

Dubins Car Challenge

August 5, 2021

DD2410 — Planning Assignment
Chris Sprague
sprague@kth.se

1 Description

In this assignment you're tasked to implement a robotic planning method to drive a Dubins car with the dynamics

$$\begin{bmatrix} \frac{dx}{dt} \\ \frac{dy}{dt} \\ \frac{d\theta}{dt} \end{bmatrix} = \begin{bmatrix} \cos \theta \\ \sin \theta \\ \tan \phi(t) \end{bmatrix}$$

from an initial position (x_0, y_0) to a target position (x_f, y_f) , while avoiding collisions with obstacles and going out of bounds.

1.1 Variables

The state variables are

$$\begin{aligned} x &= \text{horizontal position} \\ y &= \text{vertical position} \\ \theta &= \text{heading angle} \end{aligned}$$

and the control is

$$\phi(t) \in [-\pi/4, \pi/4] = \text{steering angle.}$$

2 Tasks

We'll consider two graded tasks in order of difficulty:

- **E** — reach the target with circular obstacles;
- **C** — reach the target with line obstacles.

Note:

- the line obstacles are represented by a series of circular obstacles,
- the initial and target positions are randomised,
- and the obstacle in Kattis are different.

3 Your solution file

Using the API (see `README.pdf`) and a robotic planning method, generate a sequence of steering angle commands `controls` and a sequence of times `times`, between which the commands are executed, that would yield a collision-free and task-fulfilling trajectory.

Do this by editing the function `solution(car)` contained in the file `solution.py`. If needed, supporting code can be added outside the `solution(car)` function.

The template solution looks like this:

```
def solution(car):

    ''' <<< write your code below >>> '''
    controls=[0]
    times=[0,1]

    ''' <<< write your code above >>> '''

    return controls, times
```

3.1 The solution function

The solution function `solution(car)` receives a `Car` instance `car` and returns a tuple containing

- `controls` : list: sequence of steering angles `controls[i]` : float,
- `times` : list: sequence of times at which the controls are executed `times[i]` : float.

Note: `controls[i]` is considered to be constant between `times[i]` and `times[i+1]`, hence `len(controls) == len(times) - 1`.

3.2 The Car object

The `Car` object has several attributes which you may find useful, namely:

- `x0` : float: initial x-position
- `y0` : float: initial y-position
- `xt` : float: target x-position
- `yt` : float: target y-position
- `xlb` : float: minimum x-position
- `xub` : float: maximum x-position
- `ylb` : float: minimum y-position
- `yub` : float: maximum y-position
- `obs` : list: list of tuples for each obstacle `obs[i]`, where:
 - `obs[i][0]` : float: x-position
 - `obs[i][1]` : float: y-position
 - `obs[i][2]` : float: radius

Note: these attributes should not be edited.

```
[1]: from dubins import Car
car = Car()
```

```
print(car.__dict__)
```

```
{'_environment': <dubins.Environment object at 0x7f179c4d2c50>, 'x0': 0.0, 'y0': 3.2391864432626347, 'xt': 20.0, 'yt': 6.192294996554469, 'obs': [(6.265392562094496, 6.041777229372004, 0.7163509740556866), (11.253338957069111, 6.945329079062532, 0.6590515138402037), (11.63540620099571, 4.143855055417216, 0.663082654436773), (14.366463960130238, 3.928638306976023, 0.5309048730271383), (13.360593052417883, 1.3855318481167713, 0.5209728464956396), (8.295845400408858, 3.328439611620683, 0.523820445703946), (15.419053589392373, 5.7377957493328235, 0.5507787560235685), (5.239727140428663, 2.236212798992681, 0.654161998523906), (9.159455308270687, 8.224584585768465, 0.524589635288561), (14.449178108415577, 9.161499842737221, 0.5635807320099291), (11.271015694214686, 1.004954985164139, 0.5446209607951273), (6.777727241502461, 8.752224800174412, 0.6284582987911573), (12.089677123810471, 9.143555411828443, 0.6515684558370866), (4.618899698161086, 7.908212030338597, 0.587827483548474), (7.715257169933885, 1.2752057436316044, 0.5521400804555545), (8.561981091018186, 5.6371188404126835, 0.5889112288904469), (13.556672656010928, 6.860957426729551, 0.5129774071591225), (15.417489493573708, 0.6327007297379463, 0.5240721652847098), (4.5668105984690746, 4.302153978219165, 0.5025700185424731)], 'xlb': 0.0, 'xub': 20.0, 'ylb': 0.0, 'yub': 10.0}
```

