Locomotion of a Mobile Robot in a 2D Space

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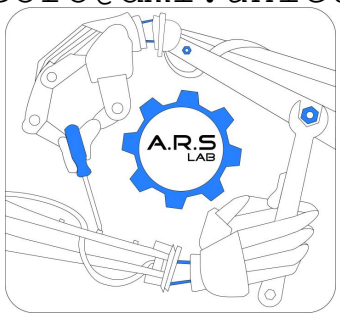


Robotic Systems

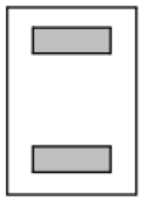
Locomotion of a Robot in a 2-Dimensional Space

Model

- According to the desired freedom degrees, a wide range of **locomotion model exists**
- Each model consider a certain number of wheels (2, 3, 4, etc.)
- and a certain kind of wheels (traction, free, steering, castor, omniball, omnidirectional, etc.)

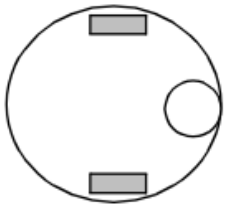
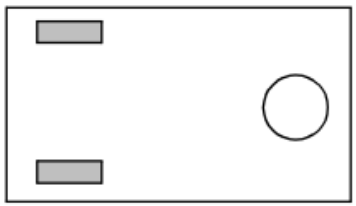
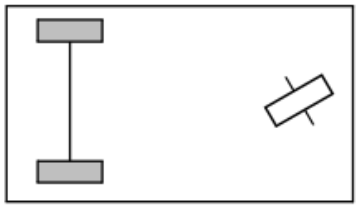
Locomotion of a Robot in a 2-Dimensional Space

Two Wheels

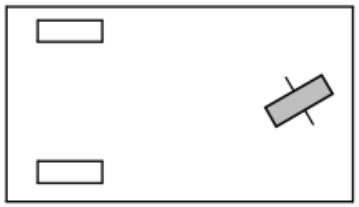
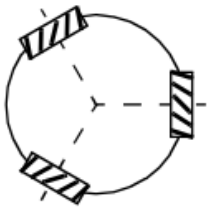
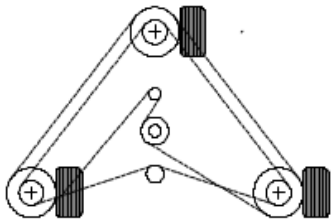
2		One steering wheel in the front, one traction wheel in the rear	Bicycle, motorcycle
		Two-wheel differential drive with the center of mass (COM) below the axle	Cye personal robot

Locomotion of a Robot in a 2-Dimensional Space

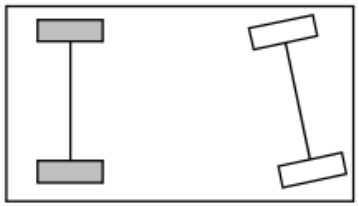
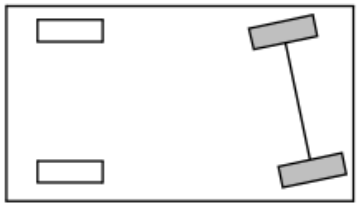
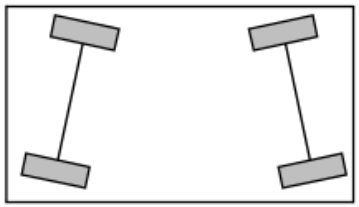
Three Wheels

3		Two-wheel centered differential drive with a third point of contact	Nomad Scout, smartRob EPFL
		Two independently driven wheels in the rear/front, 1 unpowered omnidirectional wheel in the front/rear	Many indoor robots, including the EPFL robots Pygmalion and Alice
		Two connected traction wheels (differential) in rear, 1 steered free wheel in front	Piaggio minitrucks

Three Wheels

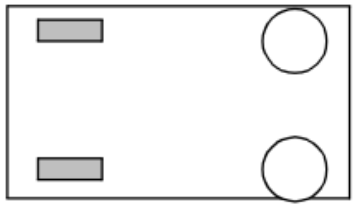
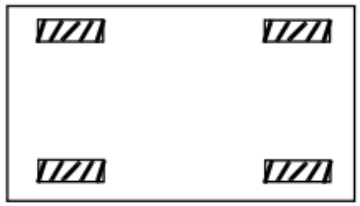
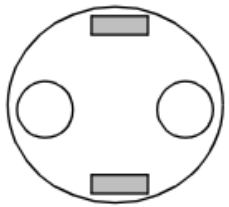
	Two free wheels in rear, 1 steered traction wheel in front	Neptune (Carnegie Mellon University), Hero-1
	Three motorized Swedish or spherical wheels arranged in a triangle; omnidirectional movement is possible	Stanford wheel Tribolo EPFL, Palm Pilot Robot Kit (CMU)
	Three synchronously motorized and steered wheels; the orientation is not controllable	"Synchro drive" Denning MRV-2, Georgia Institute of Technology, I-Robot B24, Nomad 200

Four Wheels

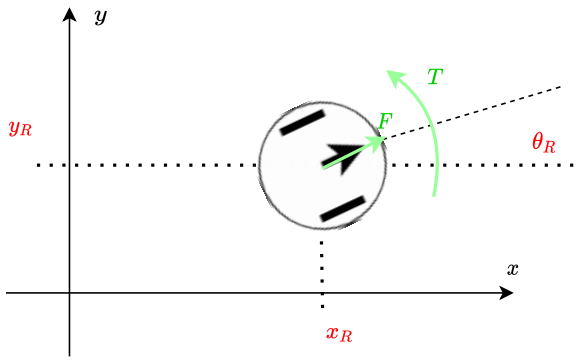
4		Two motorized wheels in the rear, 2 steered wheels in the front; steering has to be different for the 2 wheels to avoid slipping/skidding.	Car with rear-wheel drive
		Two motorized and steered wheels in the front, 2 free wheels in the rear; steering has to be different for the 2 wheels to avoid slipping/skidding.	Car with front-wheel drive
		Four steered and motorized wheels	Four-wheel drive, four-wheel steering Hyperion (CMU)

