



Vision-Guided Self-Localization for Autonomous Cars

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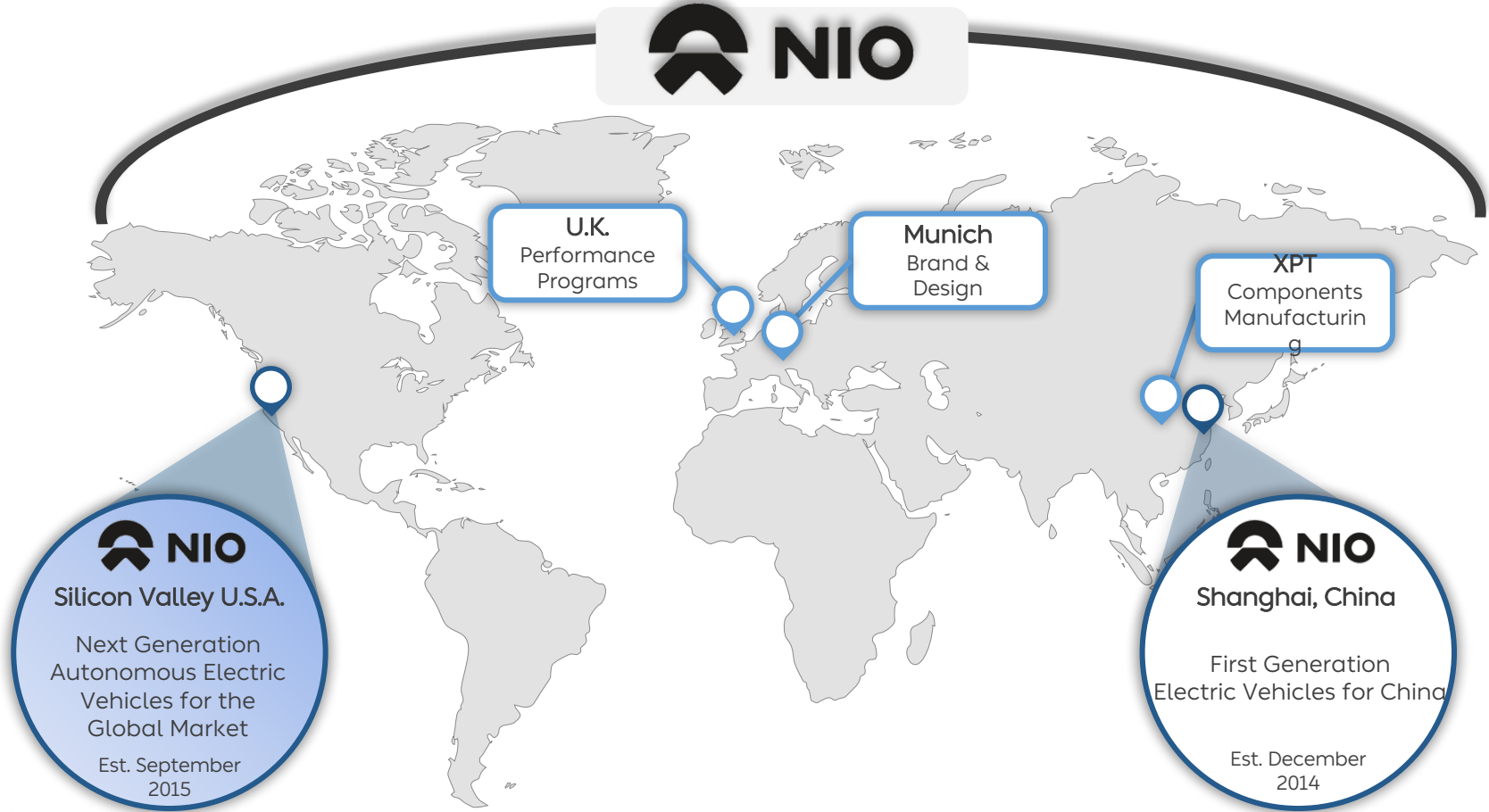
** Sathya is an intern from WPI and continuing his MS research at NIO*

