# 캡스톤 디자인

- Path Planning 14<sup>th</sup> Week Progress

A6 Blue

## **Path Planning**

### Search-based

- Dijkstra
- A\* Dijkstra + heuristic cost
- D\* Dynamic A\*

### Sampling-based

- RRT Random Tree
- RRT\* RRT + rewire

### Artificial Intelligence

- ANN Artificial Neural Network
- GA Genetic Algorithm

## **Our Algorithm**

- Design own algorithm
  - A\*, D\* 기반: 장애물 회피
- Using python & ROS
  - Simulation
- Using drone
  - 3D path planning
  - Safety distance
  - Cost function optimization

## A\* Algorithm

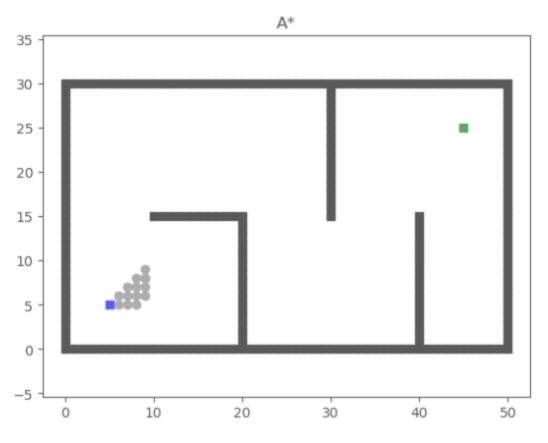
- Search-based algorithm
- Dijkstra + heuristic cost

Cost function : f(n) = g(n) + h(n)

g(n): 현재 node까지의 cost

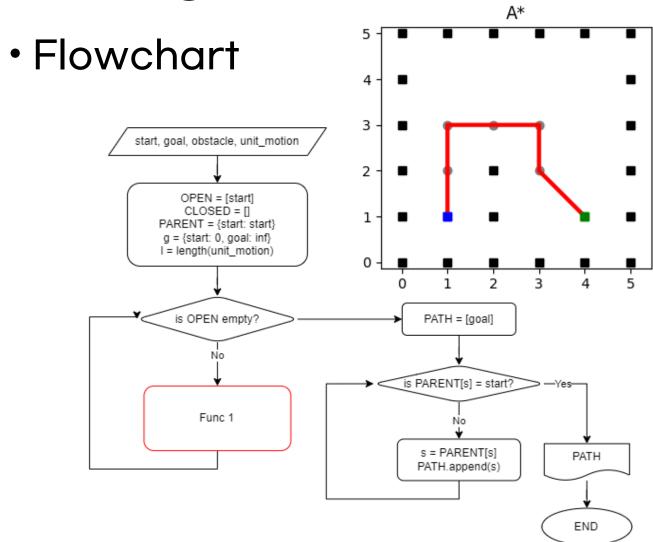
h(n): 현재 node부터 목표 node

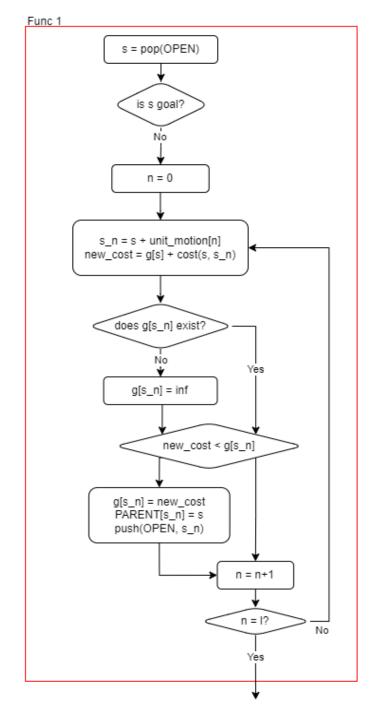
까지의 heuristic cost



### A6 Blue

# A\* Algorithm





## **D\* Algorithm**

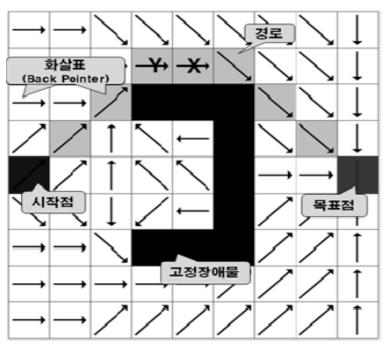
- Search-based algorithm
- Dynamic A\*
- Backward searching
- Pre-mapped global map
- Bigger memory required

