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The Scan Matchers Research and Comparison: Monte-Carlo, Olson and Hough



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A problem definition

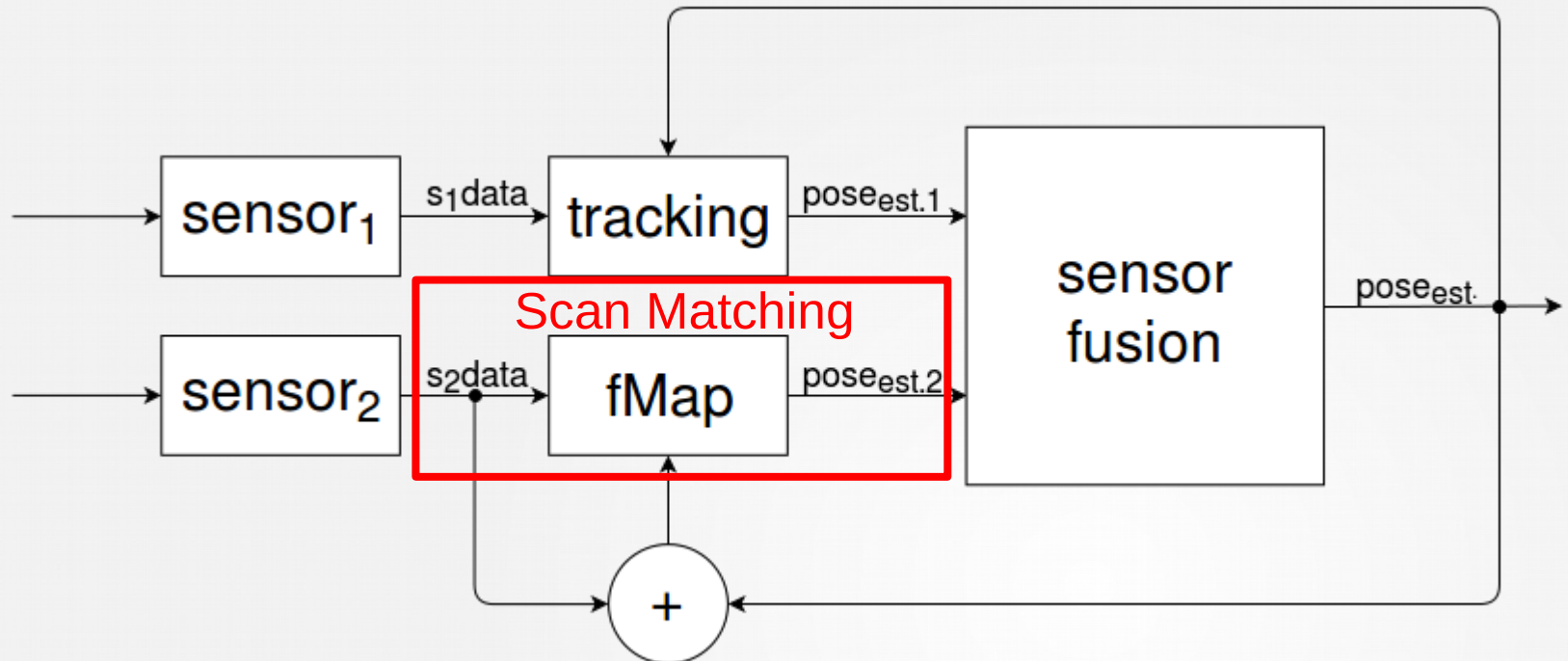


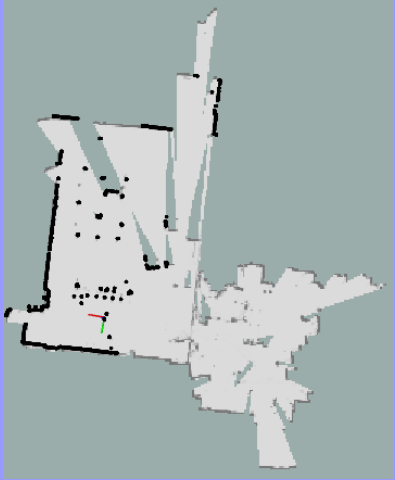


Fig. 1. Typical SLAM scheme

The problem of choosing scan matcher algorithm.

