

Robotics & XR

5 ECTS
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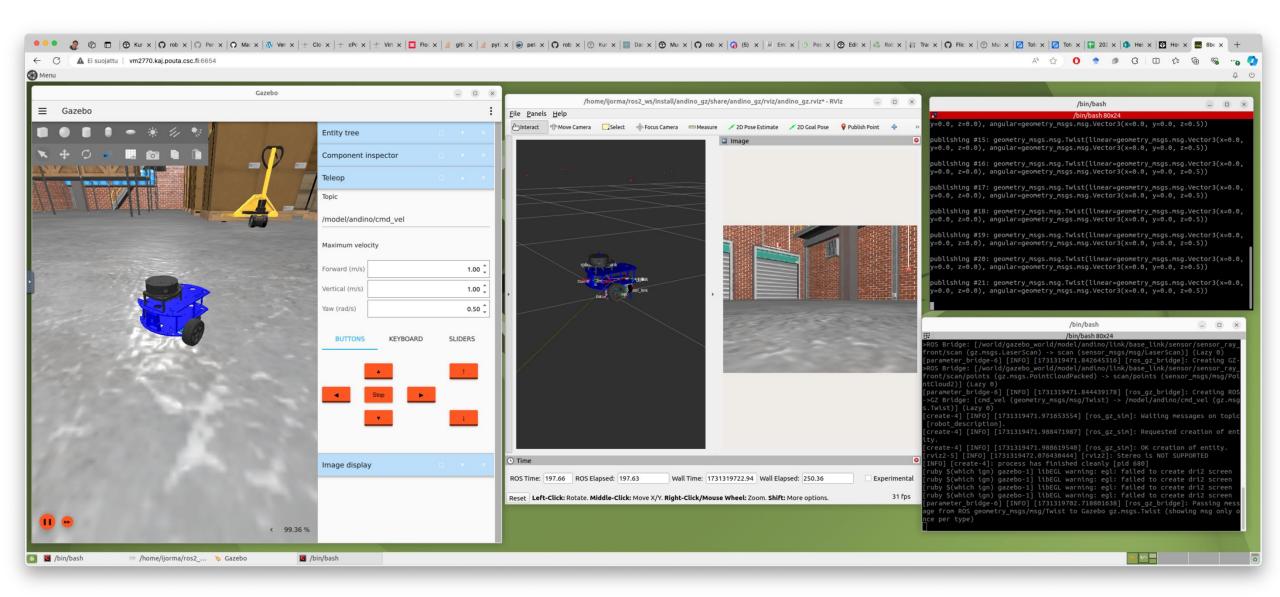
Robot control theory November 11, 2024

Course contents

- Theme 1: The role of robotics in society and an introduction to ethical issues in robotics
- Theme 2: Different types of robots and their application to realworld problems
- Theme 3: Robot control theory
- Theme 4: Navigation
- Theme 5: Robotics & Al
- Theme 6: Extended Reality applications in robotics

Some recap from the last week

- ROS 2 as framework
- Topics, nodes, actions, services
- Issues encountered
 - Docker / ROS installation and configuration
 - Running simulation is GPU intensive
 - Varying platforms
- Give us feedback: https://shorturl.at/SDaop (link is also in Moodle)
- A (partial) solution: remotely accessible sandbox
- Demo



Theme 3: Robot control theory

- Week 46 (November 11 − 17)
- Theory of robot control
- Basics of matrix computation and linear algebra
- Odometry and sensor fusion, applicable filters

Core concepts

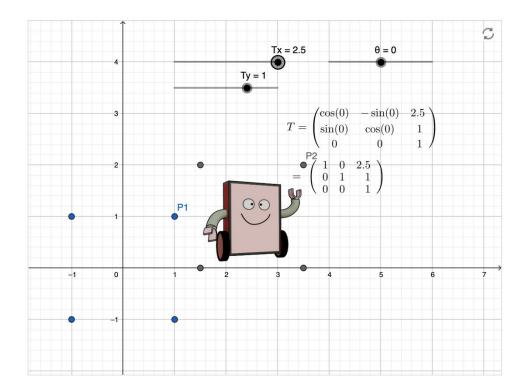
- Transformations (tf)
- Frames
- Odometry
- Mapping
- Localization
- Navigation

Transformations (tf)

