

EE4308 Turtlebot Project

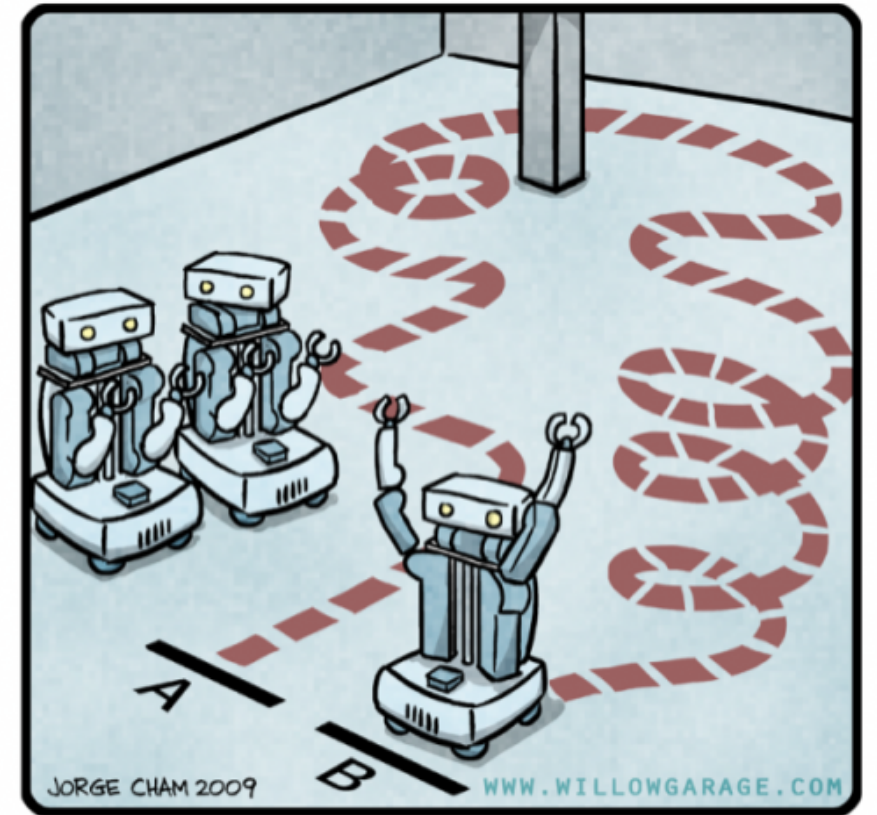
Advances in Intelligent Systems and Robotics
Academic Year 2016-2017

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What will be discussed

1. Problem statement
2. Overview of the solution
3. Information gathering
4. Global planning
5. Local planning
6. Demonstration



"HIS PATH-PLANNING MAY BE
SUB-OPTIMAL, BUT IT'S GOT FLAIR."

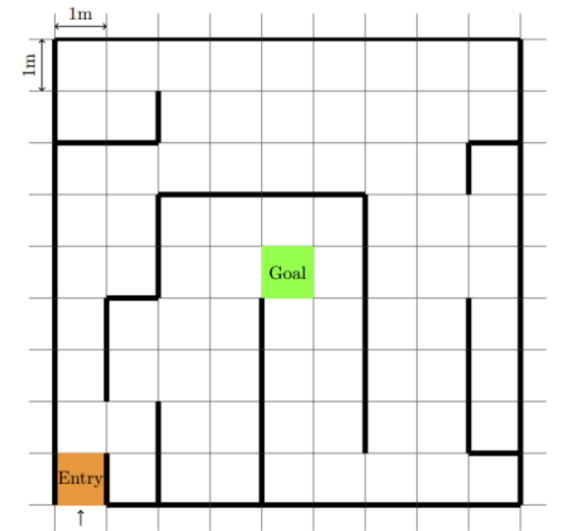
1. Problem statement

What is to be achieved

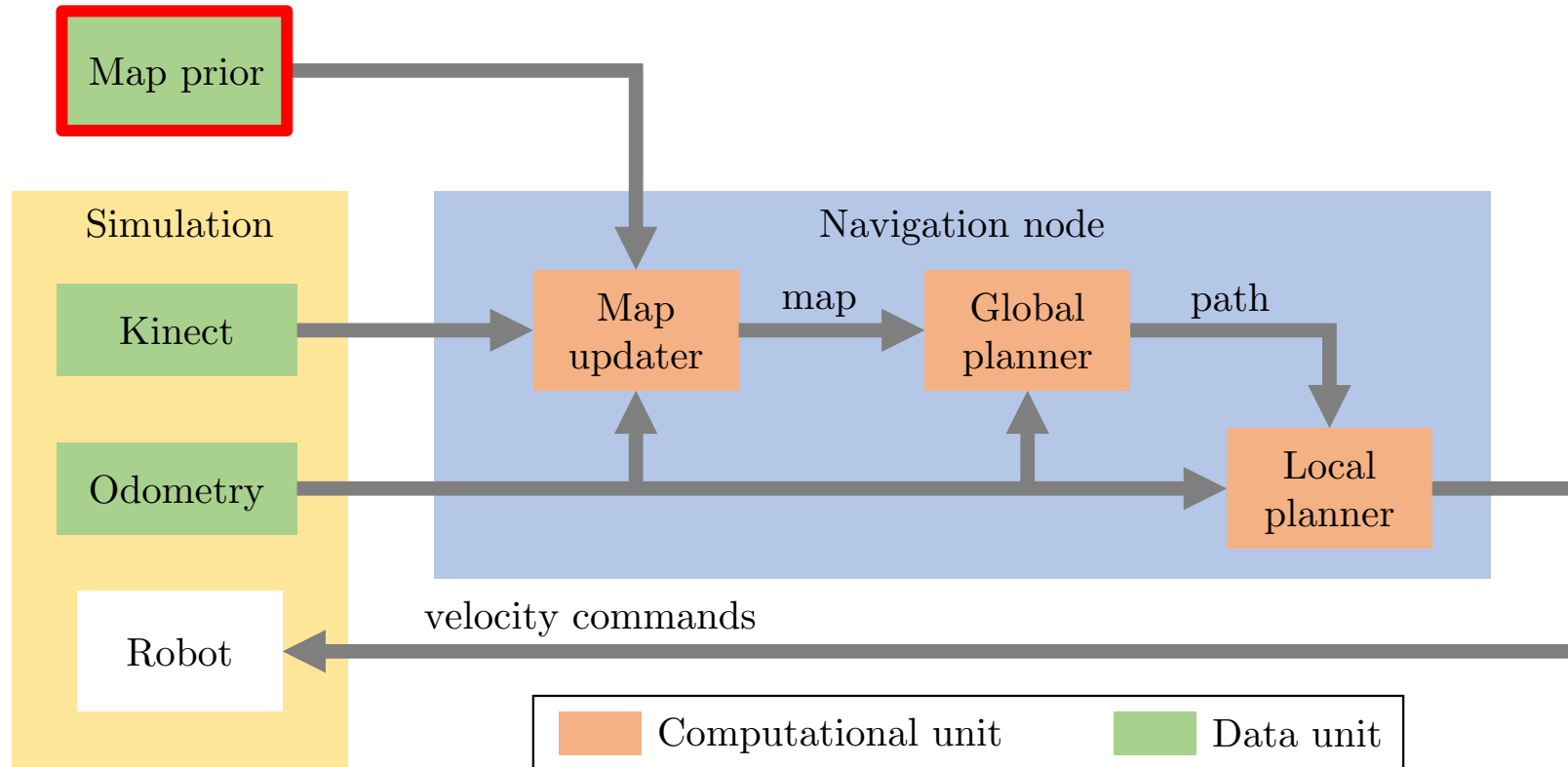
- Design a **navigation system** for a mobile robot
- Autonomously navigate from start to goal
- Environment is **not necessarily known**
- Realistic simulation