Is it possible to classify SMs?

2D Scan matching

	Input	Output
1) The map (prebuilt)	2) The laser view (from rangefinders)	1) The pose estimation
		
	3) The pose on the previous step 4) The odometry	

Problem

Methods

Results

Conclusion

Working without Scan matching

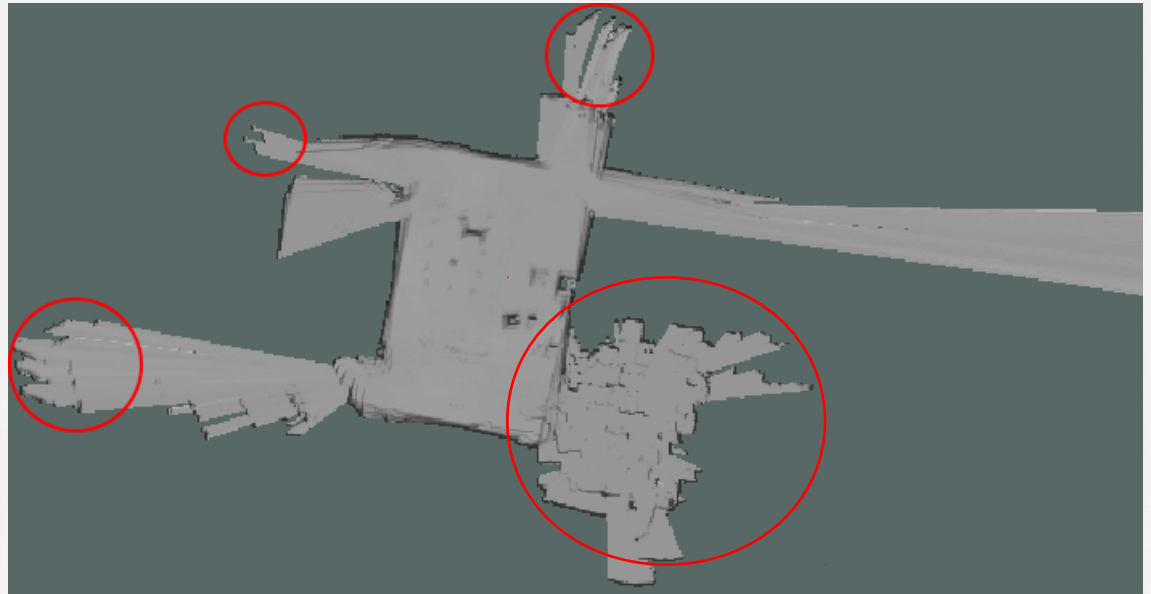
Problem

