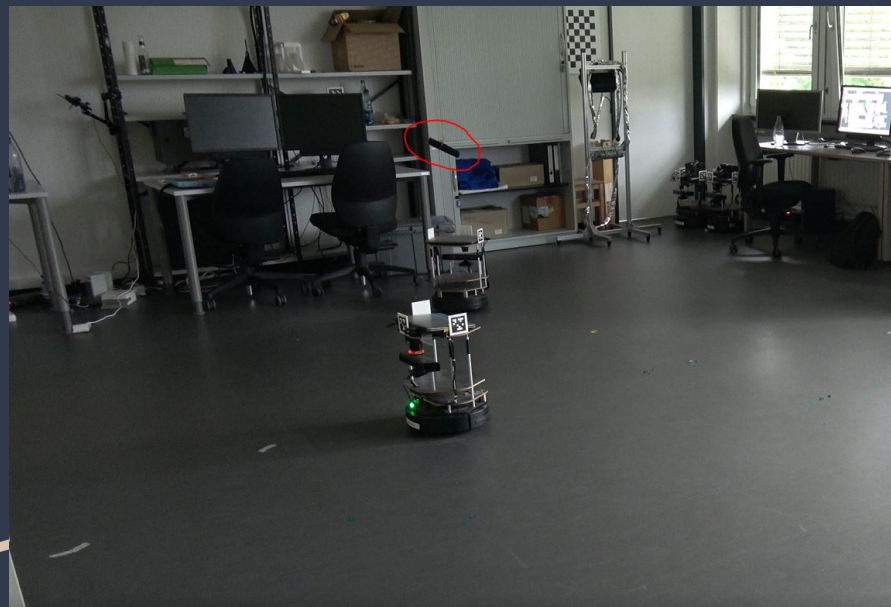


# Signal source localization using Dec-POMDP planning

By: Tobias Krüger

Supervised by: Dr. Mikko Lauri & Michael Görner

# What is the goal of this bachelor thesis ?



- Use heuristic **de**centralized - **P**artially **O**bservable **M**arkov **D**ecision **P**rocesses algorithm introduced by Dr. Lauri et al<sup>1</sup>
- Implement Information Gathering Task
  - Signal Source Localization
- Compare results and discuss viability of using Dec-POMDP on Information Gathering Tasks

<sup>1</sup>Information Gathering in Decentralized POMDPs by Policy Graph Improvement - Mikko Lauri, Joni Pajarinen and Jan Peters

# What is the goal of today's talk?

- Introduce basic principles of:
  - Dec-POMDP
  - Signal Source Localization
- Highlight interesting aspects of experiments
- Review and evaluate results
- Preview remaining work

# Presentation Agenda

1. **Related work**
2. Introduction to Dec-POMDP
3. Signal Source Localization
4. Experiments and Evaluation
5. Remaining work
6. Conclusion

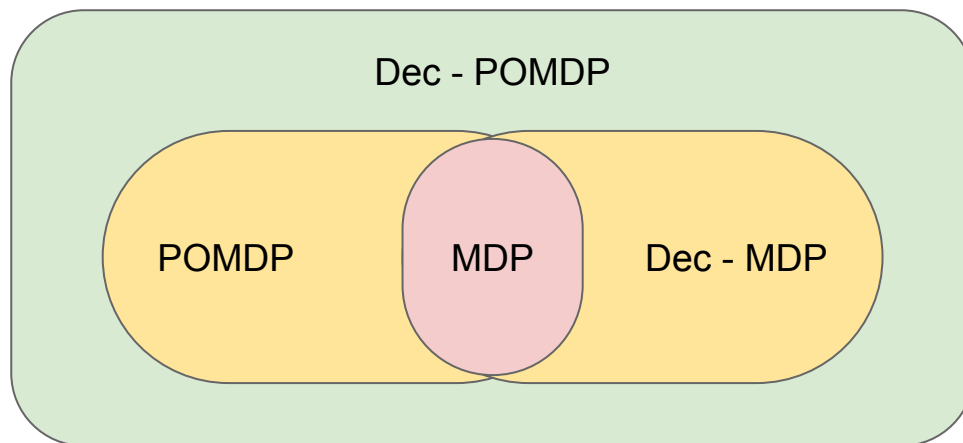
# Related Work

- Multi-agent active information gathering in discrete and continuous-state decentralized POMDPs by policy graph improvement - Paper by Lauri, M., Pajarinen, J., & Peters, J. (2020).
- Multi-modal Localization using Wi-Fi Signal Strength and 2D Range Finder - BSc by Benjamin Scholz (TAMS Group)
- Solving Multi-agent Decision Problems Modeled as Dec-POMDP: A Robot Soccer Case Study - By Okan Asik and H. Levent Akin
- Distributed Algorithms for Stochastic Source Seeking with Mobile Robot Networks: Technical Report - by Nikolay A. Atanasov, Jerome Le Ny and George J. Pappas

