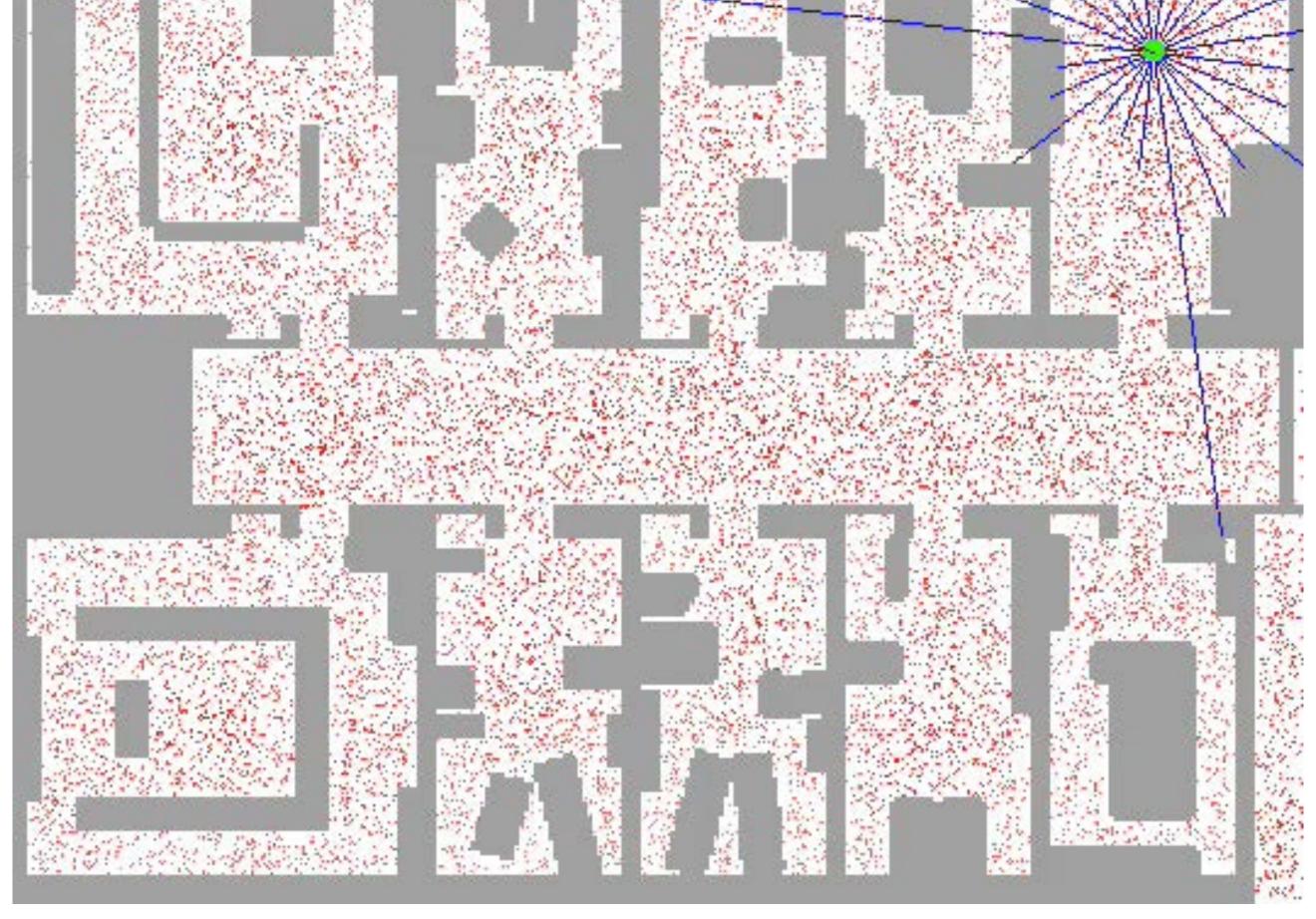
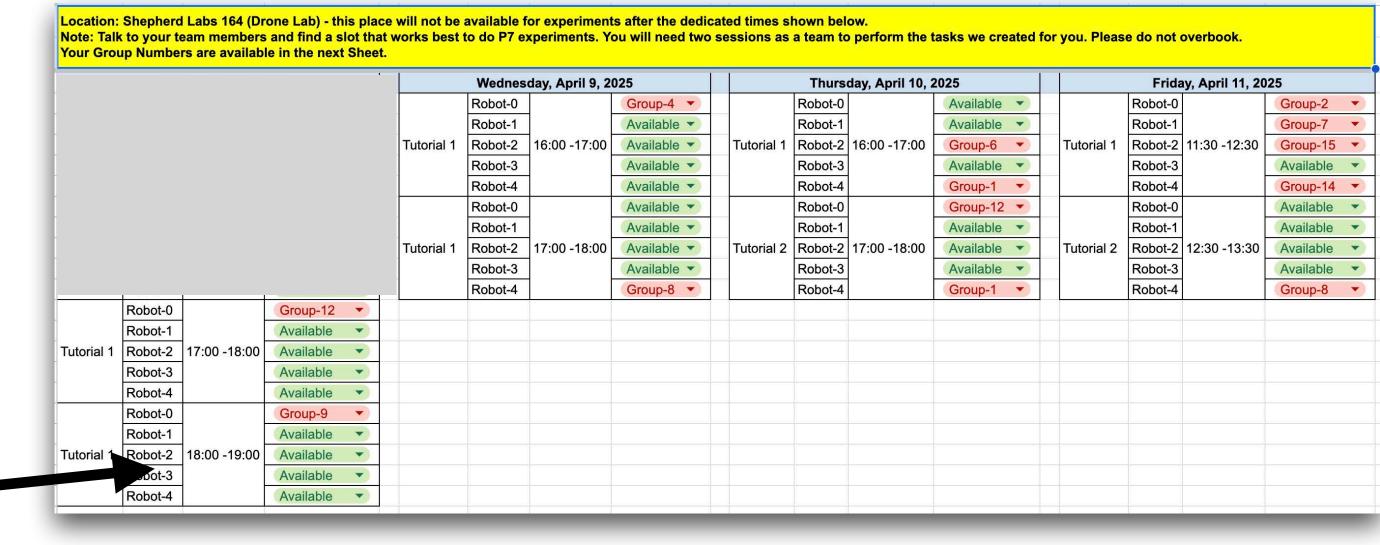
## Lecture 20 Mobile Robotics - V Localization





