

Modeling and Kinematic Control of an Omnidrive indoor Robot Platform

Objective

- Realizing an autonomous Omnidrive platform
- Modeling the kinematic control
- Identifying the system parameters
- Data fusion for path tracking

Description of the system

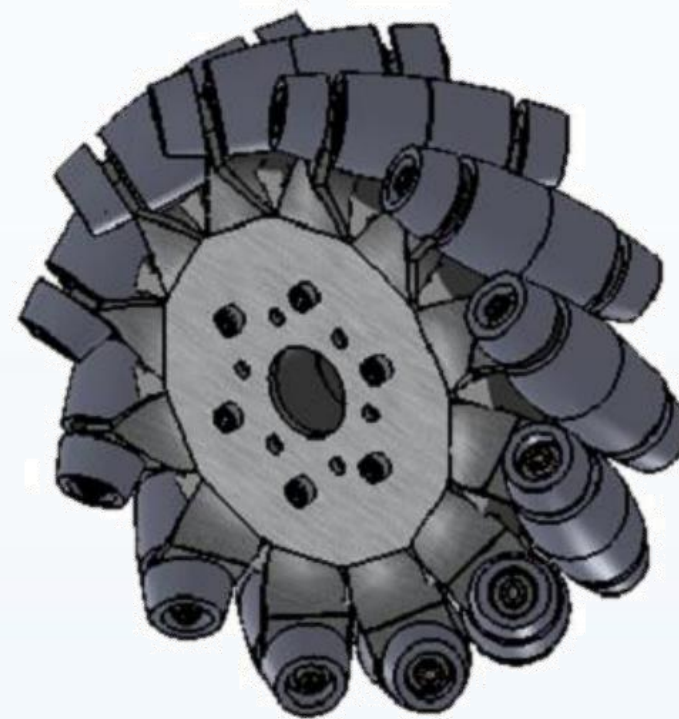
We choose to develop an omni-directional platform that will be added to the robot arm used for a medical application of Curietherapy.



InTraDE omnidrive robot platform.

Mechanical Design

- Made of aluminum profile,
- A rectangle configuration with a 900 [mm] long and 700 [mm] wide,
- Two mecanum wheels on each side of the chassis, actuated by its own DC MAXON motor.



Mecanum wheel

Electronic Design

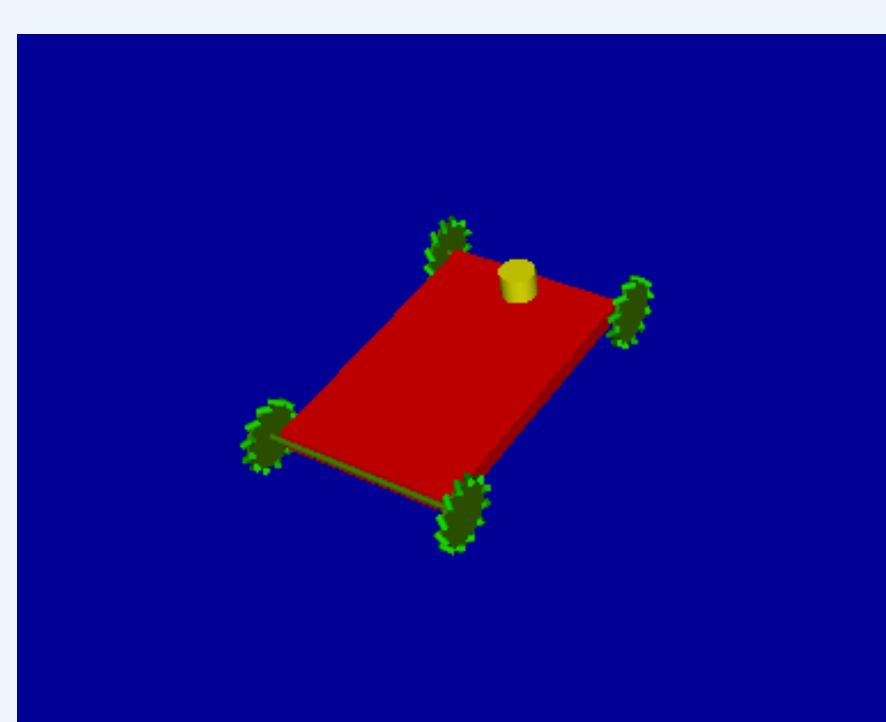
- compactRIO: real-time embedded industrial controller made by National Instruments,
- ESCON Servo Controller 50/5: equipped with a PI regulator and an auto tuning functionality,
- Xsens MTi: Inertial which contains accelerometer, gyroscope, magnetometer.



