

# Autonomous and mobile robot for hydroponic greenhouse picking tasks

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# Introduction

- Agriculture 4.0:
  - new technologies
  - the necessity of more products in a minor time.
- Different solutions presented during these years to overcome population growth.
- Use of robots to perform hard labor in less time with more productivity



# Case of study

- Development of a zucchini flower picking robot for the SolarFields hydroponic greenhouses.
- SolarFields company is a clean energy productor which is investing in hydroponic greenhouse start-ups like Idroluppolo
- These companies are working together to find healthy solution to provide fresh and biological fruits on the market
- Robotics and automation aims to make the production processes fully autonomous



# Existing robots



Root AI



Rubion



Iron Ox



Sweeper Robot

# Project specifications

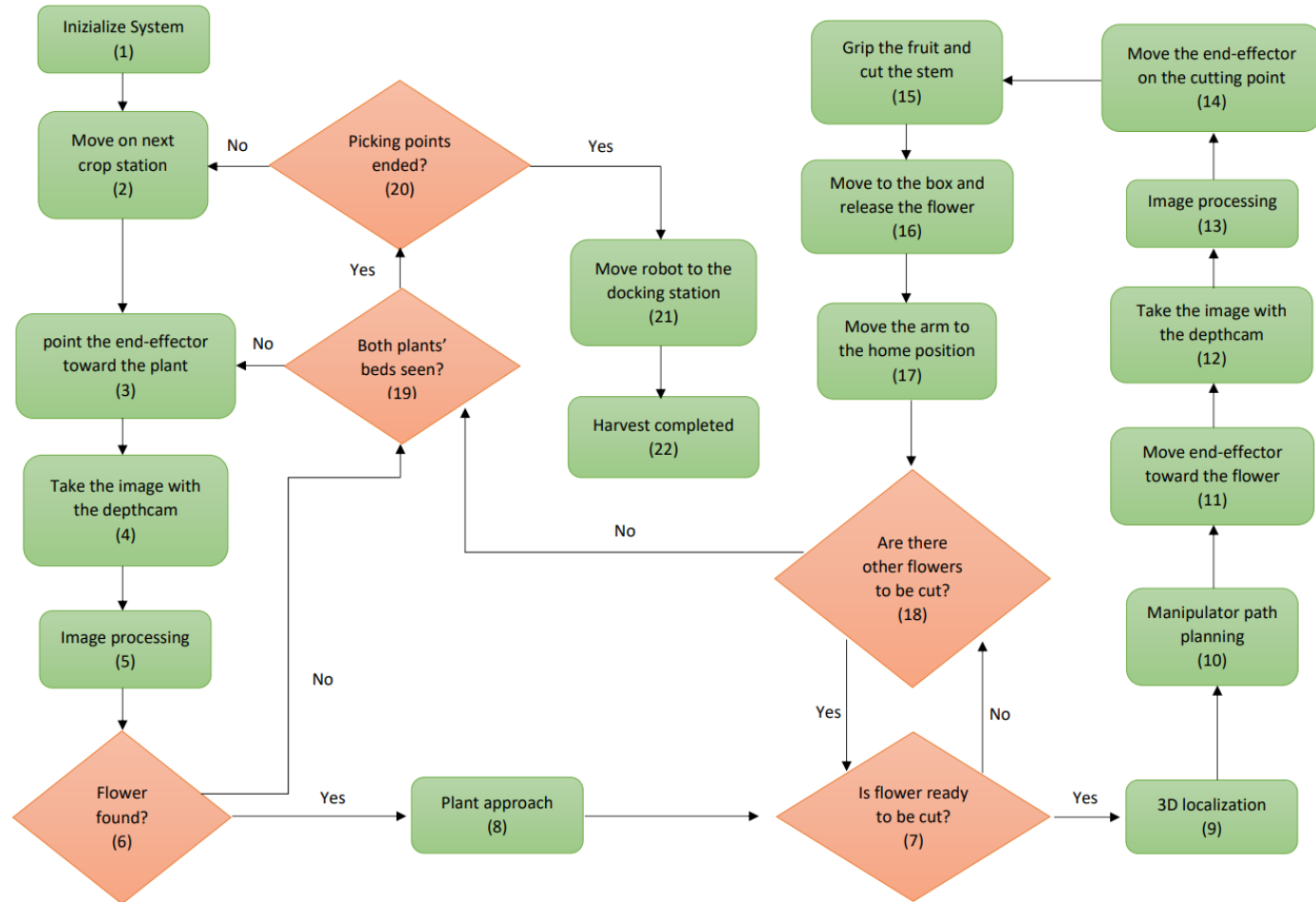
- The robot is composed of three main parts:
  - Mobile platform
  - Robotic arm
  - Vision system
- The aim of the robot is to move inside the greenhouse and pick the zucchini flowers on the plants
- The following work is divided in two main parts:

Feasibility study and  
design of algorithm  
and robot structure

Simulation and control  
of the robot to prove the  
project feasibility



# Algorithm

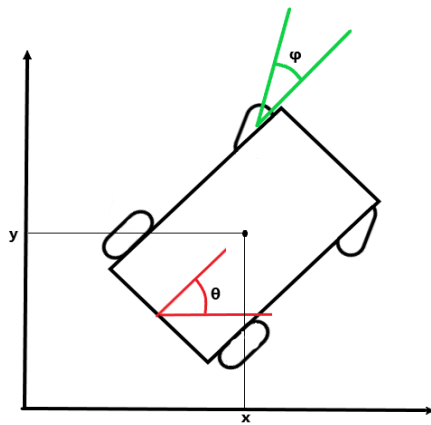


# Greenhouse environment

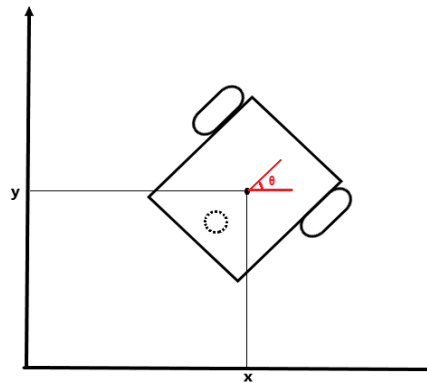


# Mobile Platform

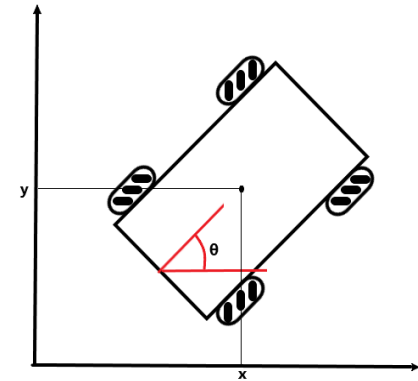
- Three different mobile platforms have been considered:



Car-like robot



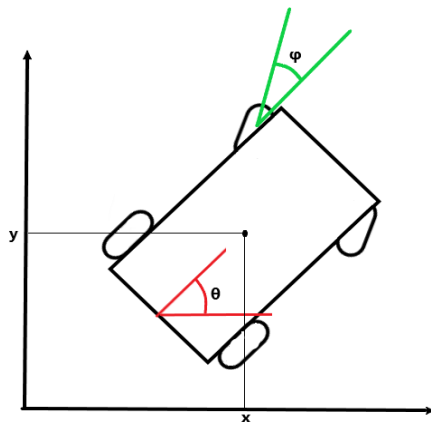
Differential drive  
robot



Mecanum wheels  
robot



# Car-like robot



Car-like robot

- Pro
  - Good stability
  - Good traction force
- Cons
  - Difficult to control in narrow passages
  - Necessity of maneuvers

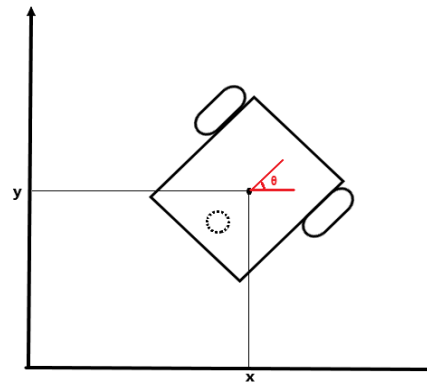
$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \\ \dot{\phi} \end{bmatrix} = \begin{bmatrix} \cos \theta & 0 \\ \sin \theta & 0 \\ \tan \frac{\phi}{l} & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ \omega \end{bmatrix}$$