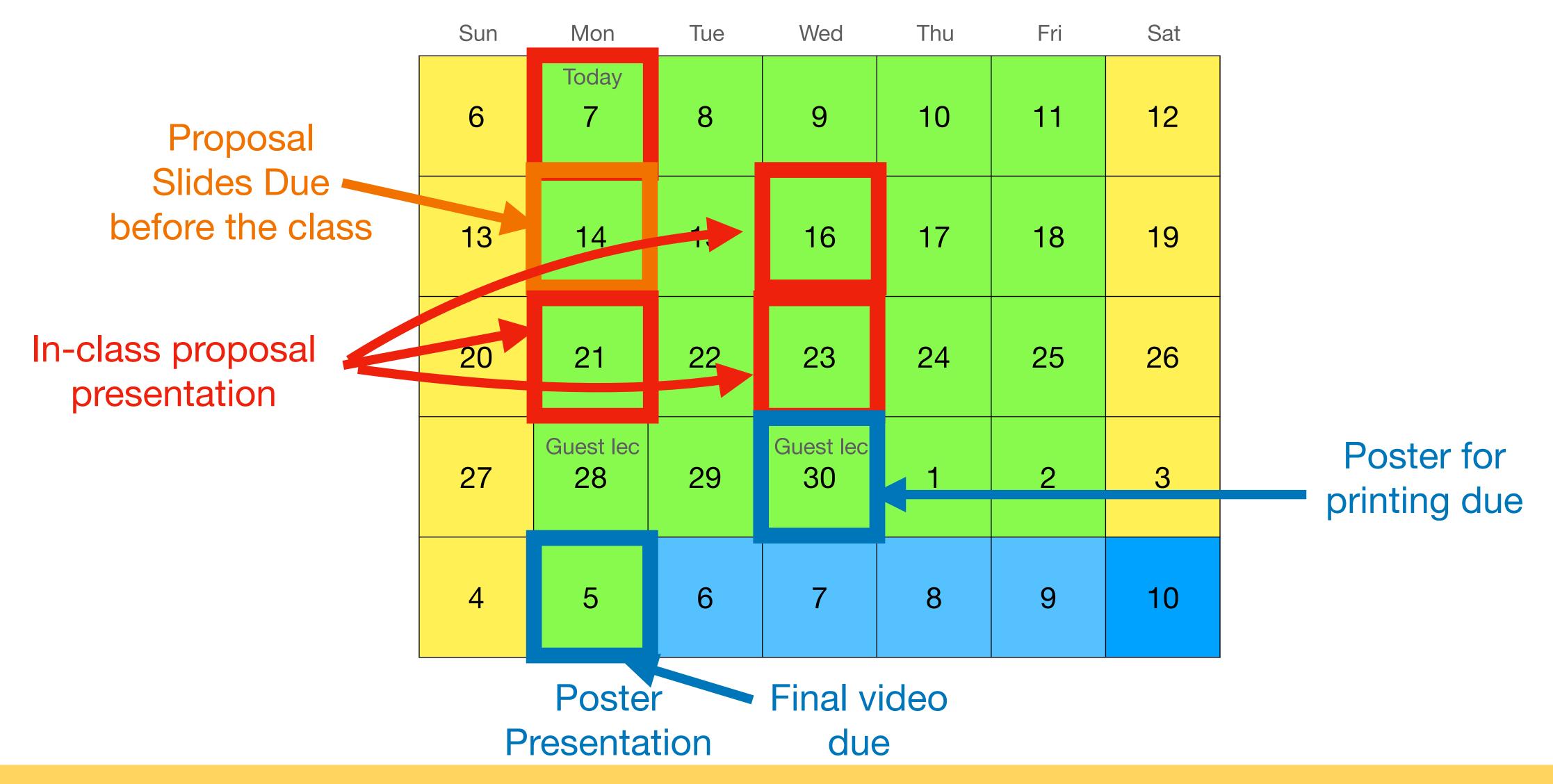
### Course logistics

- Quiz 10 will be posted tomorrow and will be due on Wed noon.
- Project 7:
  - Groups are formed.
  - Two parts (~1 hr each) Instructions will be provided.
    - 1. Tutorial 1
    - 2. Tutorial 2
  - Scheduler is shared with the class.
    - Please book your 2 1-hour sessions ASAP.
    - Both the parts needs to be completed by 04/23 as a team.
- No TA OHs between 04/07 and 04/23.
  - They will be available on demand.
  - Karthik's OH will be available to discuss final projects.
- Final Poster Session: 05/05/2025 Monday 12:30pm 2:30pm, Shepherd Labs 164 mark your calendars





### Final (Open) Project timeline





### Final (Open) Project timeline

- Proposal Slides: (template is provided)
  - 1-4 Slides
  - Title, Motivation, Input Output, Evaluation, Deliverables, Timeline, Who is doing what?
  - Where does your project stand not the 3-axes (robots, objects, tasks)?
  - Backup plan
- In-class proposal presentation (<8mins):</li>
  - Teams will get feedback from the class
- Final video:
  - Describing the project idea and the outcome.
- Poster presentation: (template will be provided)
  - Presenting the project idea and the outcome to audience.

#### Final Project: 15%

- Project proposal slides + presentation: 3%
- Final project video: 6%
- Poster presentation (evaluation by judges): 6%



### Continuing previous Lecture PF and localization



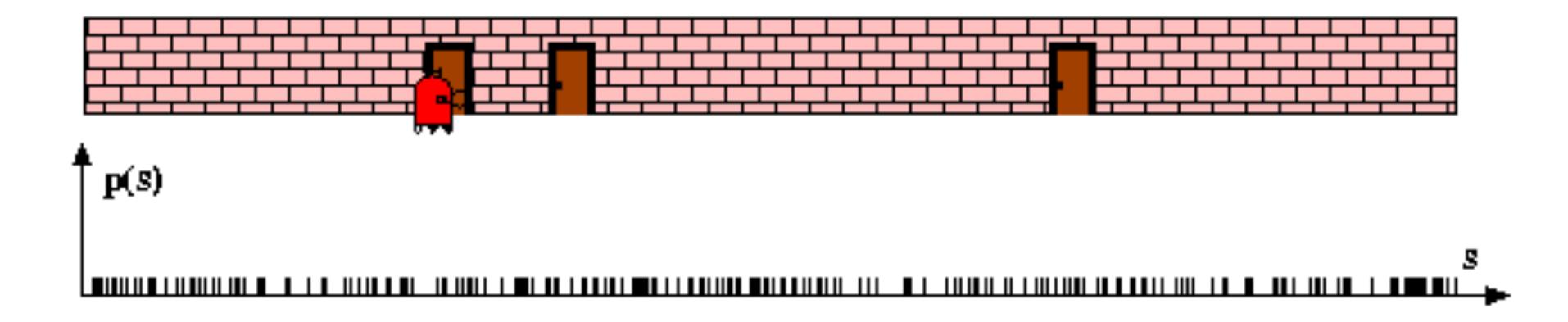
#### **Particle Filter**

```
Particle_filter(\mathcal{X}_{t-1}, u_t, z_t):
        ar{\mathcal{X}}_t = \mathcal{X}_t = \emptyset
     for j = 1 to J do
          sample x_t^{[j]} \sim \pi(x_t)
                 w_t^{[j]} = \frac{p(x_t^{[j]})}{\pi(x_t^{[j]})}
                  \bar{\mathcal{X}}_t = \bar{\mathcal{X}}_t + \langle x_t^{[j]}, w_t^{[j]} \rangle
5:
6:
            endfor
            for j = 1 to J do
                   draw i \in 1, \ldots, J with probability \propto w_t^{[i]}
8:
                   add x_t^{[i]} to \mathcal{X}_t
9:
             endfor
             return \mathcal{X}_t
```