Agenda

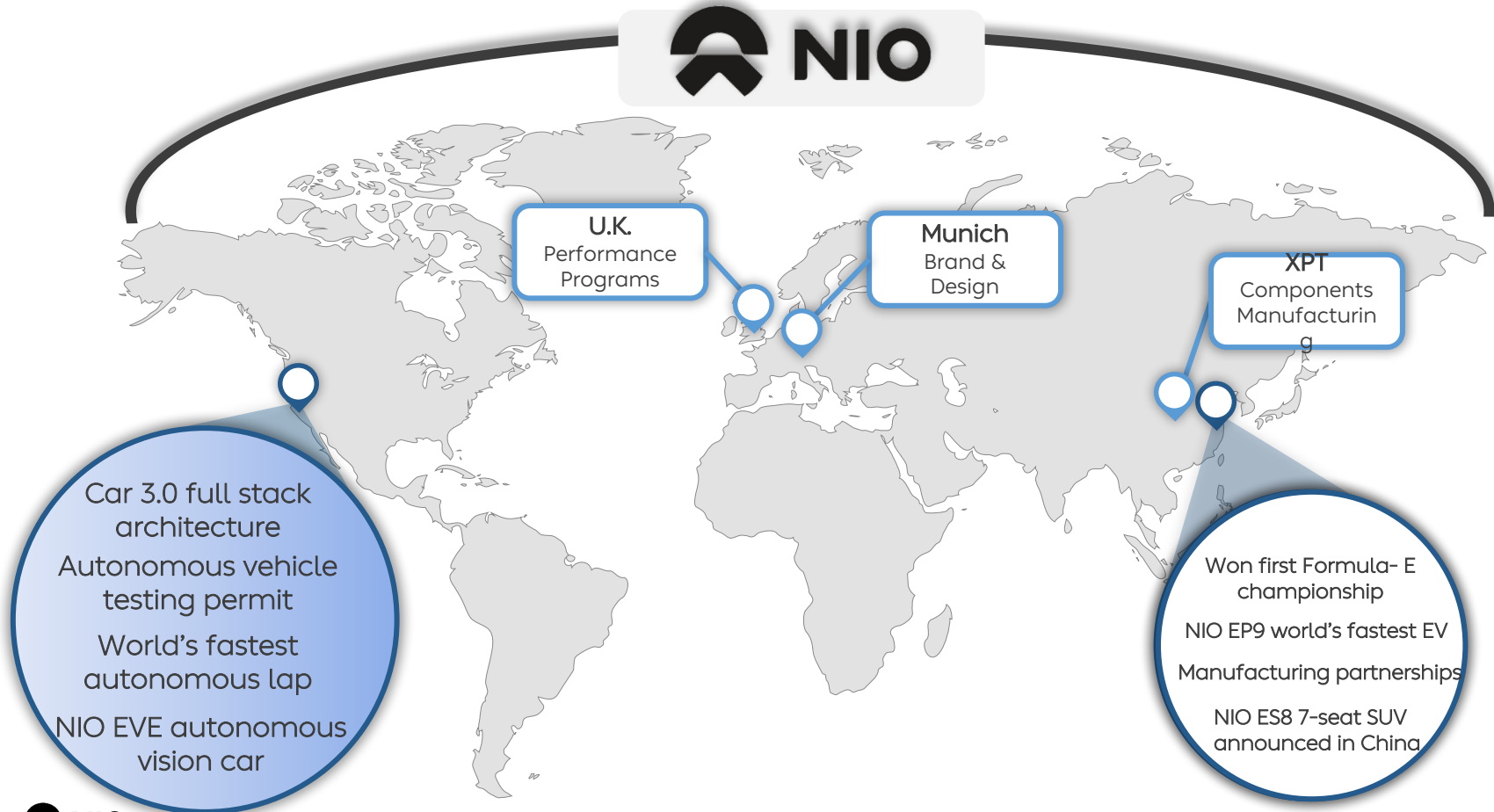
- About NIO
- Use of Various MATLAB Toolboxes at NIO
 - Project #1: EP9 Driving Autonomously at COTA
 - Control Systems Toolbox
 - MATLAB Coder
 - SIMULINK Coder
 - Project #2: Camera Calibration
 - Computer Vision System Toolbox
 - Project #3: Vision-Guided Self-Localization for Autonomous Cars
 - Image Processing Toolbox
 - Computer Vision System Toolbox
- Conclusions

About NIO

We are a Global Startup



Significant Progress To Date



EP9 : Autonomous Mode at COTA

Key Personnel:

Aaron Bailey (Head of Driving Dynamics)

Dennis Polischuk (Head of Platform Engineering, Firmware)

Kamran Turkoglu (Head of Controls/Algorithms)