Easy task for a human,
but complex for a robot...
→ need to **divide in sub-problems**

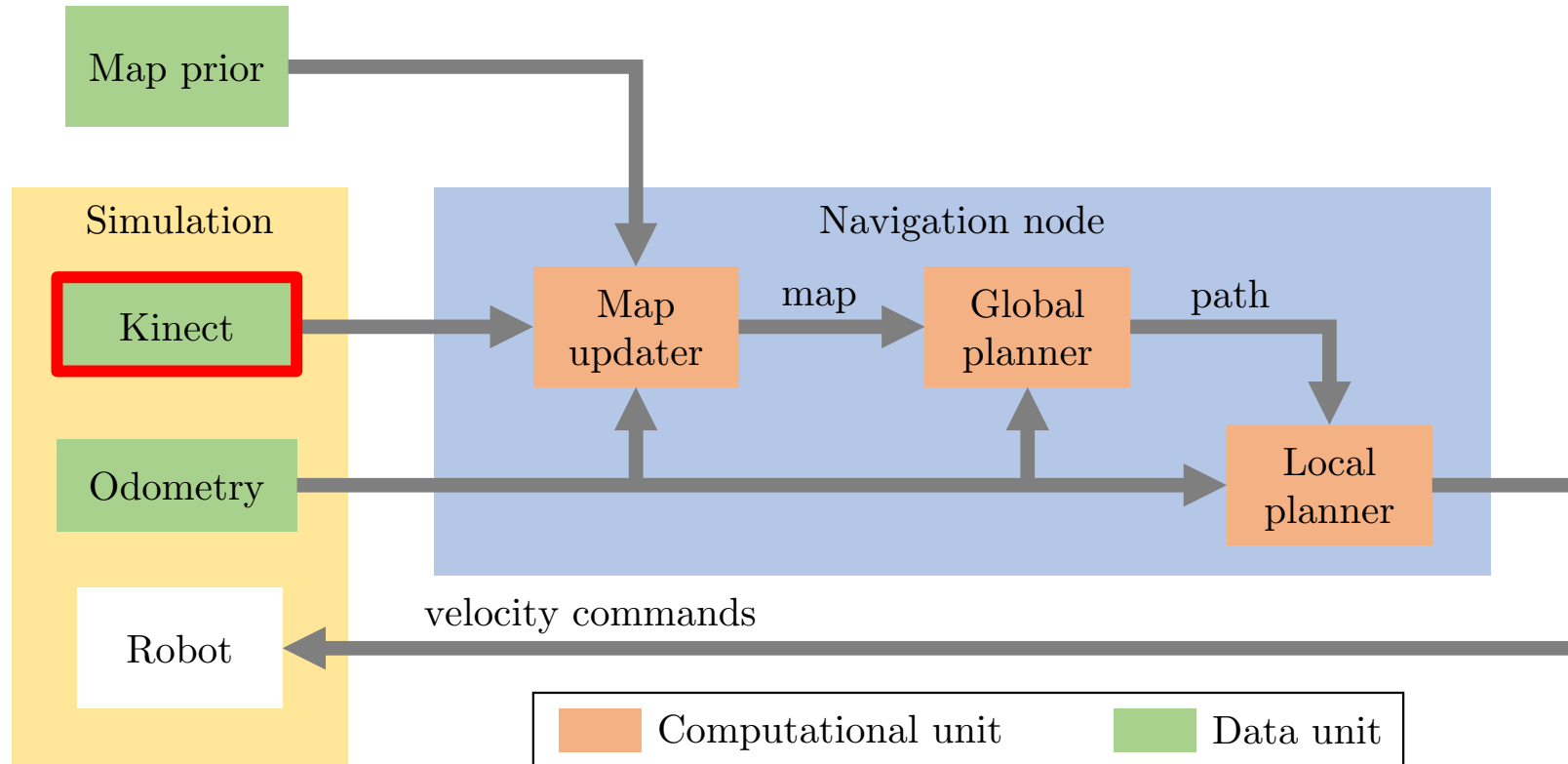


Sub-systems



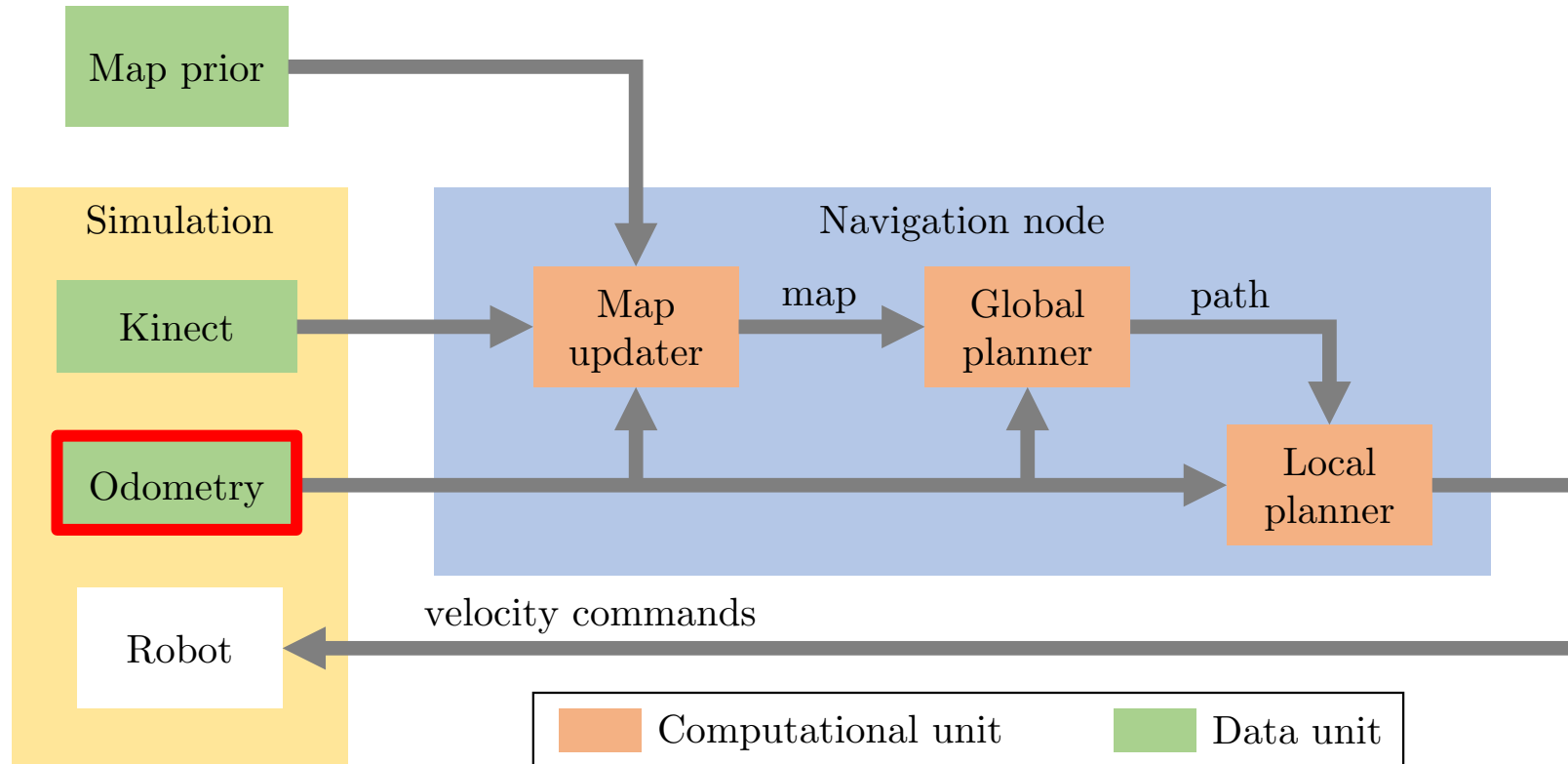