Methods

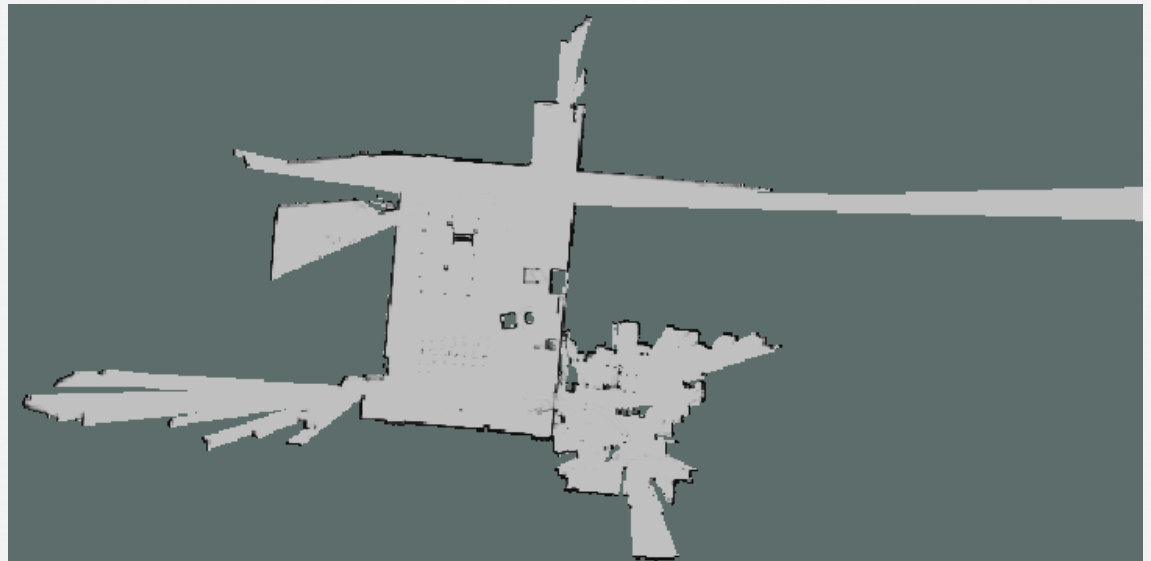
Results

Conclusion

Without SM

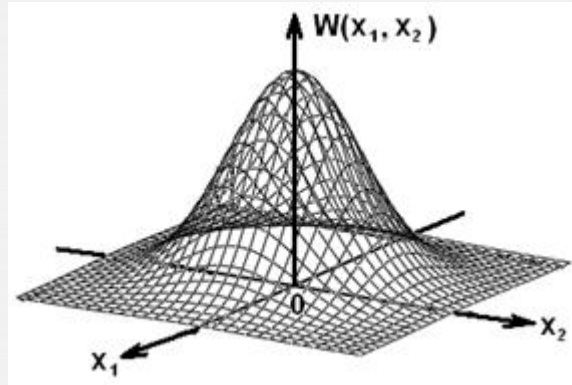


The absolute

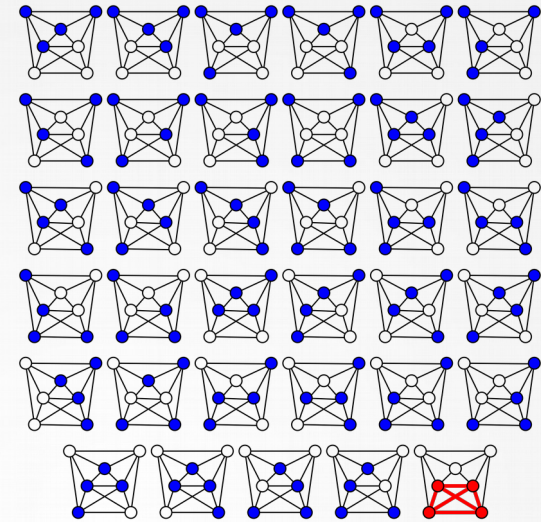


Tested Scan Matchers

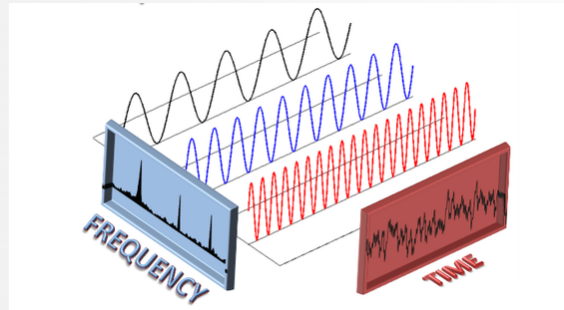
Monte-Carlo
(stochastic)



Olson
(brute-force)



Hough
(analysis-oriented)