Front wheel drive kinematic model

# Differential drive robot

- First configuration → two actuated wheels and an omnidirectional wheel

- Pro
  - Simple model
  - Good controllability
- Cons
  - Bad stability due to manipulator arm attached on



Differential drive  
robot

$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} \cos \theta & 0 \\ \sin \theta & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ \omega \end{bmatrix}$$

Unicycle model

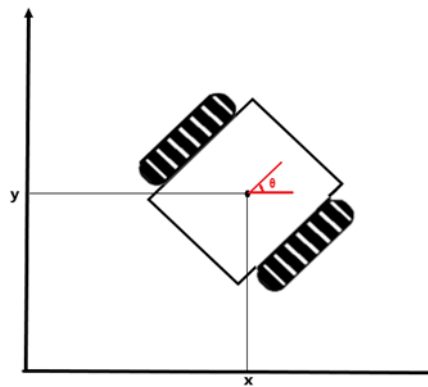
$$\begin{bmatrix} v \\ \omega \end{bmatrix} = \begin{bmatrix} \frac{r}{2}(\omega_r + \omega_l) \\ \frac{r}{d}(\omega_r - \omega_l) \end{bmatrix}$$

Differential drive  
input transformation

# Differential drive robot

- Second configuration → two actuated crawler

- Pro
  - Simple model
  - Good controllability
  - Good stability
  - Good traction
- Cons
  - Expensive



Differential drive  
robot

$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} \cos \theta & 0 \\ \sin \theta & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ \omega \end{bmatrix}$$

Unicycle model

$$\begin{bmatrix} v \\ \omega \end{bmatrix} = \begin{bmatrix} \frac{r}{2}(\omega_r + \omega_l) \\ \frac{r}{d}(\omega_r - \omega_l) \end{bmatrix}$$

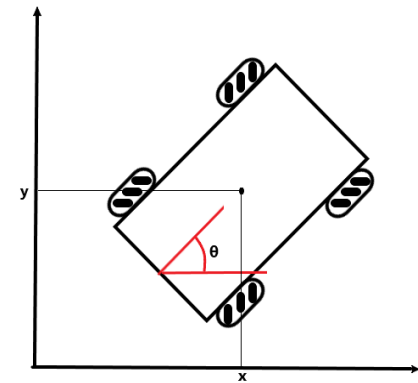
Differential drive  
input transformation

# Mecanum wheel robot

- Pro
  - Omni-directionality
  - Good controllability
  - Good stability
- Cons
  - Bad traction on rough terrain

$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = \frac{r}{4} \begin{bmatrix} 1 & 1 & 1 & 1 \\ -1 & 1 & 1 & -1 \\ -\frac{1}{(d_x+d_y)} & \frac{1}{(d_x+d_y)} & -\frac{1}{(d_x+d_y)} & \frac{1}{(d_x+d_y)} \end{bmatrix} \begin{bmatrix} \omega_{FL} \\ \omega_{FR} \\ \omega_{RL} \\ \omega_{RR} \end{bmatrix}$$