Locomotion of a Robot in a 2-Dimensional Space

Four Wheels

	Two traction wheels (differential) in rear/front, 2 omnidirectional wheels in the front/rear	Charlie (DMT-EPFL)
	Four omnidirectional wheels	Carnegie Mellon Uranus
	Two-wheel differential drive with 2 additional points of contact	EPFL Khepera, Hyperbot Chip

Locomotion of a Robot in a 2-Dimensional Space

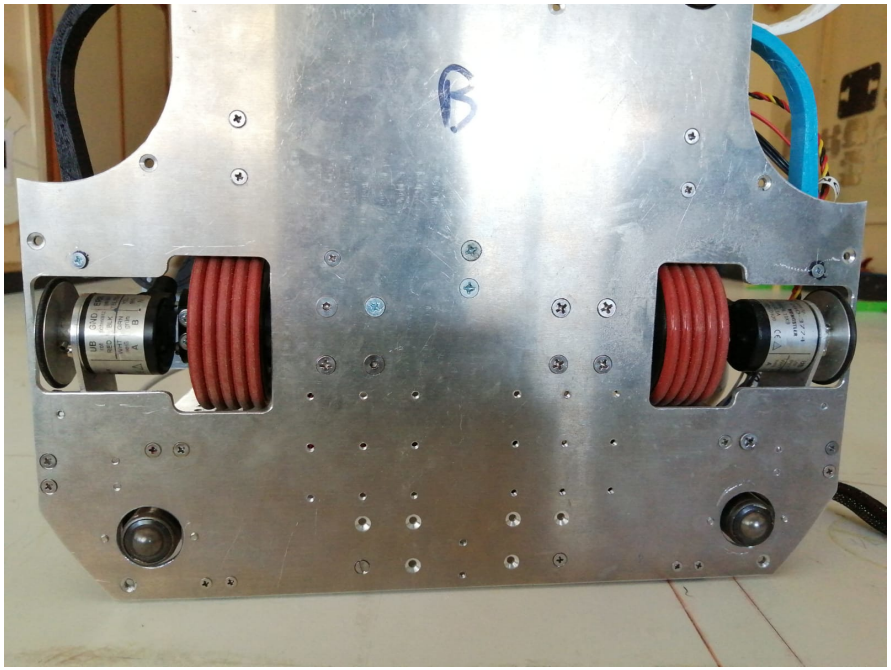


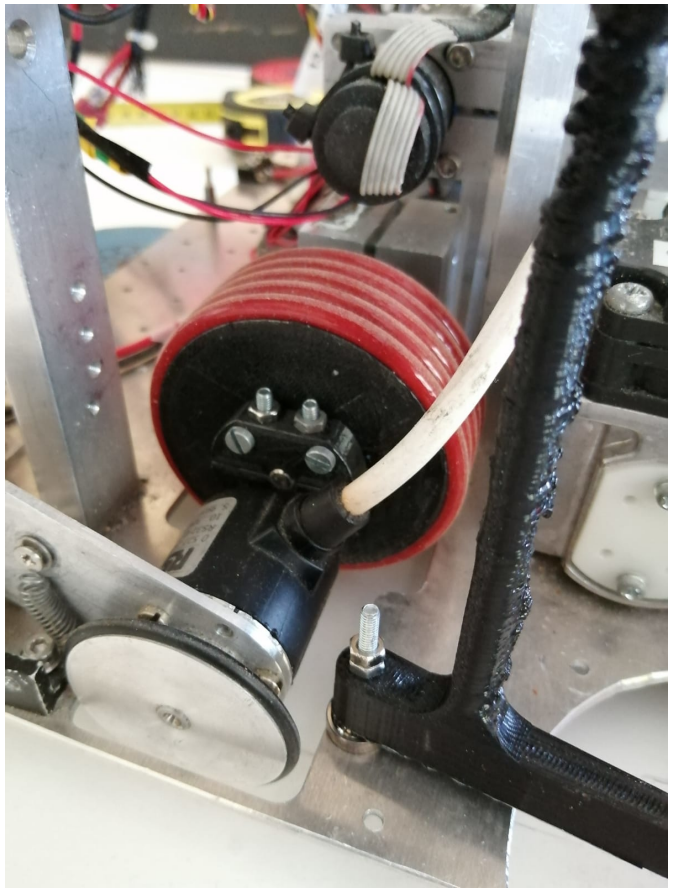
Wheels and Dynamic Model

- For each type of locomotion system, the kinematic and dynamic model must consider the **forces** generated by the **traction wheels**
- Such forces must then be transformated into F and T according to the "generic" 2D robot model
- In a similar way, according to the model, **position sensors** are tied to the wheels
- so they do not generate directly $\{x_R, y_R, \theta_R\}$ and a proper transformation is needed also in this case

Two Indipendent Wheels

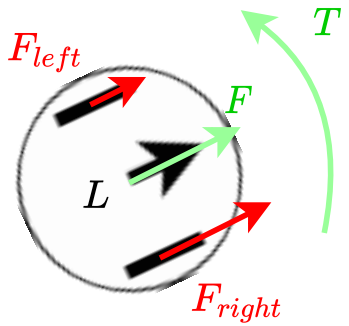
Two Independent Wheels





- Two indipendent **traction** wheels (in red)
- Two indipendent **encoders** (with free wheels, in black) to track the position