• Cost function : f(n) = g(n) + h(n)

g(n): 현재 node까지의 cost

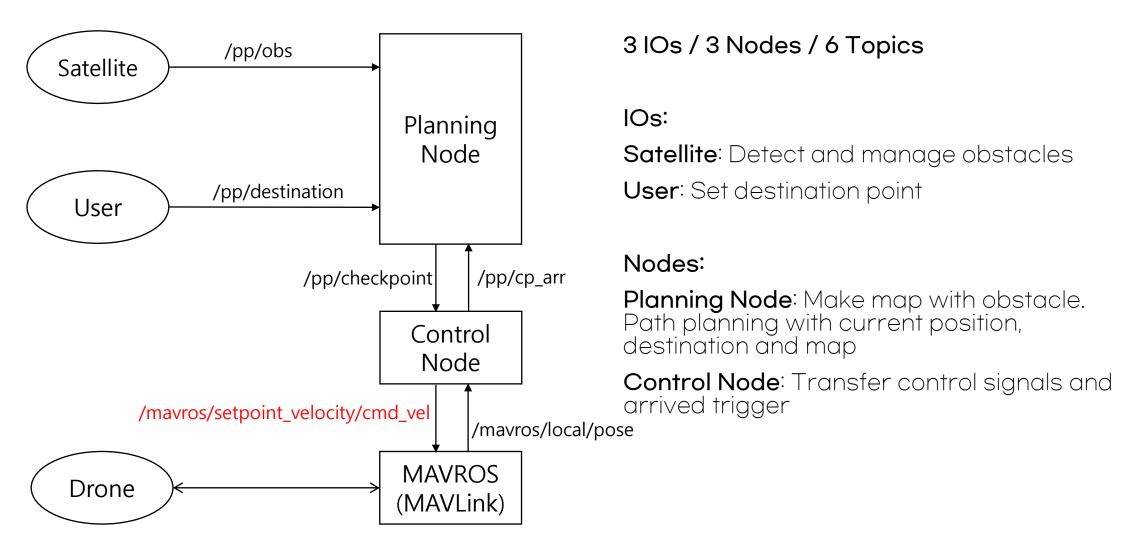
h(n): 현재 node부터 목표 node

까지의 heuristic cost

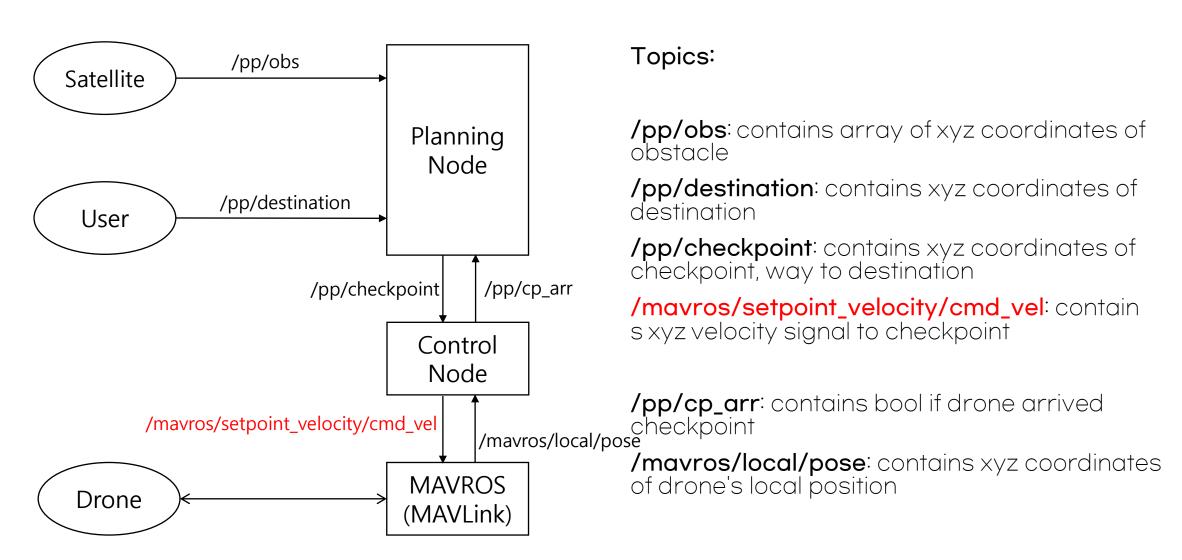


Development of a Navigation Control Algorithm for Mobile Robots Using D\* Search and Fuzzy Algorithm Yun-Ha Jung, Hyo-Woon Park, Sang-Jin Lee and Moon-Cheol Won