# Presentation Agenda

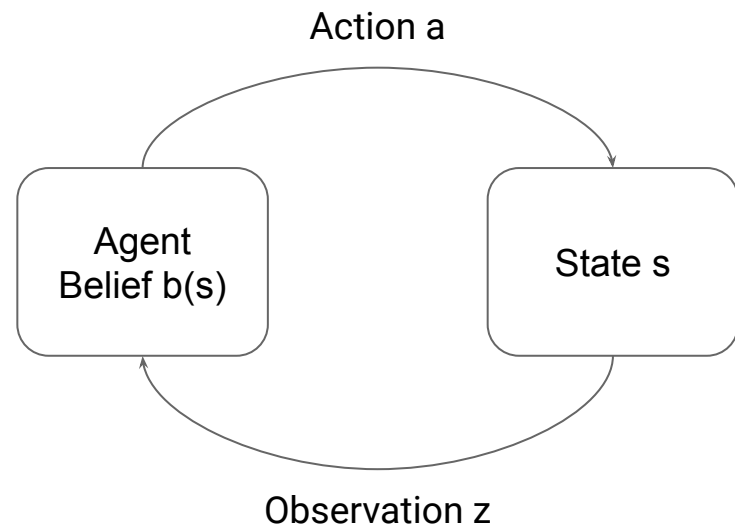
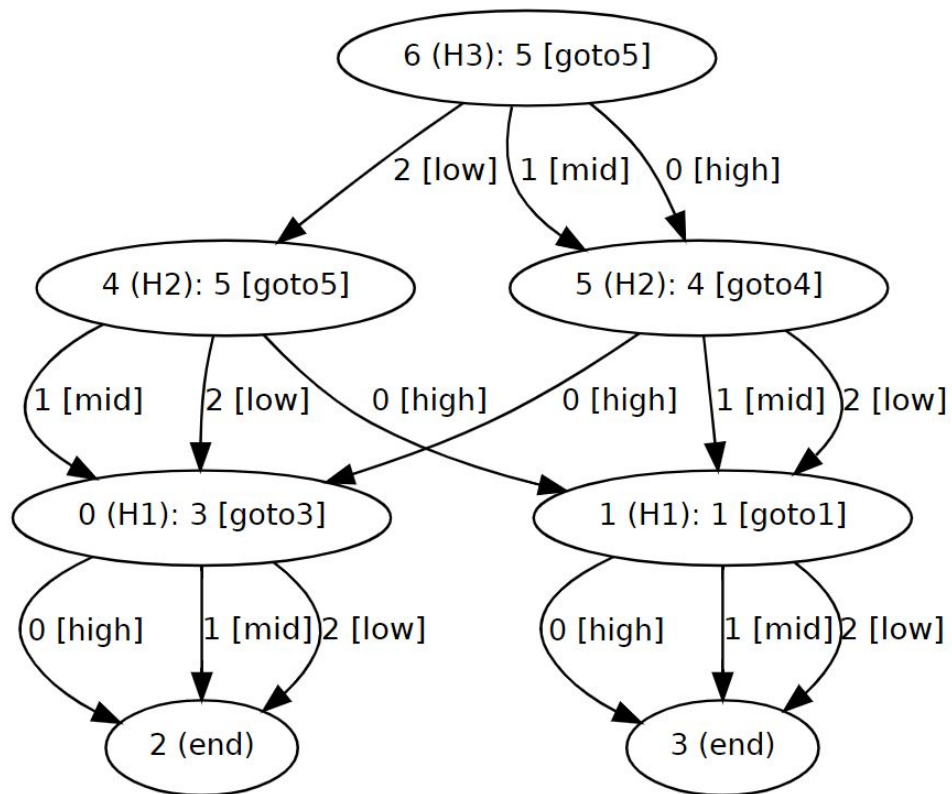
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# Decentralized Partially Observable Markov Decision Processes



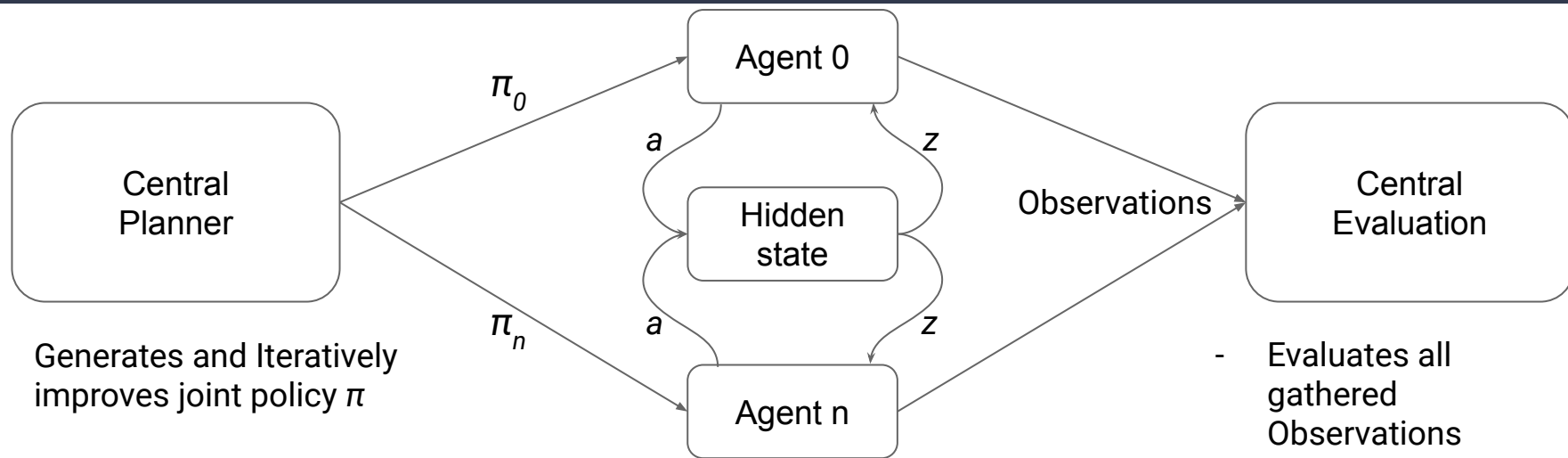
- Problem solving frameworks for one or more agents acting in an environment
- Known State-Space  $S$  - Set of all possible states
- Agents  $I = \{1, \dots, n\}$ 
  - Knows current state  $s$  or estimates it as a belief state
  - Can take action  $a_i^t \in A_i = \{a_i^1, \dots, a_i^t\}$  according to policies  $\pi_i$  at time step  $t$
  - Perceive observation  $z_i^t \in Z_i = \{z_i^1, \dots, z_i^t\}$  at time step  $t$
  - Receive Rewards according to reward function  $R(s^t, a^t)$