- Provide information on how a robot could move in n-dimensional space
- A mobile robot typically moves in 2D space and is able to rotate (3 DOF)
- Translation: Changing position in Cartesian 2D coordinate system
- Rotation: Rotating the object in the space
- Scaling, shearing, mirroring (manipulating size and shape of the object)
- Useful not only in robotics, but also in computer graphics and game design

Transformations

• Transformation matrix – rotate and translate



https://articulatedrobotics.xyz/tutorials/coordinate-transforms/transformation-matrices

Breaking down the matrix

• Representing a point in 2D / 3D Cartesian space

$$\mathbf{p}_{ ext{2D}} = egin{bmatrix} x \ y \end{bmatrix}, \mathbf{p}_{ ext{3D}} = egin{bmatrix} x \ y \ z \end{bmatrix}$$

$$\mathbf{p}_k = egin{bmatrix} x_k \ y_k \end{bmatrix}$$

Linear transformation and identity matrix

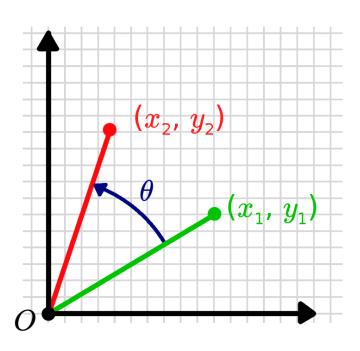
- *f*(**p**)=**Ap**
 - where A is an n x n matrix, n is number of dimensions in p
- If the function is something else than two matrices multiplied, it is non-linear
- When multiplying a point matrix with identity matrix, we get the same result

$$\begin{bmatrix} a \end{bmatrix}, \begin{bmatrix} a & b & c \ c & d \end{bmatrix}, \begin{bmatrix} a & b & c \ d & e & f \ g & h & i \end{bmatrix}$$

$$egin{aligned} \mathbf{p}_2 &= \mathbf{A}\mathbf{p}_1 \ egin{bmatrix} x_2 \ y_2 \end{bmatrix} &= egin{bmatrix} a & b \ c & d \end{bmatrix} egin{bmatrix} x_1 \ y_1 \end{bmatrix} \ &= egin{bmatrix} ax_1 + by_1 \ cx_1 + dy_1 \end{bmatrix} \end{aligned}$$

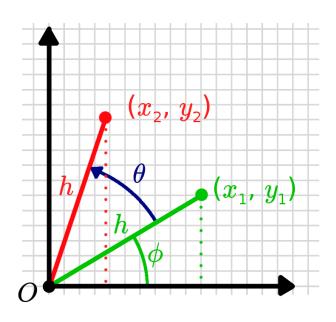
$$egin{bmatrix} x_2 \ y_2 \end{bmatrix} = egin{bmatrix} 1 & 0 \ 0 & 1 \end{bmatrix} egin{bmatrix} x_1 \ y_1 \end{bmatrix} \ = egin{bmatrix} x_1 \ y_1 \end{bmatrix}$$

Rotation – defining transformation matrix



$$egin{bmatrix} x_2 \ y_2 \end{bmatrix} = egin{bmatrix} a & b \ c & d \end{bmatrix} egin{bmatrix} x_1 \ y_1 \end{bmatrix} = egin{bmatrix} ax_1 + by_1 \ cx_1 + dy_1 \end{bmatrix}$$

Rotation – defining transformation matrix



$$egin{bmatrix} x_2 \ y_2 \end{bmatrix} = egin{bmatrix} a & b \ c & d \end{bmatrix} egin{bmatrix} x_1 \ y_1 \end{bmatrix} = egin{bmatrix} ax_1 + by_1 \ cx_1 + dy_1 \end{bmatrix}$$

$$egin{bmatrix} x_1 \ y_1 \end{bmatrix} = egin{bmatrix} h\cos(\phi) \ h\sin(\phi) \end{bmatrix}$$