Mecanum wheel robot kinematic model

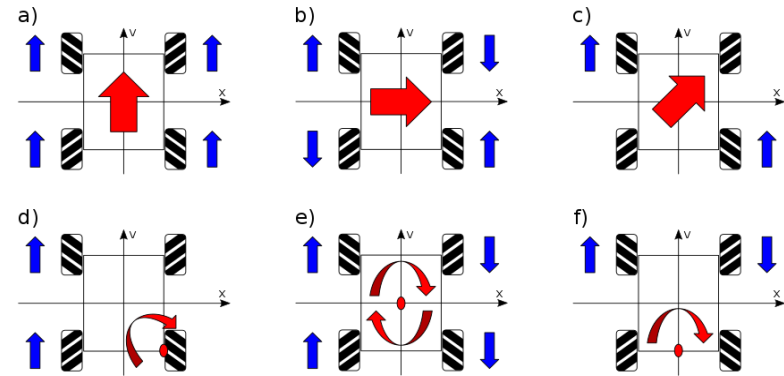


Mecanum wheels  
robot

# Mecanum wheel robot



Mecanum wheel



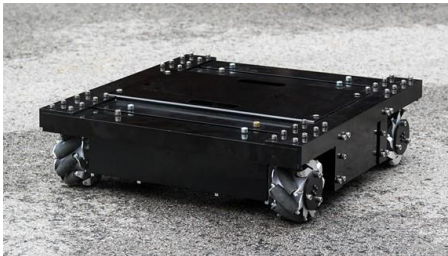
Omni-directionality

$$\begin{bmatrix} \omega_{FL} \\ \omega_{FR} \\ \omega_{RL} \\ \omega_{RR} \end{bmatrix} = \frac{1}{r} \begin{bmatrix} 1 & -1 & -(d_x + d_y) \\ 1 & 1 & (d_x + d_y) \\ 1 & 1 & -(d_x + d_y) \\ 1 & -1 & (d_x + d_y) \end{bmatrix} \begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix}$$

Inverse kinematic model

# Mobile platform prototypes

*OMR10*



- Dimensions(mm):  
480 x 450 x 134
- Load capacity:  
20kg
- Max speed:  
0.6m/s
- Battery life:  
8h
- Charging time:  
4h

*4WD mecanum*



- Dimensions(mm):  
400 x 307 x 123
- Load capacity:  
10kg
- Max speed:  
0.6m/s
- Battery life:  
6h
- Charging time:  
4h
- Sensors:  
4 ultrasonic

