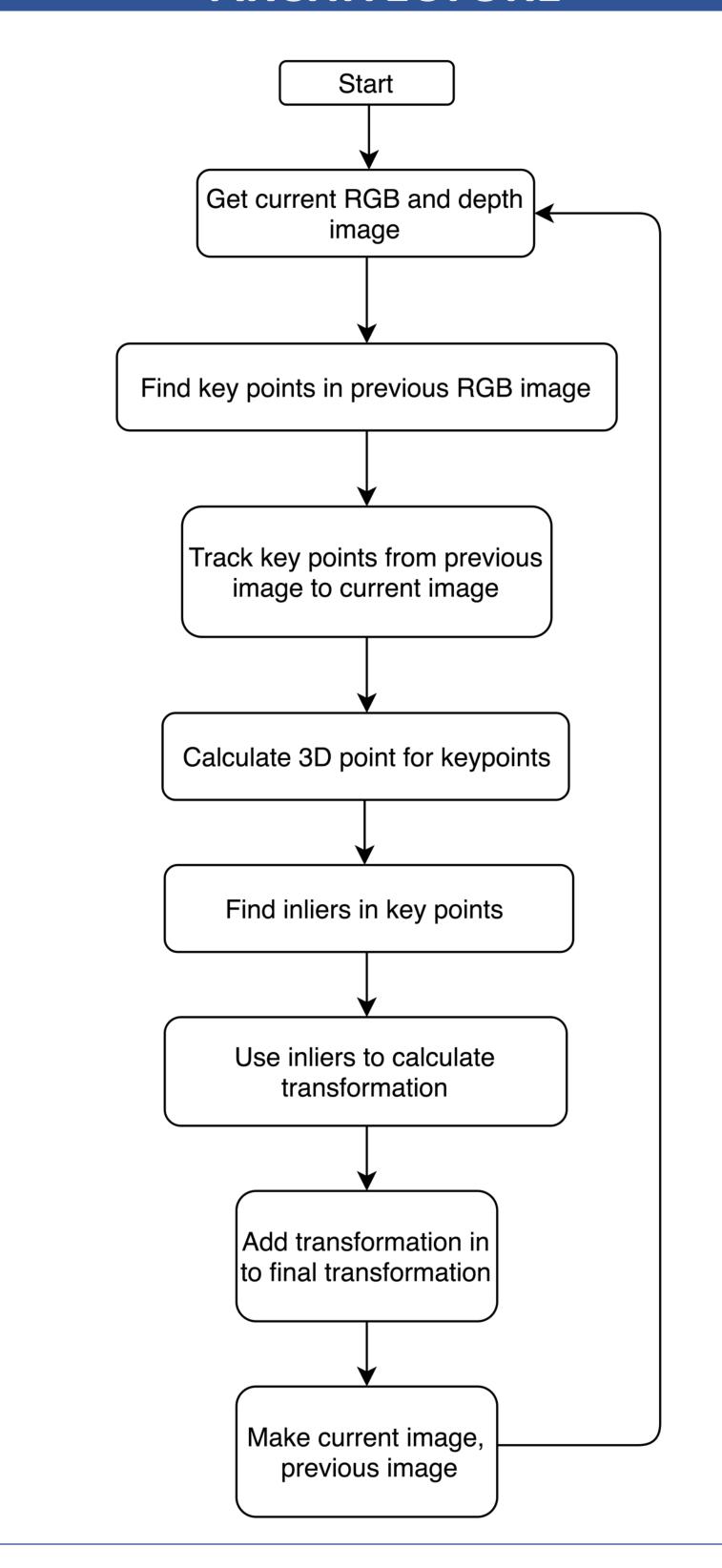
# INTRODUCATION

- The goal of this project is to use visual odometry technique on corobots using Kinect depth and RGB camera.
- Currently corobots uses encoders which face precision problems, due to wheel slips and slide.

## **BACKGROUND**

- Visual odometry is process of determining position of robot using associated camera images.
- Corobots have Kinect sensor which provides RGB and depth images.
- Using consecutive images transformation, change in camera position. is calculated in terms of rotation matrix and translation vector.
- Visual odometry computation is done either using images or 3D point cloud.
- Use of monocular camera lacks scale information while operations on 3D point cloud is computationally heavy.
- Proposed approach combines use of 2D images and selected 3D points to estimate transformation with faster computation time than traditional point cloud methods.

# ARCHITECTURE



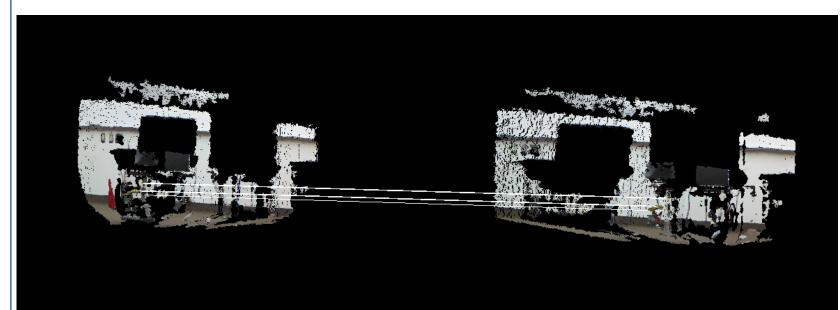
# **ALGORITHM**

# Finding and tracking Key points

- Key points are points which are distinctive, easier to track in consecutive images. FAST corner detector was used in this approach.
- Tracking is method of finding corresponding points from first image in to second. Kanade-Lucas-Tomasi feature tracker which uses optical flow was used in this approach.

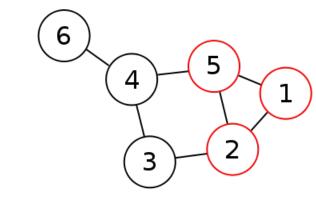
#### 3D Key point estimation

The points in images are converted to real world 3D points using depth data and intrinsic camera parameters obtained from calibration.



#### **Inlier Detection**

All 3D point coordinates and correspondences might not be accurate. This was solved by constructing 3D point graph by connecting stable points and finding maximum clique sub-graph.



#### **Transformation estimation**

http://www.pointclouds.org/

Transformation matrix between scenes consists of relative rotational matrix and translation vector. RANSAC based approach was used.

## **RESULTS**

Method	Average Time
Monocular	0.21 Sec
3D Point Cloud	4.89 Sec
Our Approach	1.02 Sec

Method	Accuracy with features > 100
Monocular	N. A.
3D Point Cloud	+- 5%
Our Approach	+- 5%

Place	Number of features	Inlier Features
Hallway	5 - 30	0 - 5
Lab	40-250	30 - 150

# CONCLUSION

- Finding and tracking features in 2D images is faster, accurate and requires much less computational time than 3D point cloud.
- Visual odometry performs better than encoders when view has good lightning, enough features to track and accurate depth values.
- Places like hallways where light is inconsistent, surface is plain and only side walls can be tracked, visual odometry fails.

#### References

- http://www.ros.org/
- Fusing points and lines for high performance tracking, Edward Rosten and Tom Drummond
- An Iterative Image Registration Technique with an Application to Stereo Vision, Bruce D. Lucas and Takeo Kanade.
- Registration with the Point Cloud Library PCL A Modular Framework for Aligning 3D Point Clouds Dirk Holz, Alexandru E. Ichim, Federico Tombari, Radu B. Rusu and Sven Behnke