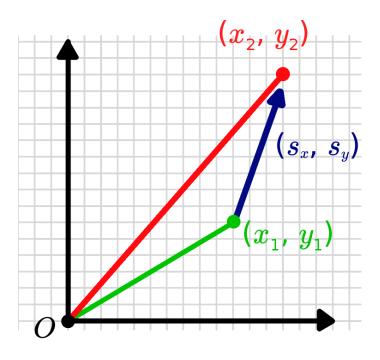
$$\begin{bmatrix} x_2 \\ y_2 \end{bmatrix} = \begin{bmatrix} h\cos(\phi + \theta) \\ h\sin(\phi + \theta) \end{bmatrix}$$

$$= \begin{bmatrix} h\cos(\phi)\cos(\theta) - h\sin(\phi)\sin(\theta) \\ h\sin(\phi)\cos(\theta) + h\cos(\phi)\sin(\theta) \end{bmatrix}$$

$$= \begin{bmatrix} x_1\cos(\theta) - y_1\sin(\theta) \\ x_1\sin(\theta) + y_1\cos(\theta) \end{bmatrix}$$

$$\begin{bmatrix} x_2 \\ y_2 \end{bmatrix} = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} \begin{bmatrix} x_1 \\ y_1 \end{bmatrix}$$

Translation – moving in coordinates



- Non-linear function f(p)=p+b
- Use of homogenous coordinates help to derive translation matrix (linear function) – affine transformation

$$egin{bmatrix} ar{\mathbf{p}}_2 &= \mathbf{A}ar{\mathbf{p}}_1 = ar{\mathbf{p}}_1 + \mathbf{s} \ egin{bmatrix} x_2 \ y_2 \ 1 \end{bmatrix} &= egin{bmatrix} ? & ? & ? \ ? & ? & ? \ ? & ? & ? \end{bmatrix} egin{bmatrix} x_1 \ y_1 \ 1 \end{bmatrix} &= egin{bmatrix} x_1 + s_x \ y_1 + s_y \ 1 \end{bmatrix}$$

$$egin{bmatrix} 1 & 0 & s_x \ 0 & 1 & s_y \ 0 & 0 & 1 \end{bmatrix} egin{bmatrix} x_1 \ y_1 \ 1 \end{bmatrix} = egin{bmatrix} x_1 + s_x \ y_1 + s_y \ 1 \end{bmatrix}$$

Transformation matrices

- First, we need to augment rotation matrix to homogenous coordinates
- Integrating with translation matrix

$$\begin{bmatrix} x_2 \\ y_2 \\ 1 \end{bmatrix} = \begin{bmatrix} \cos(\theta) & -\sin(\theta) & 0 \\ \sin(\theta) & \cos(\theta) & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ y_1 \\ 1 \end{bmatrix}$$

$$\mathbf{T} = ar{\mathbf{B}}ar{\mathbf{R}} = egin{bmatrix} 1 & 0 & s_x \\ 0 & 1 & s_y \\ 0 & 0 & 1 \end{bmatrix} egin{bmatrix} \cos(heta) & -\sin(heta) & 0 \\ \sin(heta) & \cos(heta) & 0 \\ 0 & 0 & 1 \end{bmatrix} = egin{bmatrix} \cos(heta) & -\sin(heta) & s_x \\ \sin(heta) & \cos(heta) & s_y \\ 0 & 0 & 1 \end{bmatrix}$$

Core concepts

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Frame

- A reference point in 2D or 3D space that helps to understand and describe position and orientation of an object
- Every object in a robotics application has its own frame
 - Coordinate system to measure object's poosition and orientation
- Frames are either fixed (e.g. map) or mobile (e.g. robot's base frame)
- Frames hierarchy
 - Robot's frame attached to its base -> frame moves when the robot moves
 - Robot's accessories are mounted to this frame -> they also move
- ROS2: tf2 library keeps track on frames and how they move in the space
- Transformation tree: relations between the frames (e.g. map->robot)

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Odometry

- Robot tracks its position based on previous positions and its movements (translations and rotations in space)
- Multiple sensor inputs (e.g. wheel encoders, IMU, gyroscope)
- Accumulating movements in the space
- Relying only one type of sensor (for example wheel encoder) is usually unreliable (why?)
- How to improve odometry?

Sensor fusion

- Combining data from various sources
 - Reducing uncertainty and increasing robustness of the system
- Camera, LIDAR, GPS, IMU, encoders, ...
- Kalman filters (and other algorithms) and/or sensor data weighting to combine data efficiently
- If one sensor fails, the others continue providing information
- Noise reduction and increased precision
- Consider autonomous cars: What sensors might be included in fusion?