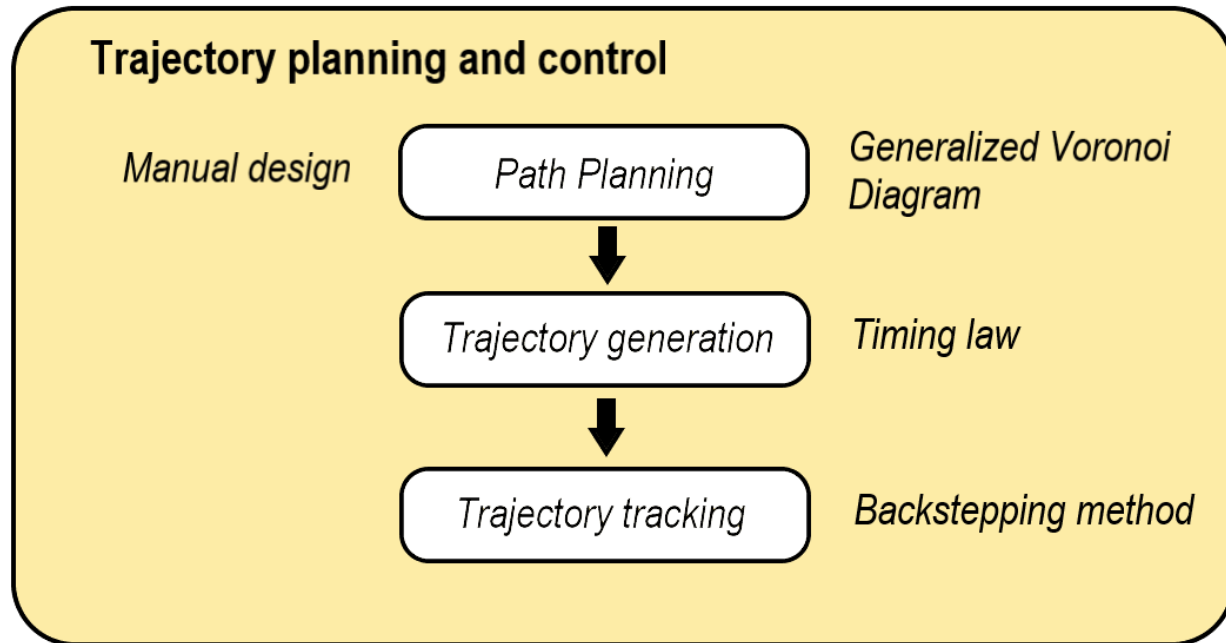
*MP-500*



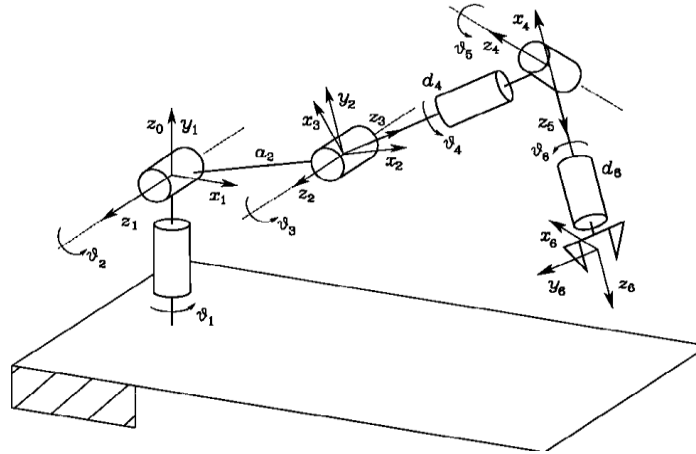
- Dimensions(mm):  
986 x 662 x 409
- Load capacity:  
80kg
- Max speed:  
0.8m/s
- Battery life:  
7h
- Charging time:  
5h
- Sensors:  
2 lidar - 4 ultrasonic



# Trajectory planning and control



# 6 DOF robotic arm



6 DOF robotic arm

- 6 DOF robotic arm has been proven to be enough for the picking task
- The forward and inverse kinematics have been computed for simulation purposes using Denavit-Hartenberg convention
- The differential kinematics has been derived by the forward kinematics and it has been used for the robot control

# 6 DOF Manipulator arm prototypes

*Gen3*



- Payload:  
0.5kg
- Max e-e speed:  
50 cm/s
- Max reach:  
706mm
- Weight:  
5.4kg

*Gen3 Light*



- Payload:  
2kg
- Max e-e speed:  
50 cm/s
- Max reach:  
891mm
- Weight:  
7.2kg

*Kinova RGB-D*



- RGB sensor:
  - Resolution:  
1980x1080 at  
30fps
- Depth sensor
  - Resolution:  
480x270 at  
30fps
  - Min distance:  
18cm

# Vision System

- One of the fundamental modules for the picking task.
- RGB-D camera chosen based on the ToF technology
- Recognition, identification and localization of the flowers inside the environment.
- Two objectives:
  - Flower recognition
  - Flower separation and localization
- Two algorithm have been developed to accomplish these tasks:
  - K-mean algorithm for color clustering
  - Moore-Neighbor tracing algorithm modified by Jacob's stopping criteria for flower separation