Two Independent Wheels



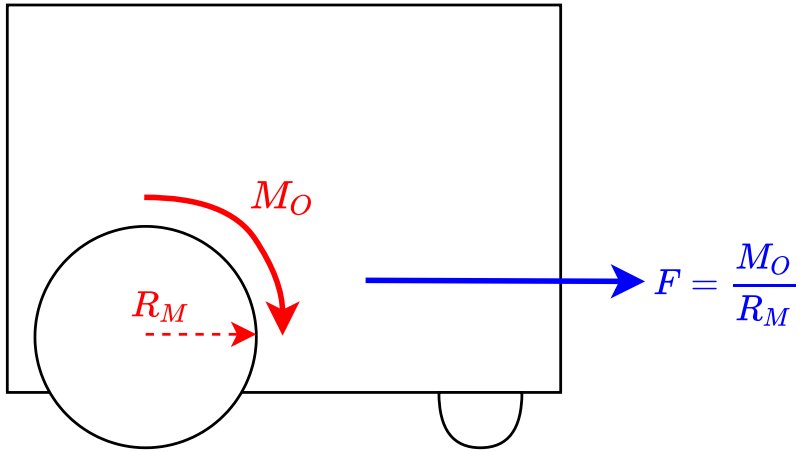
Dynamic Model

- Since we drive the traction wheels, we need a transformation from (F_{left}, F_{right}) to (F, T)
- We have:

$$\begin{aligned} F &= F_{left} + F_{right} \\ T &= L\,(F_{right} - F_{left}) \end{aligned}$$

where L is the (estimated) distance between the two traction wheels

Two Independent Wheels



Dynamic Model

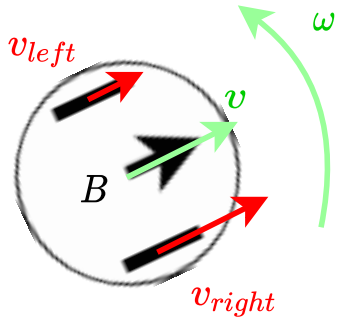
- Indeed, wheel motors generate **torques** that then are transformed in forces on the basis of the radius the traction wheels:

$$F_{left} = \frac{M_{O_{left}}}{R_{M_{left}}}$$

$$F_{right} = \frac{M_{O_{right}}}{R_{M_{right}}}$$

where $R_{M_{xxxx}}$ is the radius of the wheel and $M_{O_{xxxx}}$ is the torque generated by the motor

Two Independent Wheels



Kinematic Model

- Since we measure the two wheels, we need a transformation from (v_{left}, v_{right}) to (v, ω)
- We have:

$$v = \frac{v_{left} + v_{right}}{2} \qquad \omega = \frac{v_{right} - v_{left}}{B}$$

$$v_{left} = v - \frac{\omega B}{2} \qquad v_{right} = v + \frac{\omega B}{2}$$

where B (wheelbase) is the (estimated) distance between the **two measurement wheels**

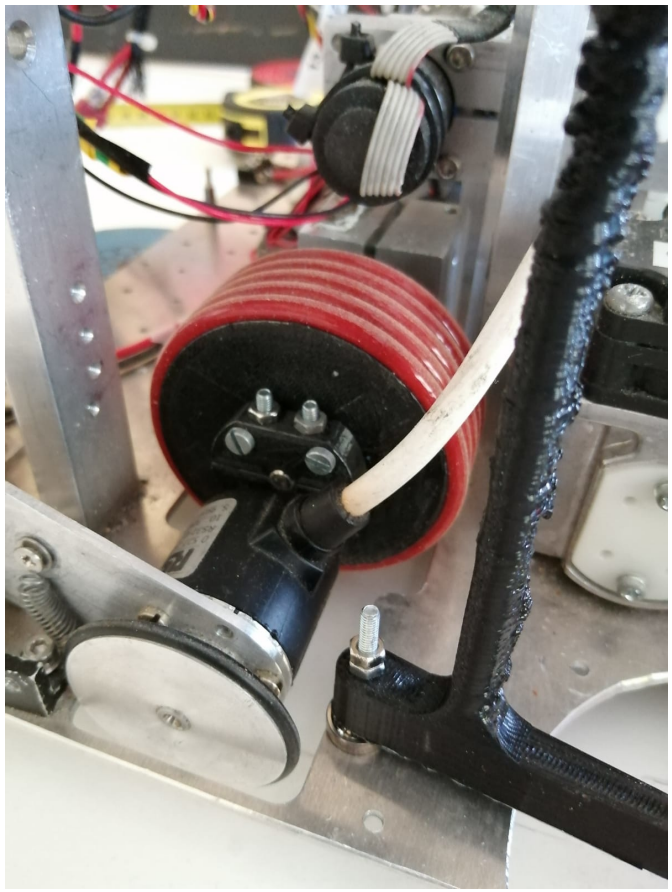
Measurement Wheels

- **Encoders** are digital sensors able to measure **rotation angles** (rather than speed)
- For each ΔT each encoder can provide the (relative) rotation angle of the wheel, i.e. $\Delta\theta_{left}$, $\Delta\theta_{right}$
- Given r_{left}, r_{right} the **radius** of each wheel we can compute the **distance** travelled by each wheel:

$$\Delta p_{left} = \Delta\theta_{left}\, r_{left} \qquad \Delta p_{right} = \Delta\theta_{right}\, r_{right}$$

- Speed can be thus computed as:

$$v_{left} = \frac{\Delta p_{left}}{\Delta T} \qquad v_{right} = \frac{\Delta p_{right}}{\Delta T}$$



Kinematic Model (Odometry or Dead-reckoning)

The final kinematics is given by:

$$\Delta p_{left} = \Delta\theta_{left}\ r_{left} \qquad \Delta p_{right} = \Delta\theta_{right}\ r_{right}$$

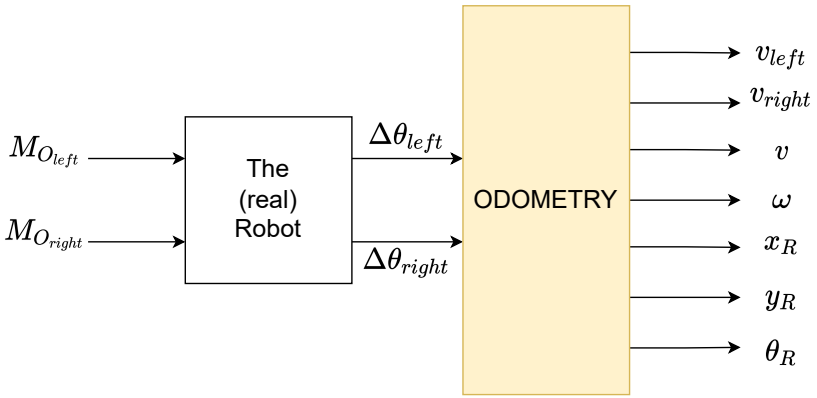
$$v_{left} = \frac{\Delta p_{left}}{\Delta T} \qquad v_{right} = \frac{\Delta p_{right}}{\Delta T}$$

$$v = \frac{v_{left} + v_{right}}{2} \qquad \omega = \frac{v_{right} - v_{left}}{B}$$

$$\Delta p = \frac{\Delta p_{left} + \Delta p_{right}}{2} \qquad \Delta\theta = \frac{\Delta p_{right} - \Delta p_{left}}{B}$$

$$\begin{aligned} x_R &= x_R + \Delta p \cos(\theta_R + \frac{\Delta\theta}{2}) \\ y_R &= y_R + \Delta p \sin(\theta_R + \frac{\Delta\theta}{2}) \\ \theta_R &= \theta_R + \Delta\theta \end{aligned}$$