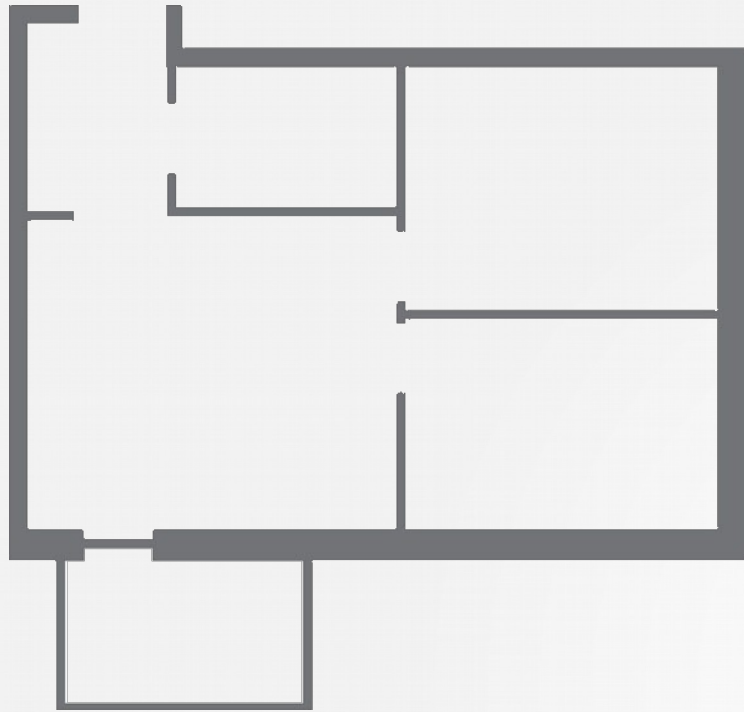
Problem

Methods

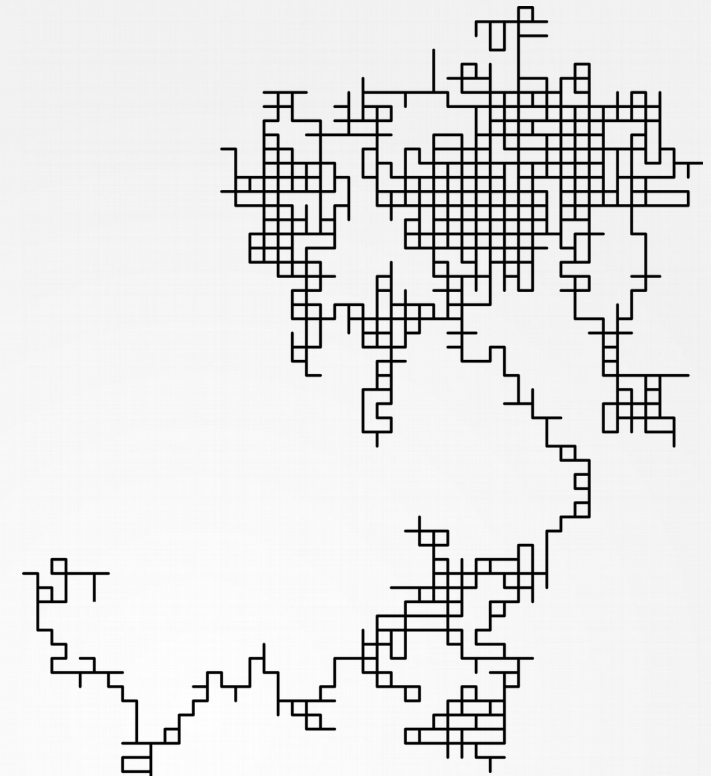
Results

Conclusion

The environment



a) The room-like world



b) The random world

c) Also tests were completed on a real data

Problem

Methods

Results

Conclusion

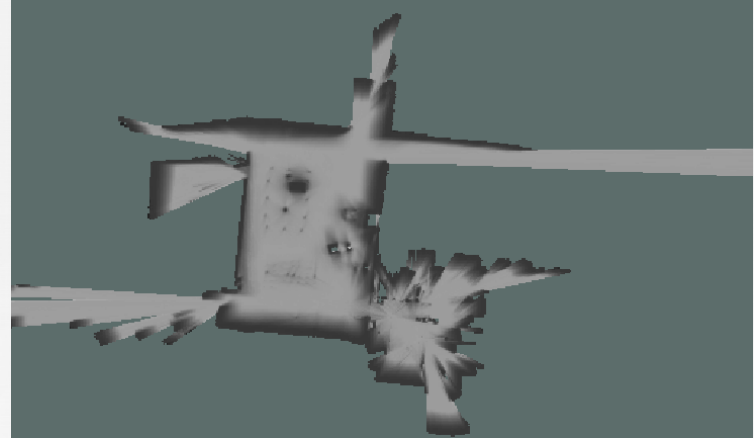
Results

Table II. The quantitative SM estimation

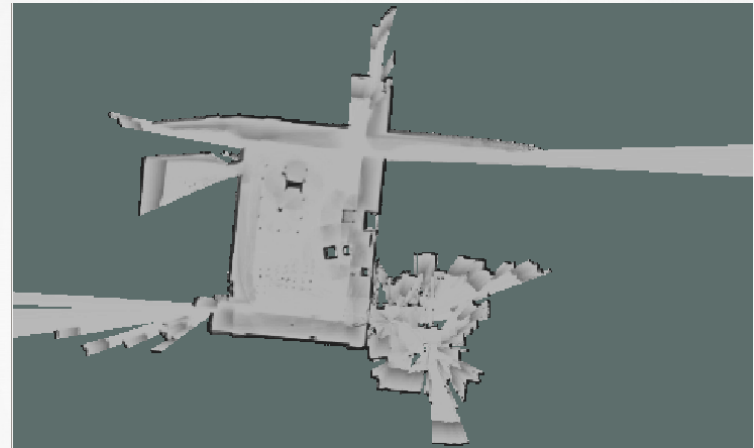
The scan matcher	The indoor world		The random world	
	Time, s	Error, m	Time, s	Error, m
