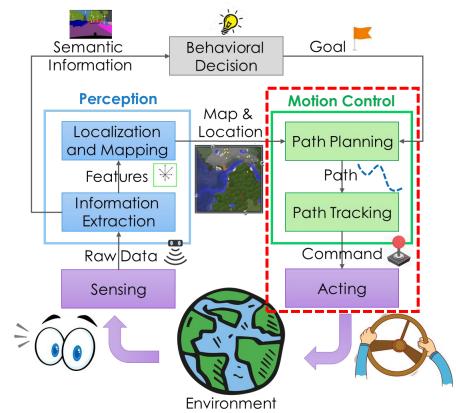
Robotic Navigation and Exploration

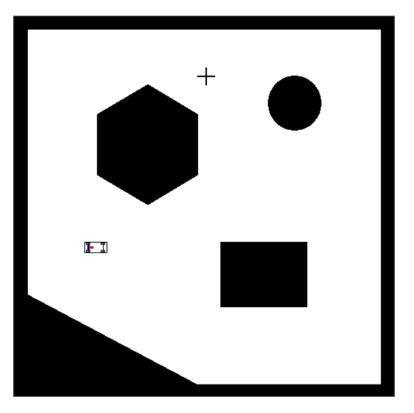
Lab1: Kinematic Model and Path Tracking Control

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Goal

 In Homework 1, you are going to complete the "Motion Control" part given a known map and localization information. In lab1 and lab2, we will lead you through the control and planning parts, respectively.





Requirement

- Python >= 3.6
- Numpy
- Opency-Python

Lab Hint

- The pages with the title contains "[Practice]" mean that there are codes that you should complete.
- The pages with the title contains "[Run]" mean that you have to run the code for testing.
- The bottom right will show the path of the codes that are related to that page.
- The codes we give are half-complete, you have to implement the codes after the comment of "TODO".

[Practice] Basic



./Simulation/utils.py

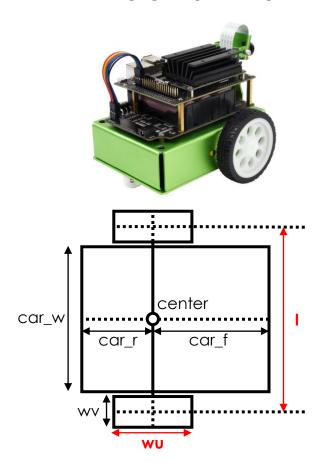
7

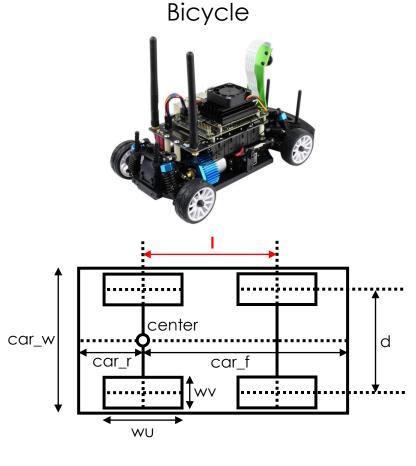
```
def step(self, state:State, cstate:ControlState) -> State:
    # TODO: Basic Kinematic Model
    state_next = state
    return state_next
```

Kinematic Models

Parameter Settings of Vehicles

Differential Drive





States

- Kinematic-Related Physical Parameters
 - $-x,y,\theta,v,\omega$ (position, angle, velocity / angular velocity)
 - (denoted "x", "y", "yaw", "v", "w" in the codes)