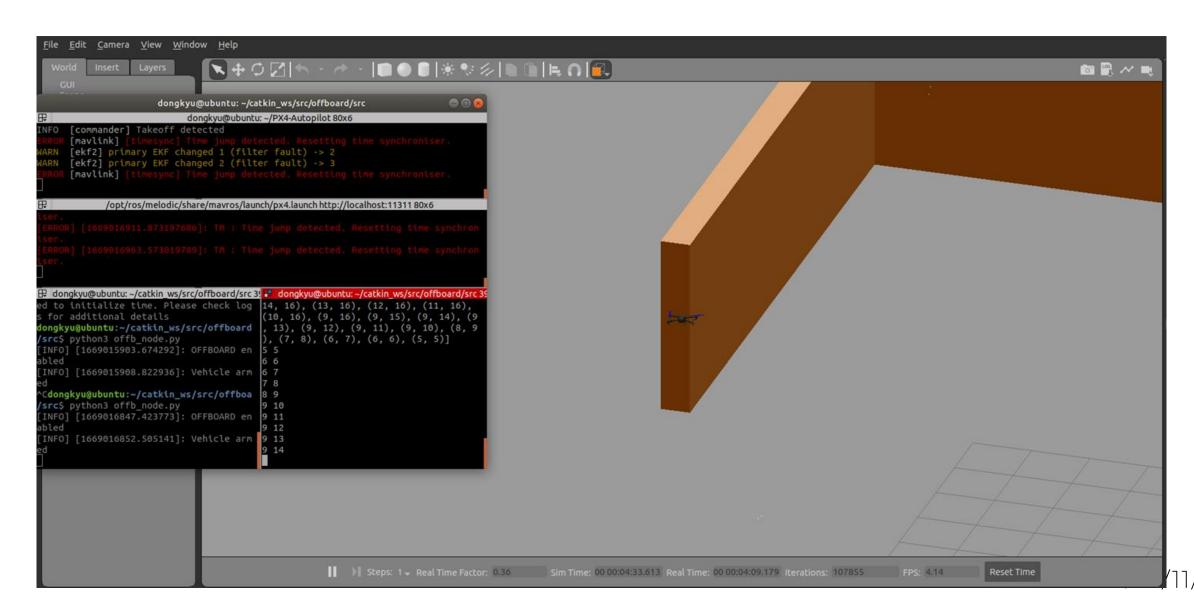
### Schematic



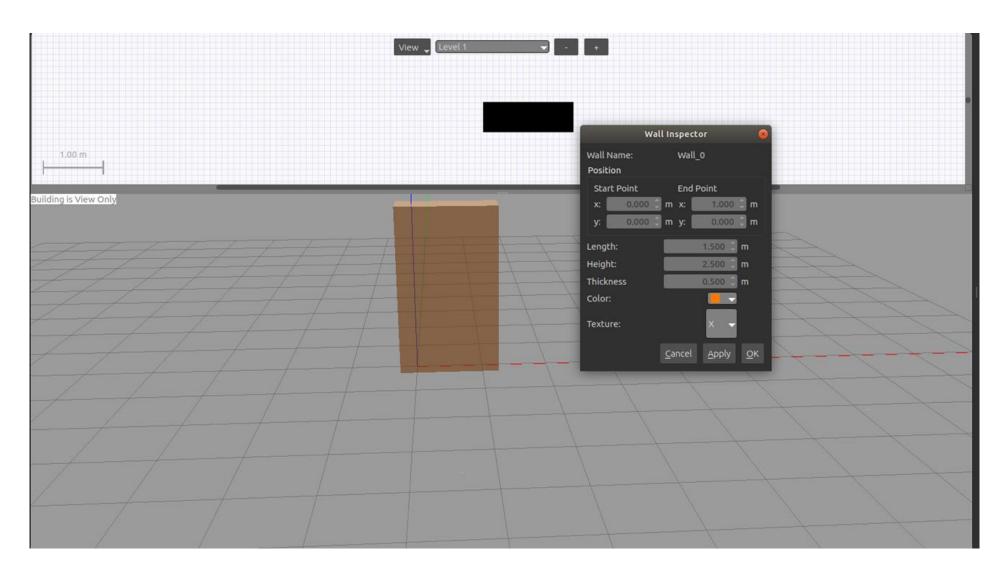
### Schematic



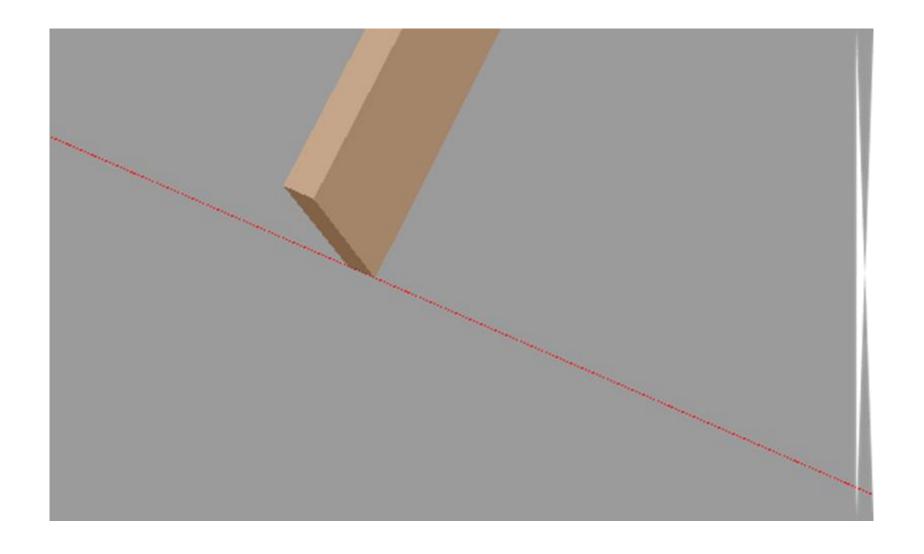
## Simulation(problem)