### Particle Filter for Localization

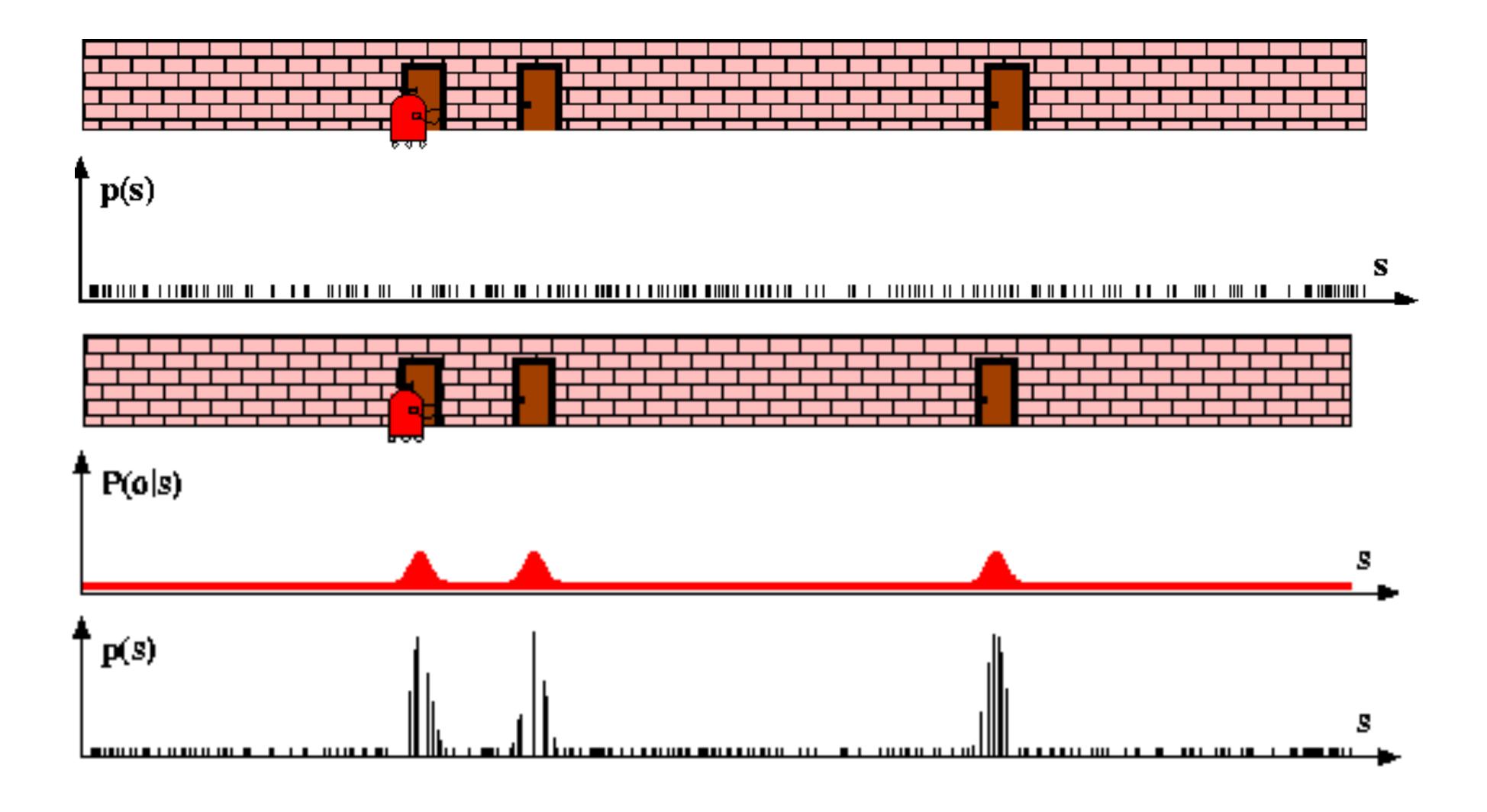
```
Particle_filter(\mathcal{X}_{t-1}, u_t, z_t):
        \mathcal{X}_t = \mathcal{X}_t = \emptyset
2: for j = 1 to J do
3: sample x_t^{[j]} \sim p(x_t \mid u_t, x_{t-1}^{[j]})
              w_t^{[j]} = p(z_t \mid x_t^{[j]})
              \bar{\mathcal{X}}_t = \bar{\mathcal{X}}_t + \langle x_t^{[j]}, w_t^{[j]} \rangle
    end for
      for j = 1 to J do
                 draw i \in 1, \ldots, J with probability \propto w_t^{[i]}
                 add x_t^{[i]} to \mathcal{X}_t
9:
10:
           endfor
           return \mathcal{X}_t
```