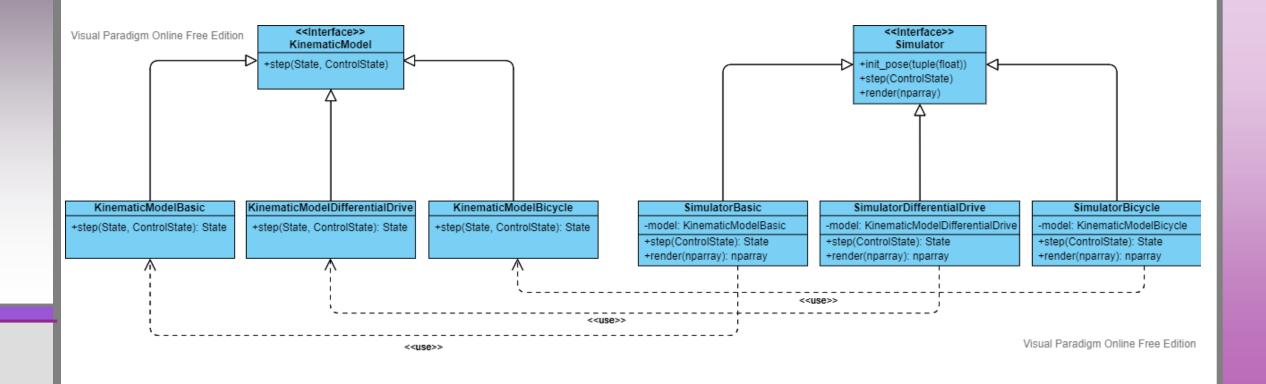
```
class State:
   def __init__(self, x=0.0, y=0.0, yaw=0.0, v=0.0, w=0.0):
       self.x = 0.0
       self.y = 0.0
       self.yaw = 0.0
       self.v = 0.0
       self.w = 0.0
       self.update(x, y, yaw, v, w)
   def update(self, x=None, y=None, yaw=None, v=None, w=None):
       if x is not None:
           self.x = x
       if y is not None:
           self.y = y
       if yaw is not None:
           self.yaw = yaw
       if v is not None:
           self.v = v
       if w is not None:
           self.w = w
```

Control States

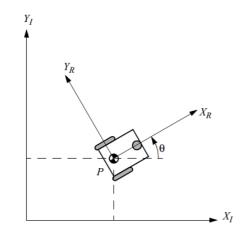
- Controlling-Related Parameter
 - Basic Model: v, ω (denoted as "v" and "w" in the codes)
 - Differential Drive: ω_{left} , ω_{right} (denoted as "lw" and "rw" in the codes)
 - Bicycle Model: a, δ (denoted as "a" and "delta" in the codes)

```
class ControlState:
   def __init__(self, control_type, *cstate):
       # Support basic/diff drive/bicycle
       self.control type = control type
       try:
           if control type == "basic":
                self.v = cstate[0]
                self.w = cstate[1]
           elif control_type == "diff_drive":
                self.lw = cstate[0]
                self.rw = cstate[1]
           elif control type == "bicycle":
                self.a = cstate[0]
                self.delta = cstate[1]
           else:
                raise NameError("Unknown control type!!")
       except NameError:
           raise
```

Class Architecture for Simulation



[Practice] Basic Kinematic Model



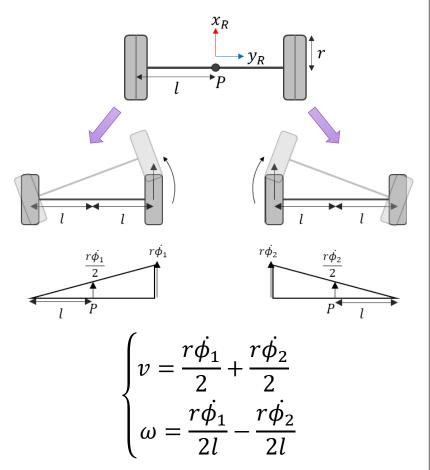
$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = R(\theta)^{-1} \begin{bmatrix} \dot{x}_R \\ \dot{y}_R \\ \dot{\theta} \end{bmatrix}$$