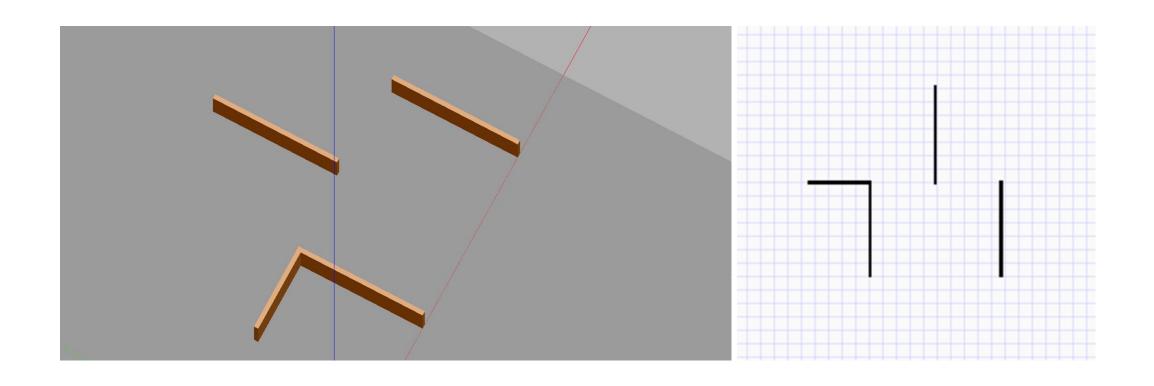
## Simulation



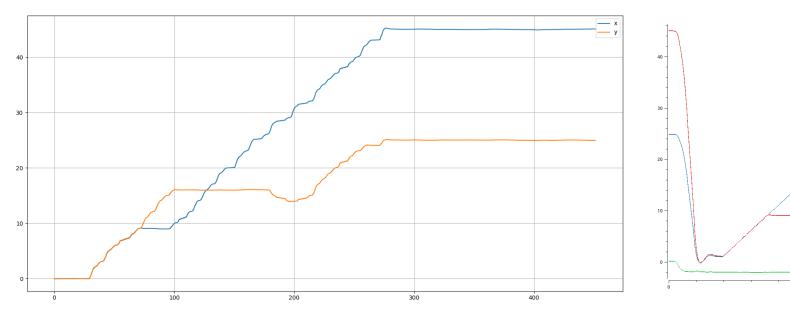
## Simulation

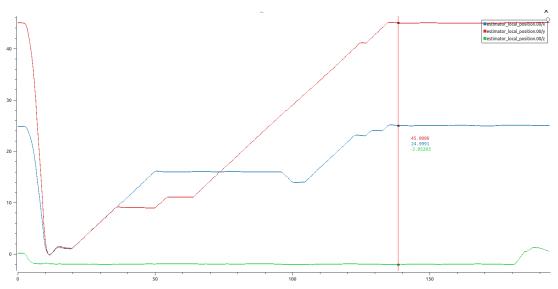


## Simulation

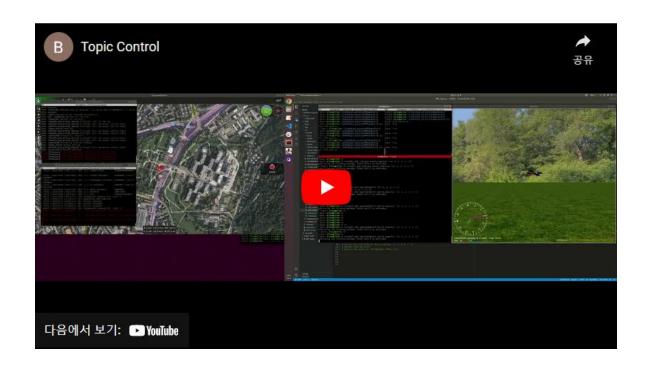


### Simulation(result)





### Control



#### Achievements:

#### PD control implemented:

- input: checkpoint, local position
- output: arrived signal, velocity

#### ROS embedded:

- Works with ROS embedded planning node
- input:/mavros/local/pose,/pp/checkpoint
- output : /mavros/setpoint\_velocity/cmd\_vel, /pp/cp\_arr

### Drone controllable via planning node Sends /pp/cp\_arr

- True if distance between checkpoint and current position is under particular distance

#### Prevent overshooting

- PD-control with x, y velocity