NIO eP9

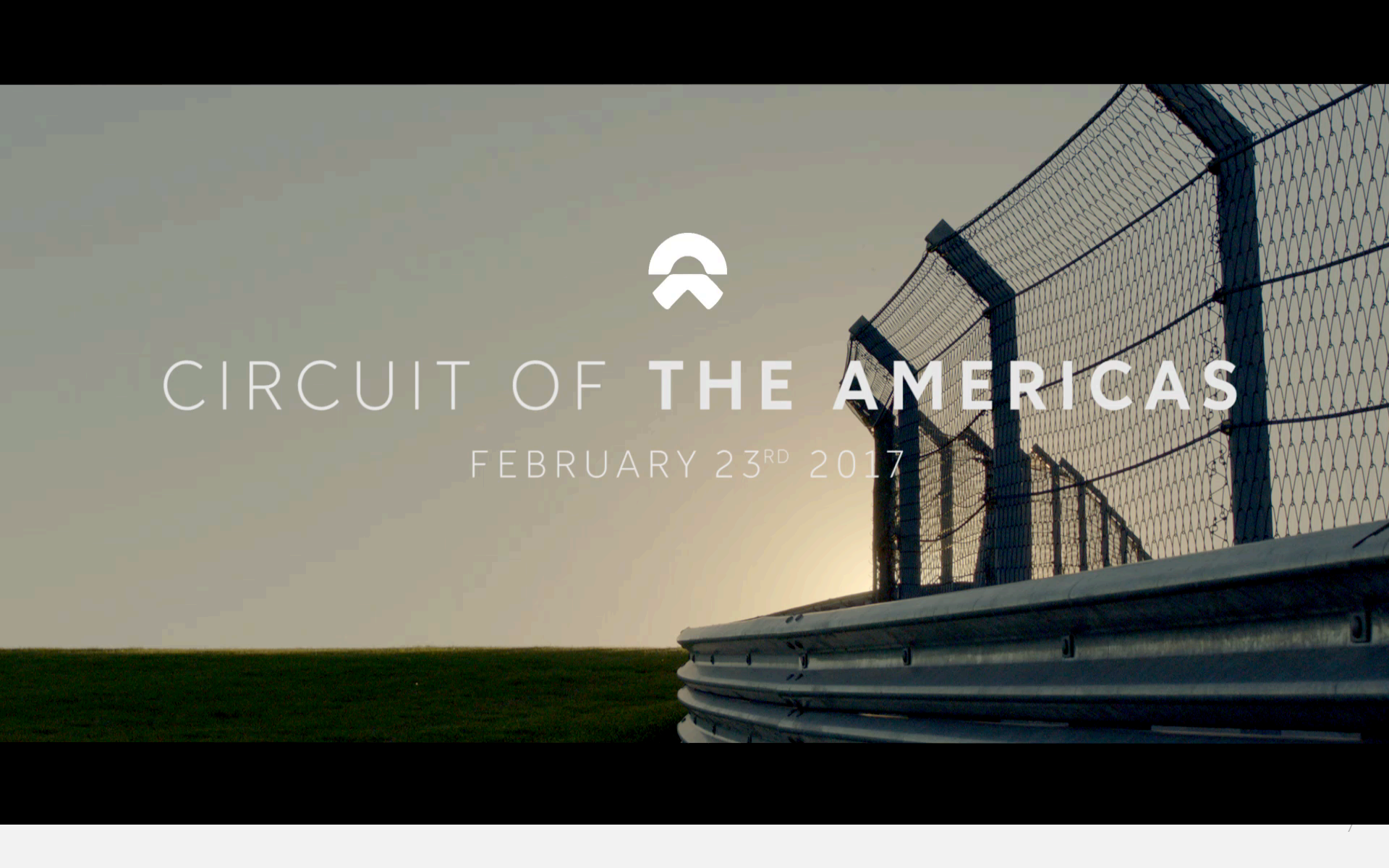
A limited edition 1 megawatt, 3G capable, electric supercar, born to push limits.





CIRCUIT OF THE AMERICAS

FEBRUARY 23RD 2017



Camera Calibration

Camera Calibration

Motivation

- Rapid prototyping of calibration methods for an ADAS tri-focal camera system with three different FoVs.
 - Narrow Field of View (28 degree)
 - Wide Field of View (150 degree) *
 - Main Field of View (52 degree)
- Calibrate a stereo camera system for a mapping pilot project.

**Special Thanks to Avi Nehemiah & Mathworks Team for providing NIO access to pre-release version of MATLAB for the Wide FoV Calibration*

Single Camera Calibration – cameraCalibrator APP

The screenshot displays the cameraCalibrator APP interface. At the top, the title bar reads "Camera Calibrator - Reprojection Errors". Below it is a menu bar with "CALIBRATION" and various icons for file operations (New, Open, Save, Add Images), options (Radial Distortion, Compute, Skew, Tangential Distortion), optimization (Optimization Options), calibration (Calibrate), zoom (Zoom In, Zoom Out, Pan), layout (Default Layout), export (Export Camera Parameters), and help (Help).

The main interface is divided into several panels:

- Data Browser:** A list of 11 images, with "1: img_31.jpg" selected.
- Image:** The main view showing "img_31.jpg". It features a checkerboard target in the foreground with green dots at the corners of the squares. A legend indicates: "○ Detected points", "□ Checkerboard origin", and "+ Reprojected points". The origin is labeled "(0,0)". Axes are labeled "X" and "Y".
- Reprojection Errors:** A histogram showing the distribution of mean error in pixels across 30 images. The y-axis is "Mean Error in Pixels" (0 to 0.6) and the x-axis is "Images" (0 to 30). A red line at the top indicates "Drag to select outliers". A dashed horizontal line at approximately 0.23 pixels is labeled "Overall Mean Error: 0.23 pixels".
- Extrinsics:** A 3D plot showing the camera's position and orientation in a coordinate system. The axes are labeled "X (millimeters)", "Y (millimeters)", and "Z (millimeters)". The camera center is marked with "X_c", "Y_c", and "Z_c".

Buttons for "Show undistorted" and "Show pattern-centric view" are located at the bottom of their respective panels.