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

$$= \begin{bmatrix} v \cos(\theta) \\ v \sin(\theta) \\ \omega \end{bmatrix}$$

[Practice] Differential Drive Kinematic Model

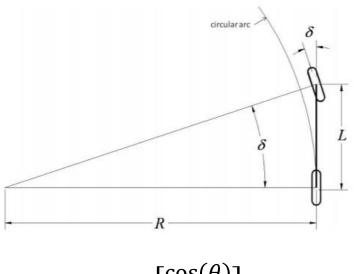
```
class KinematicModelDifferentialDrive(KinematicModel):
   def init (self, r, l, dt):
       # Simulation delta time
       self.r = r
       self.1 = 1
       self.dt = dt
   def step(self, state:State, cstate:ControlState) -> State:
       x1dot = self.r*np.deg2rad(cstate.rw) / 2
       w1 = np.rad2deg(self.r*np.deg2rad(cstate.rw) / (2*self.1))
       x2dot = self.r*np.deg2rad(cstate.lw) / 2
       w2 = np.rad2deg(self.r*np.deg2rad(cstate.lw) / (2*self.l))
       v = x1dot + x2dot
       W = W1 - W2
       x = state.x + v * np.cos(np.deg2rad(state.yaw)) * self.dt
       y = state.y + v * np.sin(np.deg2rad(state.yaw)) * self.dt
       yaw = (state.yaw + w * self.dt) % 360
       state_next = State(x, y, yaw, v, w)
       return state next
```



[Practice] Bicycle Kinematic Model

```
class KinematicModelBicycle(KinematicModel):
    def __init__(self, l, dt):
        # Distance from center to wheel
        self.l = l
        # Simulation delta time
        self.dt = dt

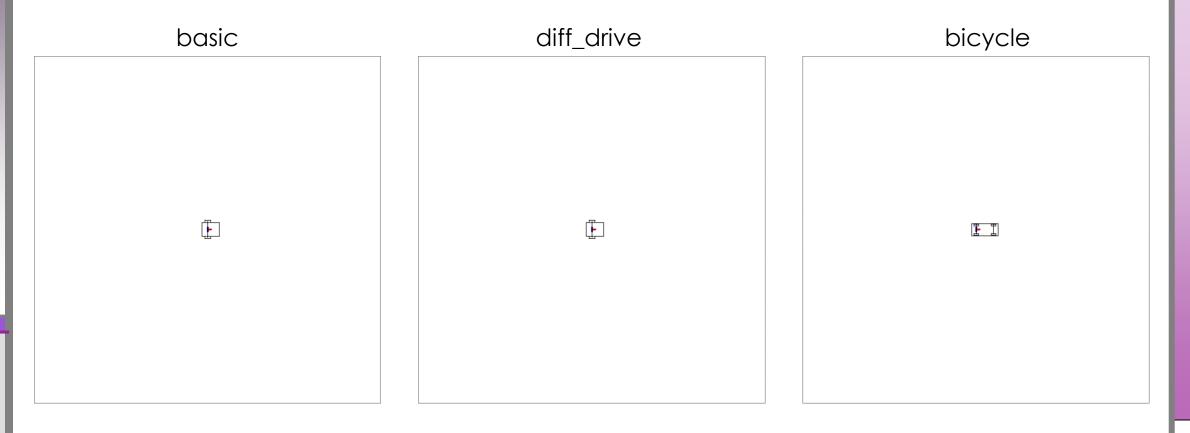
def step(self, state:State, cstate:ControlState) -> State:
        v = state.v + cstate.a*self.dt
        w = np.rad2deg(state.v / self.l * np.tan(np.deg2rad(cstate.delta)))
        x = state.x + v * np.cos(np.deg2rad(state.yaw)) * self.dt
        y = state.y + v * np.sin(np.deg2rad(state.yaw)) * self.dt
        yaw = (state.yaw + w * self.dt) % 360
        state_next = State(x, y, yaw, v, w)
        return state_next
```



$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} \cos(\theta) \\ \sin(\theta) \\ \tan(\delta) \end{bmatrix} \iota$$

[Run] Motion Model

python 01_motion_model.py -s [basic/diff_drive/bicycle]



Path Tracking Control

Path Example

 We provides two paths for testing the tracking algorithm, path1 is a line and path2 is a curve.