- **Map prior:** what is known beforehand (*a priori*) of the environment

Sub-systems



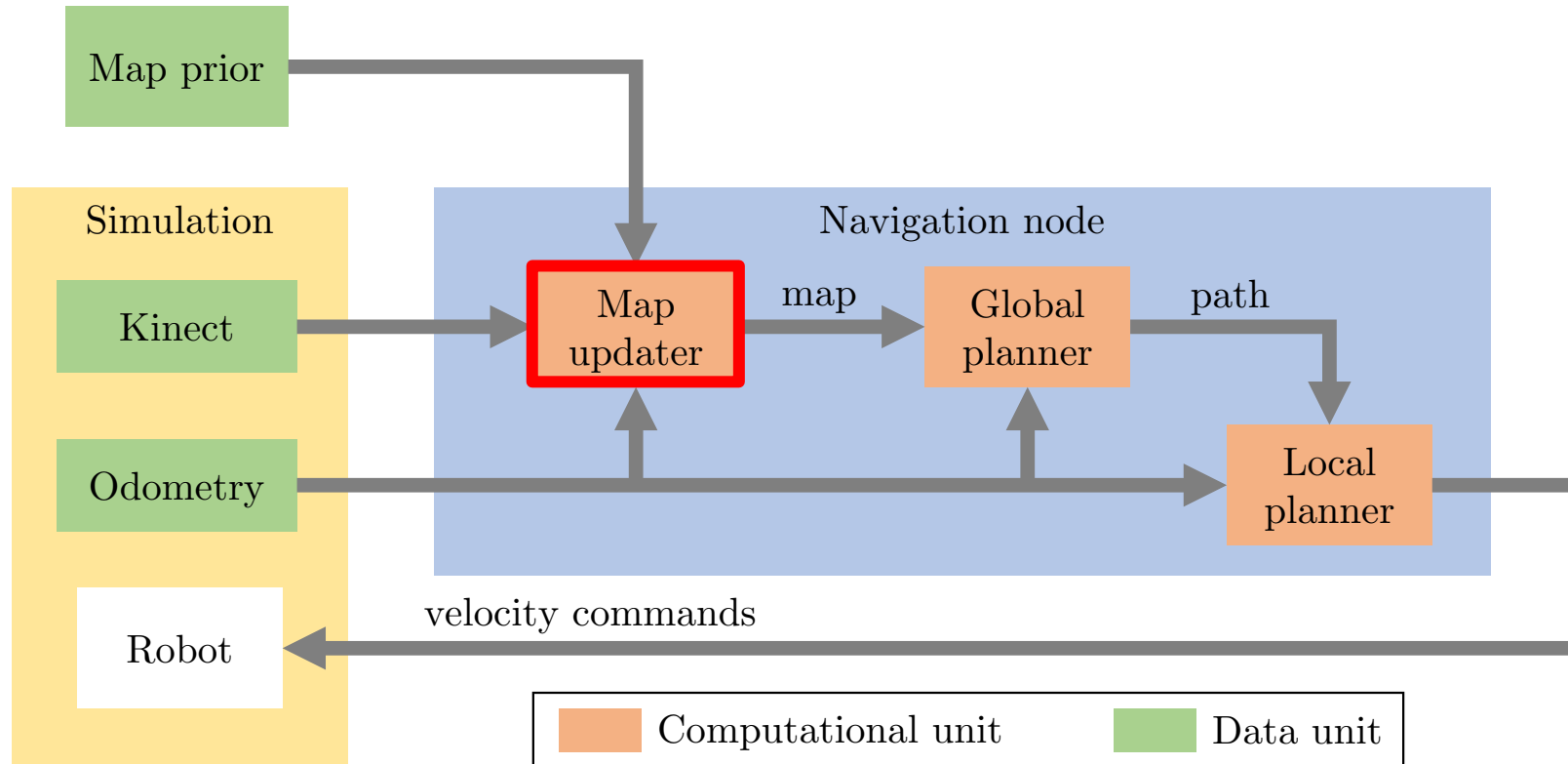
- **Kinect:** a depth camera, returns a point cloud (PC)

