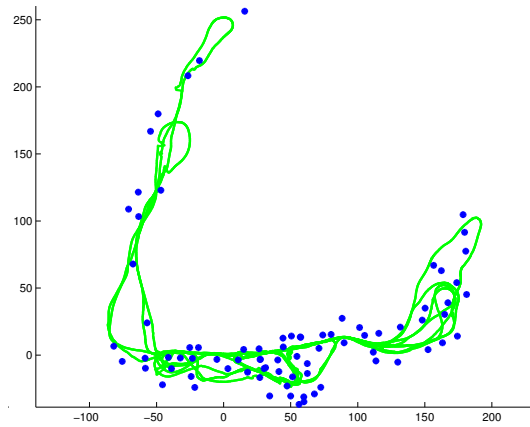


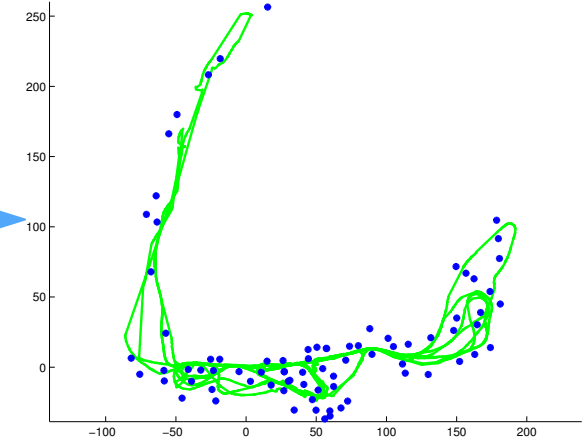
Full SLAM

151 Landmarks, 6968 Poses



After removing landmarks

74 Landmarks, 6968 Poses



After removing poses

74 Landmarks, 2971 Poses

Information based Reduced Landmark SLAM

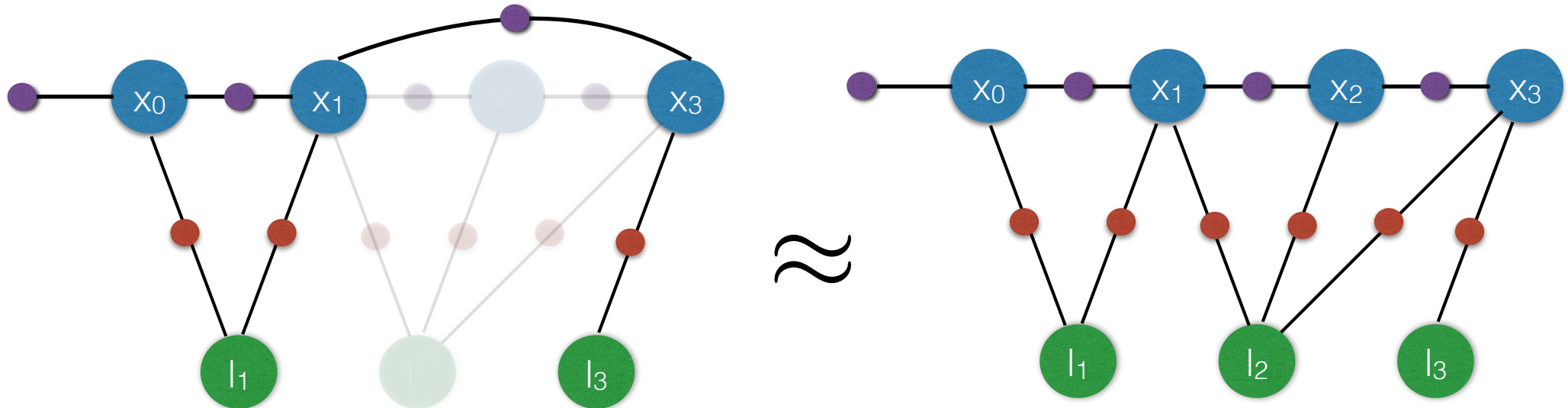
Siddharth Choudhary¹, Vadim Indelman², Henrik I. Christensen¹, Frank Dellaert¹