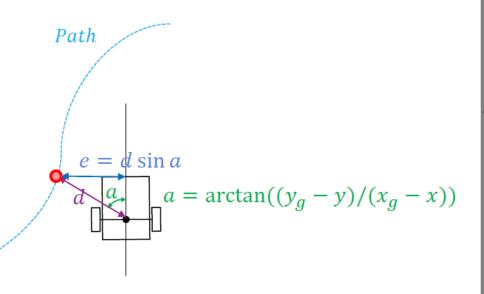
```
def path1():
   cx = np.arange(0, 500, 1) + 50
   cy = [270 \text{ for ix in } cx]
   cyaw = [0 for ix in cx]
   ccurv = [0 for ix in cx]
   path = np.array([(cx[i],cy[i],cyaw[i],ccurv[i]) for i in range(len(cx))])
   return path
def path2(p1 = 80.0):
   cx = np.arange(0, 500, 1) + 50
   cy = [np.sin(ix / p1) * ix / 4.0 + 270 for ix in cx]
   diff1 = [(np.cos(ix/p1)/p1*ix + np.sin(ix/p1))/4.0 for ix in cx]
   diff2 = [(-np.sin(ix/p1)/(p1**2)*ix + np.cos(ix/p1)/p1 + np.cos(ix/p1)/p1)/4.0 for ix in cx]
   cyaw = [np.rad2deg(np.arctan2(d1,1)) for d1 in diff1]
   ccurv = [np.abs(diff2[i])/np.power((1+diff1[i]**2),3/2) for i in range(len(cx))]
   path = np.array([(cx[i],cy[i],cyaw[i],ccurv[i]) for i in range(len(cx))])
   return path
```

Path 1

Path 2

[Practice] PID Control

Output =
$$K_p e(t) + K_i \sum_{t=0}^{t} e_t + K_d(e(t) - e(t-1))$$



Basic Model

```
min_idx, min_dist = utils.search_nearest(self.path, (x,y))
target = self.path[min_idx]
ang = np.arctan2(self.path[min_idx,1]-y, self.path[min_idx,0]-x)
ep = min_dist * np.sin(ang)
self.acc_ep += dt*ep
diff_ep = (ep - self.last_ep) / dt
next_w = self.kp*ep + self.ki*self.acc_ep + self.kd*diff_ep
self.last_ep = ep
return next_w, target
```

Bicycle Model

```
min_idx, min_dist = utils.search_nearest(self.path, (x,y))
target = self.path[min_idx]
ang = np.arctan2(self.path[min_idx,1]-y, self.path[min_idx,0]-x)
ep = min_dist * np.sin(ang)
self.acc_ep += dt*ep
diff_ep = (ep - self.last_ep) / dt
next_delta = self.kp*ep + self.ki*self.acc_ep + self.kd*diff_ep
self.last_ep = ep
return next_delta, target
```

[Practice] Pure Pursuit Control for Basic Model

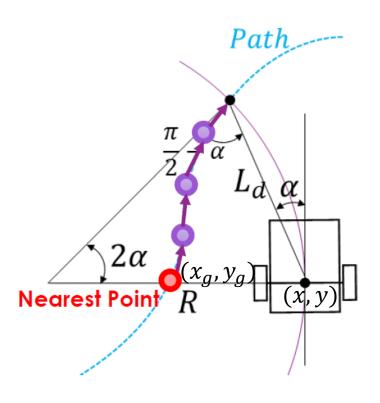
- Concept:
 - Modify the angular velocity to let the center achieve a point on path

$$\alpha = \arctan\left(\frac{y_g - y}{x_g - x}\right) - \theta$$

$$\omega = \frac{2v\sin(\alpha)}{L_d}$$

$$L_d = k * v + L_{fc} \text{, where } k, L_{fc} \text{ are parameters.}$$

- 1. Set a distance L_d .
- 2. Find the nearest point on the path.
- 3. Search the following point until the distance of the point larger than or equal to L_d .



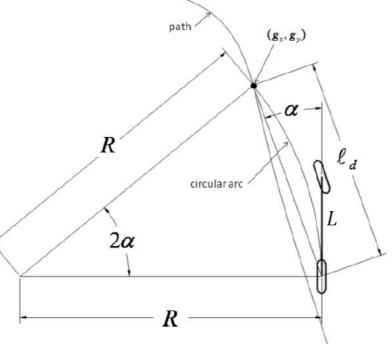
[Practice] Pure Pursuit Control for Bicycle Model

- Concept:
 - Control the steer to let the rear wheel achieve a point on the path.