Sub-systems



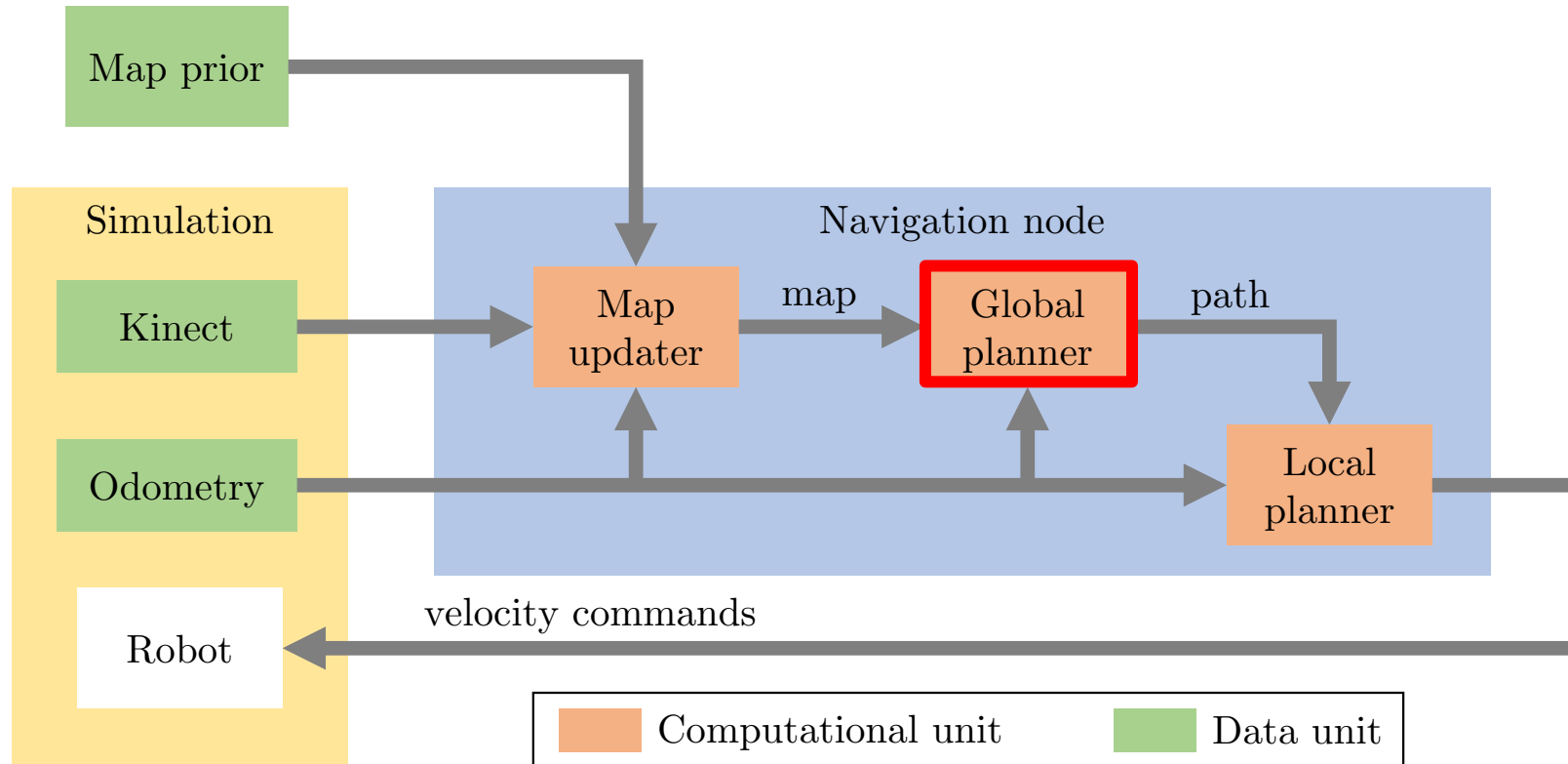
- **Odometry:** the position of the robot in the maze

Sub-systems



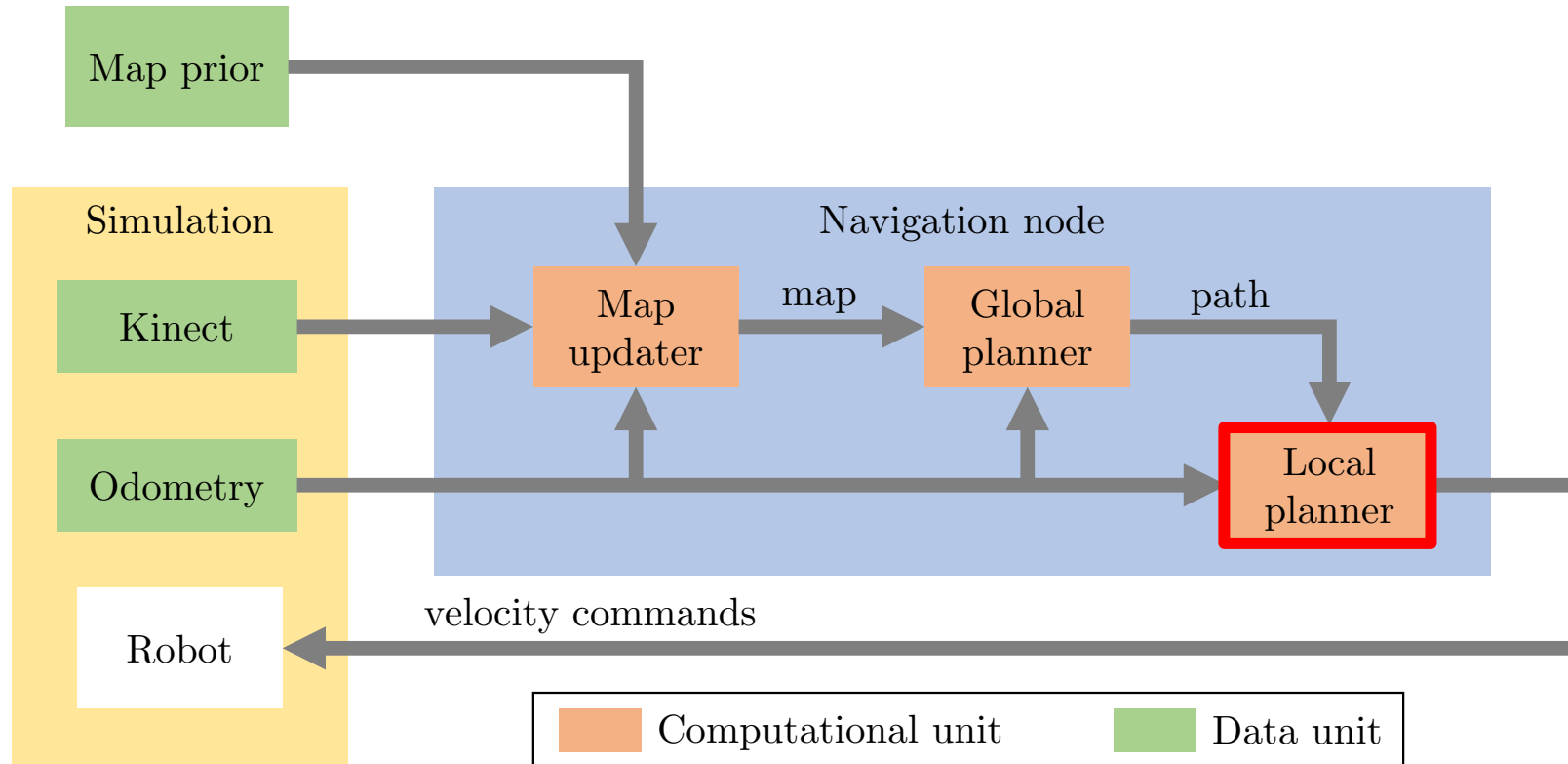
- **Map updater:** process the data to build and update an internal map of the environment

