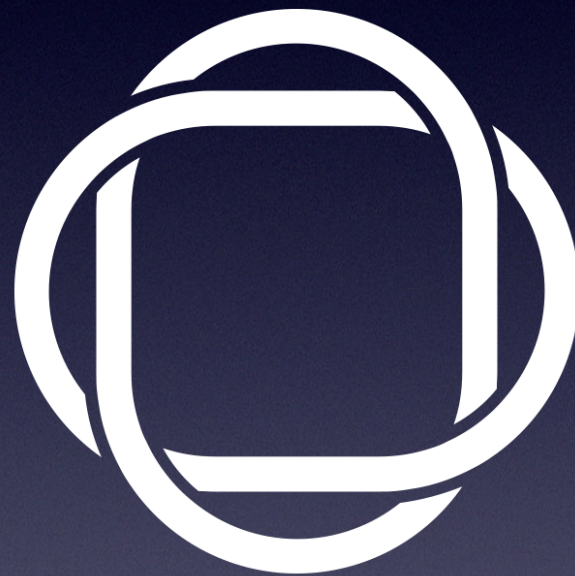


# Improve Poor GPS Area Localization for Intelligent Vehicles

Friday, 23rd February, 2018



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Dinh Van Nguyen, Fawzi Nashashibi, Trung-Kien Dao, Eric Castelli. Improving Poor GPS Area Localization for Intelligent Vehicles. MFI 2017 - IEEE International

Conference on Multisensor Fusion and Integration for Intelligent Systems, Nov 2017, Daegu, South Korea. pp.1-5. <hal-01613132>



# Abstract

- Introduction - Fusion of Laser-SLAM and GPS
- WiFi Fingerprinting Method
- Experiment and Results
- Conclusion and Limitation
- Q & A





# Laser-SLAM + GPS Fusion

- GPS is required to:
  - Map SLAM local coordinates to a global one
  - Correct SLAM accumulated error
  - Solve loop closure
- Need alternative in area with low/no GPS signal:
  - Parking structure, tunnel, urban area





# Other approaches

## Studies:

- Environment static map [2], [4], [5]
- Network of cooperating vehicle [6], [7]
- GPS + radio freq identification + vehicle to vehicle + vehicle to infrastructure [8]

## Problem:

- Fusing different sensors data
- Complicated system





# Wifi Fingerprinting Localization

- Training Phase:
  - Learn WiFi map of targeted environment
  - RSSI scanning + coordinates, at reference positions
  - Deterministic or probabilistic
- Prediction Phase:
  - Compare current signal with radio map





# Wifi Fingerprinting Localization

- Pro:
  - Less effort, data than HD map
  - Capable of mimicking GPS in fusing, smooth switching
- Con:
  - Indoor positioning
  - RSSI noise (interference, multipath propagation)
  - Prone to environment (AP, furniture) change
  - Low accuracy & speed





# Learning - RSSI recording

- Record Vector:

$$\{x_{i,1}, x_{i,2}, x_{i,3}, \dots, x_{i,n}, \rho_l\} \quad (1)$$

- Normalized in the range of  $[-1, 1)$

$$x_i = \begin{cases} -1, & AP_i \text{ undetected} \\ 1 - \frac{(-1) \times RSSI}{100}, & AP_i \text{ detected} \end{cases} \quad (2)$$





# Learning - training data

- Training Data are high variant
  - RSSI are noisy
  - Scanning frequency is low (vs movement speed)
- Method:
  - Ensemble Bagging (Bootstrap Aggregating) Neural Network

[9] L. Breiman, “Bagging Predictors,” *Mach. Learn.*, vol. 24, pp. 123–140, 1996.

[10] T. G. Dietterich, “Ensemble Methods in Machine Learning,” *Mult. Classif. Syst.*, vol. 1857, pp. 1–15, 2000.

