# sim\_traj\_planning

#### October 4, 2020

# so you don't need to restart the kernel every time

%load\_ext autoreload

%autoreload 2

In [8]: # The autoreload extension will automatically load in new code as you edit files,

```
import numpy as np
        from P1_astar import DetOccupancyGrid2D, AStar
        from P2_rrt import *
        from P3_traj_planning import compute_smoothed_traj, modify_traj_with_limits, Switching
        import scipy.interpolate
        import matplotlib.pyplot as plt
        from HW1.P1_differential_flatness import *
        from HW1.P2_pose_stabilization import *
        from HW1.P3_trajectory_tracking import *
        from utils import generate_planning_problem
        plt.rcParams['figure.figsize'] = [14, 14] # Change default figure size
The autoreload extension is already loaded. To reload it, use:
  %reload_ext autoreload
0.0.1 Generate workspace, start and goal positions
In [29]: width = 100
        height = 100
         num_obs = 25
         min_size = 5
         max_size = 30
         occupancy, x_init, x_goal = generate_planning_problem(width, height, num_obs, min_size
0.0.2 Solve A* planning problem
In [30]: astar = AStar((0, 0), (width, height), x_init, x_goal, occupancy)
         if not astar.solve():
             print "No path found"
```

# 0.1 Smooth Trajectory Generation

### 0.1.1 Trajectory parameters

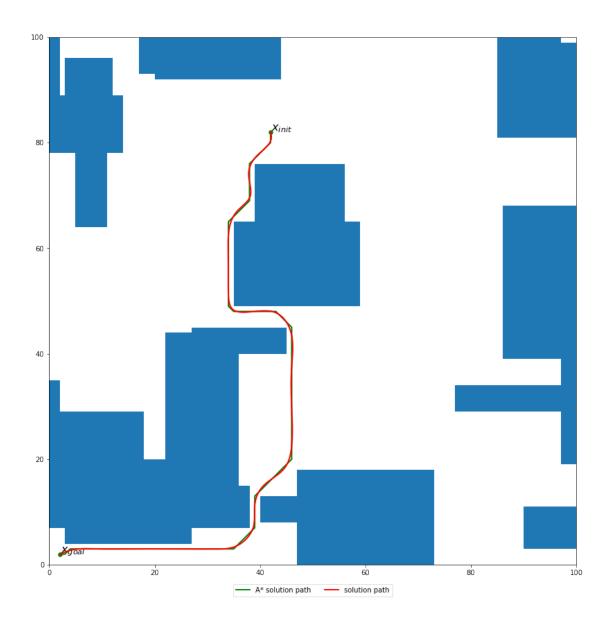
(Try changing these and see what happens)

```
In [31]: V_des = 0.3  # Nominal velocity
          alpha = 3.0  # Smoothness parameter
          dt = 0.05
```

#### 0.1.2 Generate smoothed trajectory

```
In [32]: traj_smoothed, t_smoothed = compute_smoothed_traj(astar.path, V_des, alpha, dt)

fig = plt.figure()
    astar.plot_path(fig.number)
    def plot_traj_smoothed(traj_smoothed):
        plt.plot(traj_smoothed[:,0], traj_smoothed[:,1], color="red", linewidth=2, label=
    plot_traj_smoothed(traj_smoothed)
    plt.legend(loc='upper center', bbox_to_anchor=(0.5, -0.03), fancybox=True, ncol=3)
    plt.show()
```



# 0.2 Control-Feasible Trajectory Generation and Tracking

### 0.2.1 Robot control limits

# 0.2.2 Tracking control gains

Tune these as needed to improve tracking performance.