Stereo Camera Calibration – stereoCameraCalibrator APP

STEREO CAMERA CALIBRATOR - REPROJECTION ERRORS

CALIBRATION

Radial Distortion: 2 Coefficients Skew
 3 Coefficients Tangential Distortion

Compute: Optimization Options Calibrate

Zoom In Zoom Out
Pan

Default Layout Export Camera Parameters Help

Data Browser

1: right-0000.png & left-0000.png
2: right-0001.png & left-0001.png
3: right-0002.png & left-0002.png
4: right-0003.png & left-0003.png
5: right-0006.png & left-0006.png
6: right-0007.png & left-0007.png
7: right-0008.png & left-0008.png
8: right-0009.png & left-0009.png

Image

Camera 1 Camera 2

Legend:
○ Detected points
+ Reprojected points
■ Checkerboard origin

Reprojection Errors

Image Pairs	Camera 1 Mean Error (pixels)	Camera 2 Mean Error (pixels)
5	1.05	1.05
6	0.35	0.35
7	0.42	0.42
8	0.25	0.25
9	0.22	0.22
10	0.58	0.58
11	0.65	0.58
12	0.90	0.90
13	0.42	0.42
14	0.32	0.32
15	0.30	0.30
16	0.45	0.45
17	0.55	0.42
18	0.42	0.42
19	0.25	0.25
20	0.22	0.22
21	0.20	0.20
22	0.25	0.25
23	0.42	0.42
24	0.55	0.42
25	0.60	0.42
26	0.65	0.42
27	0.55	0.42
28	0.55	0.42
29	0.55	0.42
30	0.55	0.42
31	0.55	0.42
32	0.55	0.42
33	0.55	0.42
34	0.55	0.42
35	0.55	0.42
36	0.55	0.42
37	0.55	0.42
38	0.55	0.42
39	0.55	0.42
40	0.55	0.42
41	0.55	0.42
42	0.55	0.42
43	0.55	0.42
44	0.55	0.42
45	0.55	0.42
46	0.55	0.42
47	0.55	0.42
48	0.55	0.42
49	0.55	0.42
50	0.55	0.42

Overall Mean Error: 0.40 pixels

Extrinsics

Vision-Guided Self-Localization for Autonomous Cars

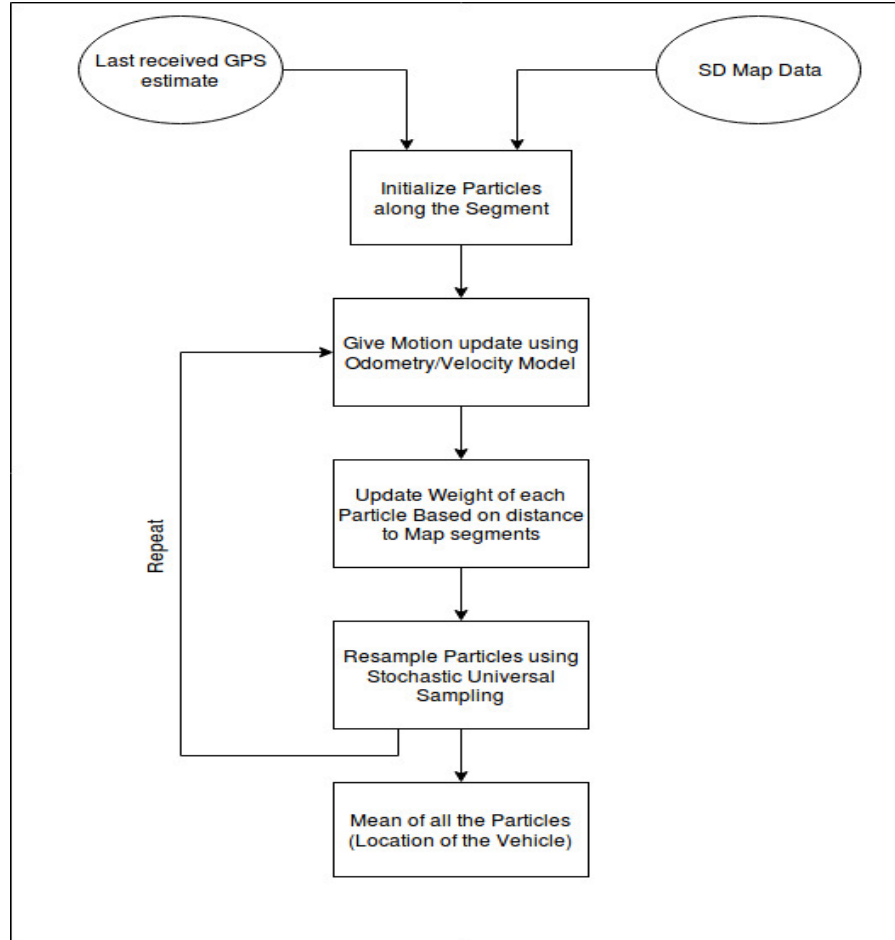
Overview

- **Motivation**
 - Limited Accuracy of GPS in Production Car (> 5 meters).
 - Poor GPS Signal Reception problems
 - Urban Areas / Urban Canyons.
 - Inside a Tunnel etc.