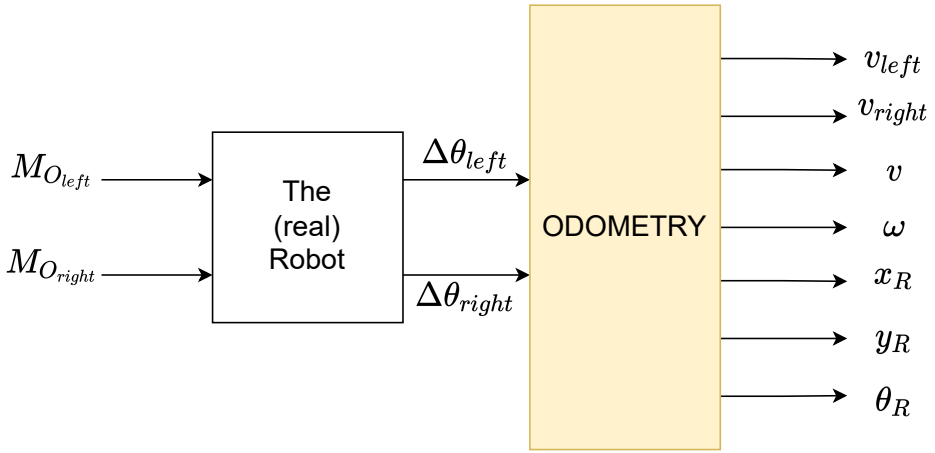
Two Independent Wheels



The Robot and the Sensors

- We can intervene on the robot by modulating the **torques** generated by left and right motors
- We can sense the robot by gathering data sampled by **encoders** in terms of angle variation of left and right measurement wheels
- The other kinematic paramaters are then computed by means of the **odometry** algorithm

Two Independent Wheels

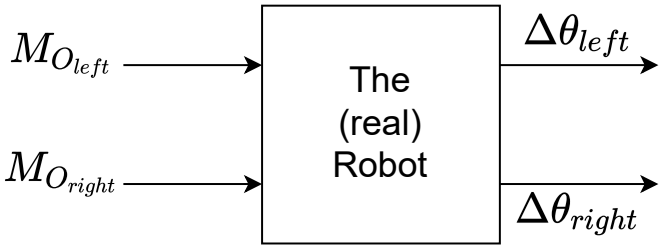


Approximation introducted by the Odometry Algorithm

- Encoders are digital sensors so $\Delta\theta_{left}$ and $\Delta\theta_{right}$ are approximated according to a certain resolution
- The algorithm assumes that r_{left} and r_{right} are **known**, but the **real radius** depends on the load on each sensing wheel
- The algorithm assumes that v_{left} and v_{right} are **constant** during ΔT, and this may not be the case
- The algorithm assumes that the mass' center of the robot is placed in the **middle point of the traction axis**, but if the weight is not uniformly distributed the reality is quite different

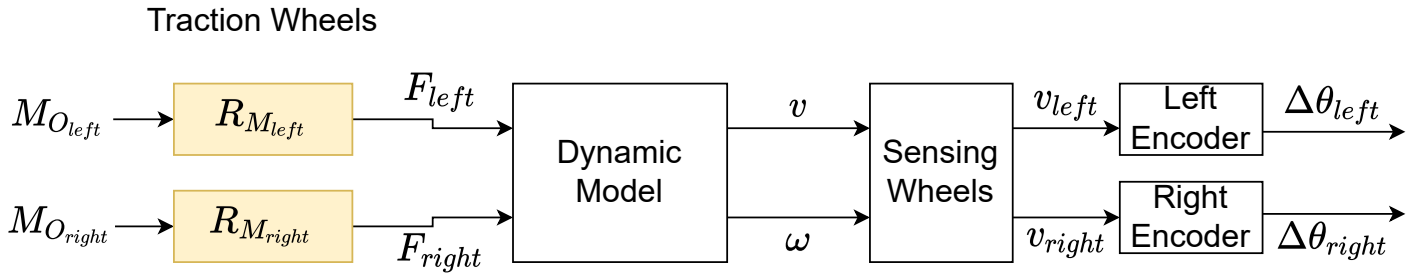
Model and Simulation

Two Independent Wheels



- According to the study made so far we can model our robot with two independent wheels
- We consider the robot as a system with two inputs, the **torques** generated by the motors ...
- ... and two outputs, the **angles** traveled by left and right sensing wheels

Two Independent Wheels



$$F_{left} = \frac{M_{O_{left}}}{R_{M_{left}}} \qquad F_{right} = \frac{M_{O_{right}}}{R_{M_{right}}}$$

$$F = F_{left} + F_{right} \qquad T = L\,(F_{right} - F_{left})$$

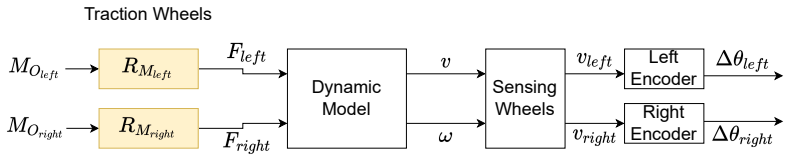
$$\dot{v} = -\frac{b}{M}v + \frac{1}{M}F$$

$$\dot{\omega} = -\frac{2\beta}{Mr^2}\omega + \frac{2}{Mr^2}T$$

$$v_{left} = v - \frac{\omega B}{2} \qquad v_{right} = v + \frac{\omega B}{2}$$

$$\dot{\theta}_{left} = \frac{v_{left}}{r_{left}} \qquad \dot{\theta}_{right} = \frac{v_{right}}{r_{right}}$$