## Contro (Overshooting)

### Solutions:

- 1. Step position input(default)
- 2. Sigmoid position input
- 3. Velocity control, P-control
- 4. Velocity control, PD-control

### Keypoints:

- 1. Faster response
- 2. Reduce overshoot (≈0%)
- 3. Reduce stuttering
- 4. Minimize acceleration

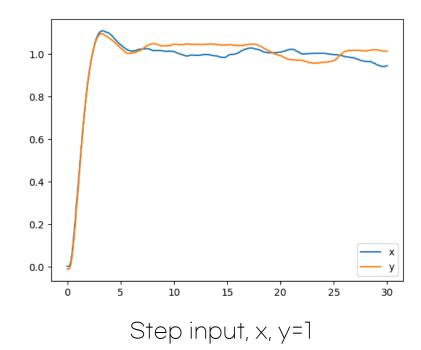
### Contro (Overshooting - default step position input)

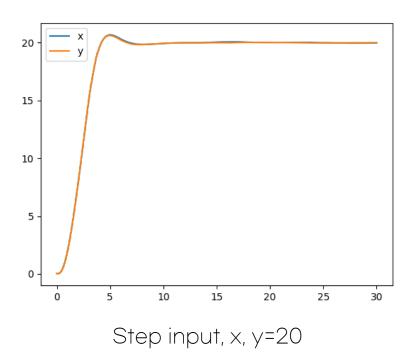
#### 1. Step position input(default)

About 15% overshoot occurs in unit step input.

About 10% overshoot occurs in 20 step input.

Settling time: about 7s

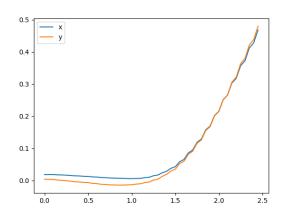


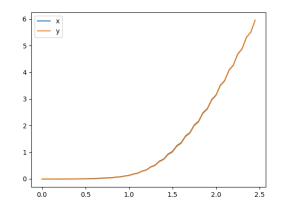


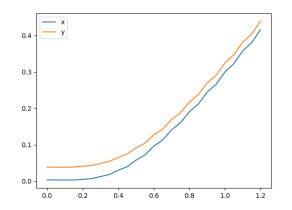
### Contro (Overshooting - sigmoid position input)