# Policies





# Solving a task using dec-POMDP



- Generates and Iteratively improves joint policy  $\pi$

- $\pi = (\pi_0, \dots, \pi_n)$

- Reward function determines value of joint Policy in Simulation

- Agent  $i$  executes Policy  $\pi_i$

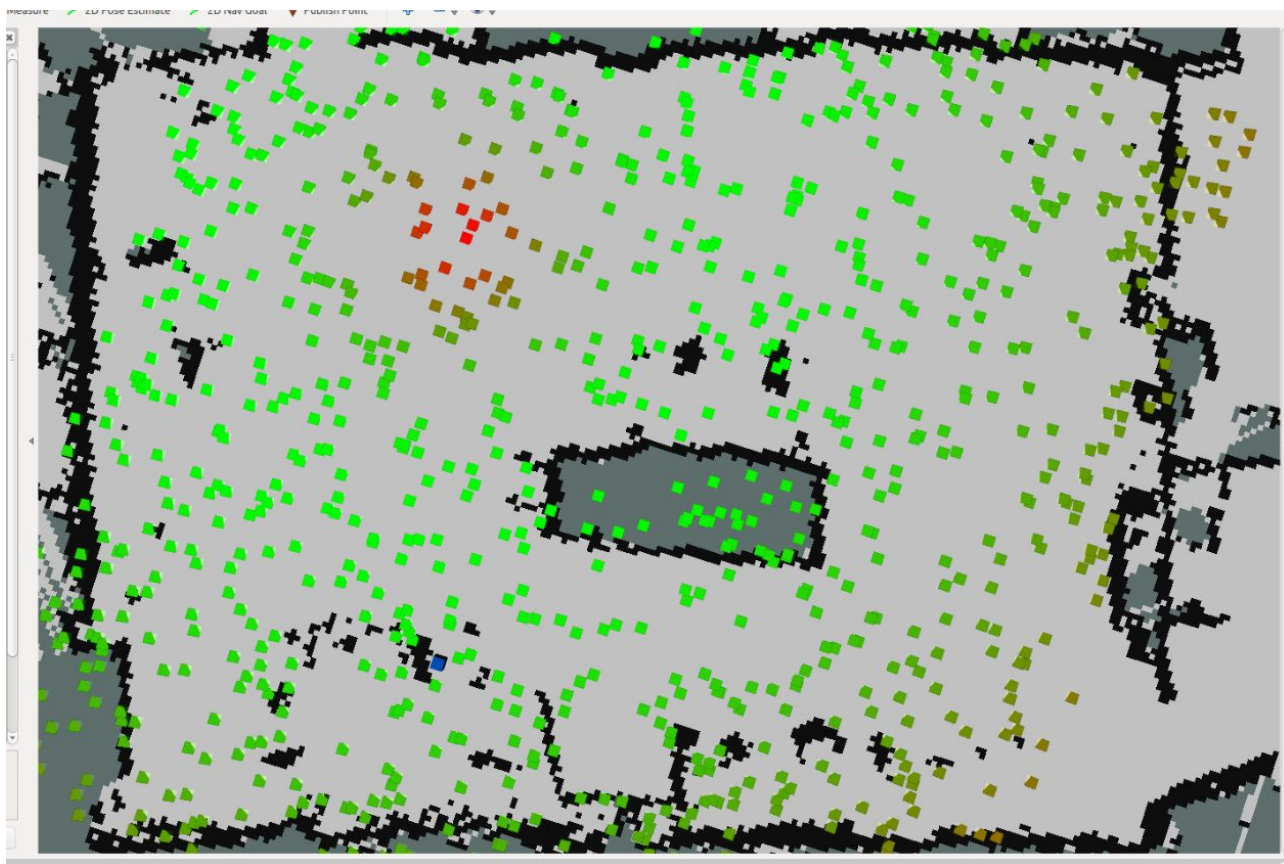
- Accumulate observations

- Evaluates all gathered Observations
- Determine result for completed task

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# Particle Filter



# Received Signal Strength (RSS) model

$$P_{rx}(x,y) = P_{tx} + G_{tx} - L_{tx} + G_{rx} - L_{rx} - L_{fs}(x,y) - L_m(x,y) - R(x,y) \quad (1)$$

$$L_{fs}(x,y) = 20 \log_{10}(d) + 20 \log_{10}(f) - 147, 55$$

$G_{rx} - L_{rx}$

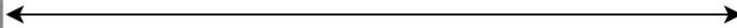
$R(x,y) = \text{Rician}(\mu, \sigma)$

$P_{tx} + G_{tx} - L_{tx}$



*Location x*

$- L_{fs}(x,y) - L_m(x,y)$



Huawei  
P20 Lite

$- R(x,y)$

*Location y*

# Received Signal Strength (RSS) model

$$P_{rx}(x,y) = P_{tx} + G_{tx} - L_{tx} - L_{fs}(x,y) - L_m(x,y) - R(x,y) \quad (1)$$