Two Independent Wheels



Encoders, the matter of resolution

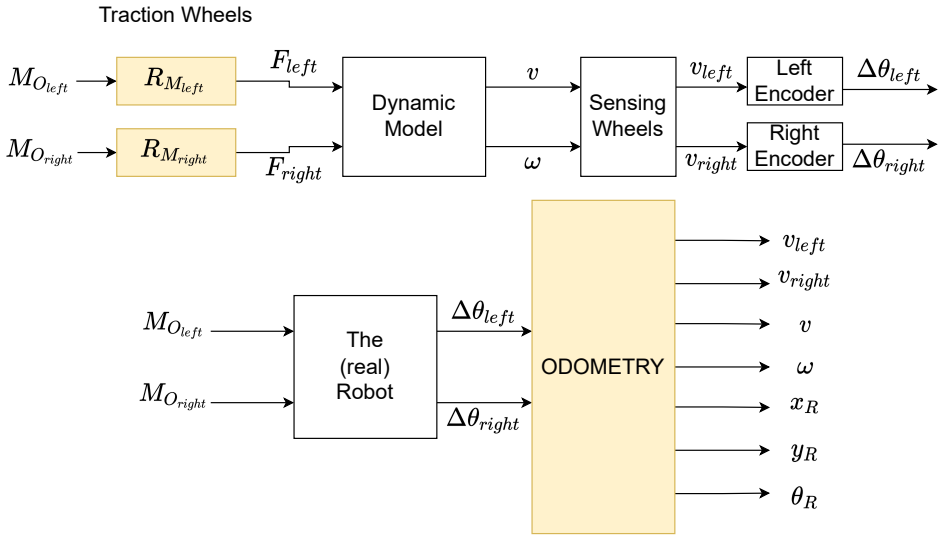
$$\dot{\theta}_{left} = \frac{v_{left}}{r_{left}} \qquad \dot{\theta}_{right} = \frac{v_{right}}{r_{right}}$$

$$\Delta\theta_{left} = \frac{v_{left}}{r_{left}}\Delta T \qquad \Delta\theta_{right} = \frac{v_{right}}{r_{right}}\Delta T$$

- Encoders are **digital sensors** so they feature a certain specific **resolution** which is given as the minimum angle (variation) ϵ they can percieve
- The **real** sensed data is therefore:

$$\Delta\theta_{left} = \left\lfloor \frac{v_{left}}{r_{left}}\Delta T \frac{1}{\epsilon} \right\rfloor \epsilon \qquad \Delta\theta_{right} = \left\lfloor \frac{v_{right}}{r_{right}}\Delta T \frac{1}{\epsilon} \right\rfloor \epsilon$$

Two Independent Wheels



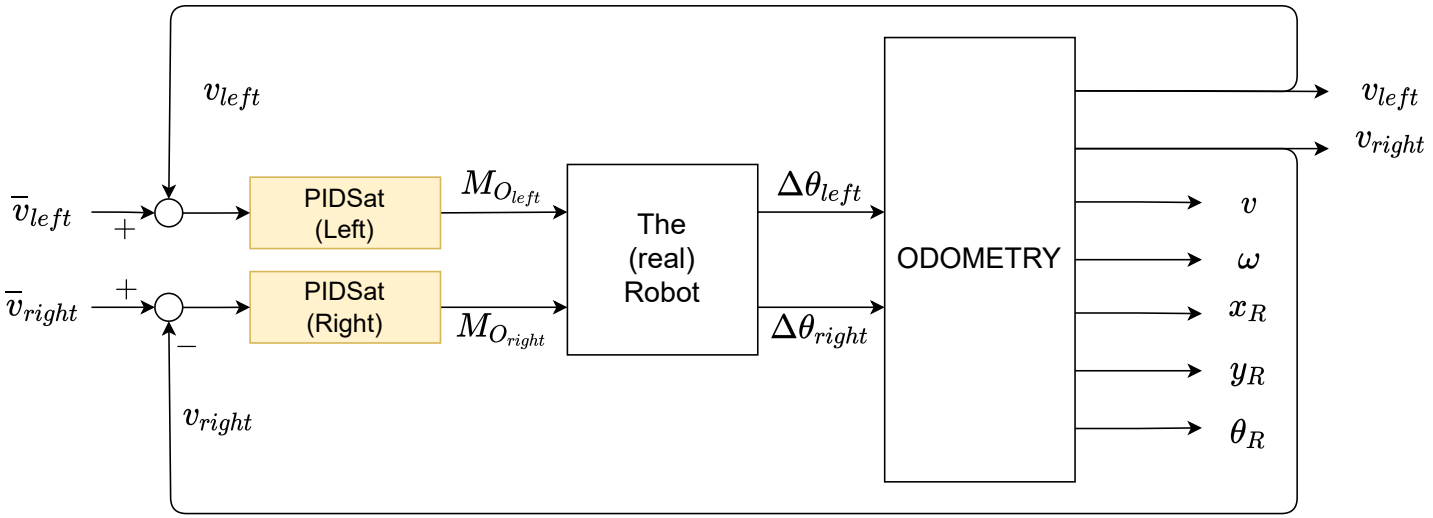
The Code

See:

- `lib/models/cart2d.py`
- `tests/cart_2d_two_wheels/test_robot_odometry_1.py`

Motion Control

Two Independent Wheels

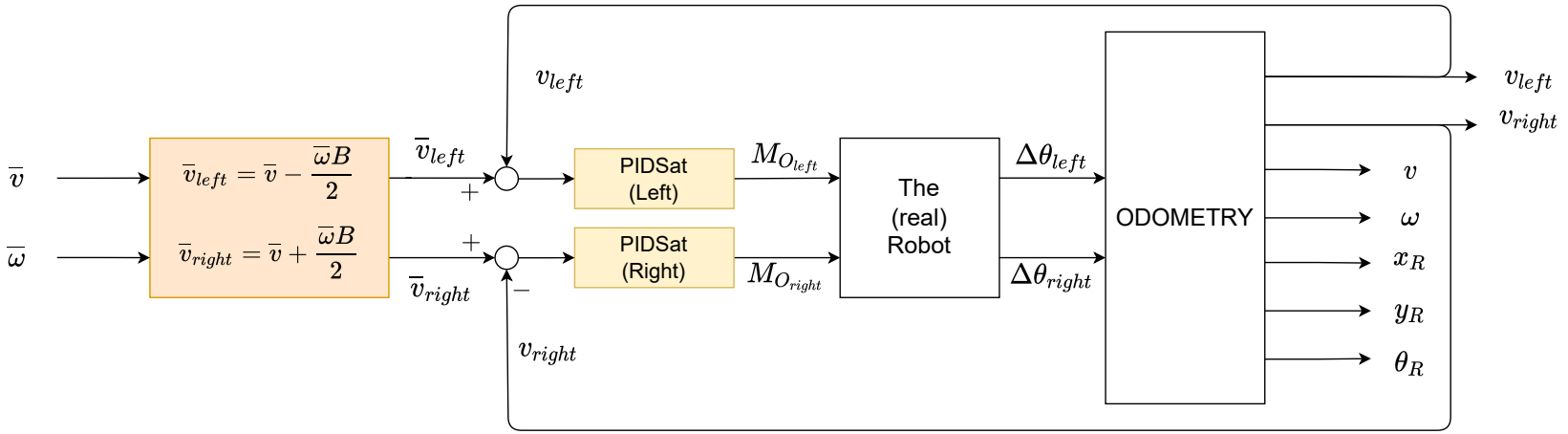


Speed Control

- We can control each wheel in **speed** by comparing the actual speed $v_{left\,/\,right}$ with the relevant reference $\overline{v}_{left\,/\,right}$
- To this aim, we use **two different PID controllers**, one for each wheel
- The output of each controller is the **command for the motors**, i.e. the torque in our case

(see `tests/cart_2d_two_wheels/test_speed_control.py` and `tests/cart_2d_two_wheels/test_speed_control_ramp.py`)

Two Independent Wheels

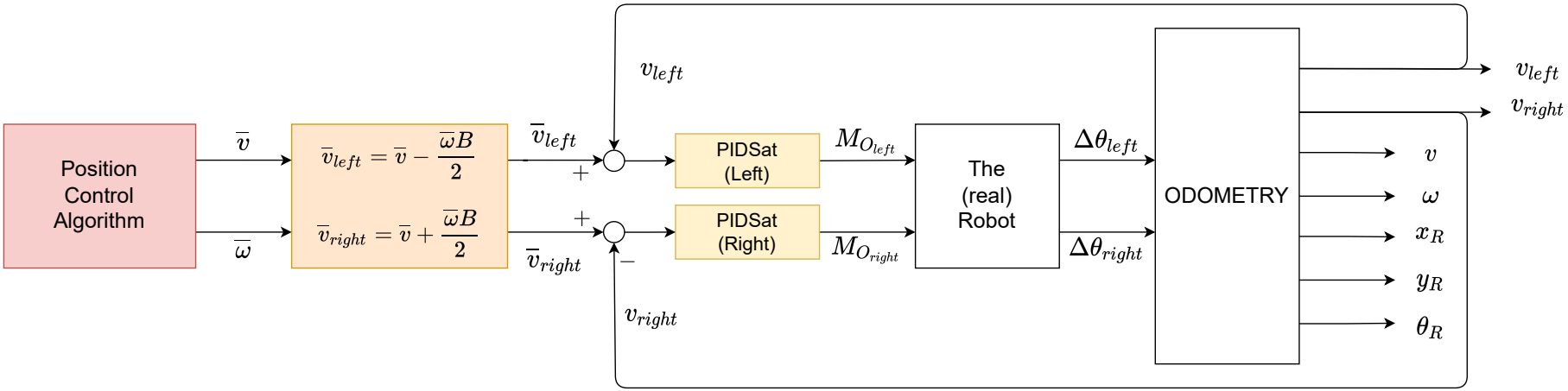


Speed Control

- If we add a transformation block from (v, ω) to (v_{left}, v_{right}), we can control the robot using the speeds of the rigid body

(see **tests/cart_2d_two_wheels/test_speed_control_polar.py**)

Two Independent Wheels



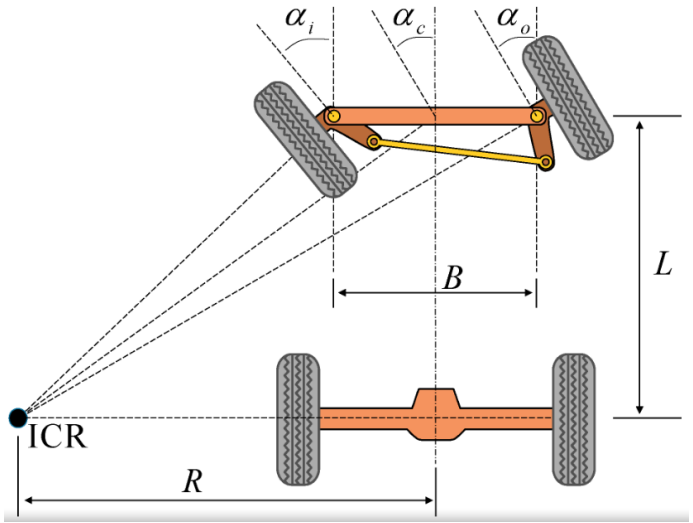
Position Control

- In turn, we can easily apply all position control algorithms we studied (polar, speed profile, etc.)