¹ Institute for Robotics and Intelligent Machines, Georgia Tech

² Technion - Israel Institute of Technology



Problem Statement

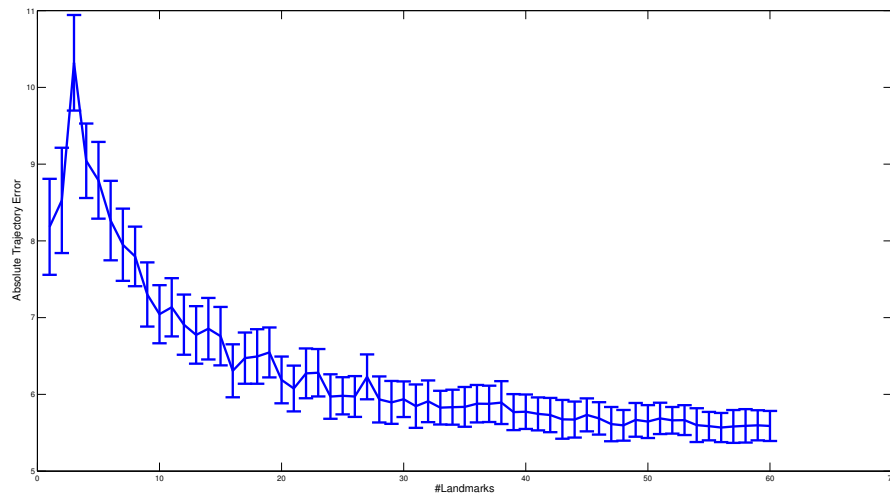


Incrementally finding a reduced subset of landmarks and poses such that the difference between the trajectory estimated using all landmarks and poses and using a subset of landmarks and poses is minimal as well

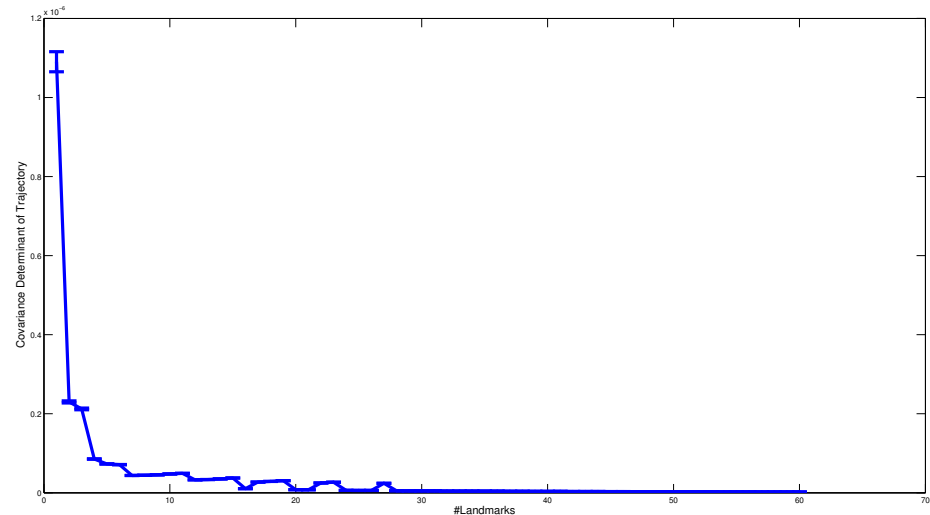


Motivation

Experiment with random sprinkling of landmarks given known ground truth trajectory



Trajectory Error



Marginal Covariance

Trajectory error and covariance stabilizes after adding a few landmarks



Related Work

Graph Reduction

- Dissanayake et al. Map management for efficient simultaneous localization and mapping (ICRA 2002)
- Strasdat et al. Which landmark is useful? learning selection policies for navigation in unknown environments. (ICRA 2009)
- Ila et al. Information based compact pose SLAM (TRO 2010)
- Eade et al. Monocular graph SLAM with complexity reduction (IROS 2010)
- Kretzschmar and Stachniss. Information theoretic compression of pose graphs for laser based SLAM (IJRR 2012)

Graph Sparsification

- Vial et al. Conservative sparsification for efficient and consistent approximation estimation (IROS 2011)
- Huang et al. Consistent sparsification for graph optimization (ECMR 2013)
- Wang et al. Kullback-leibler divergence based graph pruning in robotic feature mapping (ECMR 2013)
- Carlevaris-Bianco et al. Generic Node removal for Factor Graph SLAM (TRO 2014)
- Mazuran et al. Non linear graph sparsification for SLAM (RSS 2014)



Contributions

- Developed an **information theoretic algorithm** to efficiently **reduce the number** of landmarks and poses **without compromising the accuracy** of the estimated trajectory.
- Proposed an **incremental version** of the algorithm which can be used in a SLAM framework required for online operations.



