



SLAM-based Integrity Monitoring for Multi-Sensors and Multi-Receivers



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Introduction

- **Integrity** denotes a measure of confidence in the correctness of position estimated by the navigation system
 - Error bound of estimated position is **Protection Level (PL)**
- In **urban environments**, **measurement redundancy** plays a key role in Integrity Monitoring (IM)
- **Challenges of GPS-only integrity** in urban canyons
 - Tall buildings and thick foliage cause satellite blockage, multipath effects and satellite broadcast anomalies
- **Our solution:** Leverage urban infrastructure to introduce additional redundancy by aiding GPS receiver with sensors that include
 - Visual feature-rich surroundings using camera
 - Cooperative inter-vehicle interactions via ranging

Objectives

- Develop an **aided-GPS IM algorithm** for urban areas, which
 - Provides a flexible platform for easy scalability across varied sensors;
 - Accounts for multiple measurement faults in different sensors, not just GPS
 - Computes the PL of estimated position

Multi-sensor: GPS and fish-eye camera



Multi-receiver: Cooperative network of GPS receivers



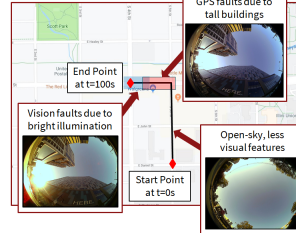
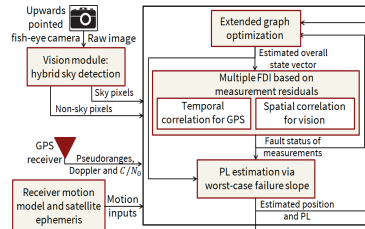
SLAM-based IM: PL Estimation

- PL is derived as **linear function** of the **worst-case failure slope of Graph-SLAM**
 - Failure slope is ratio between the position error and measurement residual
- Worst-case failure slope equals **maximum eigenvalue** of the failure slope formulation [2]
- Depends on fault status of measurements but is independent of absolute fault magnitude

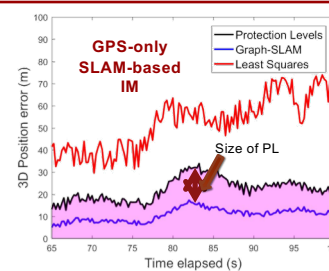
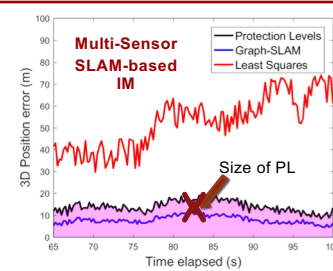
Multi-Sensor SLAM-based IM [3]

- Incorporating pixel intensities from **fish-eye camera** as additional constraints in graph optimization localize key image pixels, receiver and GPS satellites
- In multiple FDI, each measurement assigned binary fault status, i.e., reliable or unreliable
 - Temporal analysis of GPS measurement residuals
 - Analysis of spatial correlation across pixel intensity residuals

Multi-sensor SLAM-based IM



Real-world experiments for 100 s in semi-urban area of Champaign, IL experiencing both GPS and vision faults

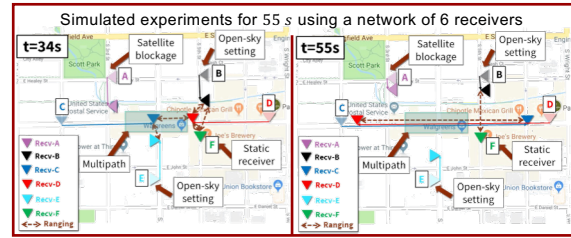
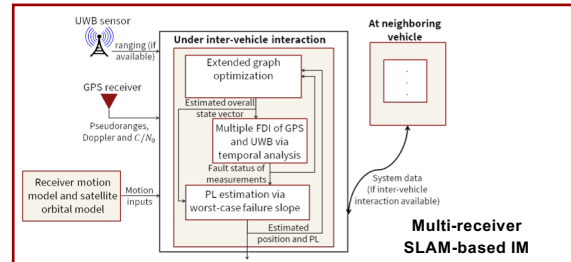


Statistics (m)	SLAM-based IM		Least Squares
	GPS-only	Multi-Sensor	
Position Error	16.2	8.8	52.44
Size of PL	10.5	6.5	--

Achieved tighter PL via multi-sensor SLAM-based IM as compared to GPS-only SLAM-based IM

Multi-Receiver SLAM-based IM [4]

- **Distributed** approach to Graph-SLAM that additionally utilizes **Ultra-Wide Band (UWB)** ranging across a network of vehicles
- At each vehicle, simultaneously localization of GPS satellites, itself and interacting vehicles



Receivers	Position Error (m)		Size of PL (m)	
	Multiple	Single	Multiple	Single
A (satellite blockage)	7.0	8.4	5.2	6.0
B (Open-sky)	5.9	7.1	4.3	5.8
C (Multipath)	9.1	12.5	5.4	8.4
D (Multipath)	7.1	11.6	4.6	9.7
E (Open-sky)	2.4	3.7	1.8	2.1

Lower position errors and tighter PL via multi-receiver SLAM-based IM

Conclusions

- Estimated PL in urban areas using SLAM-based IM via
 - Multi-sensor setup that utilizes GPS and fish-eye camera
 - Multi-receiver that include a cooperative network of vehicles
- Using real-world and simulated experiments, demonstrated higher position accuracy and associated tighter PL

References

[1] S. Bhamidipati, G. X. Gao, "Multiple GPS Fault Detection and Isolation Using a Graph-SLAM Framework," ION GNSS+, Miami, FL, Sept-2018, pp. 2672-2681
 [2] M. Joergler, F. C. Chan, and B. Pervan, "Solution separation versus residual-based RAIM," NAVIGATION: Journal of the Institute of Navigation, vol. 61, no. 4, pp. 273-291, 2014

References

[3] S. Bhamidipati, G. X. Gao, "SLAM-based Integrity Monitoring Using GPS and Fish-eye Camera," ION GNSS+, Miami, FL, Sept 2019, pp. 4116-4129
 [4] S. Bhamidipati, G. X. Gao, "Distributed Cooperative SLAM-based Integrity Monitoring Via a Network of Receivers," ION GNSS+, Miami, FL, Sept 2019, pp. 2023-2034

Acknowledgements and Contact

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