NI compactRIO

Software Design

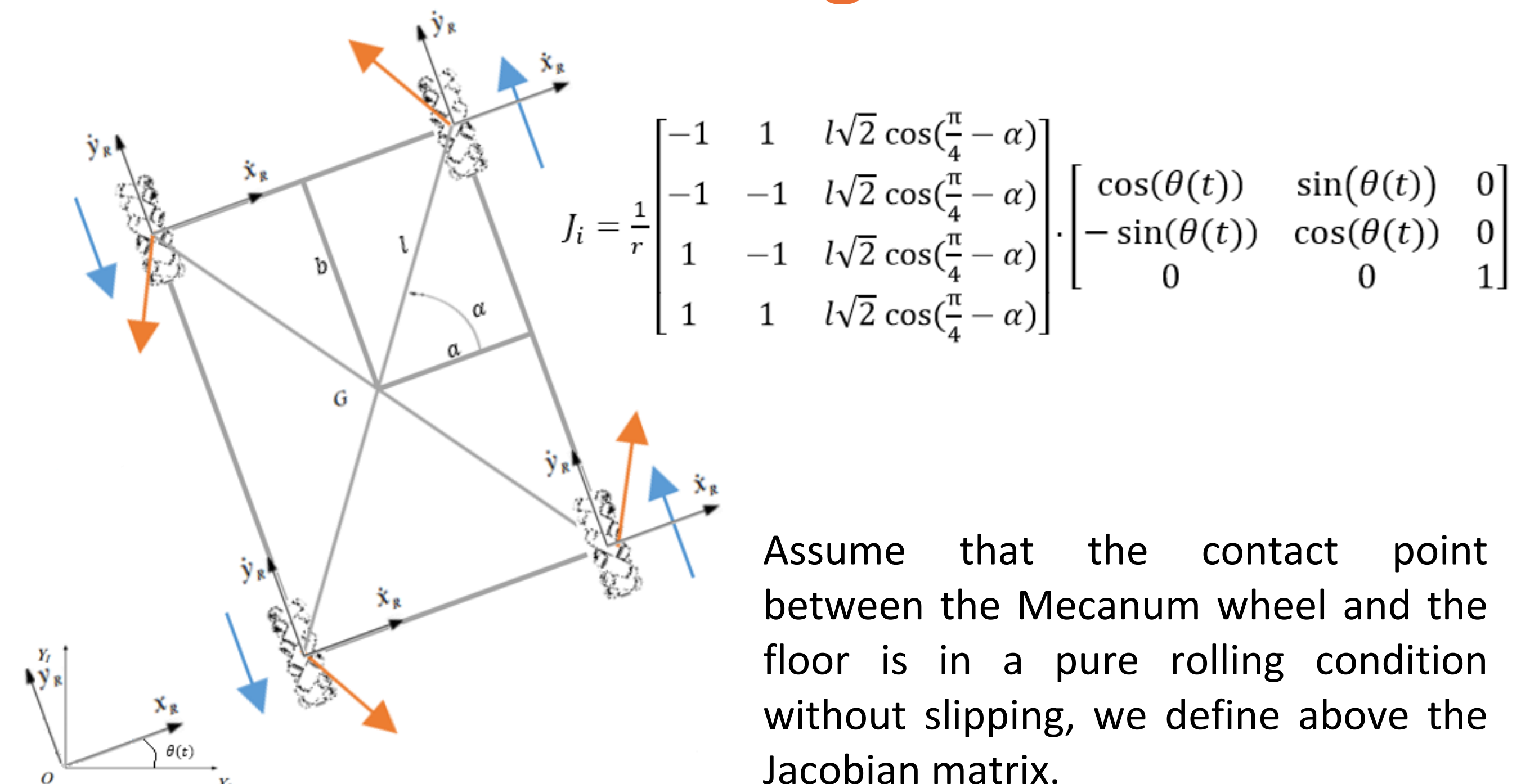
- MATLAB/Simulink: simulation of the inverse and the forward kinematic model and the trajectory tracking,
- LabVIEW Robotics: 3D simulation of the robot motion,
- LabVIEW Real Time implementation of signal processing into the compactRIO.