$$L_{fs}(x,y) = 20 \log_{10}(d) + 20 \log_{10}(f) - 147,55$$

$G_{rx} - L_{rx}$

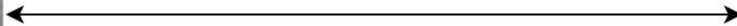
$$R(x,y) = \text{Rician}(\mu, \sigma)$$

$P_{tx} + G_{tx} - L_{tx}$



Location x

$$- L_{fs}(x,y) - L_m(x,y)$$



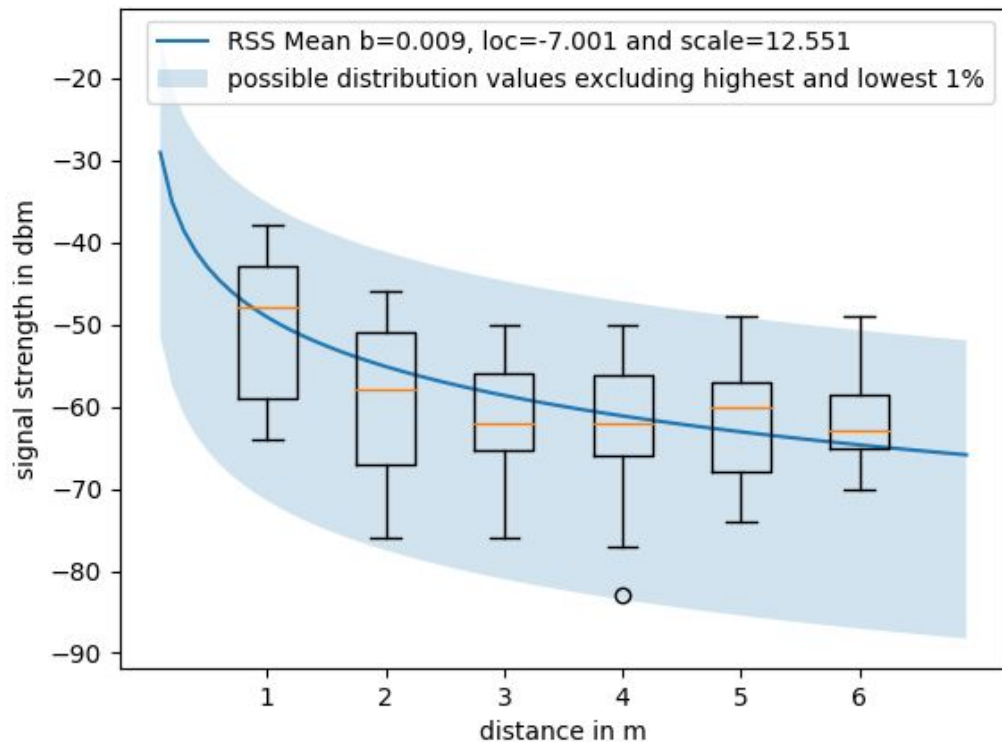
$$- R(x,y)$$



Huawei P20 Lite

Location y

# Received Signal Strength (RSS) model

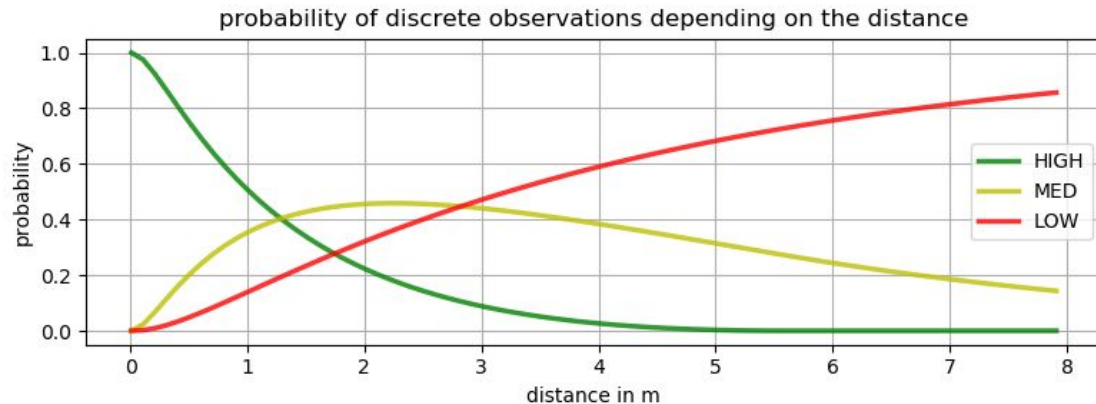
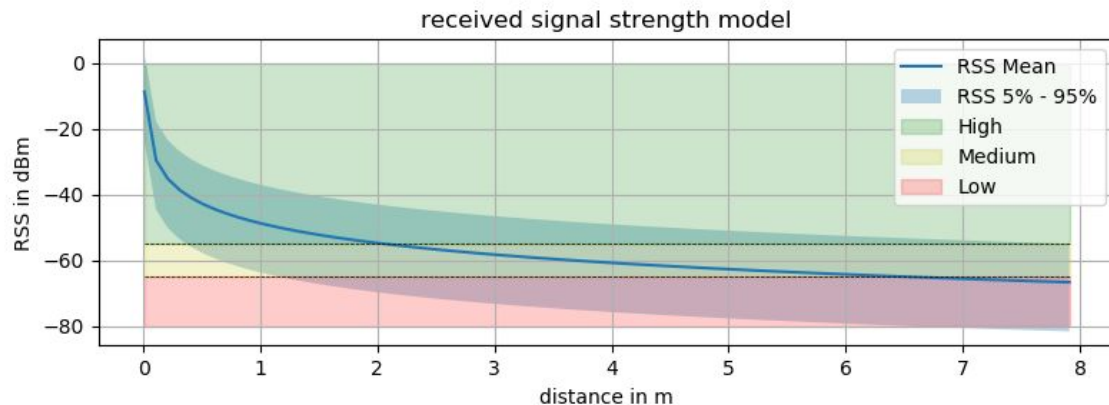


# Received Signal Strength (RSS) model

0 dBm > High  $\geq$  -55 dBm

-55 dBm > Medium  $\geq$  -65 dBm

-65 dBm > Low



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# Overview of Conducted Experiments

- **One Robot in Simulation**
  - Purpose:
    - Testing components / RSS model
  - Execution:
    - Dec-POMDP Planning for two Robots (one Policy voided)
    - Simulated measurements
- **Two Robots in a real environment**
  - Two Robot Limit given by Dec-POMDP Algorithm
  - Policies generated directly before experiment
- **Two Robots with random movements in a real environment**
  - Purpose:
    - Drawing a baseline
  - Execution:
    - Robots move between predefined locations at random
    - Predefined event horizon

# Example Experiment

- **VIDEO of Example Experiment**
  - 5 Measurements per Action
  - 3 Actions per Robot
  - 4 Improvement steps
  - 2 wide Policies