In [34]: 
$$kpx = 2$$
  
 $kpy = 2$ 

```
kdx = 2
kdy = 2
```

# 0.2.3 Generate control-feasible trajectory

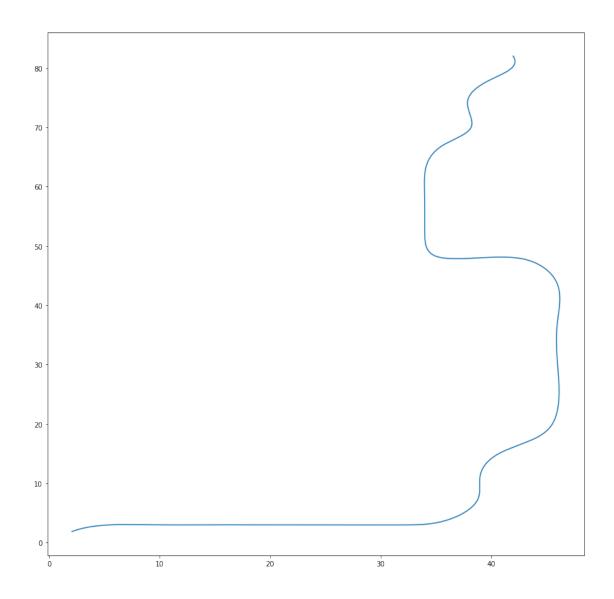
```
In [36]: t_new, V_smooth_scaled, om_smooth_scaled, traj_smooth_scaled = modify_traj_with_limit.

#TODO REMOVE BELOW
    traj_t = traj_smooth_scaled.T
    x = traj_t[0]
    y = traj_t[1]

V_t = V_smooth_scaled.T

fig = plt.figure()
    plt.plot(x, y)

Out[36]: [<matplotlib.lines.Line2D at 0x263e73c8>]
```



### 0.2.4 Create trajectory controller and load trajectory

In [45]: traj\_controller = TrajectoryTracker(kpx=kpx, kpy=kpy, kdx=kdx, kdy=kdy, V\_max=V\_max, traj\_controller.load\_traj(t\_new, traj\_smooth\_scaled)

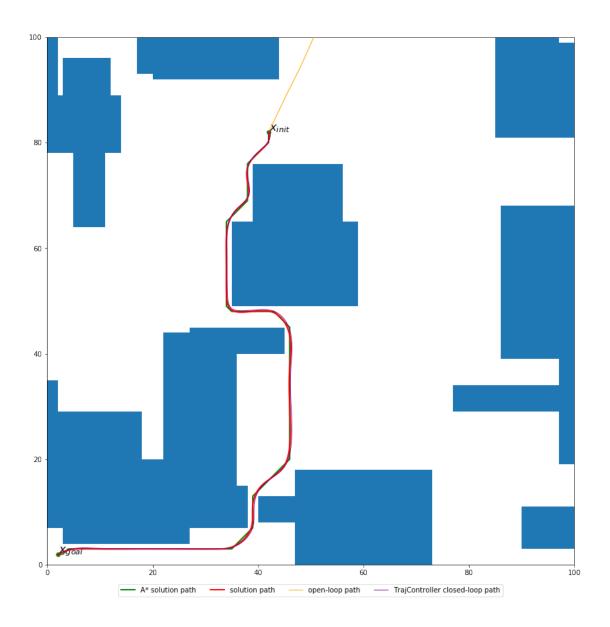
### 0.2.5 Set simulation input noise

(Try changing this and see what happens)

In [46]: noise\_scale = 0.05

### 0.2.6 Simulate closed-loop tracking of smoothed trajectory, compare to open-loop

```
s_0 = State(x=x_init[0], y=x_init[1], V=V_max, th=traj_smooth_scaled[0,2])
s_f = State(x=x_goal[0], y=x_goal[1], V=V_max, th=traj_smooth_scaled[-1,2])
actions_ol = np.stack([V_smooth_scaled, om_smooth_scaled], axis=-1)
states_ol, ctrl_ol = simulate_car_dyn(s_0.x, s_0.y, s_0.th, times_cl, actions=actions
states_cl, ctrl_cl = simulate_car_dyn(s_0.x, s_0.y, s_0.th, times_cl, controller=traj
fig = plt.figure()
astar.plot_path(fig.number)
plot_traj_smoothed(traj_smoothed)
def plot_traj_ol(states_ol):
   plt.plot(states_ol[:,0],states_ol[:,1], color="orange", linewidth=1, label="open-
def plot_traj_cl(states_cl):
   plt.plot(states_cl[:,0], states_cl[:,1], color="purple", linewidth=1, label="Traje")
plot_traj_ol(states_ol)
plot_traj_cl(states_cl)
plt.legend(loc='upper center', bbox_to_anchor=(0.5, -0.03), fancybox=True, ncol=4)
plt.show()
```