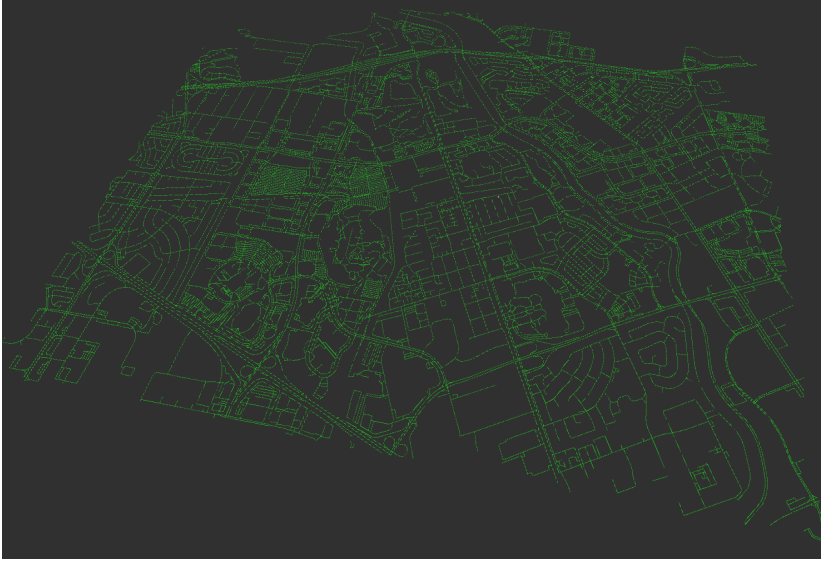
- **Available Input Signals**
 - Last received GPS estimate.
 - SD Maps
 - HD maps are currently being investigated.
 - Odometry
 - Change in position over time.
 - Camera + LiDAR
 - Work in Progress

- **Method Implemented**
 - Particle Filtering

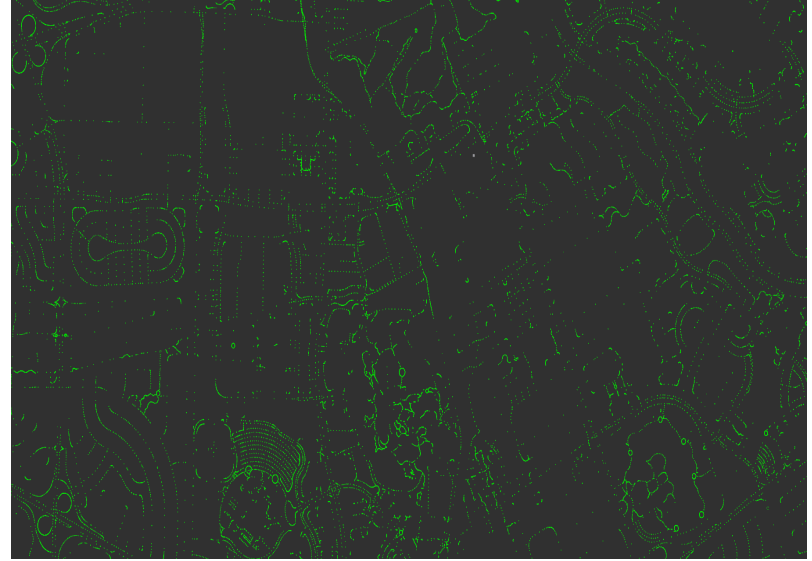
Algorithm: Using Particle Filters for Localization