(see `tests/cart_2d_two_wheels/test_position_control.py`)

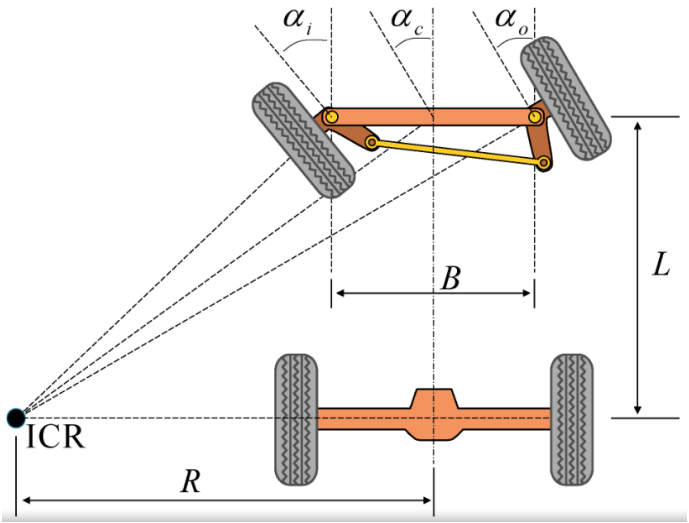
Ackermann Steering

Ackermann Steering



- It is the locomotion model of the cars
- It is based on a (rear or front) single traction actuator (motor) plus a mechanism to steer front wheels
- The traction (rear in figure) is connected to a gearbox called **differential** that allow traction wheels to rotate at different speeds during turns

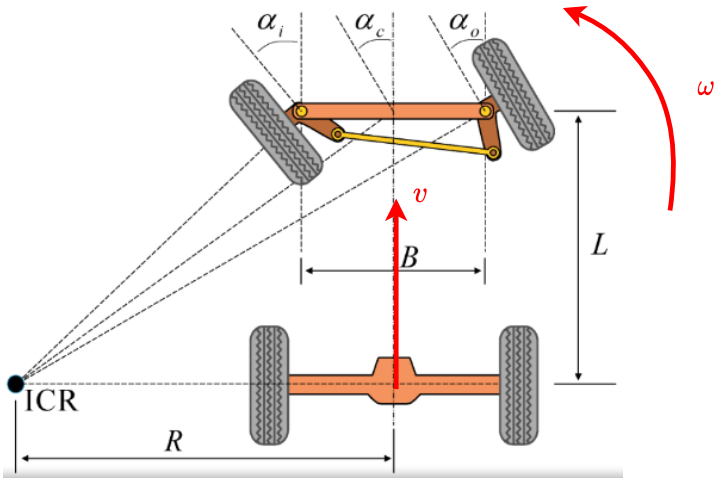
Ackermann Steering



Geometric Model

- *ICR*: Instantaneous Center of Rotation
- *R*: Radius of Rotation
- *B*: Wheelbase
- *L*: Lateral Wheelbase
- α_i: Inner wheel steering angle
- α_o: Outer wheel steering angle
- α_c: Center (virtual) wheel steering angle

Ackermann Steering

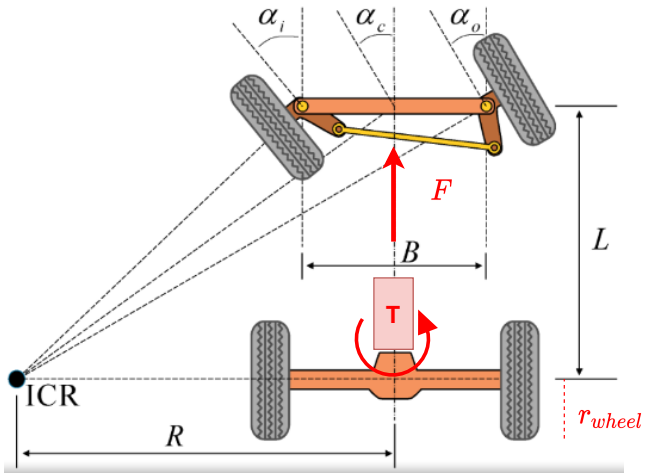


Kinematics

- In Ackermann steering vehicle the speeds of the rigid body (v, ω) **are not independent**
- ω depends on v and the steering angle α_c

$$\omega = \frac{v}{R} \qquad R = \frac{L}{\tan \alpha_c}$$

Ackermann Steering



Dynamics

- In Ackermann steering vehicle the motor generates a **torque** T that, on the basis of the radius of traction wheels, becomes a traction **force**
- The dynamics can be modeled in a similar way as to what we did for the cart

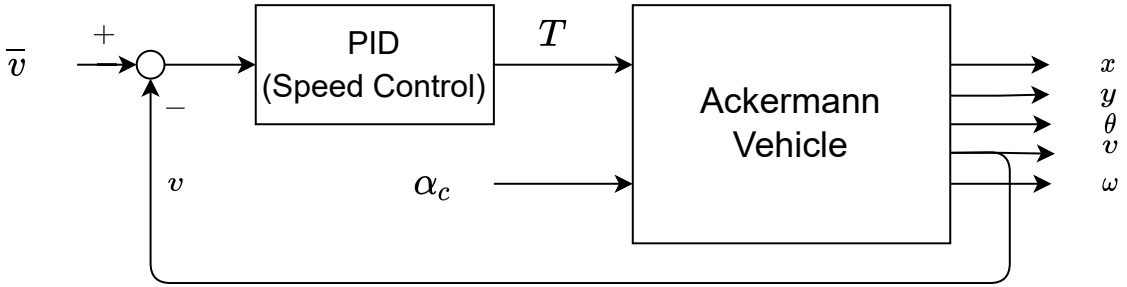
$$F = \frac{T}{r_{wheel}} \qquad \dot{v} = -\frac{b}{M}v + \frac{1}{M}F$$

Ackermann Steering

lib/models/cart2d.py

```
class AckermannSteering:

    def __init__(self, _mass, _lin_friction,
                       _r_traction, _lateral_wheelbase):
        self.M = _mass
        self.b = _lin_friction
        self.r_wheels = _r_traction
        self.l_wb = _lateral_wheelbase
        self.v = 0
        self.w = 0
        self.x = 0
        self.y = 0
        self.theta = 0

    def evaluate(self, delta_t, torque, steering_angle):
        _force = torque / self.r_wheels
        new_v = self.v * (1 - self.b * delta_t / self.M) + \
                delta_t * _force / self.M
        if steering_angle == 0:
            new_w = 0
        else:
            curvature_radius = self.l_wb / math.tan(steering_angle)
            new_w = new_v / curvature_radius

        self.x = self.x + self.v * delta_t * math.cos(self.theta)
        self.y = self.y + self.v * delta_t * math.sin(self.theta)
        self.theta = self.theta + delta_t * self.w
        self.v = new_v
        self.w = new_w
```