# Comparison between Random and dec-POMDP

- Experiment 1 using  
Dec-POMDP Planning

- 5 Steps per Robot

- 3 Planning  
Improvement Steps

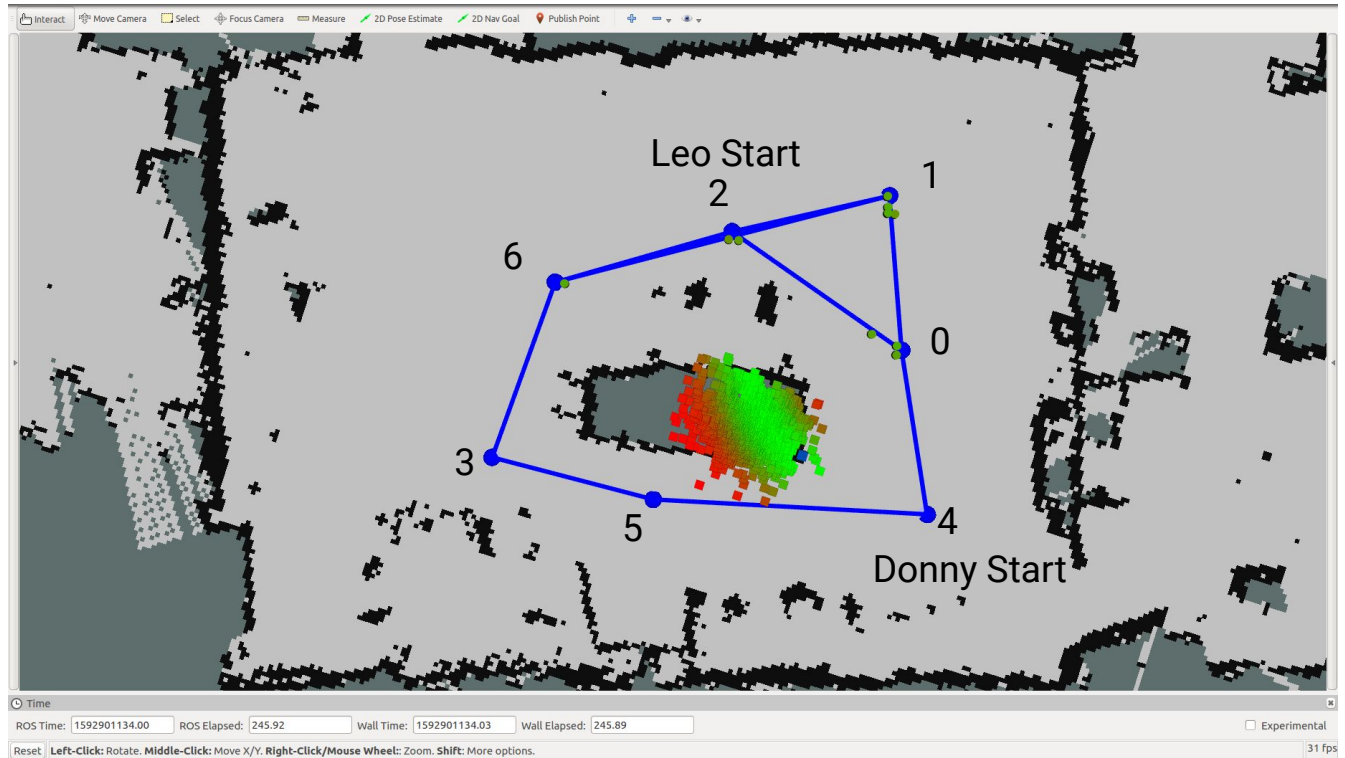
- Planning time ~ 4min

- Leo moves

$6 \rightarrow 1 \rightarrow 2 \rightarrow 1 \rightarrow 0$

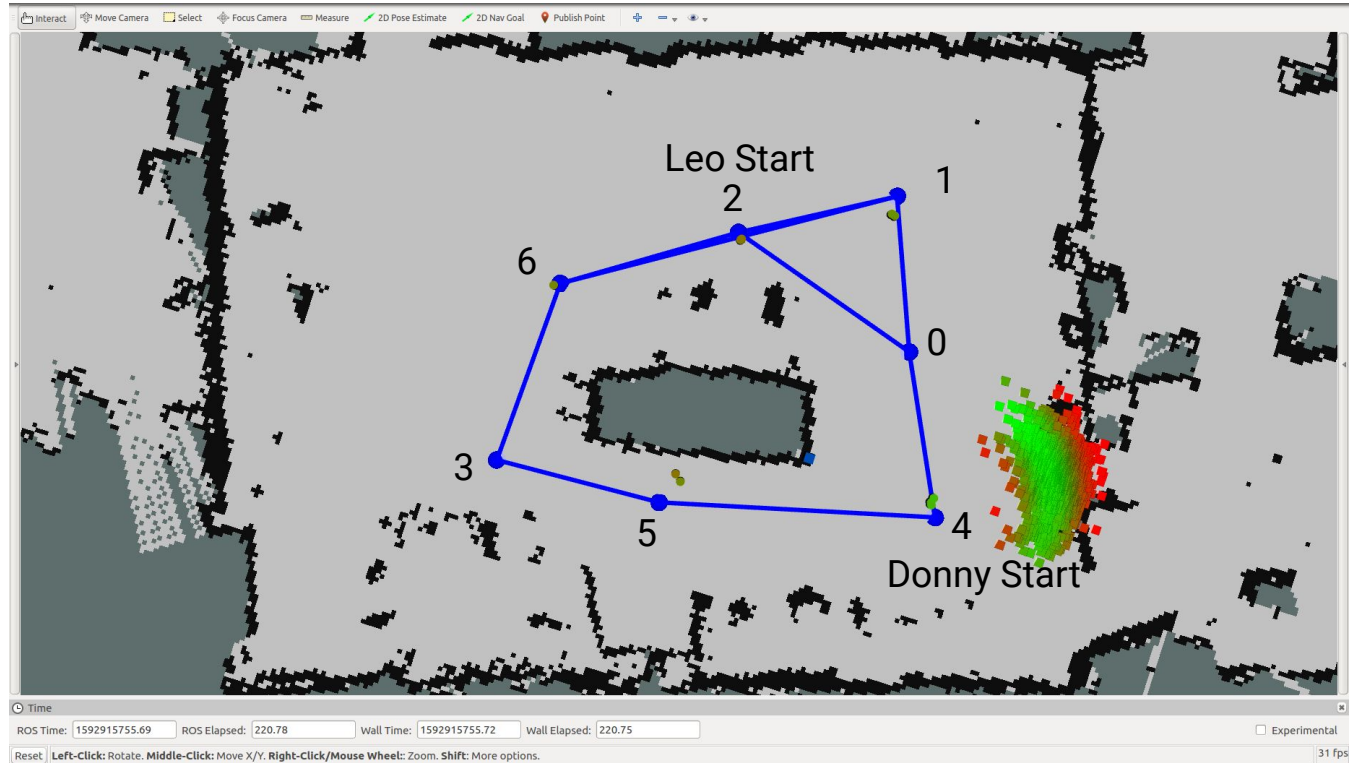
- Donny moves