SD Map



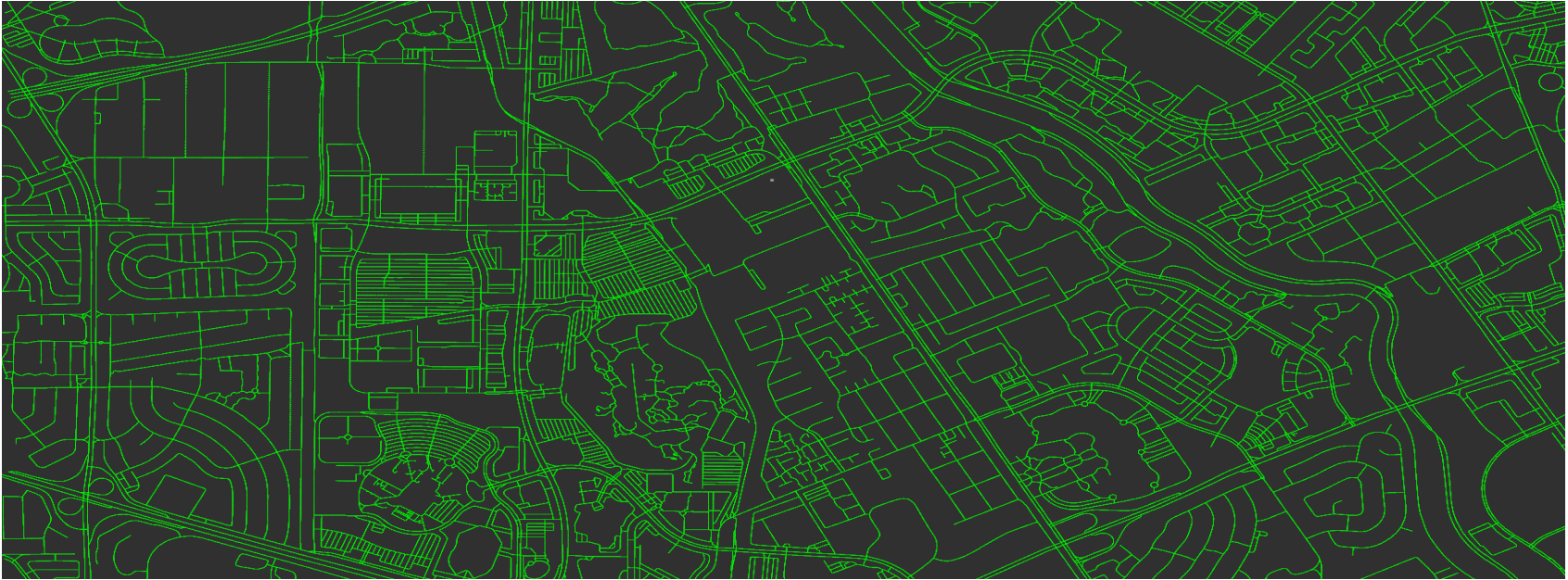
SD Map



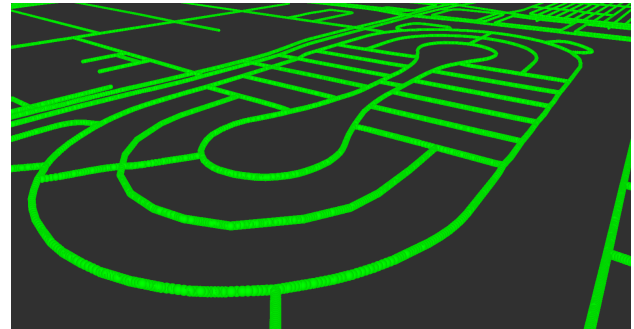
Close Up of SD Map

- The SD map information that is currently available contains only the Segment Geometry.
- Each segment has N points represented in the local frame.
- Number of points in the segment are sparse.
- For our application
 - We need closely spaced points to generate particles all along the segment.

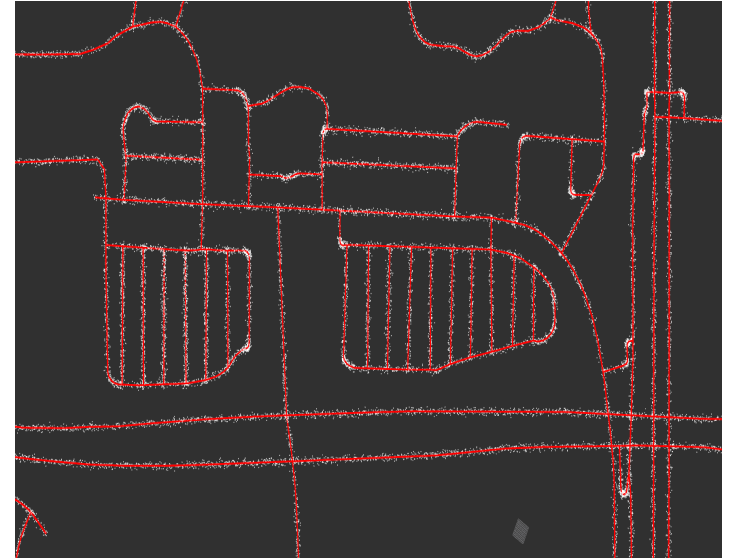
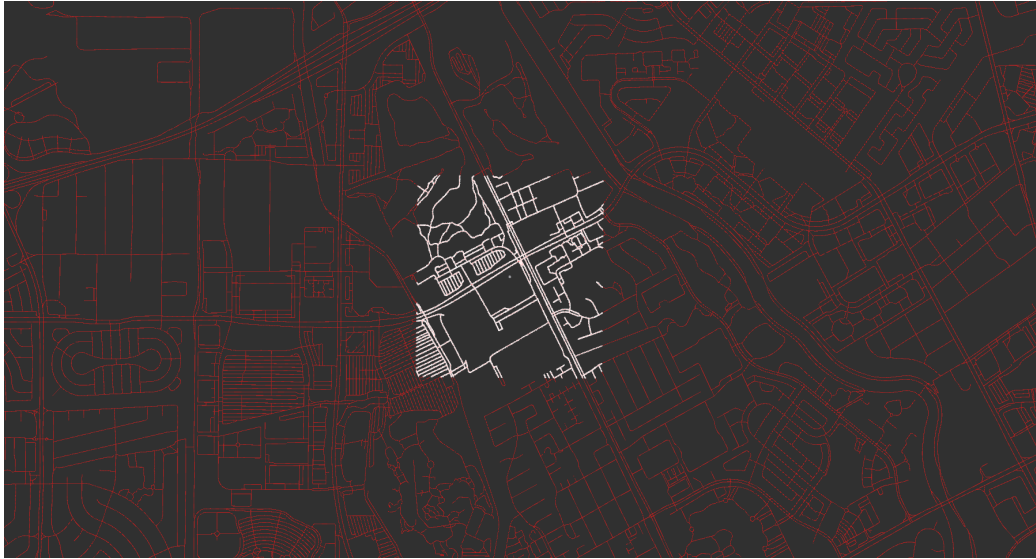
SD Map - Augmentation



- Number points along each segment along the map is augmented.
- Due to this, we have a more dense map information for localization.