3D simulation

Methods and Results

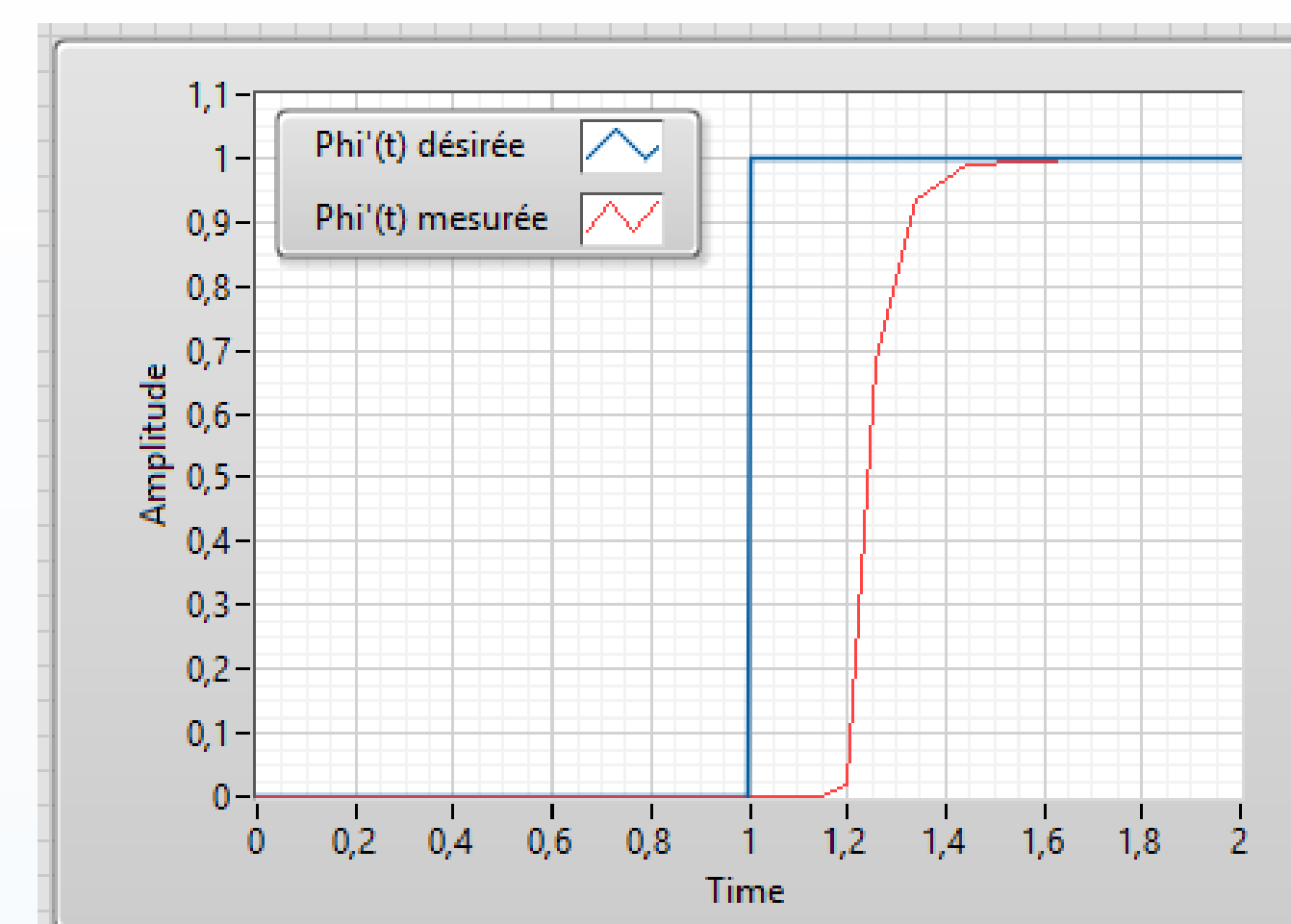
Kinematics Modeling



Schematic of the Mecanum robot

Assume that the contact point between the Mecanum wheel and the floor is in a pure rolling condition without slipping, we define above the Jacobian matrix.