Ackermann Steering



Speed Control

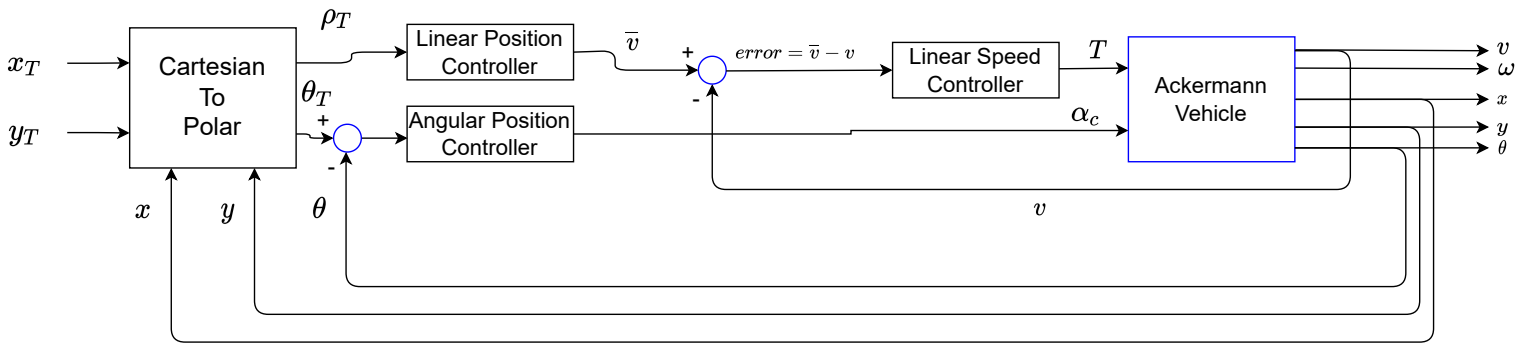
- The linear speed v can be directly controlled using a classical PID
- Angular speed ω **cannot be** directly controlled since it depends on both the steering angle and the linear speed

Ackermann Steering

tests/ackermann_2d/test_speed_ackermann.py

```
class AckermannRobot(RoboticSystem):

    def __init__(self):
        super().__init__(1e-3) # delta_t = 1e-3
        # Mass = 10kg
        # side = 15cm
        # wheels radius = 2cm
        # friction = 0.8
        self.car = AckermannSteering(10, 0.8, 0.02, 0.15)

        # 5 Nm max, antiwindup
        self.speed_controller = PIDSat(2.0, 2.0, 0, 5, True)

    def run(self):
        (v, w) = self.get_speed()
        vref = 1.5

        Torque = self.speed_controller.evaluate(self.delta_t, vref, v)
        Steering = 0

        self.car.evaluate(self.delta_t, Torque, Steering)
        ...
```

Ackermann Steering



Position Control

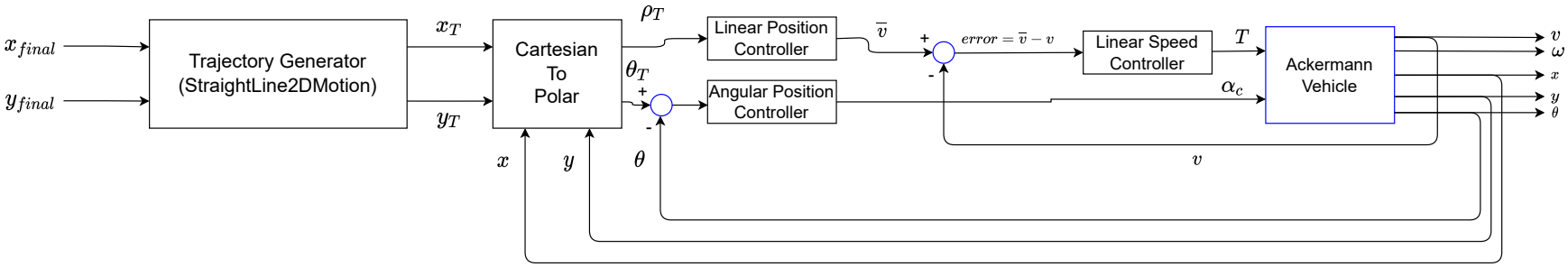
- The polar position control can be directly used by considering steering angle instead of angular speed
- The output of the angular position controller is not the target ω but the **steering angle** α_c
- Remember that that both linear and angular position controllers are P-controllers with saturation

Ackermann Steering

tests/ackermann_2d/test_polar_ackermann.py

```
class AckermannRobot(RoboticSystem):

    def __init__(self):
        super().__init__(1e-3) # delta_t = 1e-3
        # Mass = 10kg
        # side = 15cm
        # wheels radius = 2cm
        # friction = 0.8
        self.car = AckermannSteering(10, 0.8, 0.02, 0.15)

        # 5 Nm max, antiwindup
        self.speed_controller = PIDSat(2.0, 2.0, 0, 5, True)

        # kp_lin = 1, vmax = 2 m/s
        # kp_ang = 1, steering max = 45 deg
        self.polar_controller = Polar2DController(1.0, 2.0,
                                                  1.0, math.pi/4)

    def run(self):
        (vref, steering) = self.polar_controller.evaluate(self.delta_t,
                                                          0.5, 0.5,
                                                          self.get_pose())

        (v, w) = self.get_speed()

        torque = self.speed_controller.evaluate(self.delta_t, vref, v)

        self.car.evaluate(self.delta_t, torque, steering)
        ...
```

Ackermann Steering



Trajectory Following

- The polar position control can be in turn driven by the **trajectory generator** in order to follow a certain straight line towards a final point

(see **`tests/ackermann_2d/test_virtual_robot_ackermann.py`**)

Locomotion of a Mobile Robot in a 2D Space

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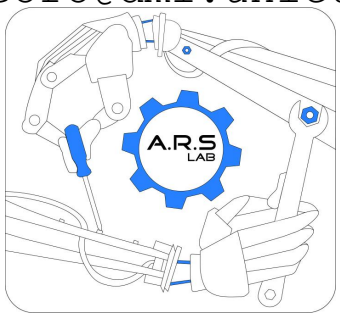

Robotic Systems