### Particle Filters



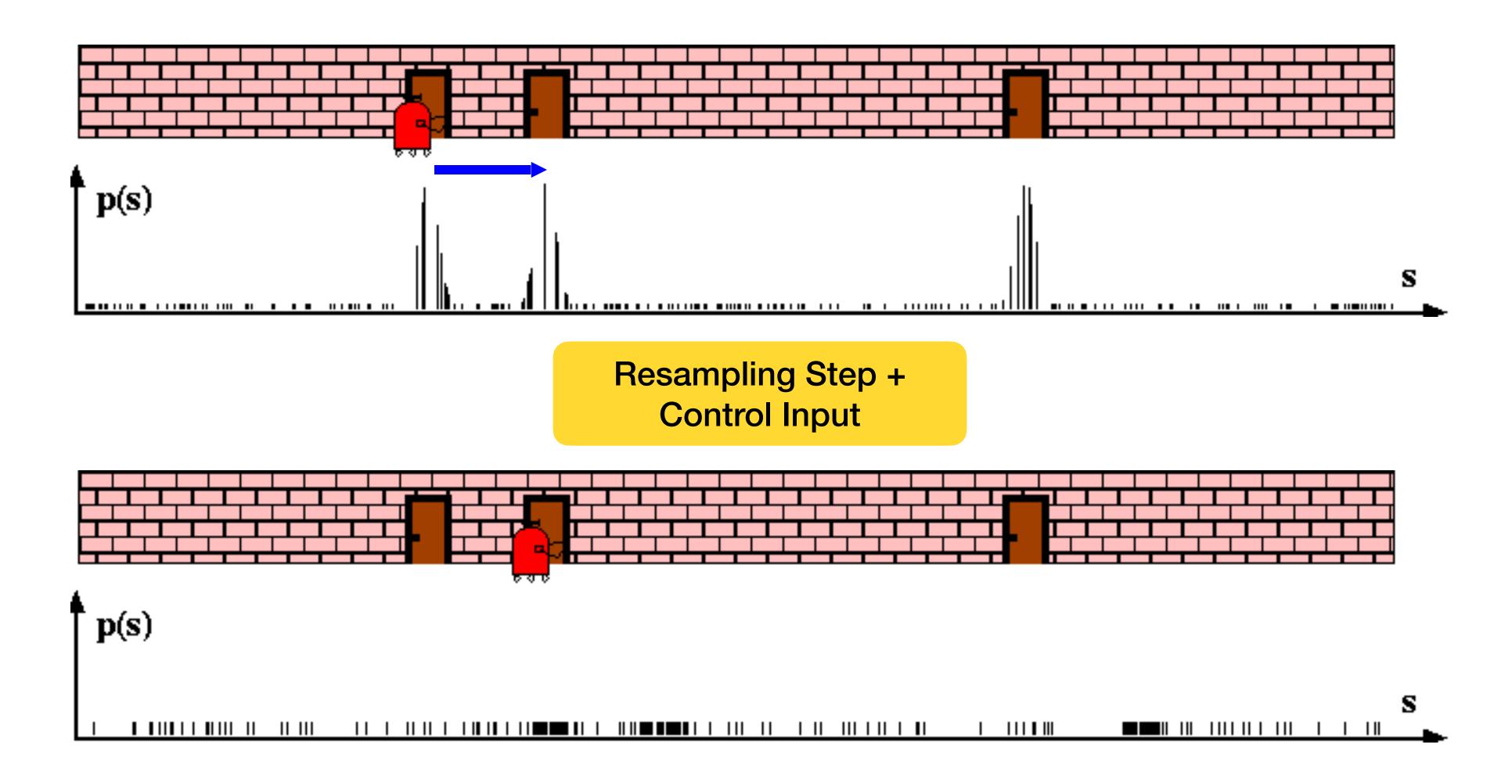


#### Sensor Information: Importance Sampling



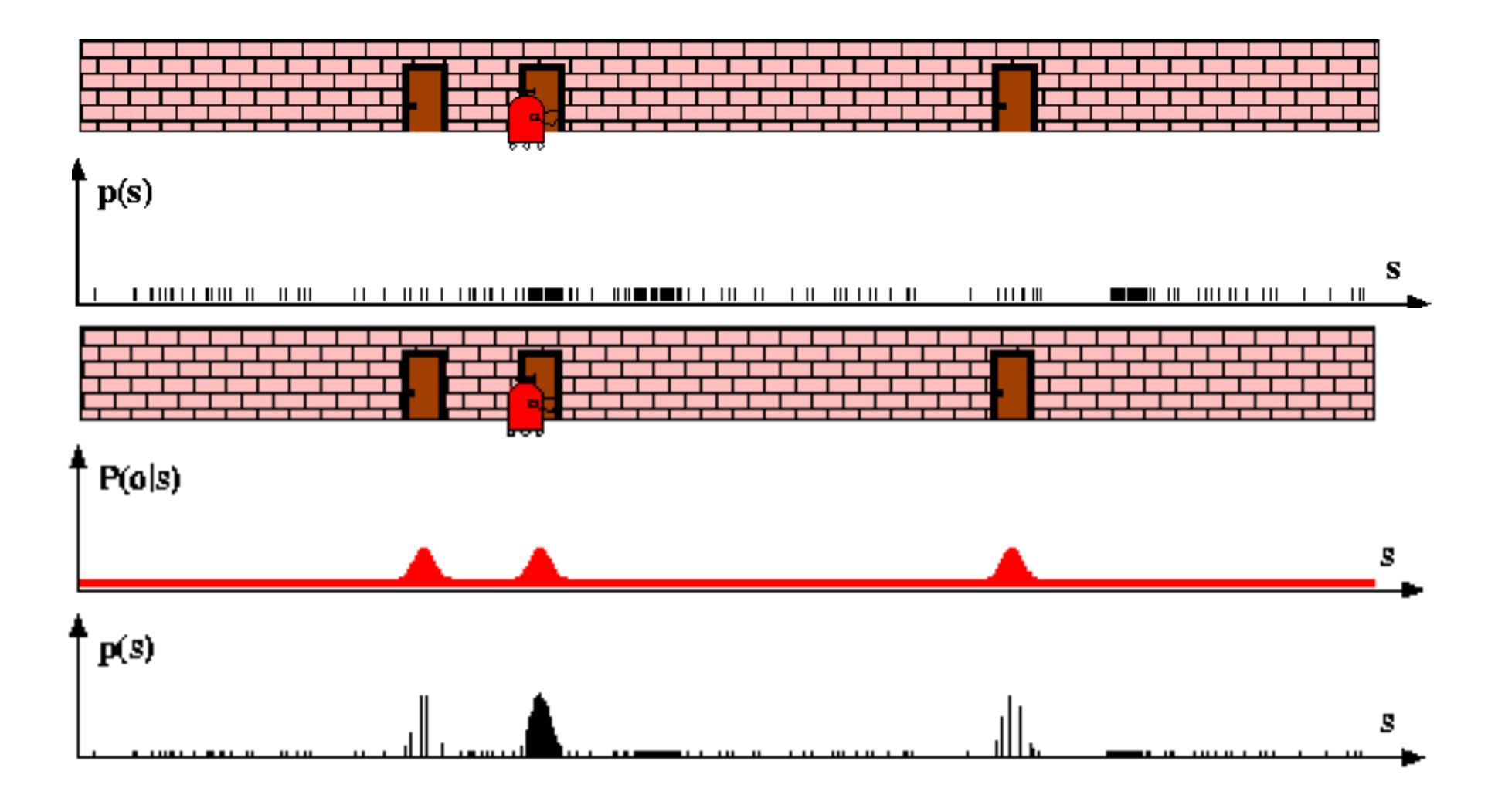


#### **Robot Motion**



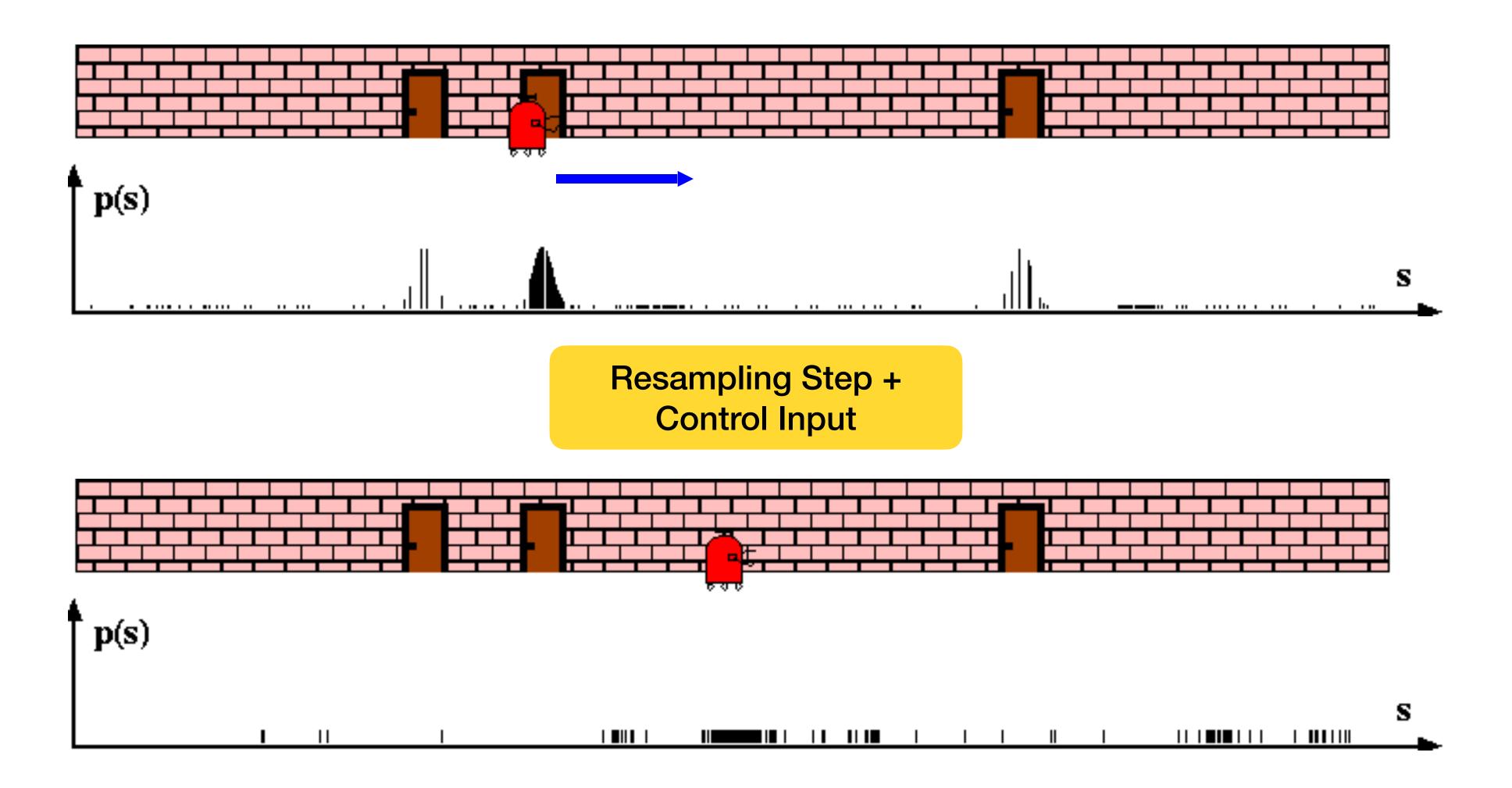


#### Sensor Information: Importance Sampling





#### **Robot Motion**

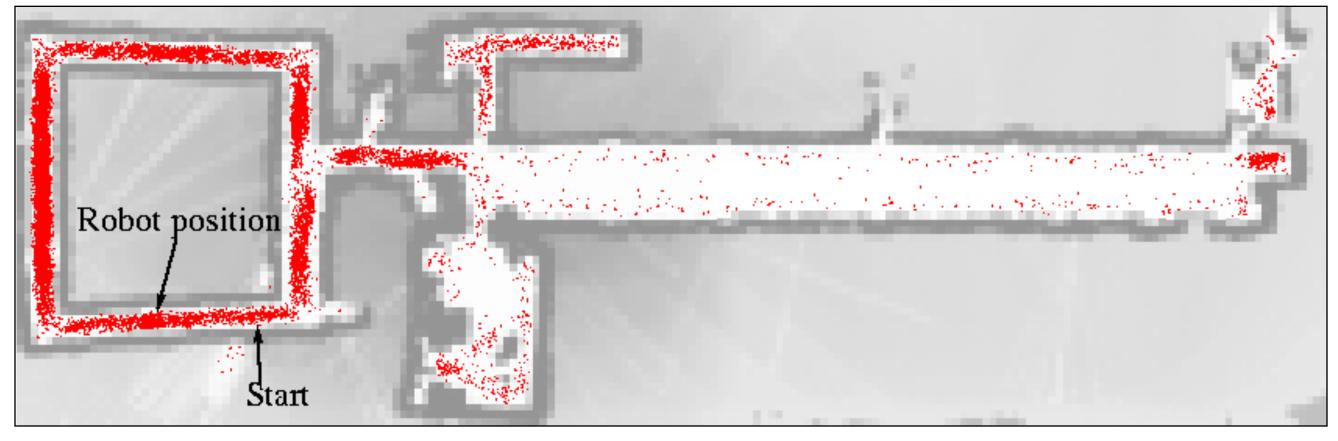