System response identifying



System step response.

Using LabVIEW, we plot the step response of the system.

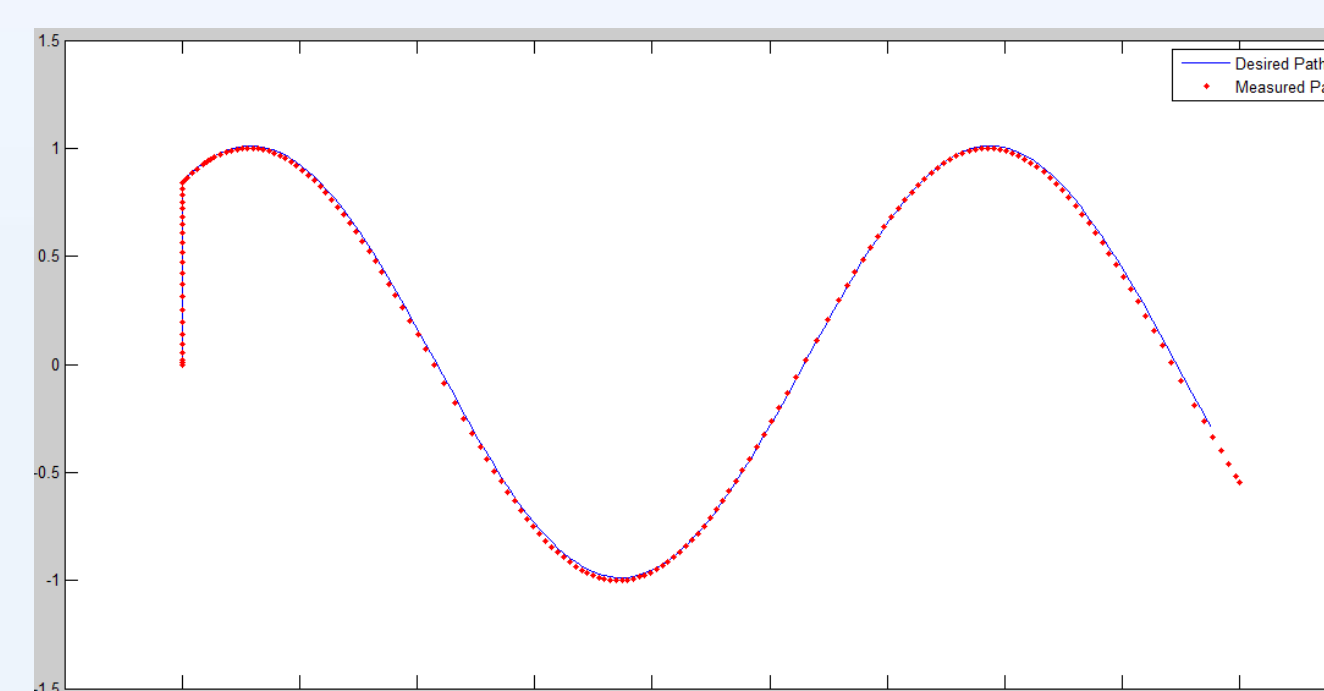
Assume we have a first order system with delay, we get:

$$\frac{\dot{\varphi}^{mesurée}(p)}{\dot{\varphi}^{désirée}(p)} = \frac{e^{-0.2p}}{1 + 0.05p}$$