$0 \rightarrow 1 \rightarrow 0 \rightarrow 2 \rightarrow 1$



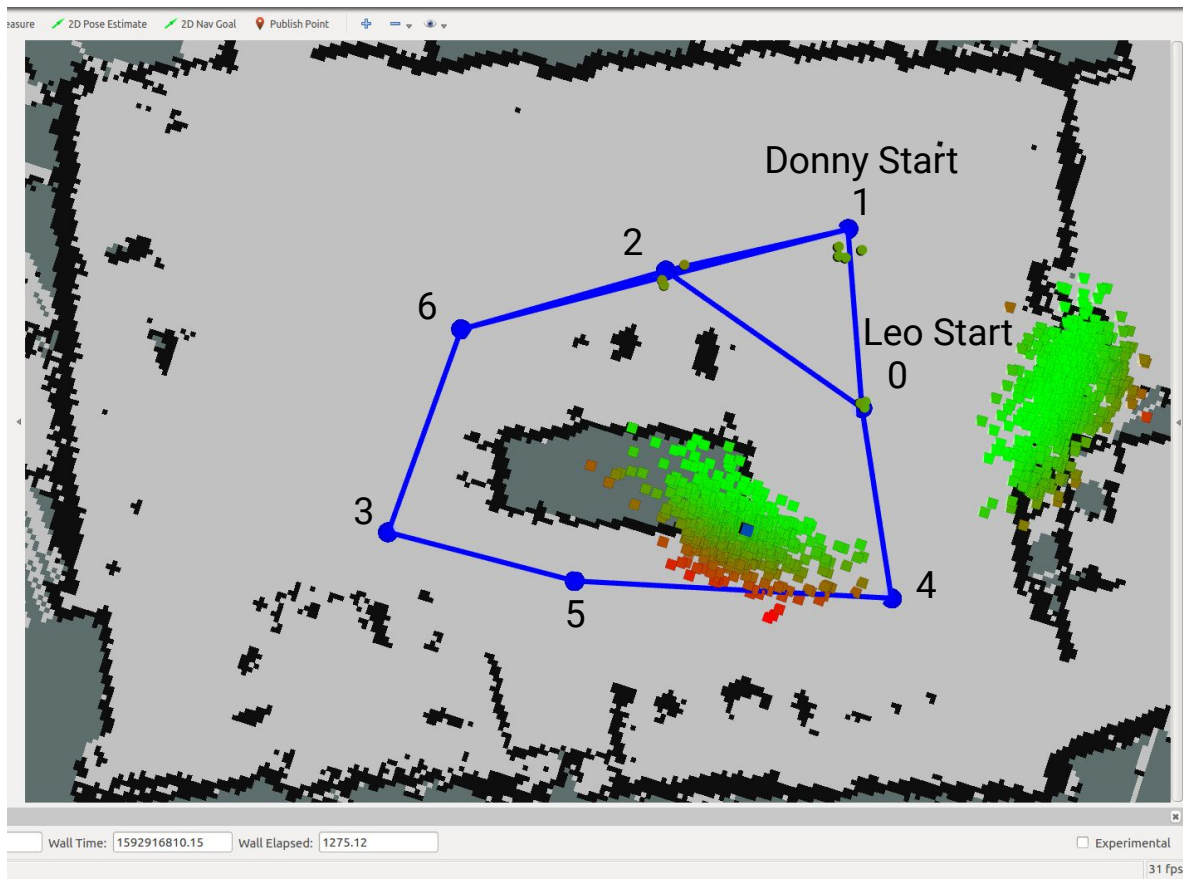
# Comparison between Random and dec-POMDP

- Experiment 1 using Random Movement
- 5 Steps per Robot
- Leo moves  
 $2 \rightarrow 6 \rightarrow 1 \rightarrow 2 \rightarrow 1$
- Donny moves  
 $4 \rightarrow 5 \rightarrow 4 \rightarrow 5 \rightarrow 4$