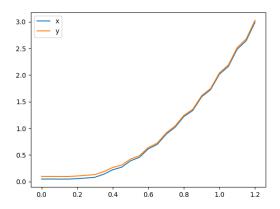
#### 2. Sigmoid position input(logistic function)

Response time with logistic position input is too long





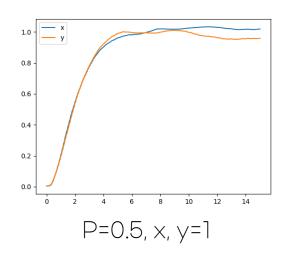


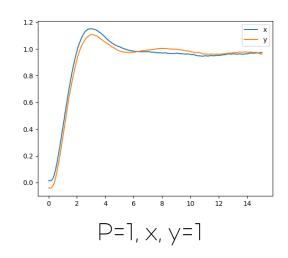


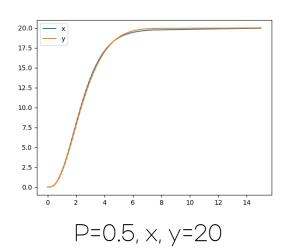
## Control (Overshooting - P-controlled velocity)

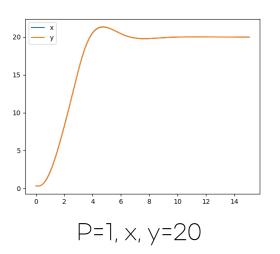
#### 3. Velocity control, P-control

No overshoot occurs with p=0.5 About 15% overshoot occurs with p=1 Settling time =  $5\sim6$ s on both p=0.5 and p=1



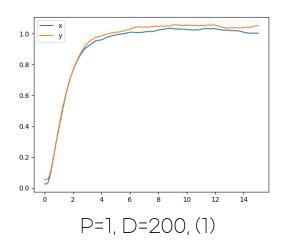


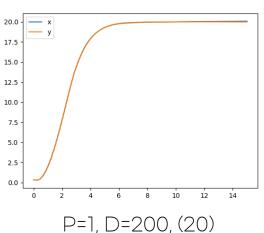


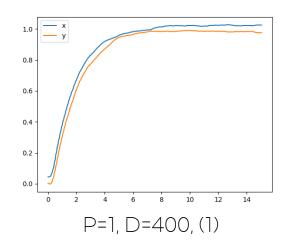


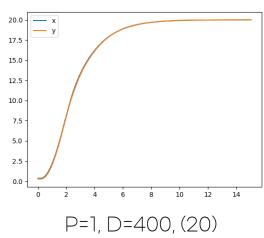
### Control (Overshooting - PD-controlled velocity)

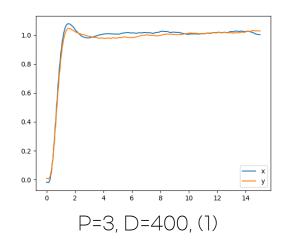
### 4. Velocity control, PD-control

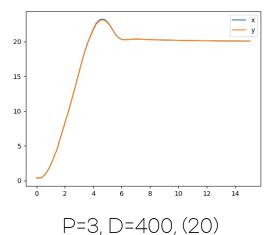


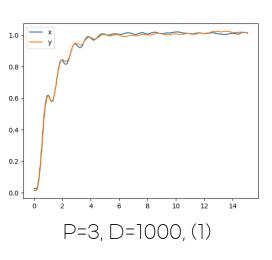


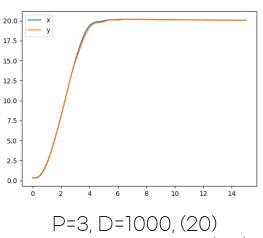












### Control (Overshooting - PD-controlled velocity)

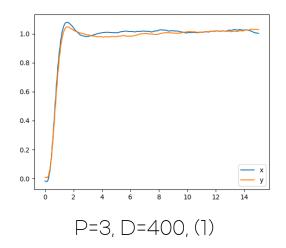
### 4. Velocity control, PD-control

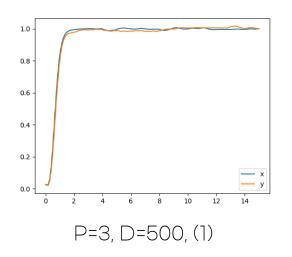
No overshoot occurs with p=1

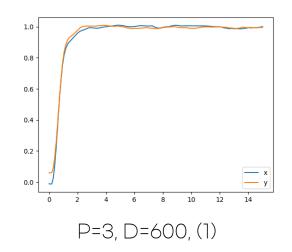
Faster settling time and response time with p=3