Sub-systems



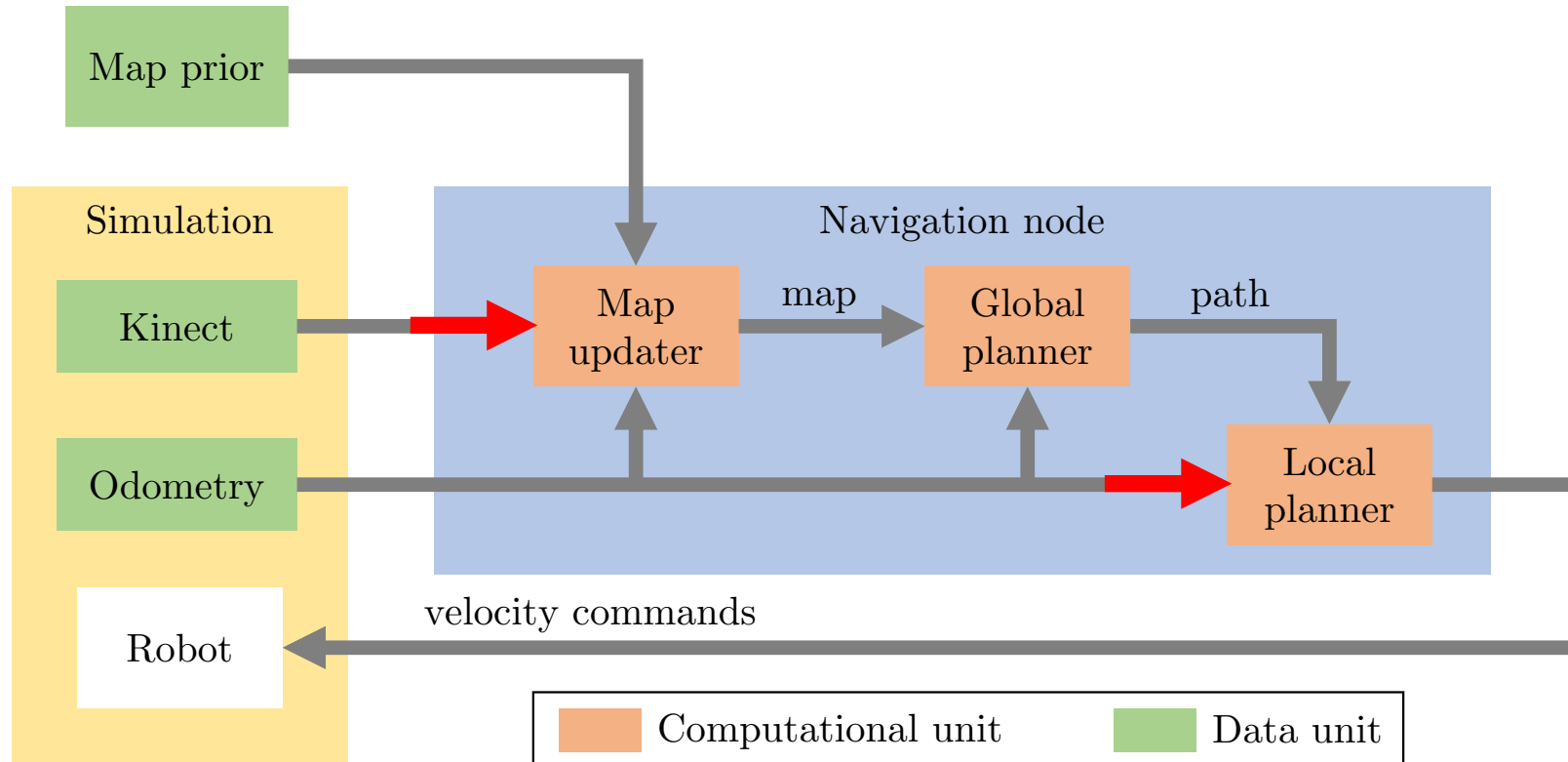
- **Global planner:** compute a path from start to goal using the map

Sub-systems



- **Local planner:** ensure that the robot actually follows the path, computes the velocity commands

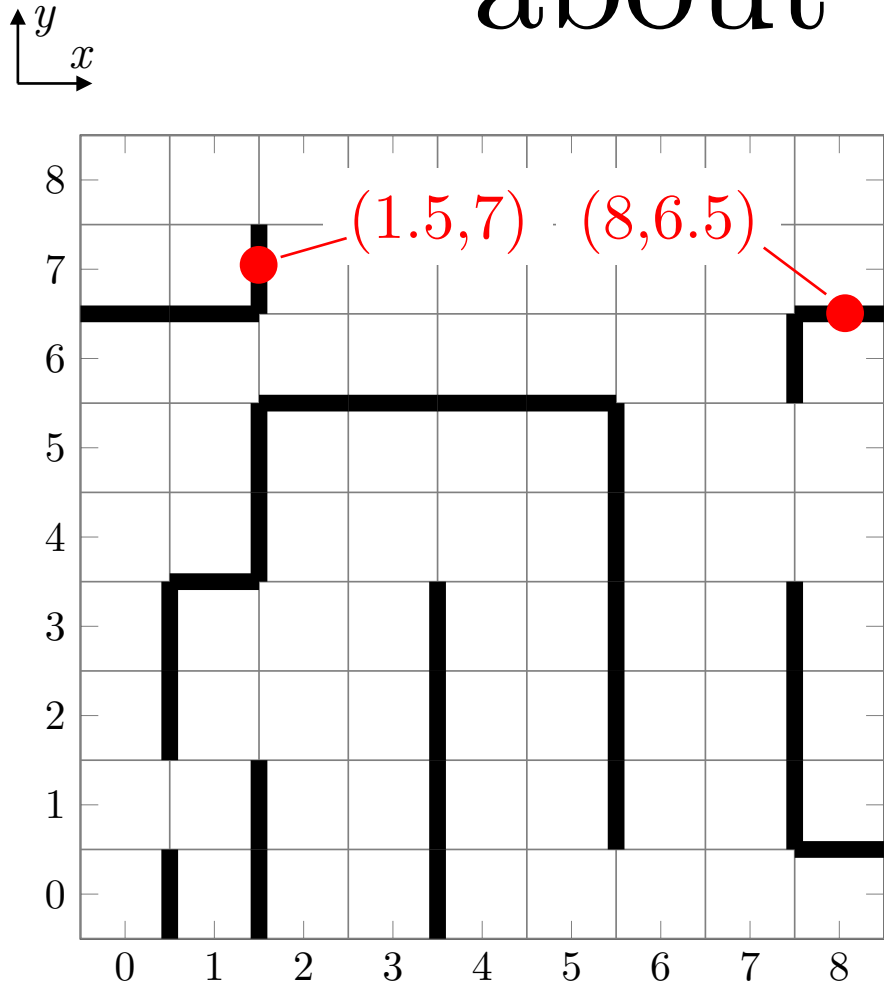
Implementation :: ROS



- **Navigation system** = ROS node, in Python
- **Callbacks** trigg. by new Kinect/Odometry data

3. Information Gathering

How to efficiently store information about the environment ?