### Why have constant number of particles through out?



### Adaptive Sampling

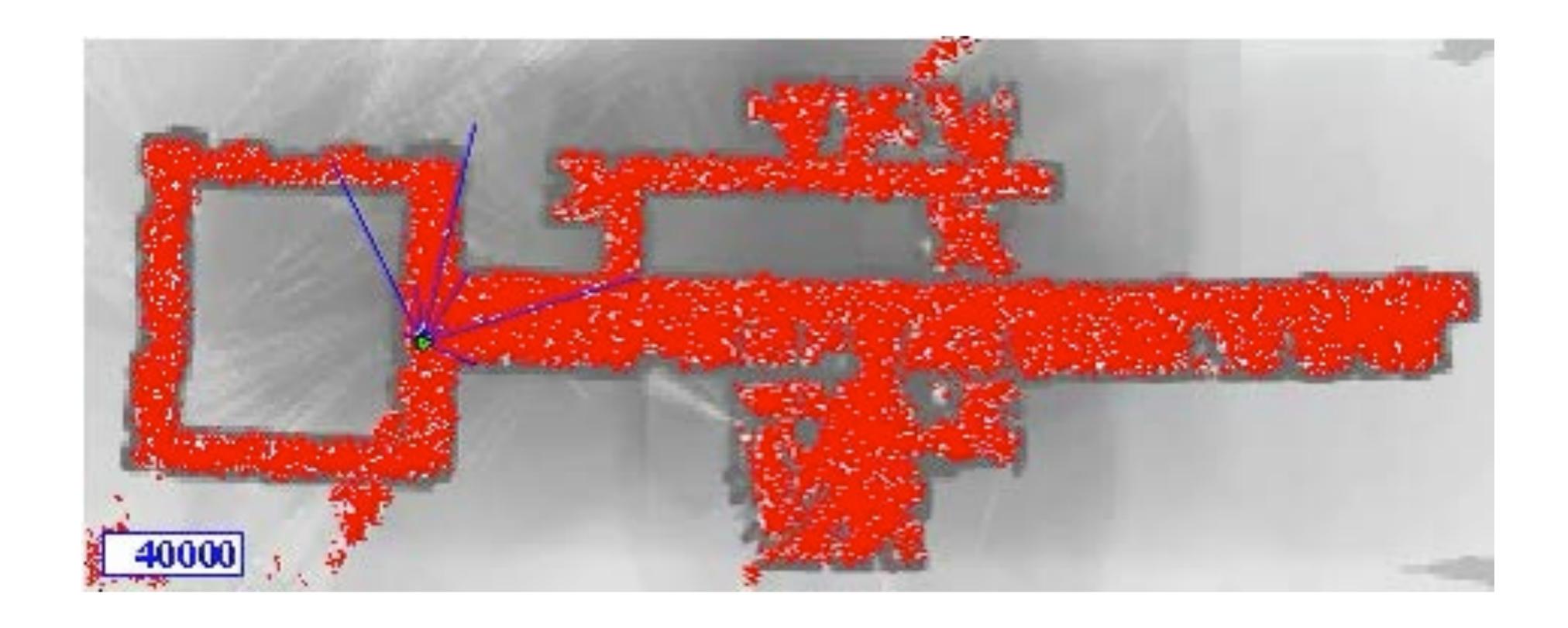








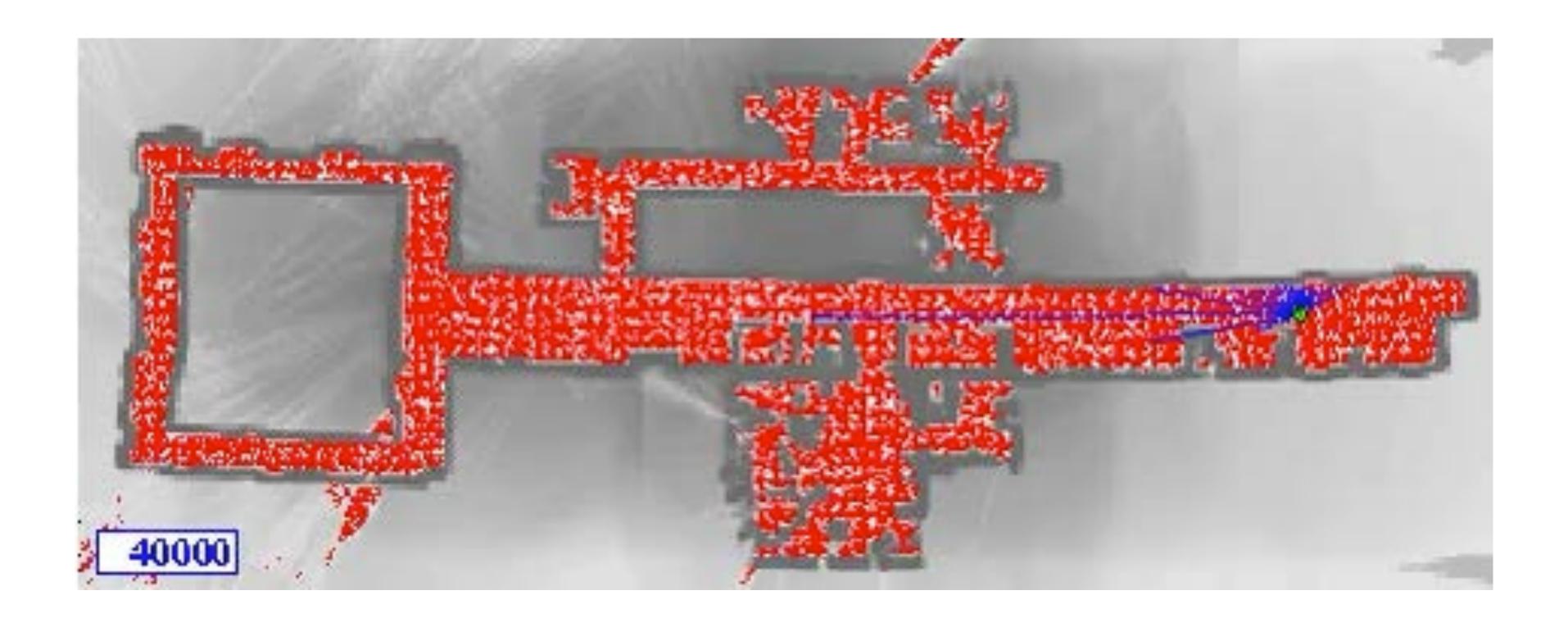
### KLD-Sampling Sonar



Adapt number of particles on the fly based on statistical approximation measure