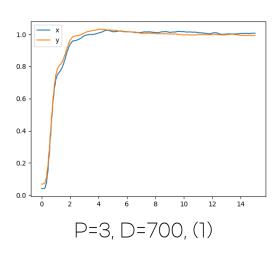
Stuttering occurs with large d(=1000) for small distance

- => p=3, d=1000 shows best performance for large distance
- -> find optimal d parameters with p=3 for small distance



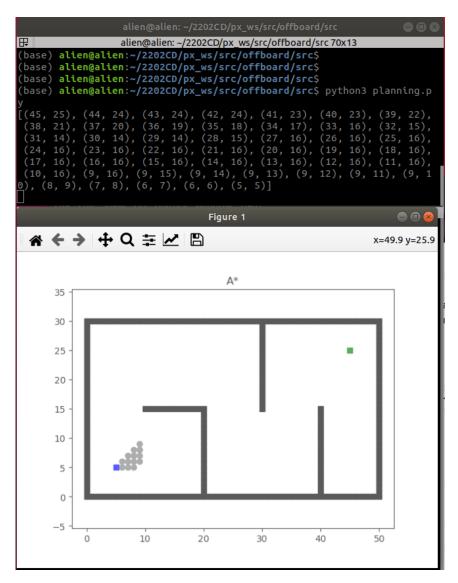






=> p=3, d=500 shows best performance for large distance

# **Planning**



#### Achievements:

#### A\* algorithm implemented

- input: map, start, destination
- output: route to destination

#### ROS embedded

- input:/pp/map,/pp/destination, /pp/cp\_arr,/pp/obs
- output : /pp/checkpoint

#### Dynamic mapping(continuous A\*)

Update map with A\* when obstacle detected

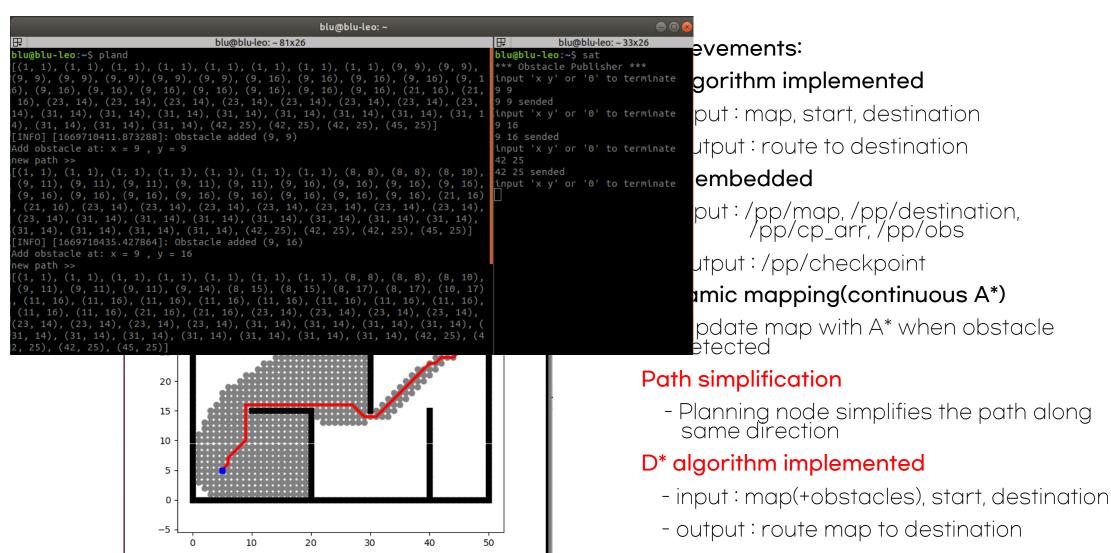
#### Path simplification

- Planning node simplifies the path along same direction

#### D\* algorithm implemented

- input: map(+obstacles), start, destination
- output: route map to destination

# Planning (Dynamic)

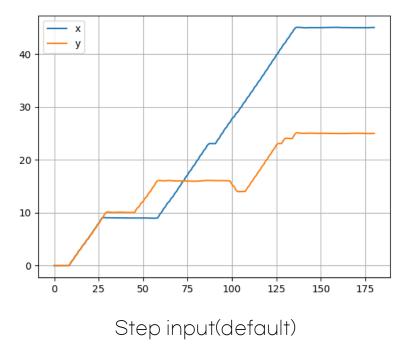


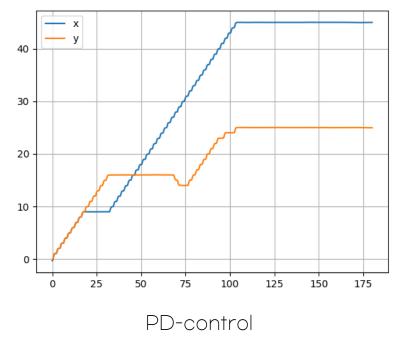