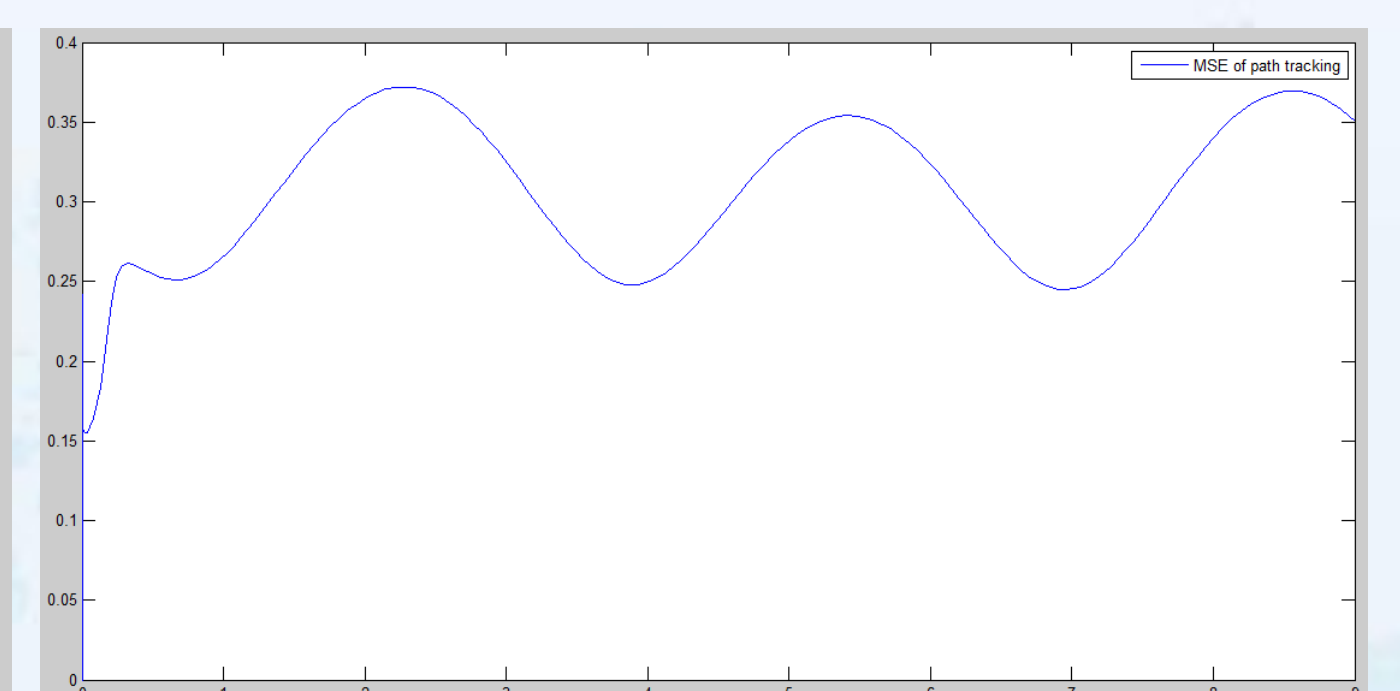
Path tracking

Using MATLAB/Simulink, we simulate the path tracking of the trajectory below:

$$x(t) = \begin{cases} 0 & \text{si } t < 1 \\ t & \text{si } t \geq 1 \end{cases} \quad y(t) = \begin{cases} 0 & \text{si } t < 0 \\ \sin(t) & \text{si } t \geq 0 \end{cases} \quad \theta(t) = 0 \quad \forall t$$



Desired and measured trajectory



Mean squared error of path tracking

Data fusion

The dead-reckoning method using rotary encoders depends on friction of ground surface and it has an accumulative error.

We estimate the position of the vehicle by fusing odometry and inertial dead-reckoning method based on Kalman filter technique as below:

$$\begin{cases} \dot{\theta}(t) = 0 \cdot \theta(t) + B \begin{pmatrix} \dot{\varphi}_1^{mesuré}(t) \\ \dot{\varphi}_2^{mesuré}(t) \\ \dot{\varphi}_3^{mesuré}(t) \\ \dot{\varphi}_4^{mesuré}(t) \end{pmatrix} + W \\ g^{mesuré} = 1 \cdot \theta(t) + V \end{cases}$$

Where
 $\dot{\varphi}_i^{mesuré}(t)$, $i = \{1,2,3,4\}$ angular velocity of each wheel measured by the encoder,
 $\theta^{mesuré}(t)$ robot orientation measured by the inertial,
 $B = \frac{r}{4l\sqrt{2} \cos(\frac{\pi}{4} - \alpha)} (1 \ 1 \ 1 \ 1)$ input matrix,
 W : state noise
 V : measurement noise.