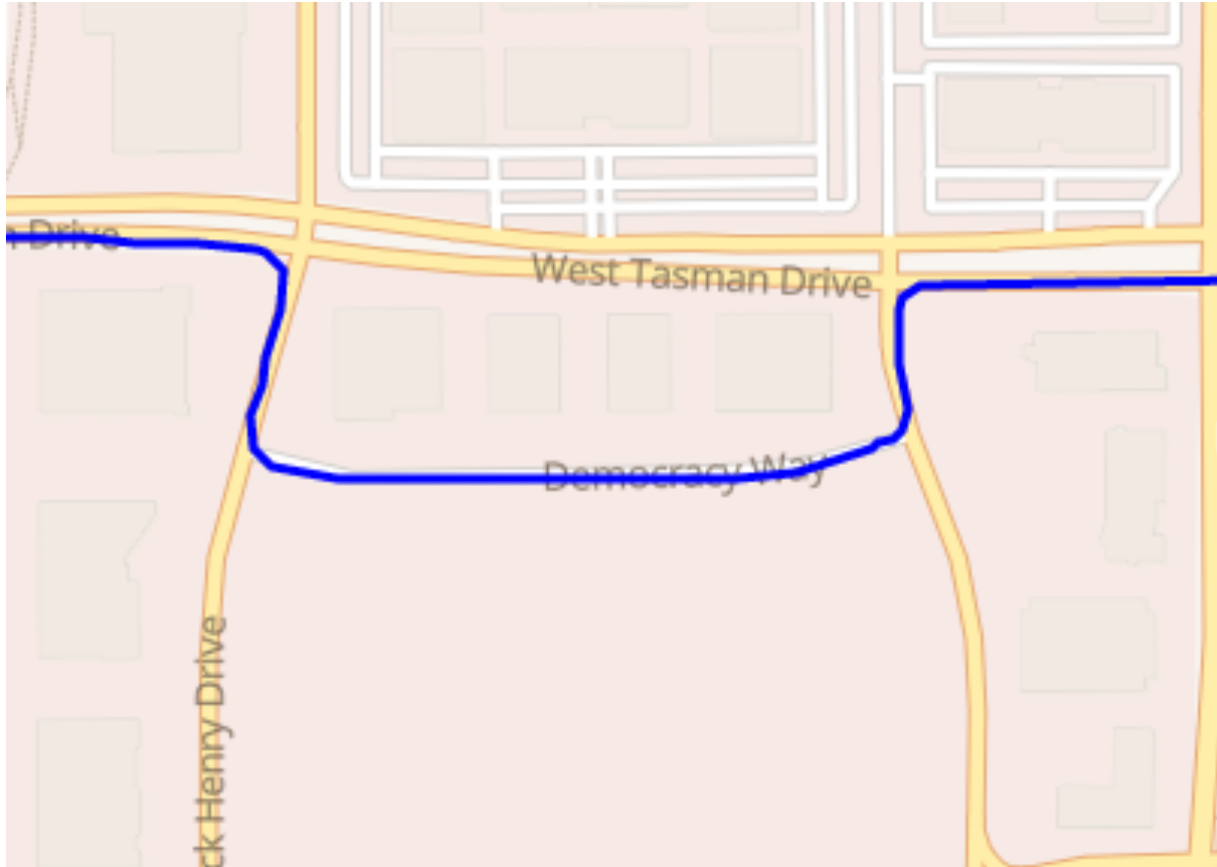


Particle Filtering for Localization



- Once we have the augmented map, we need to sample particles along the segments of the map.
 - States of the particles are x, y, θ .
- First we are extracting a 500m x 500m area using the last received GPS point.
- For augmenting, each point along the segment is considered as a possible location of the vehicle and the particle's orientation θ is the orientation of the segment.
- Using this information the particles are generated with orientations equal to $\theta, 180 + \theta$.

Experiment: Map Segment to Localize



Test Route: Trying to localize within this map segment

Results

The screenshot displays a ROS2 simulation environment. On the left, a camera view shows a black screen with the text "No Image" in white. The top toolbar includes icons for Interact, Move Camera, Select, Focus Camera, Measure, 2D Pose Estimate, 2D Nav Goal, and Publish Point. The main area is divided into several terminal windows. The top-left terminal shows the output of a particle filter node, displaying statistics such as "Number of Particles := 78576", "Temp Particles Size := 78349", and "New Particles Size := 78349". It also lists particle cloud size and velocity. The top-right terminal shows a recording interface with buttons for "Start recording", "Cancel recording", "Save recording", "Hide window", and "Quit". The middle-left terminal shows the output of a "done" command. The middle-right terminal shows the output of a "roscore" command, indicating that the core service is started. The bottom-left terminal shows the output of a "cat" command, displaying the version of the OpenCL library. The bottom-right terminal shows the output of a "cat" command, displaying the version of the OpenCL library. At the bottom of the interface, a "Time" panel shows "ROS Time: 1500398864.82", "ROS Elapsed: 93.34", "Wall Time: 1500398864.85", and "Wall Elapsed: 93.24". Below the time panel, there is a "Reset" button and a legend: "Left-Click: Rotate, Middle-Click: Move X/Y, Right-Click: Zoom, Shift: More options."