Monte-Carlo	0,02	$\Delta x=0,011$ $\Delta y=0,009$ $\Delta \theta=0,02$	0,3	$\Delta x=0,08$ $\Delta y=0,09$ $\Delta \theta=0,05$
Olson	0,3	$\Delta x=0,001$ $\Delta y=0,001$ $\Delta \theta=0,001$	2,7	$\Delta x=0,0005$ $\Delta y=0,0004$ $\Delta \theta=0,001$
Hough	0,1	$\Delta x=0,01$ $\Delta y=0,01$ $\Delta \theta=0,008$	0,8	$\Delta x=0,02$ $\Delta y=0,02$ $\Delta \theta=0,016$

Built maps

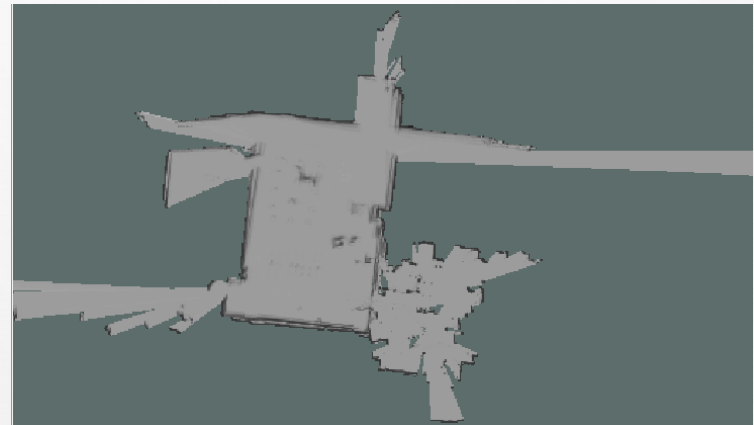
Monte-Carlo
(blured)



Olson
(weak blurred)



Hough
(not blurred)