# Flower identification



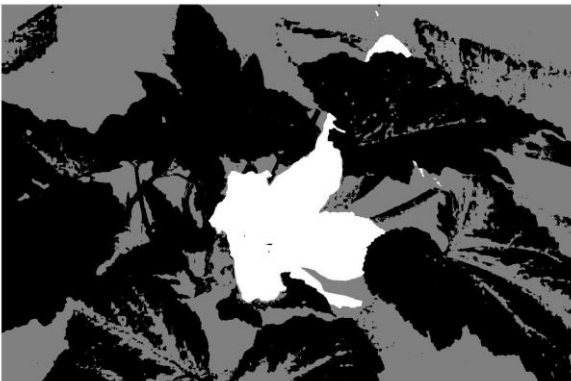
Original image



First cluster



Second cluster



Labeled image



Third cluster

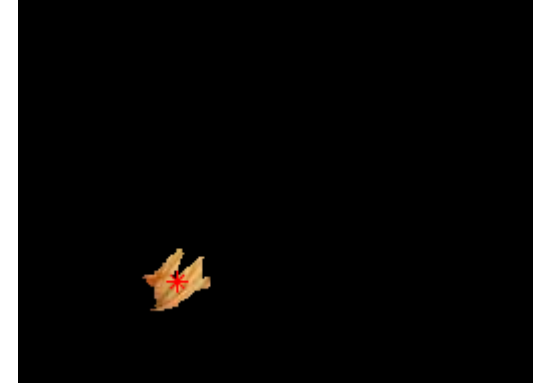
# Flower separation and localization



Flower cluster



Labeled image



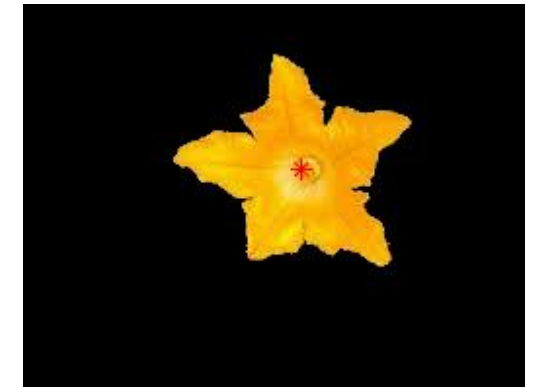
First flower



Binarization, noise  
attenuation and hole filling

$$z_{\epsilon} = D(X_I, Y_I)$$
$$x_{\epsilon} = \frac{z_{\epsilon}(X_I - C_x)}{f_x}$$
$$y_{\epsilon} = \frac{z_{\epsilon}(Y_I - C_y)}{f_y}$$

Flower 3D localization



Second flower



# Trajectory generation

## Path planning

- Position path

$$p(s) = p_i + s \frac{p_\varepsilon}{\|p_\varepsilon\|}$$

$$\dot{p}(s) = \dot{s} \frac{p_\varepsilon}{\|p_\varepsilon\|}$$

- Orientation path

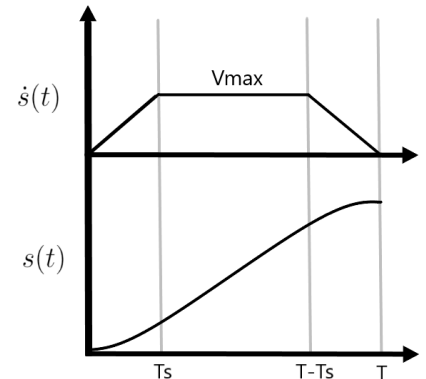
$$\phi(s) = \phi_i + s \frac{\phi_\varepsilon}{\|\phi_\varepsilon\|}$$

$$\dot{\phi}(s) = \dot{s} \frac{\phi_\varepsilon}{\|\phi_\varepsilon\|}$$

- Final path

$$\dot{x}_d(s) = \begin{bmatrix} \dot{p}(s) \\ \dot{\phi}(s) \end{bmatrix}$$

## Timing law

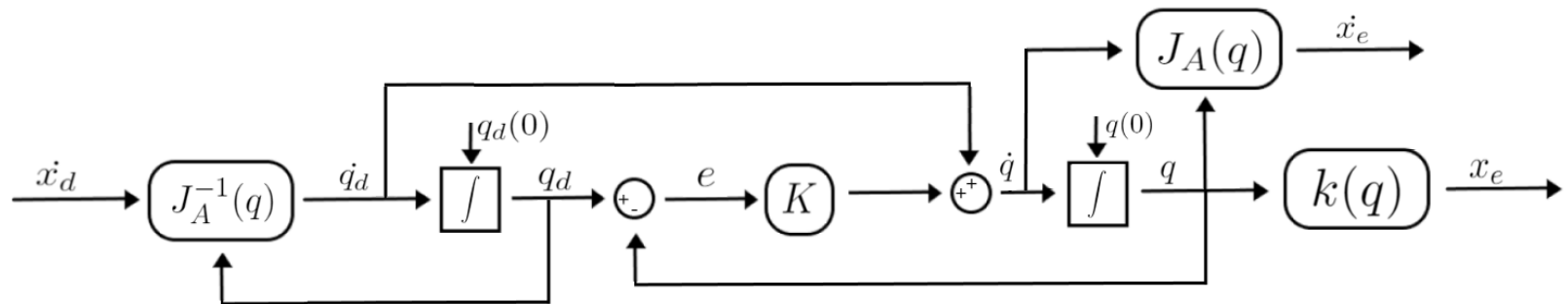


$$T_s = \frac{V_{max}}{A_{max}}$$

$$T = \frac{lA_{max} + V_{max}^2}{A_{max}V_{max}}$$

$$\dot{s}(t) = \begin{cases} A_{max}t, & t \in [0, T_s] \\ V_{max}, & t \in [T_s, T - T_s] \\ -A_{max}(t - T), & t \in [T - T_s, T] \end{cases}$$