When the Green (ground truth) and Blue dots (computed position value) overlap, the vehicle has self-localized in the map



Current Limitations

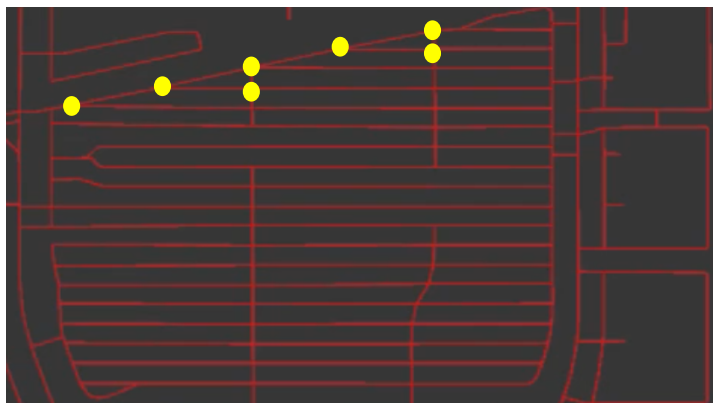
- Its an iterative approach.
- Solution takes time to converge to a location.
- Computed location can have ambiguities if the driving path is mostly along straight lines.

How can we improve ?

Proposed Solution: Add Visual Features to the Map Nodes

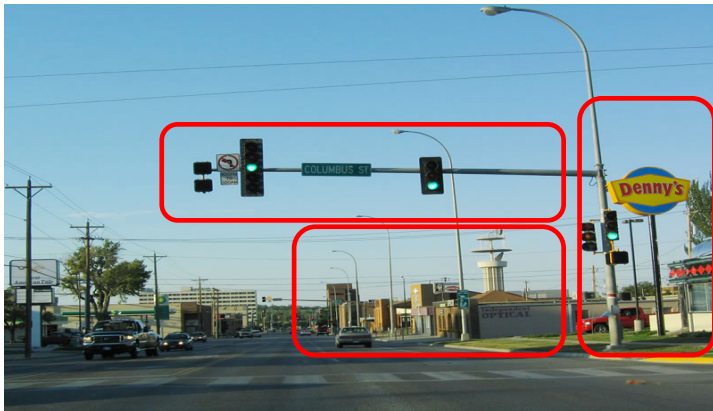
- Extract visual features from the camera sensors and send it to a map cloud.
- Visual features enhanced map data can then be used in the real-time localization.
- The localization itself can be done in real-time
 - Feature extraction and matching
 - Reduces the number of potential locations for the particle filters before we converge to a solution.
- This is a form of map crowdsourcing: every car contributes to build a better localization map

Augmenting the map with visual place features



Nodes

- The yellow dots represent the map nodes.
- For illustration, we show a few nodes. In reality each intersection in the map is a node point.
- The road segments (red lines) connect the different nodes in a map



Visual Features

$$\bar{x}_1 = [x_{1,1} \quad x_{1,2} \quad \dots \quad x_{1,p}]$$

$$\bar{x}_2 = [x_{2,1} \quad x_{2,2} \quad \dots \quad x_{2,p}]$$

$$\bar{x}_n = [x_{n,1} \quad x_{n,2} \quad \dots \quad x_{n,p}]$$

- We would store this info at each yellow dot above during mapping and also as a dynamic update.
- During localization, we extract the features and do feature matching. This will reduce the number of particle filters we need and the places we search for a matching location.

Visual image captured at an intersection

The red boxes indicate some robust visual features we would like to extract.

References

- Levinson, Jesse, Michael Montemerlo, and Sebastian Thrun. "Map-Based Precision Vehicle Localization in Urban Environments." *Robotics: Science and Systems*. Vol. 4. 2007.
- Brubaker, Marcus A., Andreas Geiger, and Raquel Urtasun. "Lost! leveraging the crowd for probabilistic visual self-localization." *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition*. 2013.
- Ruchti, Philipp, et al. "Localization on openstreetmap data using a 3d laser scanner." *Robotics and Automation (ICRA), 2015 IEEE International Conference on*. IEEE, 2015.

Conclusions

Conclusions

- Presented an overview of the various projects at NIO using Matlab Toolboxes
 - Project #1: EP9 Driving Autonomously at COTA
 - With the help of Control Systems Toolbox, MATLAB Coder and SIMULINK Coder, NIO was able to implement the autonomous driving feature (GPS waypoint following) and successfully demo EP9 at the COTA.
 - Our Controls and Algorithms team use Control Systems Toolbox, MATLAB Coder and SIMULINK Coder for the L2/L4 features currently in development.
 - Project #2: Camera Calibration
 - With the Computer Vision System Toolbox, NIO was able to do rapid prototyping of the various calibration algorithms for the ADAS trifocal camera system.
 - Project #3: Vision-Guided Self-Localization for Autonomous Cars
 - NIO currently relies on the Image Processing Toolbox and the Computer Vision System Toolbox for the visual feature recognition that we are incorporating into improving the self-localization algorithm.
- In addition,
 - NIO is currently using Automated Driving System Toolbox for some its R&D projects.

Thank you !!

NIO is hiring
<https://www.nio.io/careers>