### KLD-Sampling Laser



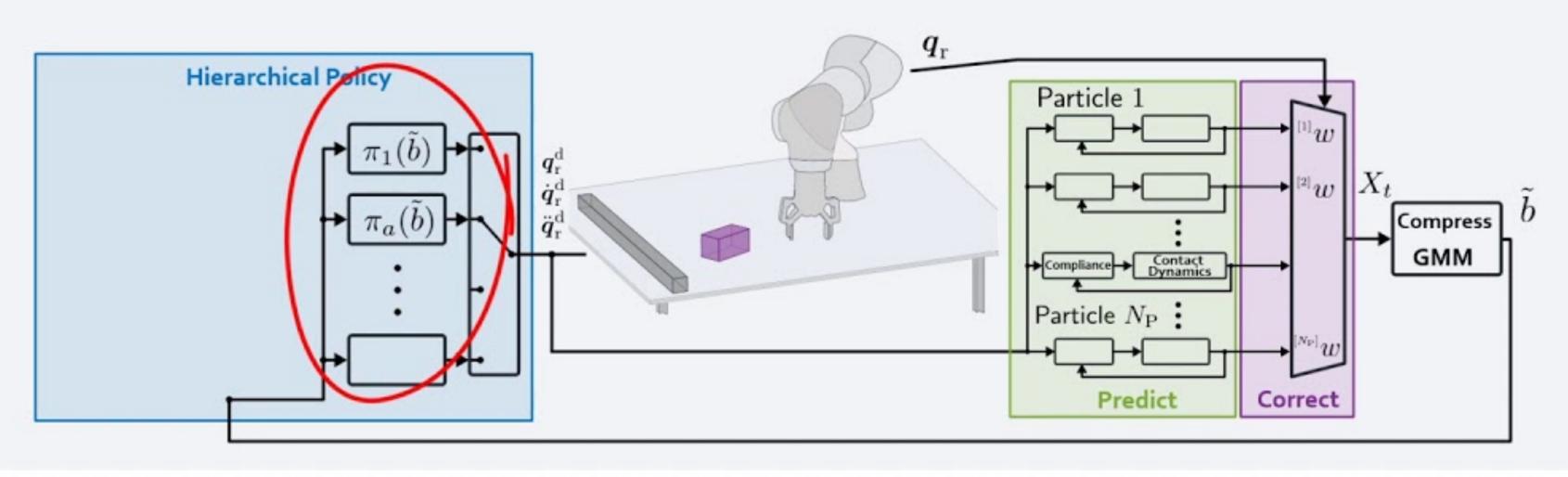


### What if the localization is not about the robot?

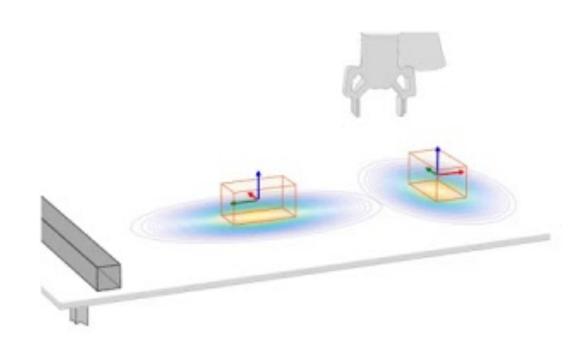








#### Hierarchical Policy - Goal-Directed Low-Level Controllers

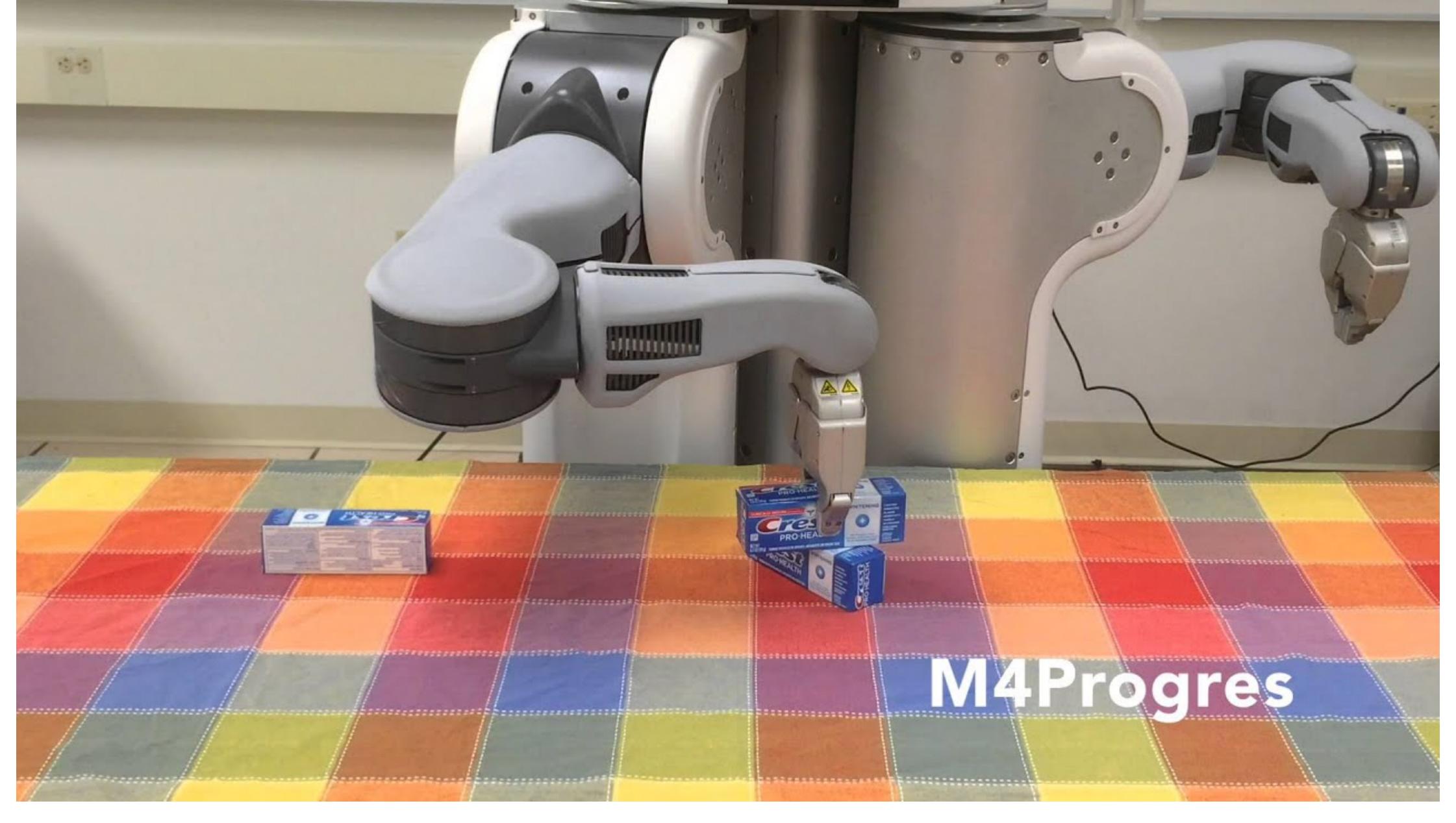


"Controlling Contact-Rich Manipulation Under Partial Observability"

Florian Wirnshofer (Siemens AG)\*; Philipp Sebastian Schmitt (Siemens AG); Georg von Wichert (Siemens AG); Wolfram Burgard (University of Freiburg)

RSS 2020





Zhiqiang Sui, Lingzhu Xiang, Odest Chadwicke Jenkins, Karthik Desingh, "Goal-directed Robot Manipulation through Axiomatic Scene Estimation," IJRR 2017.