Objective Function

Change in the
uncertainty of the
estimated trajectory

Memory consumption of
landmarks and poses

$$\rho(L_s, X_s) = (1 - \lambda) d(\Theta_s^*) + \lambda m(L_s, X_s)$$

Regularization Parameter



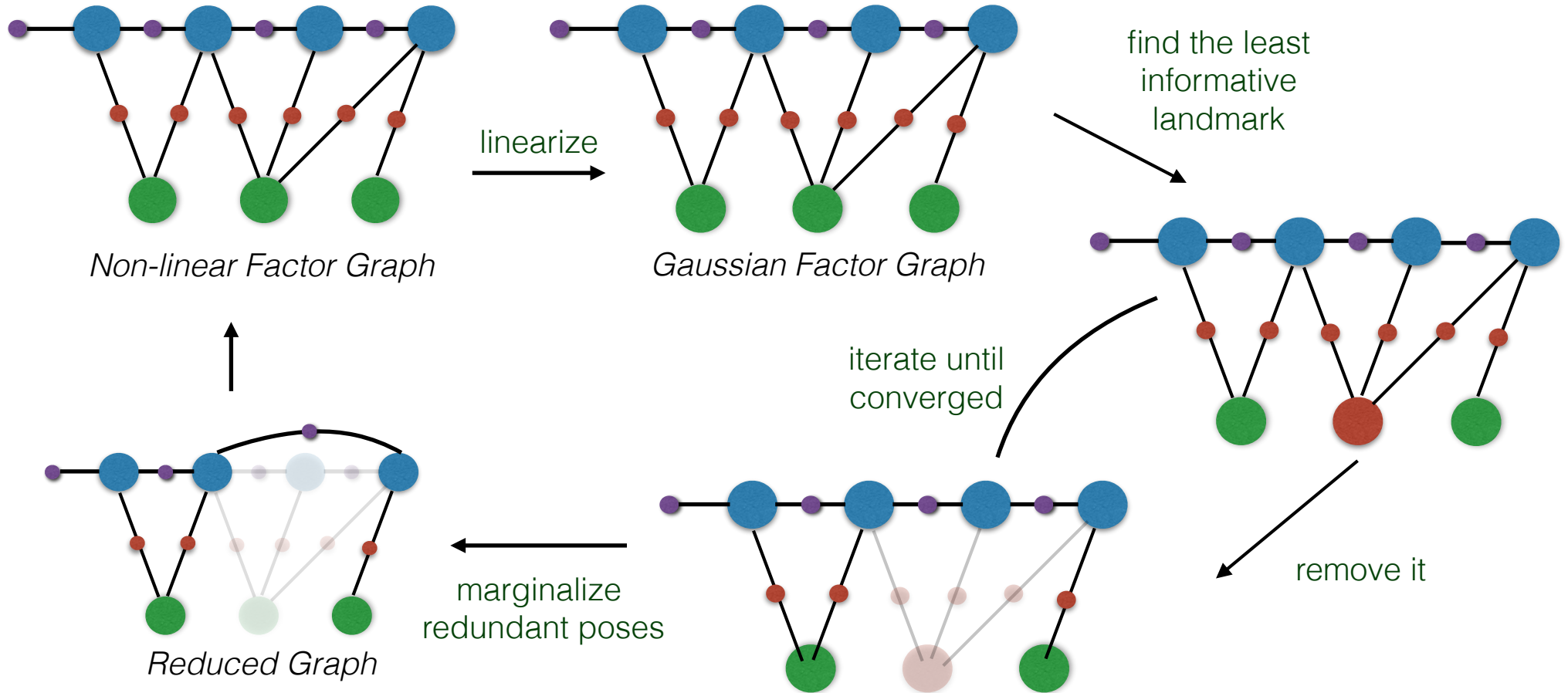
Objective Function

Find a subset of landmarks and poses that minimizes the objective function $\rho(L_s, X_s)$

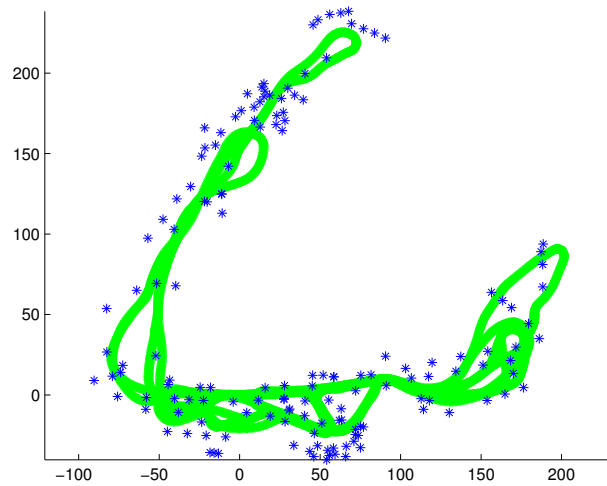
$$\{L^*, X^*\} = \arg \min_{L_s, X_s} \rho(L_s, X_s)$$