$$\alpha = \arctan\left(\frac{y_g - y}{x_g - x}\right) - \theta$$

$$\delta = \arctan\left(\frac{2L\sin(\alpha)}{L_d}\right)$$

$$L_d = k * v + L_{fc} \text{, where } k, L_{fc} \text{ are parameters.}$$



[Practice] Stanley Control for Bicycle Model

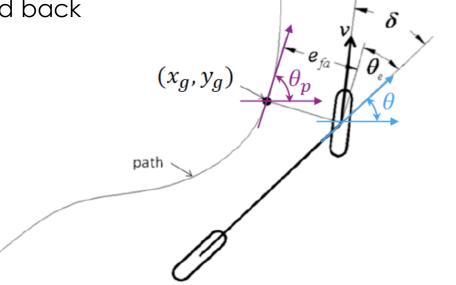
- Concept:
 - Exponential stability for front wheel feed back
- Some Implementation Details

$$\theta_e = \theta_p - \theta$$

$$\dot{e} = v_f \sin(\delta - \theta_e)$$

$$\delta = \arctan(-\frac{ke}{v_f}) + \theta_e$$

$$e = \begin{bmatrix} x - x_g \\ y - y_g \end{bmatrix} \begin{bmatrix} \cos(\theta_p + 90) \\ \sin(\theta_p + 90) \end{bmatrix}$$



Hint: Beware to transform the angle into the boundary of -180~+180

[Practice] LQR Control (Bonus)

- We have already completed the part of DARE solving.
- Following the steps:
 - Construct the matrix A, B, X of the linear approximation model.

$$Ax + Bu = \begin{bmatrix} 1 & dt & 0 & 0 \\ 0 & 0 & v & 0 \\ 0 & 0 & 1 & dt \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} e \\ \dot{e} \\ \theta \\ \dot{\theta} \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ \frac{v}{L} \end{bmatrix} \delta \text{ (Bicycle Model)}$$

$$A \qquad X \qquad B \qquad U$$

- Solve DARE and get the matrix P of the value function. $P = Q + A^T P A - A^T P B (R + B^T P B)^{-1} B^T P A$

- Compute the optimal control. $u_t^* = -(R + B^T P_{t+1} B)^{-1} B^T P_{t+1} A x_t$

[Practice] Control of Differential Drive

We only implement the controller for basic and bicycle kinematic model.
 The control of the differential drive can be simply modified by the controller of basic model.

$$\dot{\phi_2} = \left(v - \frac{r\dot{\phi_1}}{2}\right)\frac{2}{r} = \frac{2v}{r} - \dot{\phi_1}$$

$$\omega = \frac{r\dot{\phi_1}}{2l} - \frac{r\left(\frac{2v}{r} - \dot{\phi_1}\right)}{2l} = \frac{r\dot{\phi_1} - v}{l}$$

$$\dot{\phi_1} = \frac{v}{r} + \frac{\omega l}{r}$$

$$\dot{\phi_2} = \frac{v}{r} - \frac{\omega l}{r}$$

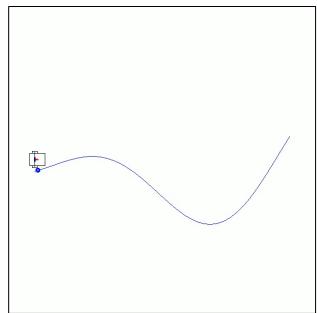
[Run] Path Tracking

python 02_path_tracking.py -s [basic/diff_drive/bicycle] (simulator)

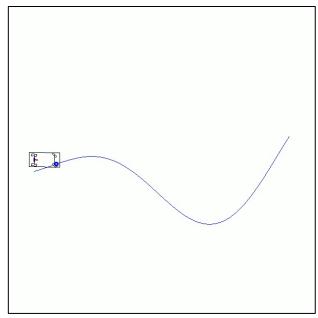
-c [pid/pure_pursuit/stanley/lqr] (controller)

-t [1/2] (path type)

basic + pure_pursuit



bicycle + stanley



Remind

- Check if you complete the "TODO" in the following files:
 - ./Simulation/kinematic_basic.py
 - ./Simulation/kinematic_differential_drive.py
 - ./Simulation/kinematic_bicycle.py
 - ./PathTracking/controller_pid_basic.py
 - ./PathTracking/controller_pid_bicycle.py
 - ./PathTracking/controller_pure_pursuit_basic.py
 - ./PathTracking/controller_pure_pursuit_bicycle.py
 - ./PathTracking/controller_stanley_bicycle.py
 - ./02_path_tracking.py
 - ./PathTracking/controller_lqr_basic.py (bonus)
 - ./PathTracking/controller_lqr_bicycle.py (bonus)

Q&A 24