Use the constraints on the maze:
walls only at boundaries of cells

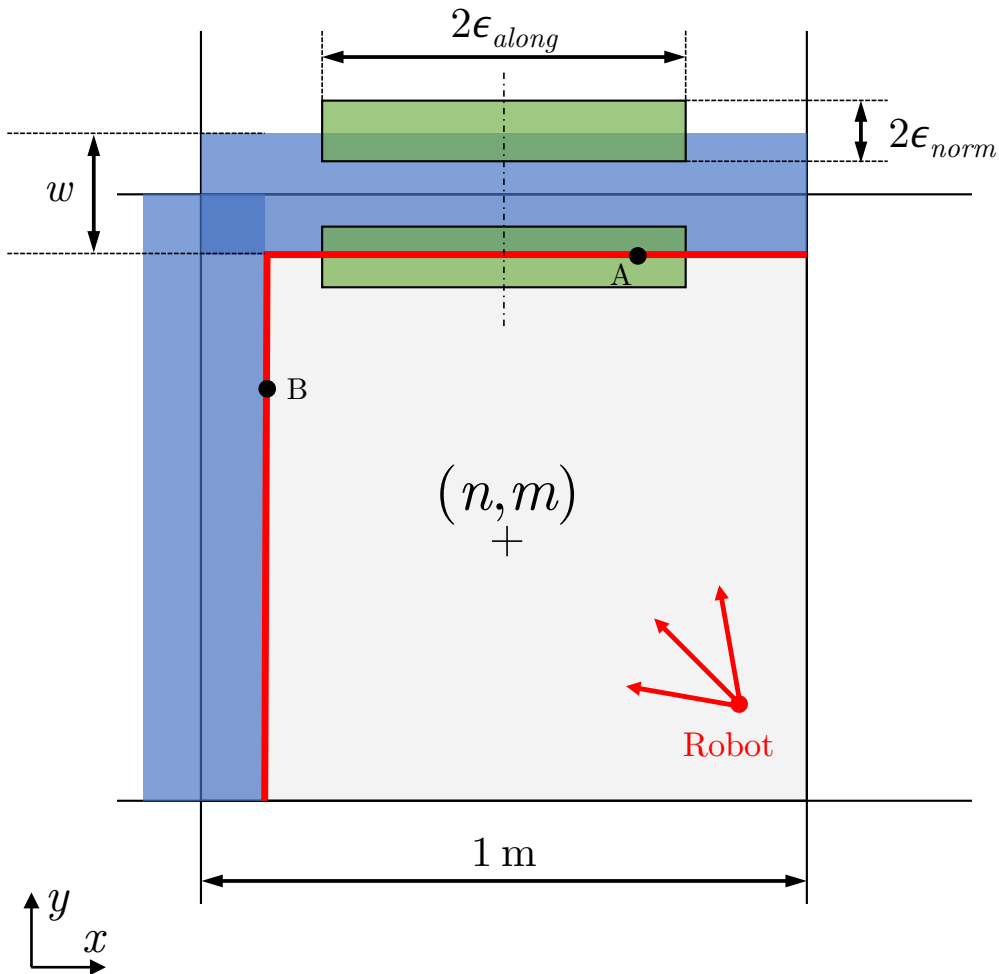
→ only store the coordinates
of the centres of the walls !

One of the two coordinates is an integer:

x \rightarrow horizontal wall

$y \rightarrow$ vertical wall

How to extract walls from the Kinect ?



1. Extract a row of the 2D PC
2. Define a **ROI** with tolerances:
 $(n \pm \epsilon_{along}, m + 0.5 \pm w/2 \pm \epsilon_{norm})$ horizontal
 $(n + 0.5 \pm w/2 \pm \epsilon_{norm}, m \pm \epsilon_{along})$ vertical
3. For each point: **check** if in any possible ROI
4. If enough points are assumed to belong to a wall
→ **add it to the map**

4. Global planning

Global planner:

1. Find a path from **S** to **G** with A^*

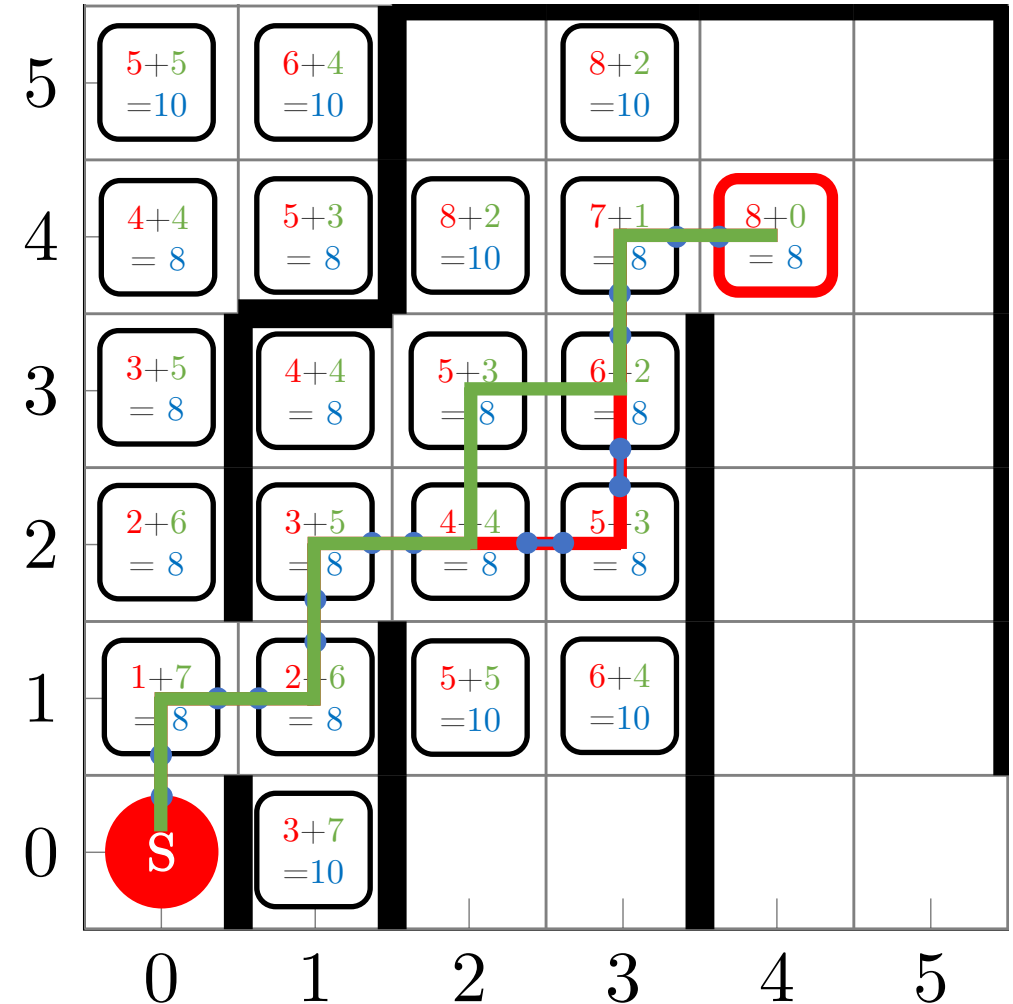
Expand cost from **S**:

$$f = g + h$$

g move cost from **S**

h Manhattan dist. to **G**

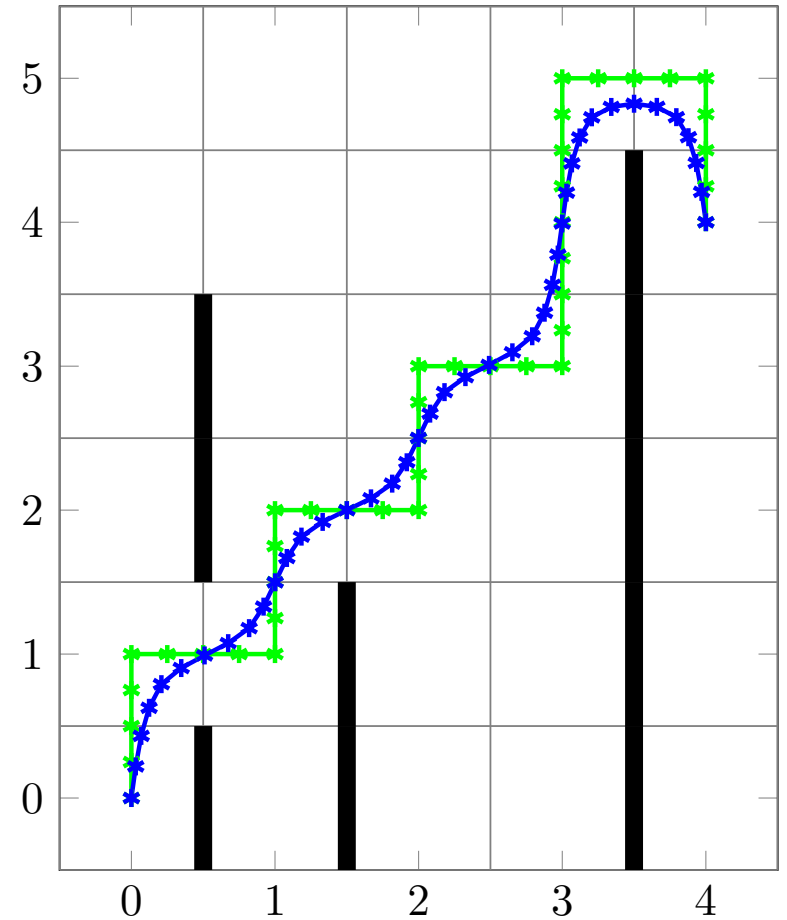
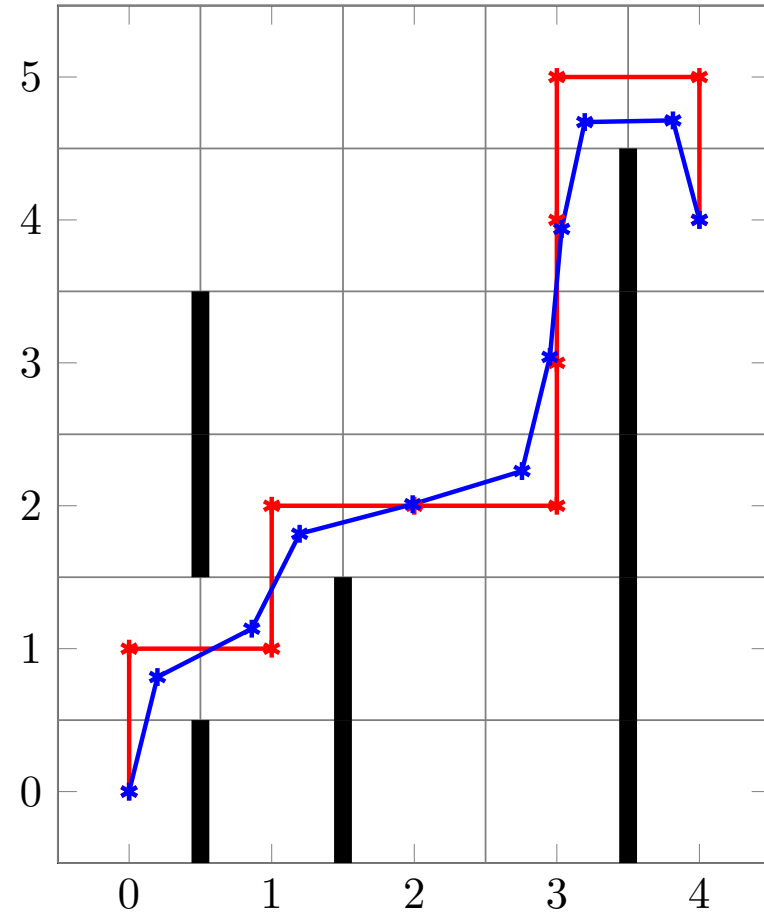
Can dis/encourage
turns/lines by tuning g