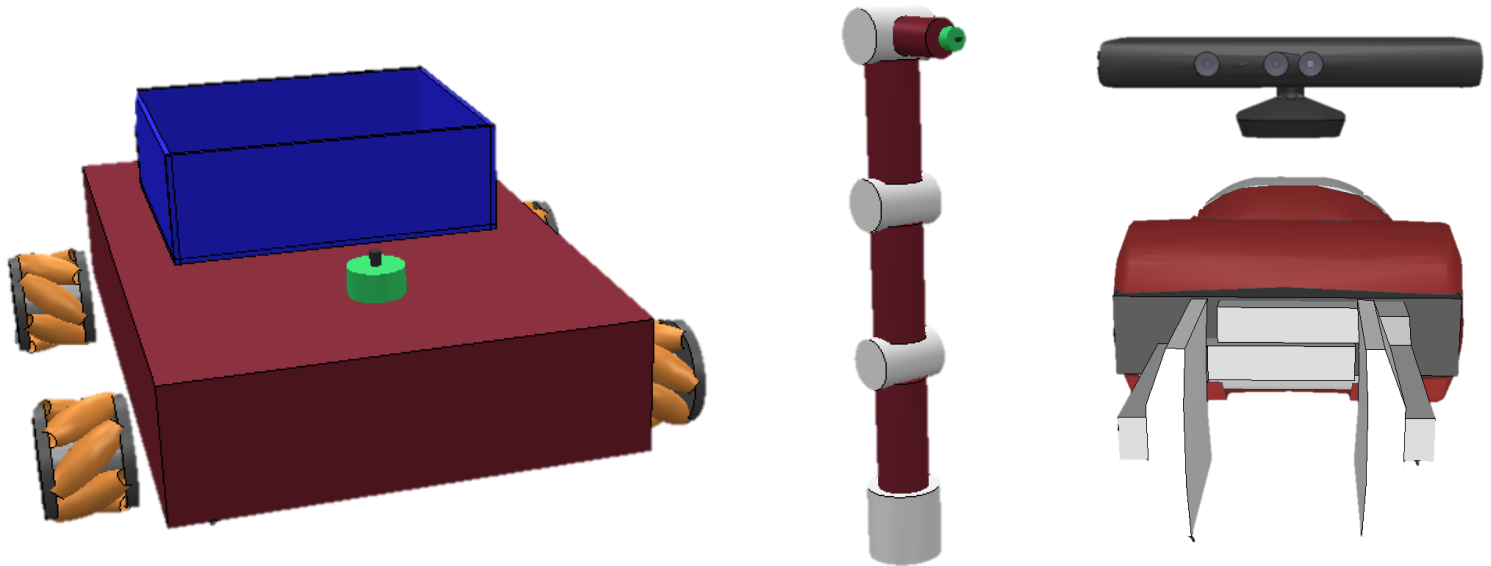
# Robotic arm kinematic control



- A kinematic controller has been used to control the robot
- The input of the control loop is the cartesian speed trajectory of the robot's end-effector
- The trajectory is converted from the speed cartesian space to the speed joint space using the robot inverse analytic Jacobian
- Two contribution for the robot input:
  - Feedforward  $\rightarrow \dot{q}_d$
  - Feedback  $\rightarrow K(q_d - q)$