- Can overcome high variant and noisy data



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# Experiment - environment



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# Experiment - System



- IbeoLux LIDAR
- Credibilist SLAM
- Standard 2.4GHz Wifi antenna
- IMU
- RTK GPS for ground truth





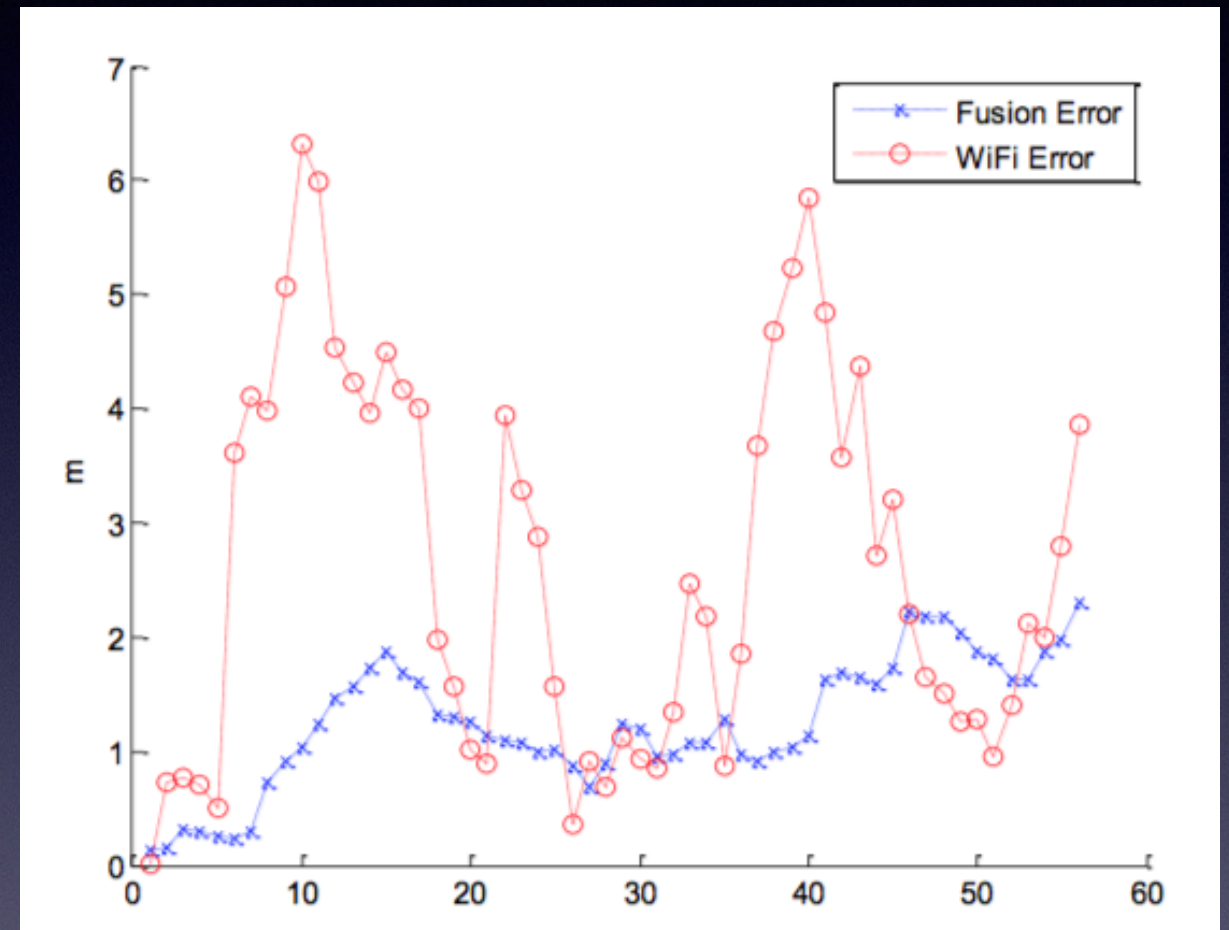
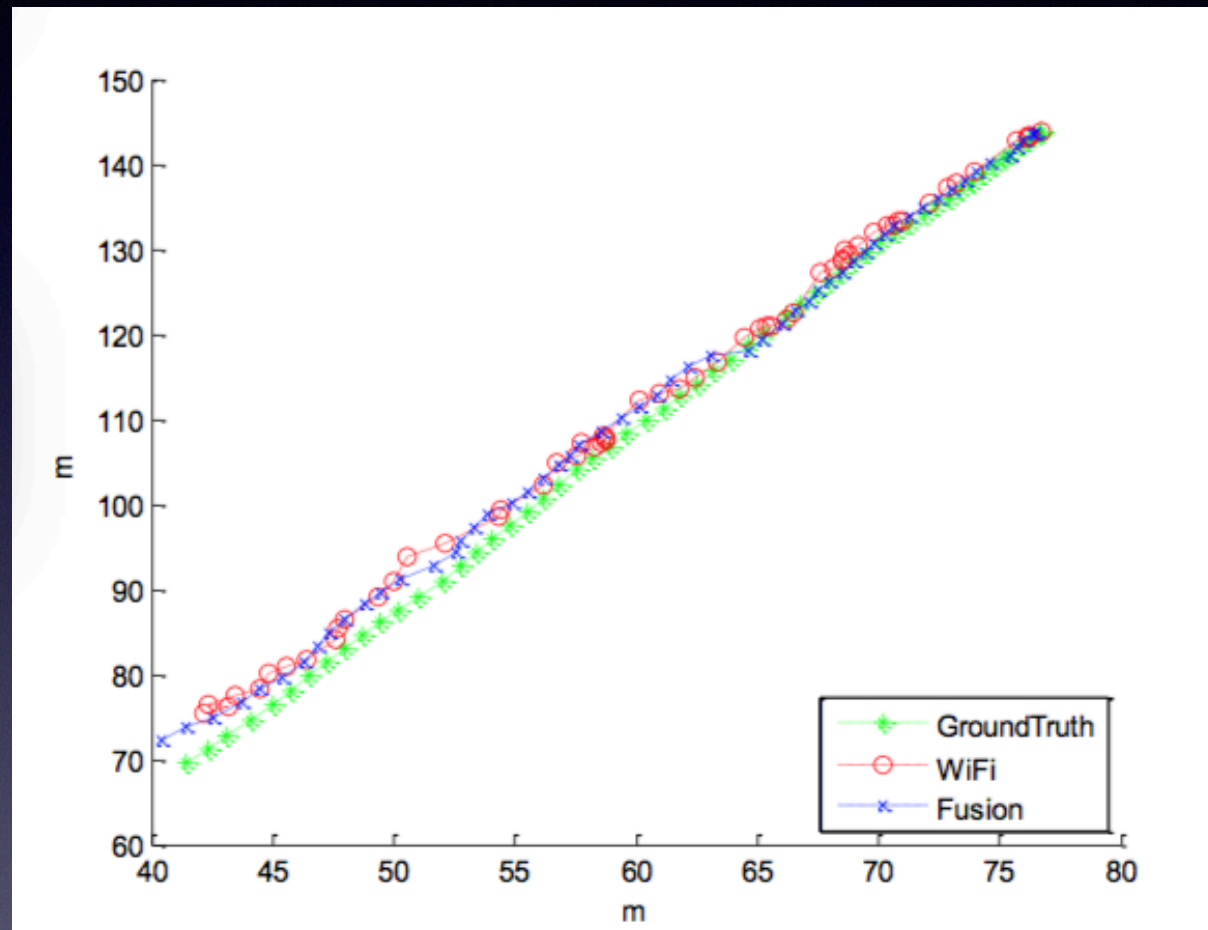
# Experiment - radio map