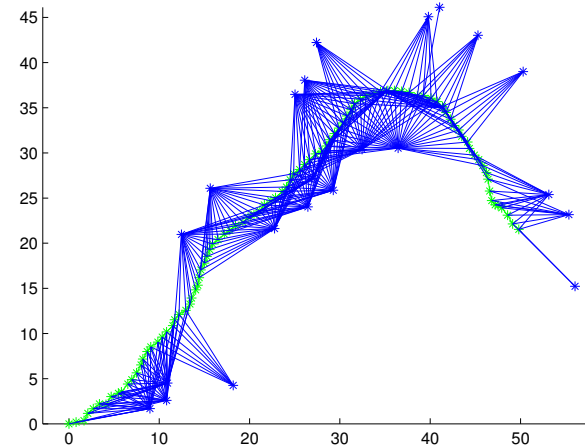
Algorithm



Datasets



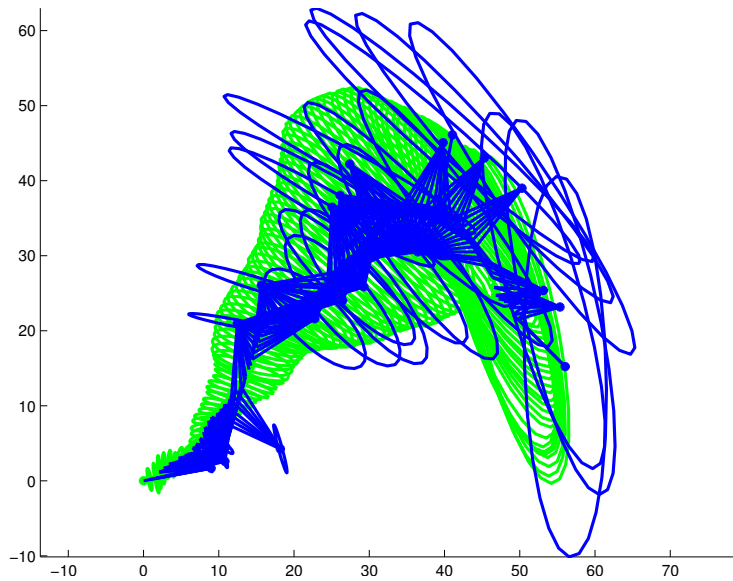
(a) Victoria Park dataset



(b) Synthetic dataset

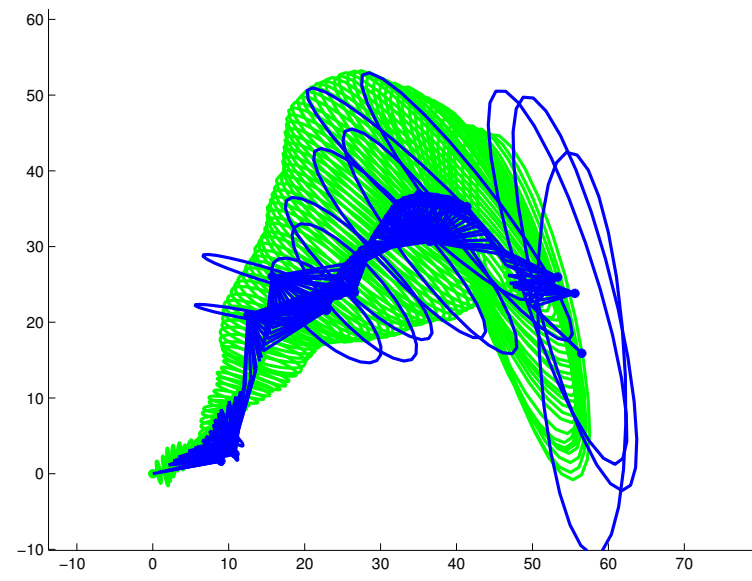
	L	P	Z	Area Covered
Victoria Park dataset	151	6969	10608	200×250 sq. units
Synthetic dataset	24	95	422	50×50 sq. units