# Simulation

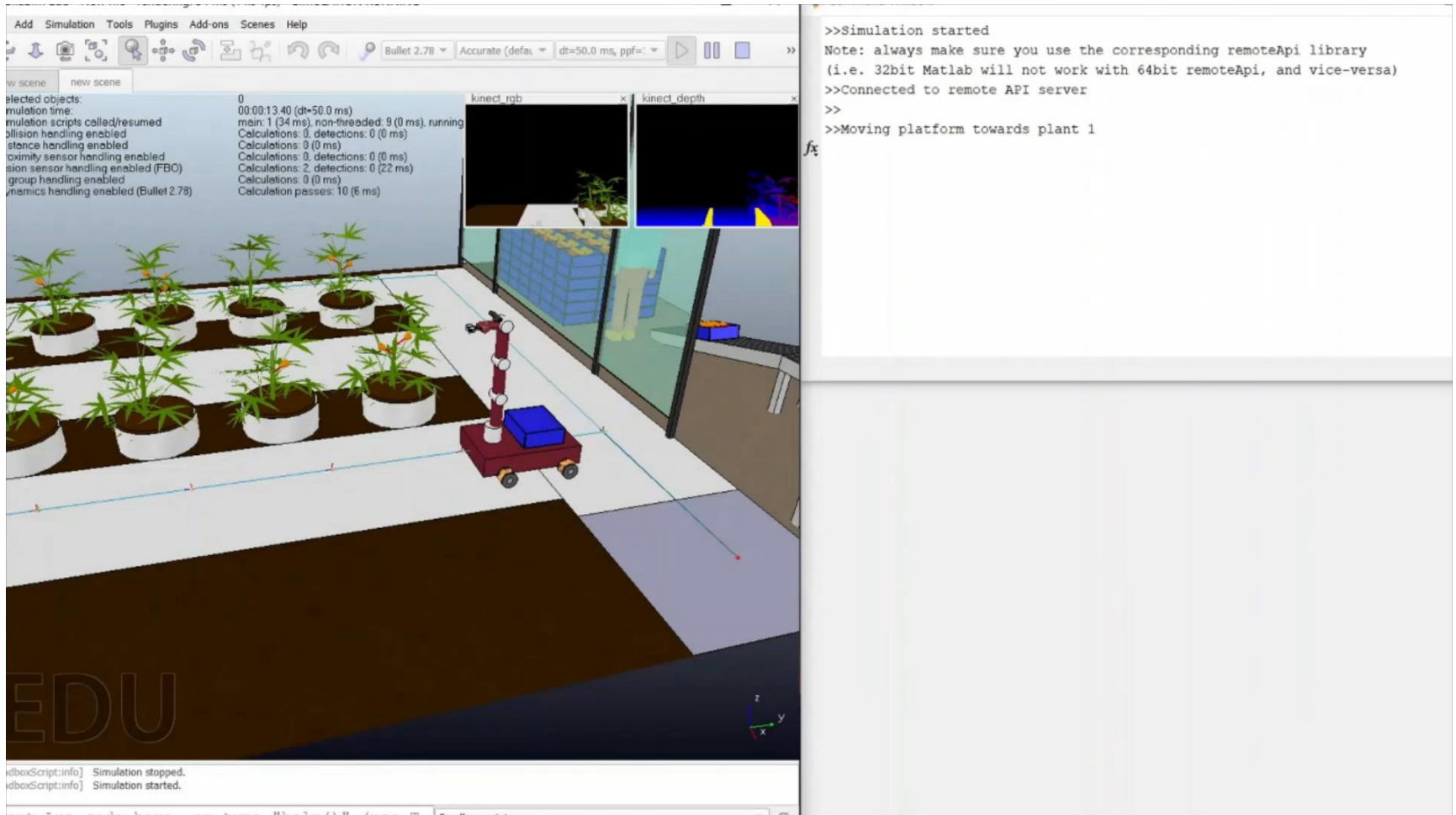
- Tools
  - Matlab
  - Vrep
- The robot mobile platform and its 6 DOF manipulator arm have been developed on Vrep using joints and primitive shapes



# Simulation



# Simulation



# Conclusion

- Although the project has been simplified, the simulation had proven that the algorithms and the structure used works. This project could be applied to a real application scenario in an agriculture 4.0 hydroponic greenhouse.
- Future improvements:
  - Implement a localization systems
  - Improve the machine learning algorithms
  - Use multiple robots - distributed systems
  - Use a more sophisticated robot controller
  - Others