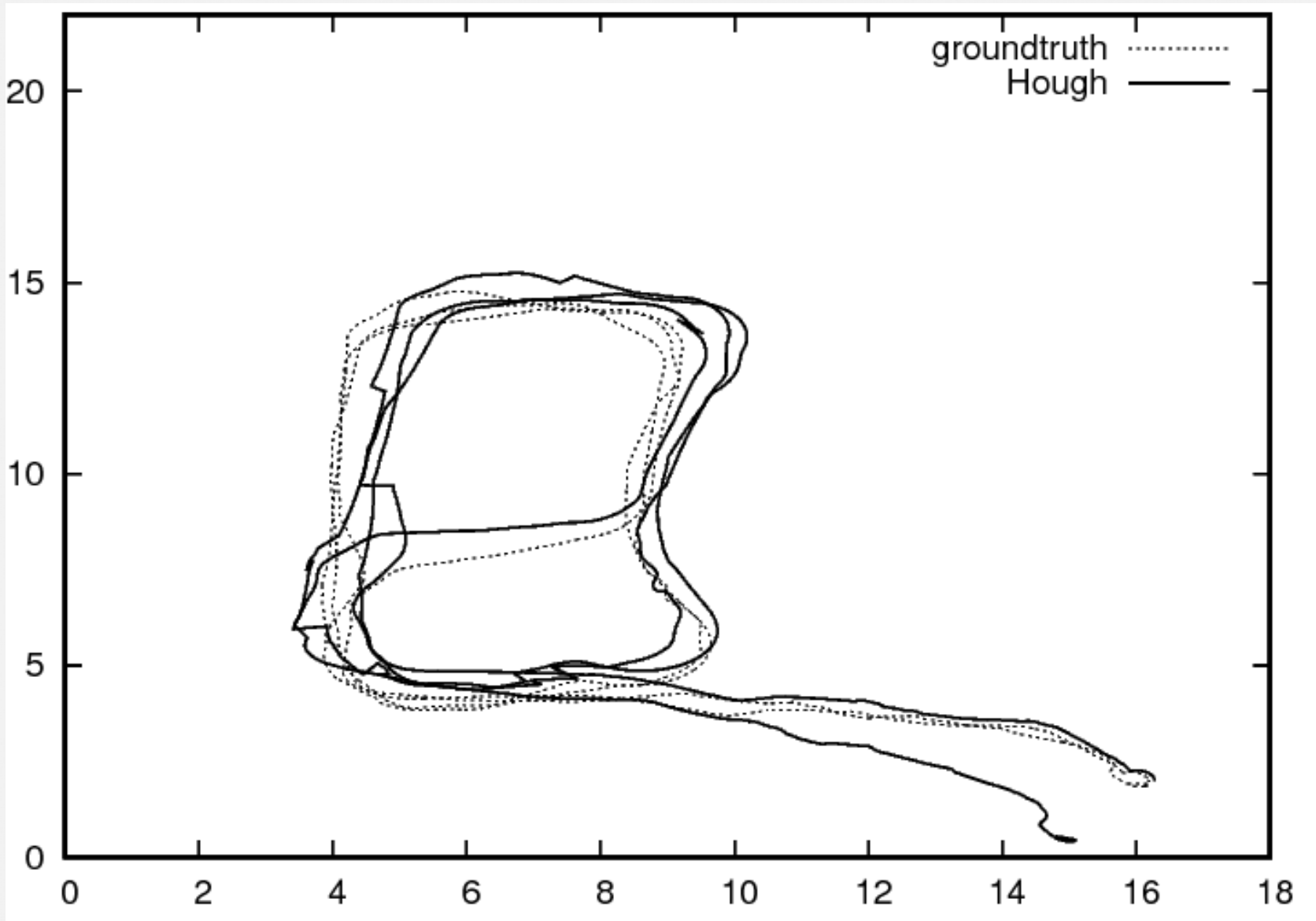
Problem

Methods

Results

Conclusion

The Hough Trajectory



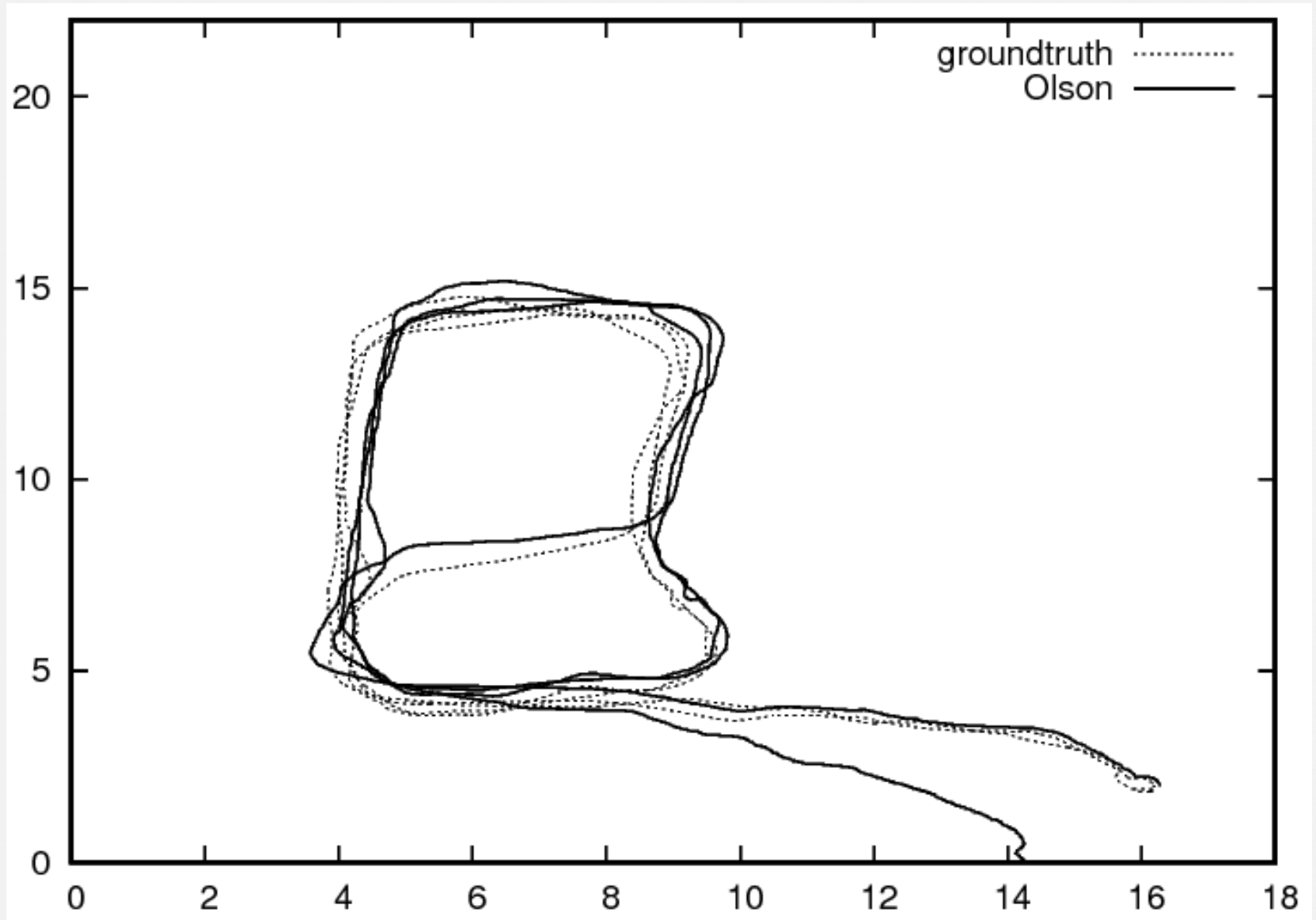
Problem

Methods

Results

Conclusion

The Olson Trajectory



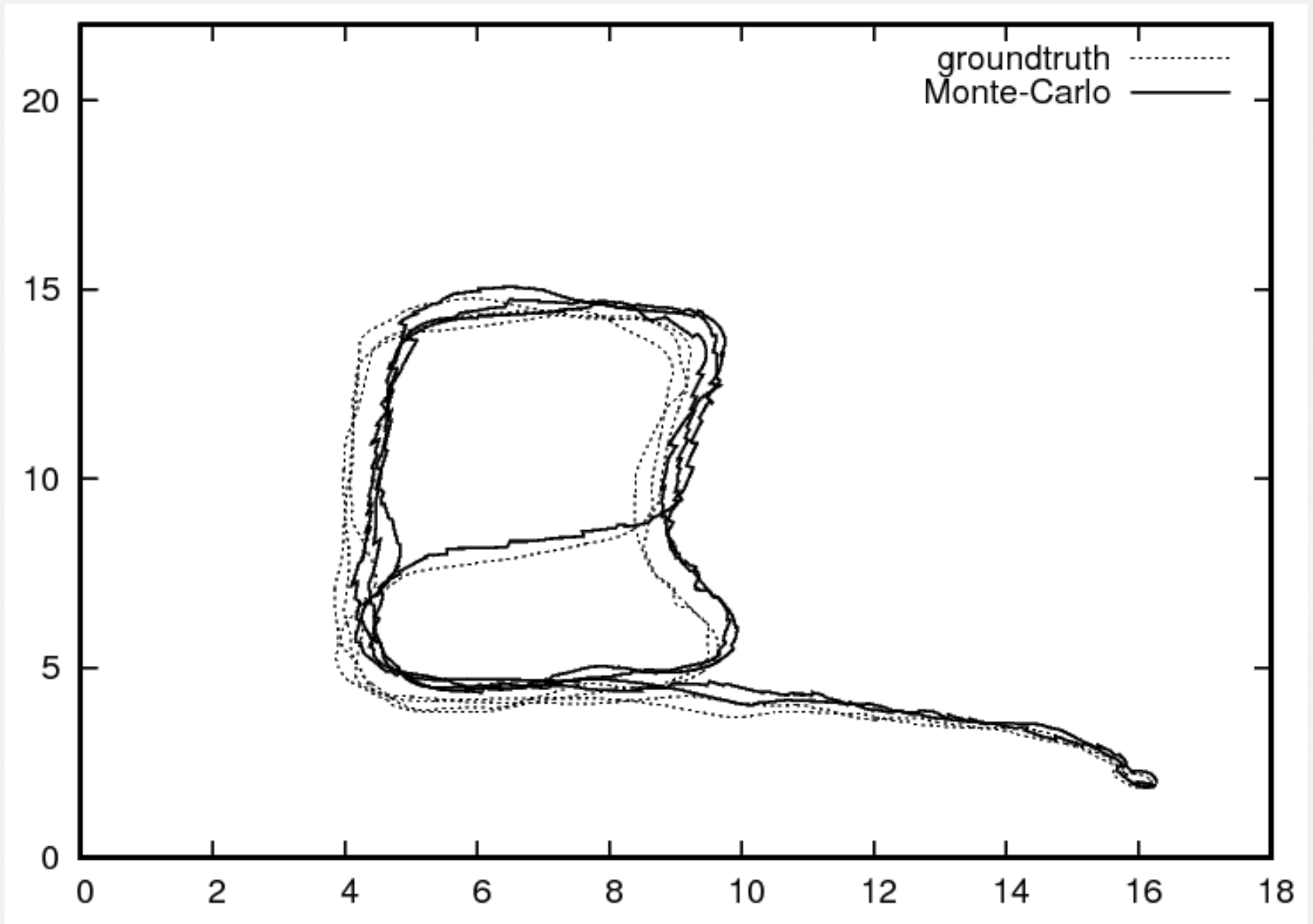
Problem

Methods

Results

Conclusion

The Monte-Carlo Trajectory



Problem

Methods

Results

Conclusion

Conclusion

Table III. SM comparison

	Monte-Carlo	Olson	Hough	
Problem				
Methods	Time	Fast	Slow	Moderate
Results	Teoretical estimation	Stochastic, unpredictable	Accurate	Error & time trade off
Conclusion	Real aplication	Accurate with a blurred world	Accurate with an own blurred table	Accumulates error

Future work

Problem

Methods

Results

Conclusion

- Include in SLAM constructor
github.com/OSLL/slam-constructor
- Apply to other SLAM algorithms
- Test on real robots

Thank you!