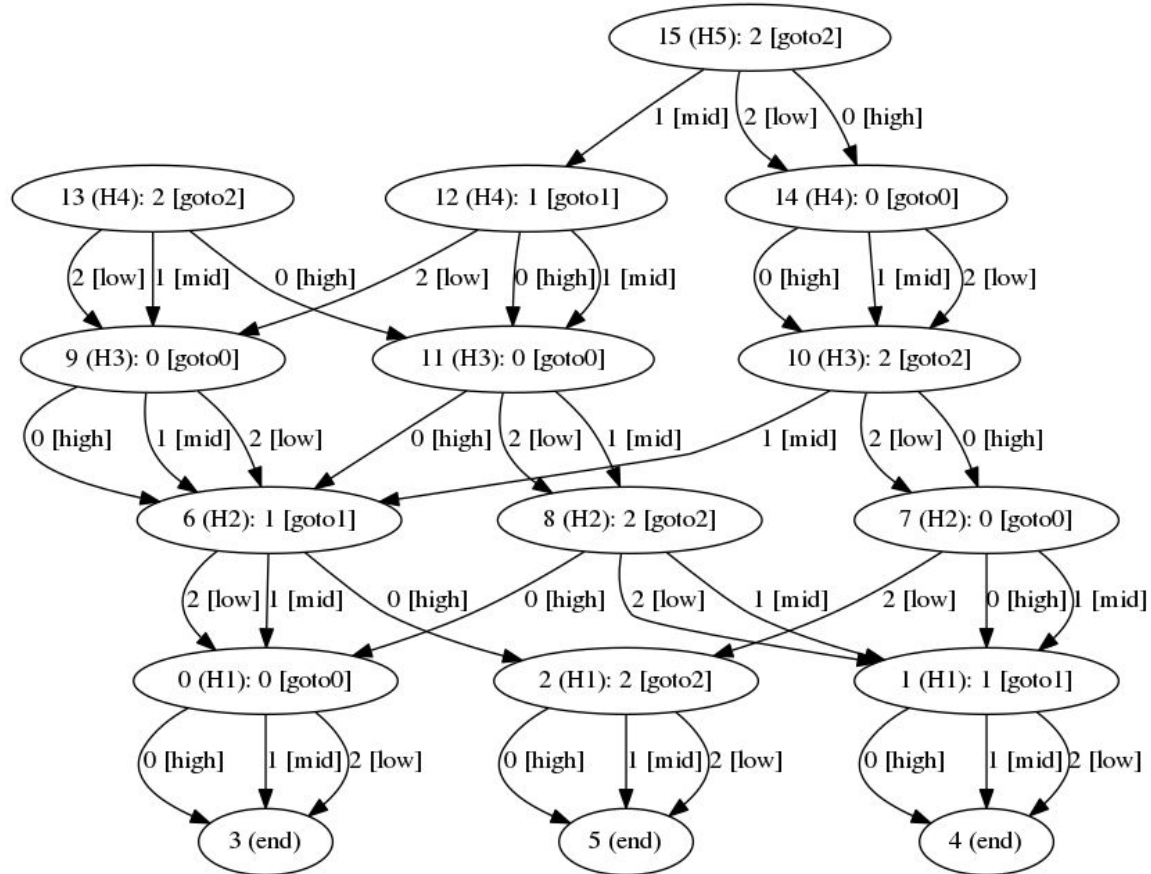


# Comparison between Random and dec-POMDP

- Experiment 2 using Dec-POMDP Planning
- 5 Steps per Robot
- 9 Planning Improvement Steps
- Planning time ~ 10min
- Leo moves  
 $2 \rightarrow 1 \rightarrow 0 \rightarrow 1 \rightarrow 0$
- Donny moves  
 $2 \rightarrow 1 \rightarrow 0 \rightarrow 2 \rightarrow 1$

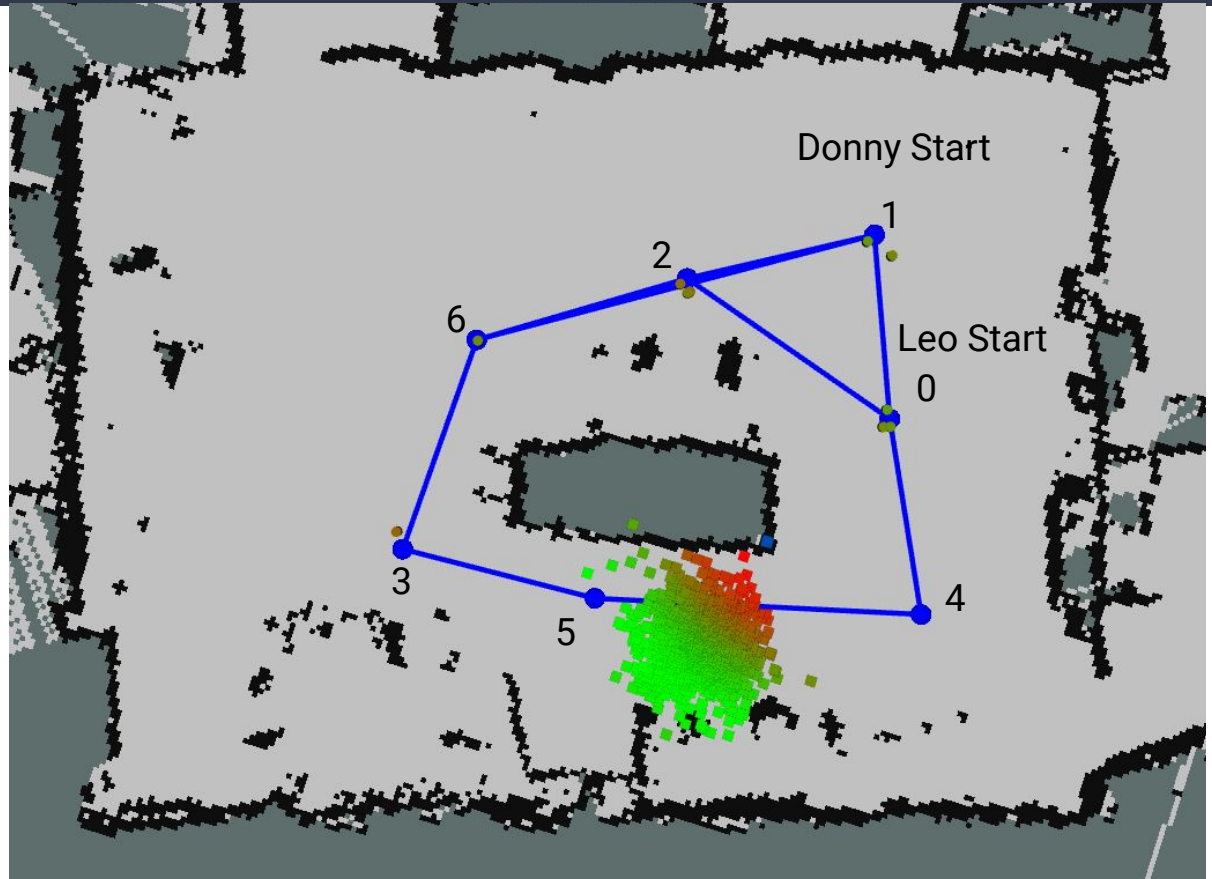


# Policy for Donny in Experiment 2



# Comparison between Random and dec-POMDP

- Experiment 2 using Random Movement
- 5 Steps per Robot
- Leo moves  
 $0 \rightarrow 1 \rightarrow 2 \rightarrow 6 \rightarrow 3$
- Donny moves  
 $1 \rightarrow 2 \rightarrow 0 \rightarrow 2 \rightarrow 0$