- 15 reference points
- Training data: 30 scans/point of WiFi signal collected
- ensemble of 50 neural networks, 170 input neurons each, 90 nodes at hidden layer, 15 outputs trained





# Wifi Localization vs SLAM



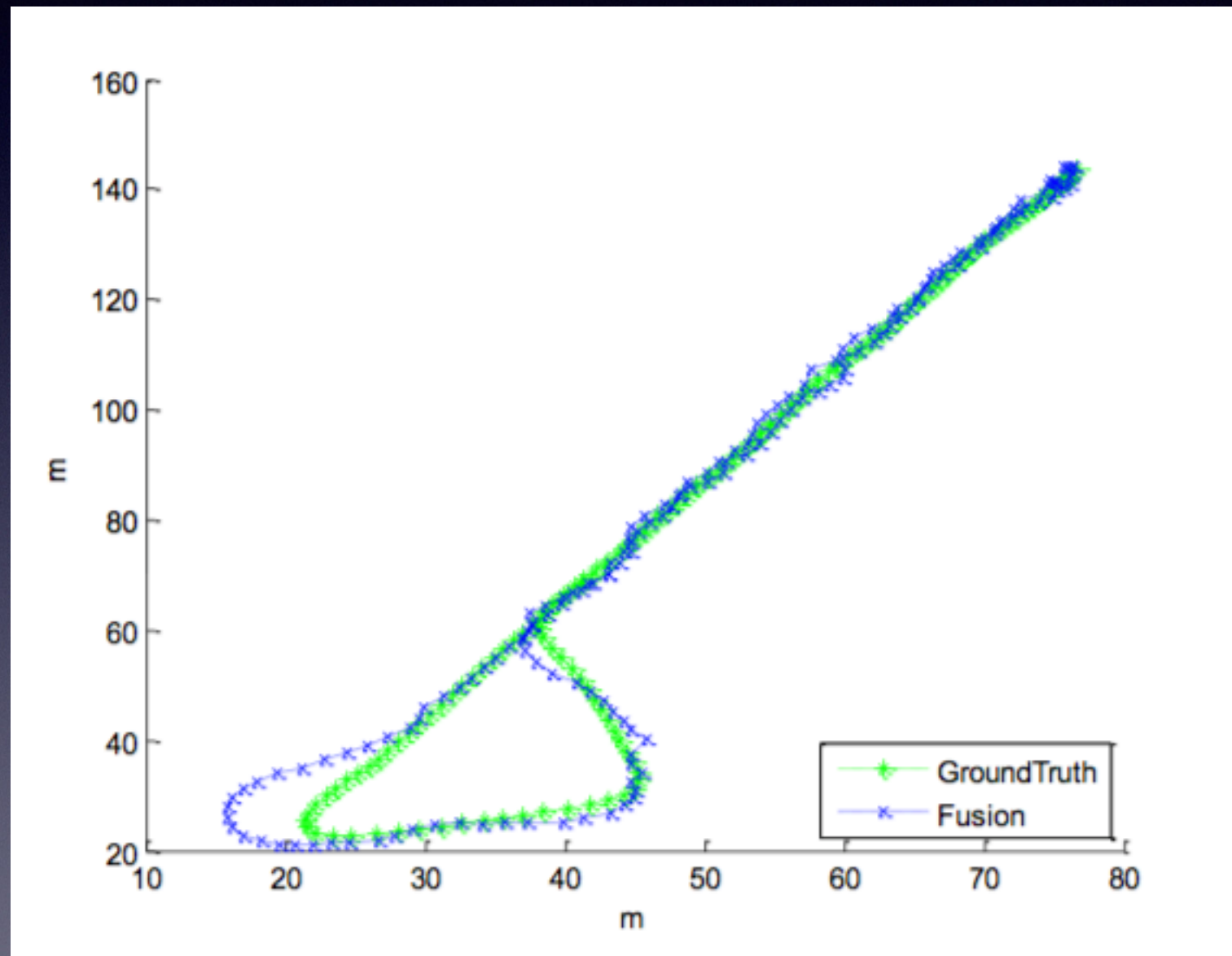
- average error = 3.328m, 98% errors < 6m