3.3 The step function

The method that you'll need to utilise in your implementation of robotic planning methods is `step(car, x, y, theta, phi)` (imported from `dubins`), which takes as its arguments:

- `car` : `Car`: instance of `Car`
- `x` : `float`: x-position
- `y` : `float`: y-position
- `theta` : `float`: heading angle
- `phi` : `float`: steering angle
- `dt=0.01`: `float`: time-step size

and returns a tuple of the form `(xn, yn, thetan)` after `dt` seconds, containing:

- `xn` : `float`: new x-position
- `yn` : `float`: new y-position
- `thetan` : `float`: new heading angle

Note: `dt` should not be below 0.01s.

After computing the new state `xn, yn, thetan = step(car, x, y, theta, phi)`, check `car.obs` to see if the new position is within any obstacles, `(car.xlb, car.xub, car.ylb, car.yub)` to see if it is out of bounds, and `(car.xt, car.yt)` to see if it is close the the target position.

3.4 Taking a single step

```
[2]: from dubins import step

# heading and arbitrary steering angle
theta, phi = 0.0, 0.1

# take a step
step(car, car.x0, car.y0, theta, phi, dt=0.1)
```

```
[2]: (0.1, 3.2391864432626347, 0.010033467208545055)
```

3.5 Recording multiple steps

```
[3]: # trajectory: x, y, theta, phi, time
xl, yl, thetal, phil, tl = [car.x0], [car.y0], [0.0], [], [0.0]

# simulate for 1 seconds with constant steering angle
phi = 0.1
for _ in range(10):
    xn, yn, thetan = step(car, xl[-1], yl[-1], thetal[-1], phi, dt=0.1)
    xl.append(xn)
    yl.append(yn)
    thetal.append(thetan)
    phil.append(phi)
    tl.append(tl[-1] + 0.1)

print('The state after 10s is (x={:.3f}, y={:.3f}, theta={:.3f})'.format(
    xl[-1], yl[-1], thetal[-1]
))
print('The controls and times were:\n phi={} \n t={}'.format(phil, tl))
```

The state after 10s is (x=0.999, y=3.284, theta=0.100)

The controls and times were:

```
phi=[0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1]
t=[0.0, 0.1, 0.2, 0.30000000000000004, 0.4, 0.5, 0.6, 0.7, 0.7999999999999999,
0.8999999999999999, 0.9999999999999999]
```

3.6 Creating a solution

```
[4]: def solution(car):

    # trajectory: x, y, theta, phi, time
    x, y = car.x0, car.y0
    xl, yl, thetal, phil, tl = [x], [y], [0.0], [], [0.0]

    # simulate for 10 seconds with constant steering angle
```

```

phi = 0.1
for _ in range(1000):
    xn, yn, thetan = step(car, xl[-1], yl[-1], thetal[-1], phi, dt=0.1)
    xl.append(xn)
    yl.append(yn)
    thetal.append(thetan)
    phil.append(phi)
    tl.append(tl[-1] + 0.1)

# return controls and times
return phil, tl