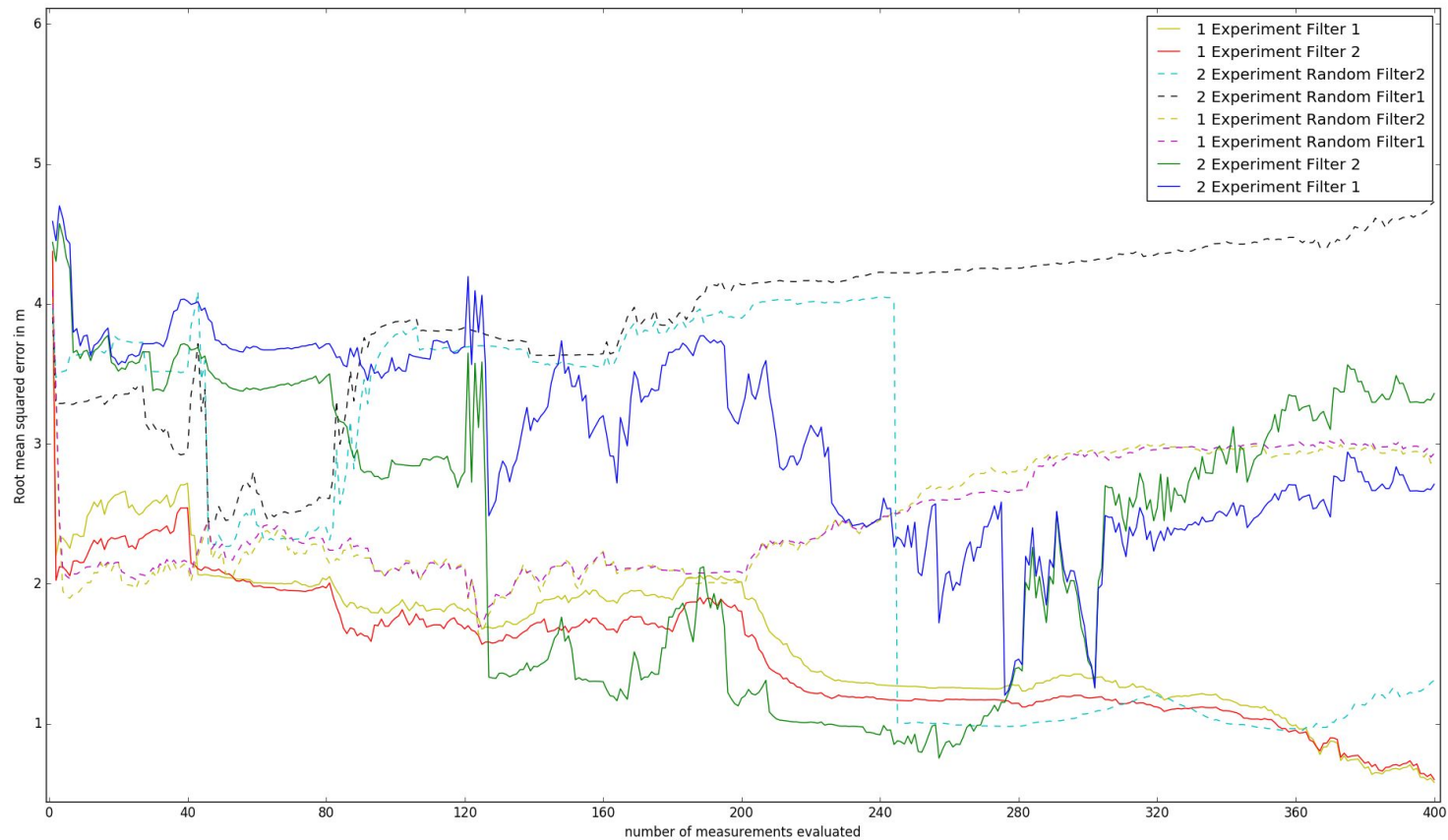


# Comparison Criteria of Particle Filter Results

$$\textit{Weighted\_Error} = \sqrt{\frac{\sum_{i=1}^n w_i d(\textit{loc}_{real}, \textit{loc}_i)^2}{\sum_{i=1}^n w_i}}$$



# Comparison between Random and dec-POMDP



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# Remaining work

- Conduct more Experiments to compare Random Movement to dec-POMDP Planning
  - Less steps more possible locations
- Try out alternatives to the particle Filter for Evaluation
  - Fixed Grid
- Potentially change the variance of the RSS Model to reduce the effects of noise on the particle filter

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# Conclusion

- The heuristic dec-POMDP algorithm introduced by Mikko Lauri has successfully been applied to a real world example
- Two Turtle Bots can be used to successfully approximate the position of a Wi-Fi Signal Source
- Experiments so far indicate an advantage of using dec-POMDP over random movement

# Sources

## Dec-POMDP Information:

- D.S. Bernstein, S. Zilberstein and N. Immerman “The Complexity of Decentralized Control of Markov Decision Processes” in Journal of Mathematics of Operations Research, vol 27, no. 4, pp819-840, November 2002
- Lauri, M., Pajarinen, J. & Peters, J. Multi-agent active information gathering in discrete and continuous-state decentralized POMDPs by policy graph improvement. Auton Agent Multi-Agent Syst 34, 42 (2020).  
<https://doi.org/10.1007/s10458-020-09467-6>
- Thrun, Sebastian, Wolfram Burgard, and Dieter Fox (2005). Probabilistic Robotics. The MIT Press.

## Signal Source Localization:

- Atanasov, Nikolay, Jerome Le Ny, and George J. Pappas (2014). Distributed Algorithms for Stochastic Source Seeking with Mobile Robot Networks: Technical Report.

## Related Works:

- Multi-modal Localization using Wi-Fi Signal Strength and 2D Range Finder - BSc by Benjamin Scholz (TAMS Group)
- Solving Multi-agent Decision Problems Modeled as Dec-POMDP: A Robot Soccer Case Study - By Okan Asik and H. Levent Akin

## Tools Used:

- Quigley, M.; Gerkey, B.; Conley, K.; Faust, J.; Foote, T.; Leibs, J.; Berger, E.; Wheeler, R.; Ng, A. ROS: an open-source Robot Operating System. In Proceedings of the ICRA workshop on open source software,2009.

# Impact of robot orientation