$$\sigma_{Wifi} = 6m$$



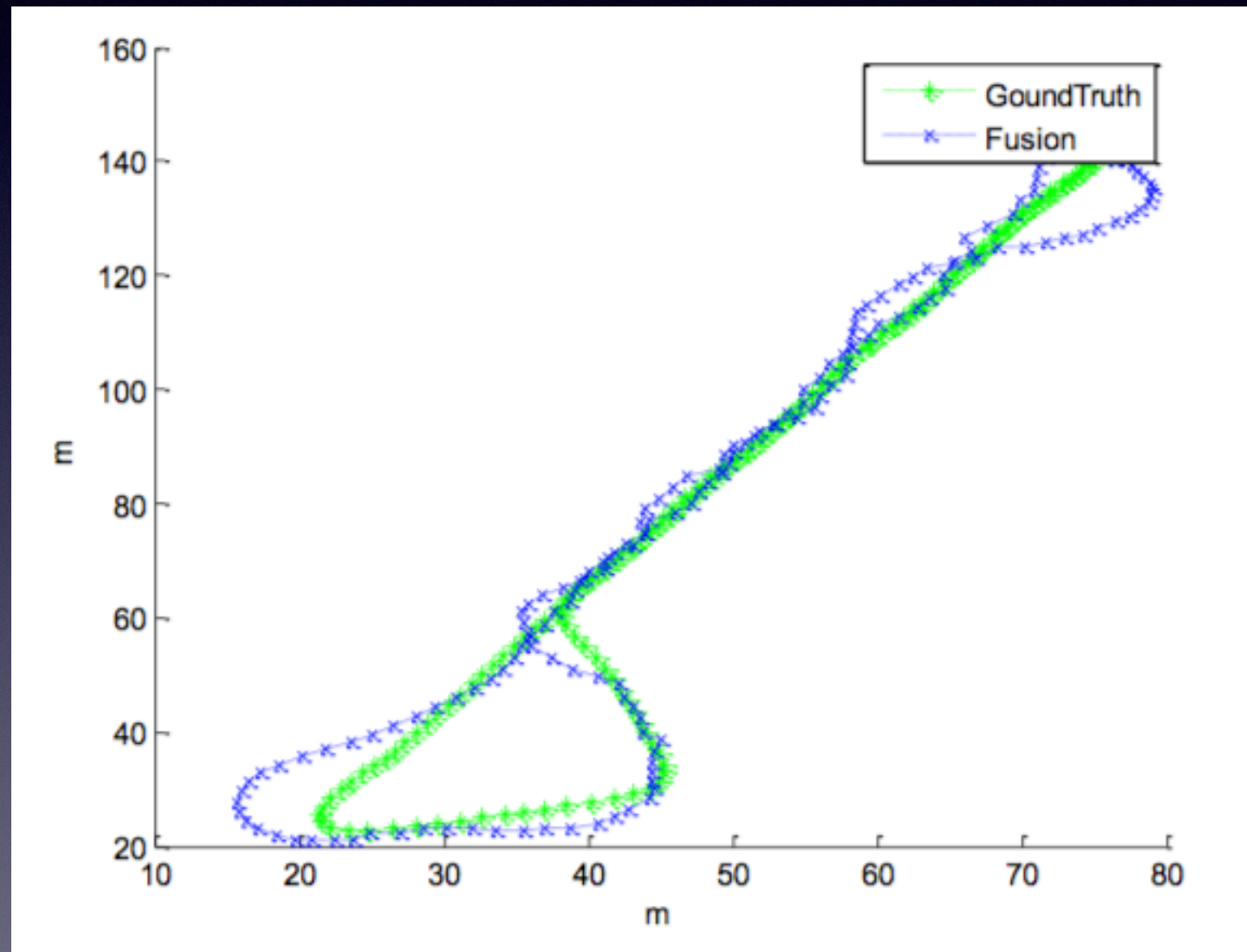


# Fusion of GPS and SLAM ( $\sigma_{GPS} = 6$ )





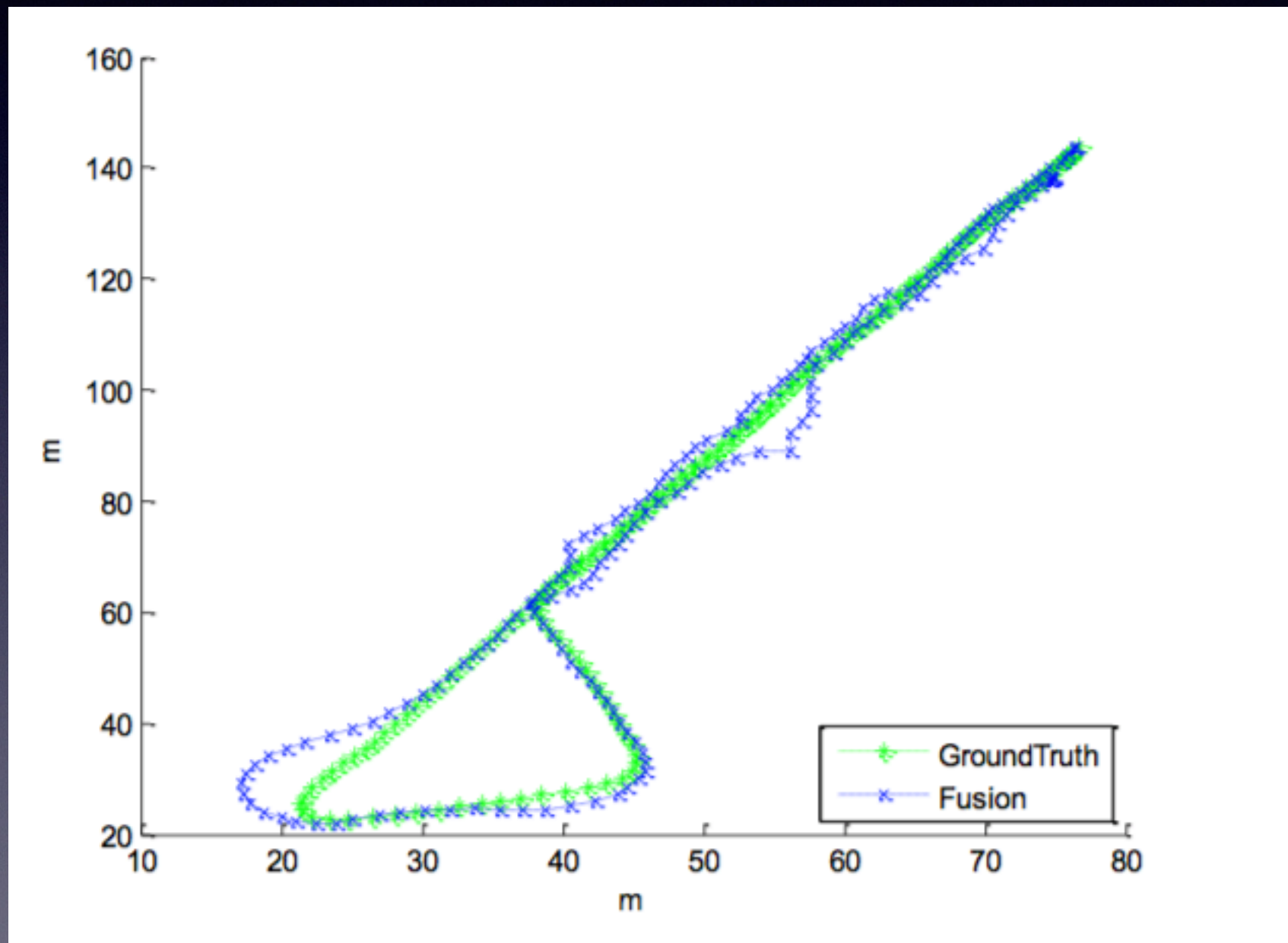
# Fusion of GPS and SLAM ( $\sigma_{GPS} = 10$ )





# Fusion of Wifi, GPS and SLAM

$(\sigma_{GPS} = 6, \sigma_{WiFi} = 6)$





# Conclusion

- “alternative solution for weak GPS area with WiFi Fingerprinting localization technique”
- future improvement: “deep learning... mutli-receiver”