```

3.7 Evaluating your solution

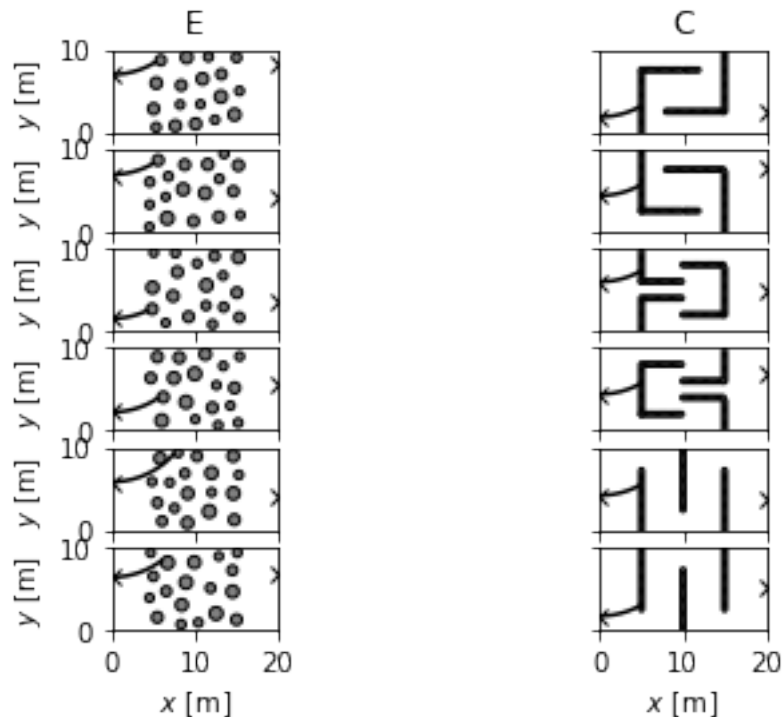
```

[5]: %matplotlib inline
from main import main
main(solution, plot=True, verbose=False)

```

Grade E: 0/6 cases passed.

Grade C: 0/6 cases passed.



Once you're done editing `solution.py`, you can evaluate how well your solution did in the terminal by executing

- `python3 main.py`,
- `python3 main.py -p` for plotting,
- `python3 main.py -v` for step feedback,
- or `python3 main.py -p -v` for both.

Note:

- you must install `matplotlib` for plotting to work,
- simulation is done at `dt=0.01` between `times[i]` and `times[i+1]`.

A succesful solution will generate something like this:

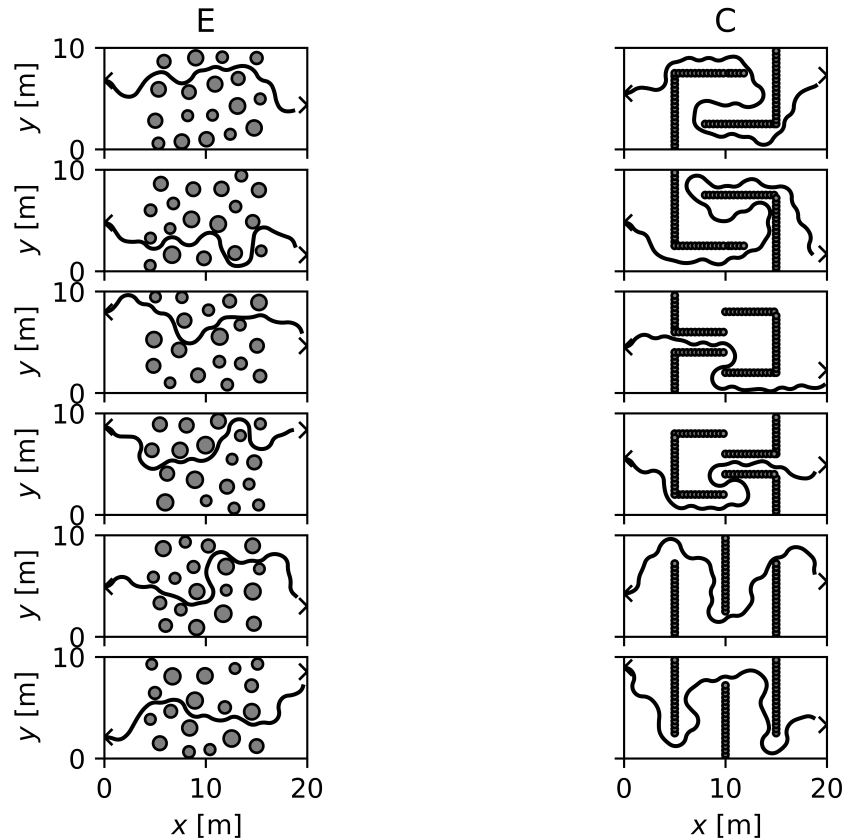
```
python3 main.py -p
```

Grade E: 6/6 cases passed.

Grade C: 6/6 cases passed.

```
[6]: from IPython.display import Image
      Image(filename='plot.png', width=350)
```

[6]:



4 Useful resources

- [PythonRobotics](#) for planning algorithms.