Synthetic dataset



24 Landmarks,
94 Poses

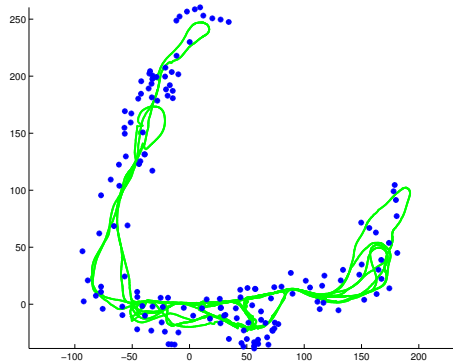
After
Minimization
→



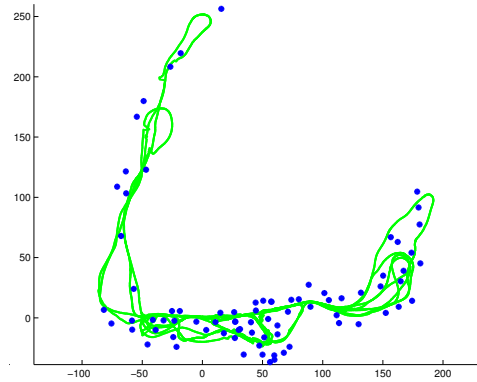
14 Landmarks,
94 Poses



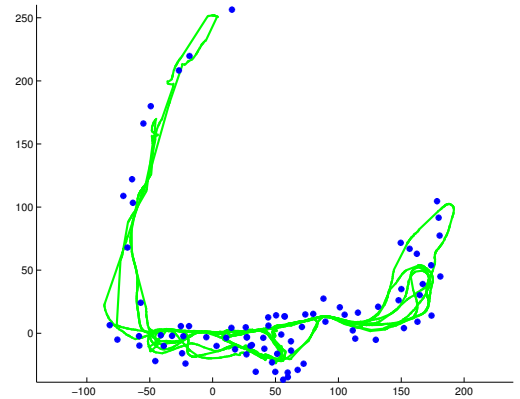
Victoria Park dataset



➔
Removing
redundant
landmarks



➔
Removing
redundant
poses



151 Landmarks,
6968 Poses

74 Landmarks,
6968 Poses

74 Landmarks,
2971 Poses

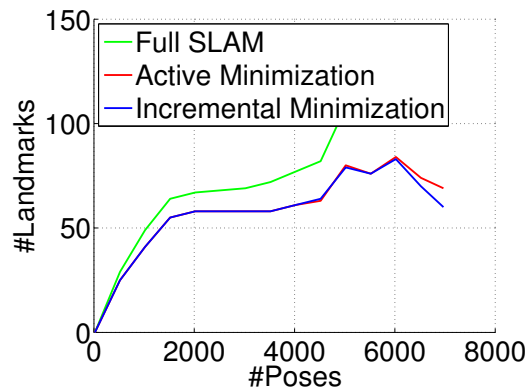


Information based Reduced Landmark SLAM

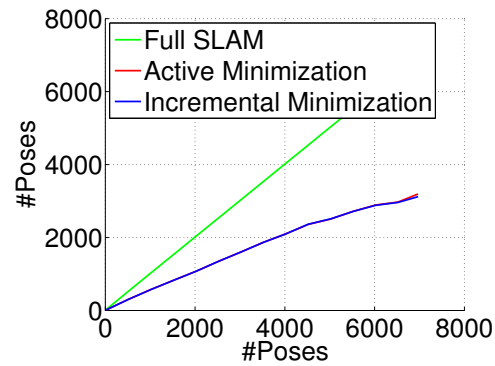
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Georgia Tech Institute for Robotics and Intelligent Machines

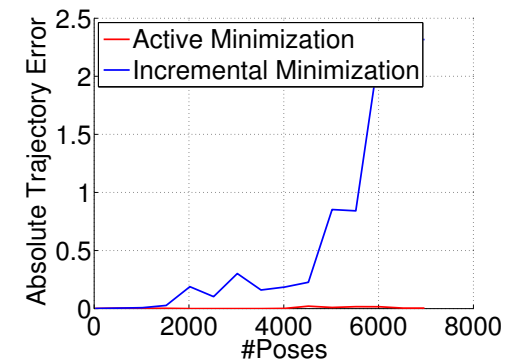
Victoria Park dataset



Number of Landmarks



Number of Poses



Absolute trajectory error



Conclusions

- Proposed an **incremental and active minimization** algorithm that can be used in a SLAM framework resulting in reduced landmark based SLAM.
- Showed a reduction of **40-50%** in the number of landmarks and around **55%** in the number of poses.



Thank You.
cogrob.org



Information based Reduced Landmark SLAM

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