# Planning (Path simplification)

#### Why is path simplification required?

To prevent overshoot and to improve response time, PD-control applied.

But path with unit length(1 or 1.41) made drone continuously stops on each checkpoints.

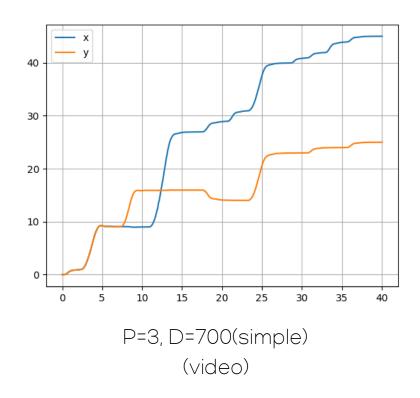


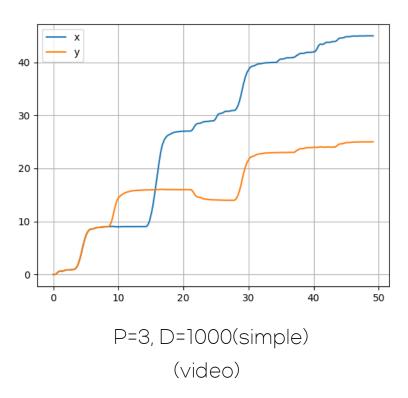


- -> Response time shortened with PD-control: 130s -> 110s
- --> Apply path simplification: Planning node simplifies the path along same direction

# Planning (Path simplification)

Path simplification made checkpoints with large distance; PD parameters had to changed P=3, D=700 / P=3, D=1000 showed best performances.

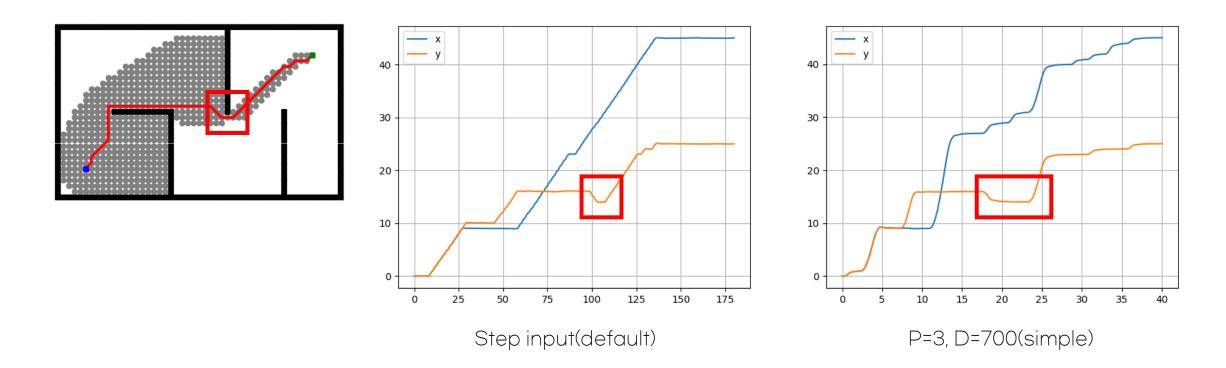




- **=> Response time shortened with simplification**: 110s->37s, 45s
- => No stop-and-go occurred.

# Planning (Path simplification - comparison)

Comparison between step and PD simplified control.



Linear motion is faster with PD-simplified, zig-zag motion is faster with step position control. .

## Demo