### Physically Plausible Scene Estimation for Manipulation in Clutter

Karthik Desingh<sup>1</sup>, Odest Chadwicke Jenkins<sup>1</sup>, Lionel Reveret<sup>2</sup>, Zhiqiang Sui<sup>1</sup>

> <sup>1</sup>University of Michigan, Ann Arbor, USA <sup>2</sup>INRIA Rhône-Alpes, Saint Ismier, France

Karthik Desingh, Odest Chadwicke Jenkins, Lionel Reveret, Zhiqiang Sui, "Physically Plausible Scene Estimation for Manipulation in Clutter," Humanoids 2016.



# Redojgntzekla@kajedt@withBebises Raw Object Detection

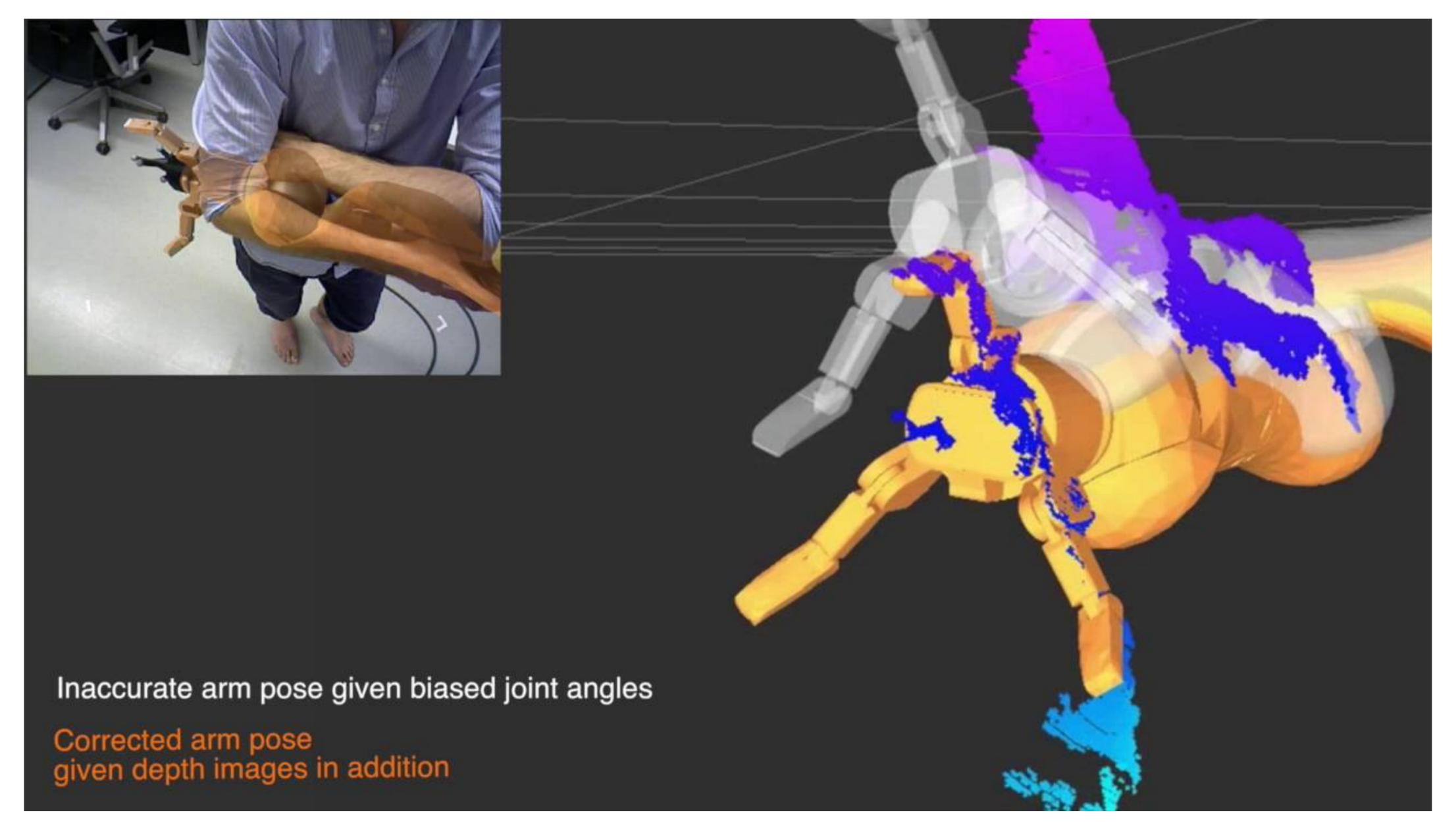
Zhen Zeng, Yunwen Zhou, Odest Chadwicke Jenkins, Karthik Desingh, "Semantic Mapping with Simultaneous Object Detection and Localization," IROS 2018





Zhen Zeng, Adrian Röfer, Odest Chadwicke Jenkins, "SLiM: Semantic Linking Maps for Active Visual Object Search.," ICRA 2020





Probabilistic Articulated Real-Time Tracking for Robot Manipulation. Garcia Cifuentes, Cristina and Jan Issac and Manuel Wüthrich and Stefan Schaal and Jeannette Bohg.

IEEE Robotics and Automation Letters (RA-L) 2017.





Karthik Desingh, Shiyang Lu, Anthony Opipari, Odest Chadwicke Jenkins, "Efficient Nonparametric Belief Propagation for Pose Estimation and Manipulation of Articulated Objects," Published: Science Robotics Journal May 2019



### Next Lecture: Mapping



### Talk to me about your final project ideas!

### Book your P7 lab sessions ASAP See the post on Edstem