4. Global planning

Global planner: 2. Smooth

Minimise J
with Gradient
Descent

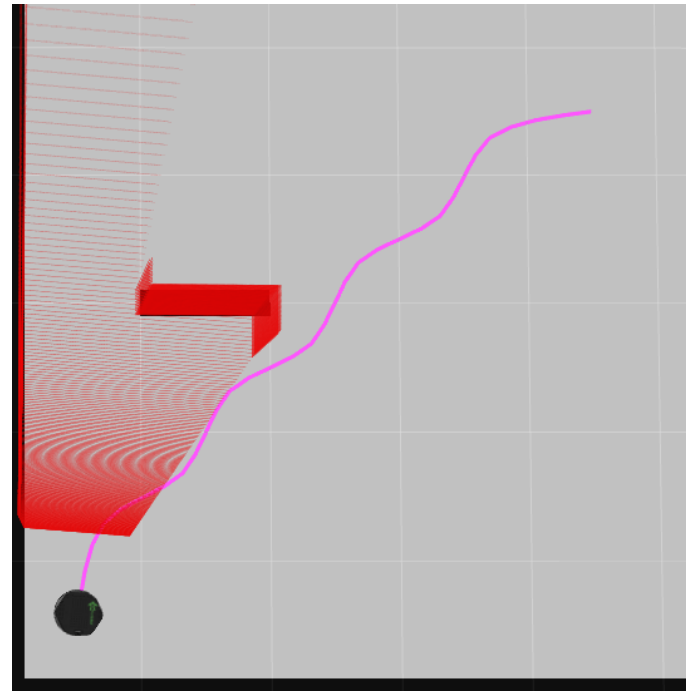


$$J = \frac{1}{2} \sum_{i=1}^n \underbrace{\alpha \left((x_i - x'_i)^2 + (y_i - y'_i)^2 \right)}_{\text{original path}} + (1 - \alpha) \underbrace{\left((x'_i - x'_{i+1})^2 + (y'_i - y'_{i+1})^2 \right)}_{\text{shortened smooth path}}$$

Known or unknown environment ?

If no map prior:

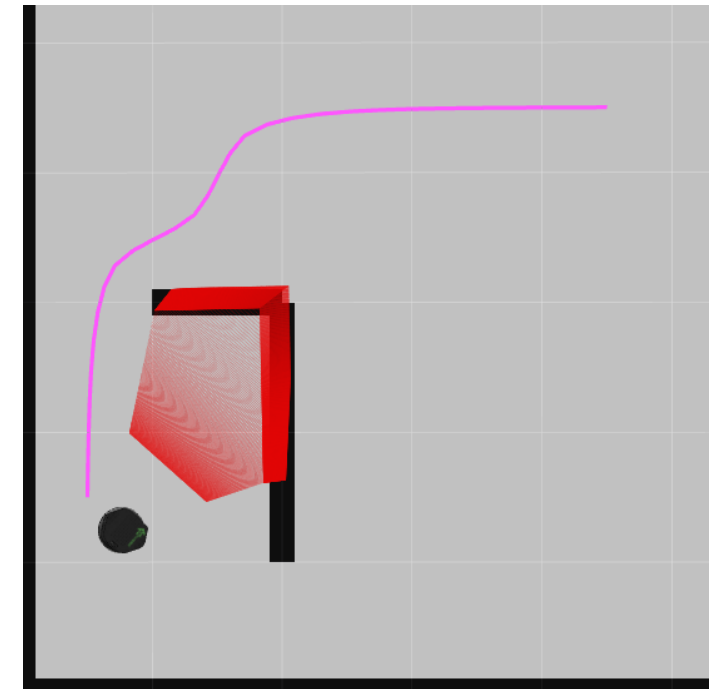
1. Assume there are no walls (1)
2. Add some as you discover them
3. Update the path if necessary (2)



(1)



(2)



Local planner: following the path

- PI controller for linear and angular velocities
- Weighted average of the errors on the next k points:

$$\theta_{PI} = \frac{\sum_{i=1}^k \theta_{j+i} d_{j+i} w_i}{\sum_{i=1}^k d_{j+i} w_i}$$

Orientation error

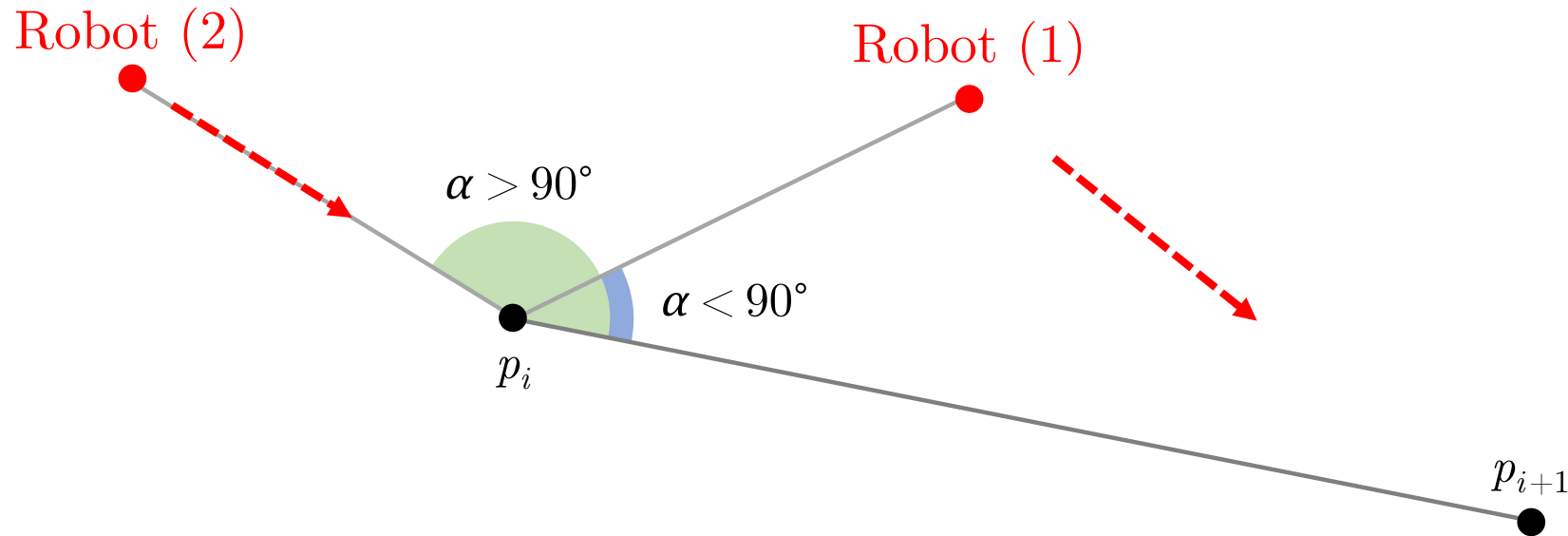
$$d_{PI} = \frac{\sum_{i=1}^k d_{j+i} w_i}{\sum_{i=1}^k w_i}$$

Distance error

- Act as a **local smoothing**
 - Limit speed + anti wind-up
- j current point
 w_i weights

Local planner: choosing the next point

1. Find the closest point to the current position (here i)
2. Compute the angle α :
 - If smaller than 90° then $i+1$ is the next
 - Otherwise i



6. Demonstration

Show time!