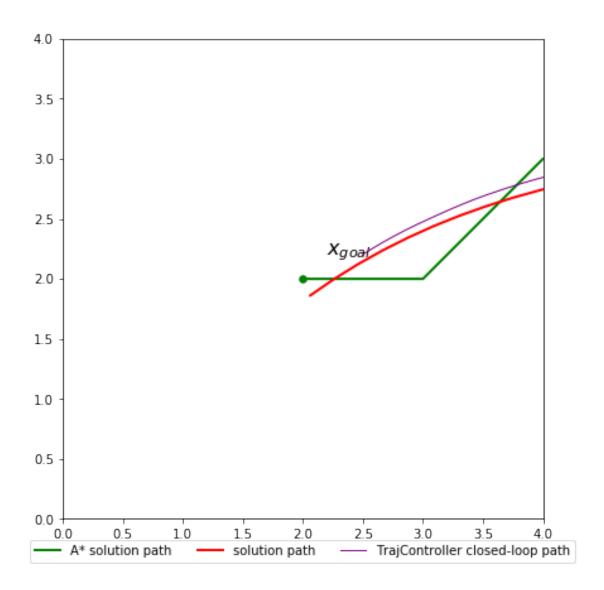


# 0.3 Switching from Trajectory Tracking to Pose Stabilization Control

### 0.3.1 Zoom in on final pose error

```
In [48]: l_window = 4.

fig = plt.figure(figsize=[7,7])
    astar.plot_path(fig.number)
    plot_traj_smoothed(traj_smoothed)
    plot_traj_cl(states_cl)
    plt.legend(loc='upper center', bbox_to_anchor=(0.5, -0.03), fancybox=True, ncol=3)
    plt.axis([x_goal[0]-l_window/2, x_goal[0]+l_window/2, x_goal[1]-l_window/2, x_goal[1]-plt.show()
```



### 0.3.2 Pose stabilization control gains

Tune these as needed to improve final pose stabilization.

In [49]: 
$$k1 = 1$$
.  $k2 = 1$ .  $k3 = 1$ .

### 0.3.3 Create pose controller and load goal pose

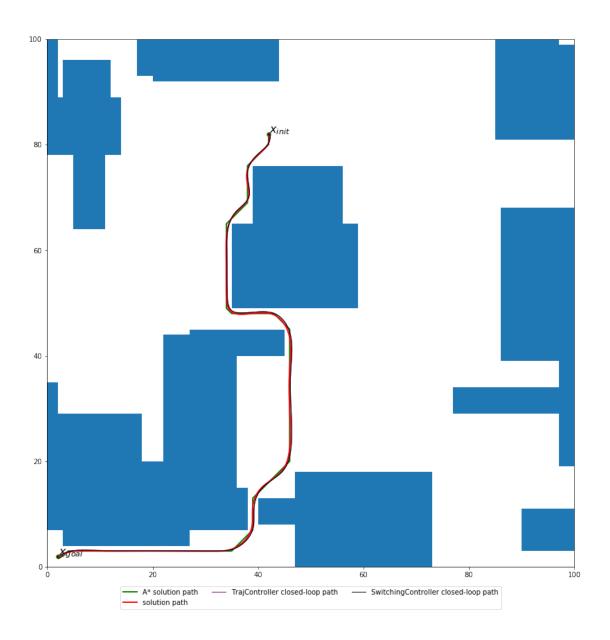
Note we use the last value of the smoothed trajectory as the goal heading  $\theta$ 

#### 0.3.4 Time before trajectory-tracking completion to switch to pose stabilization

Try changing this!

```
In [58]: t_before_switch = 3.0
```

#### 0.3.5 Create switching controller and compare performance

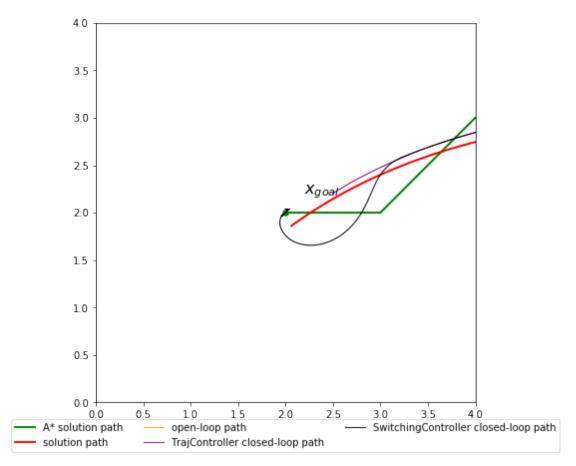


# 0.3.6 Zoom in on final pose

```
In [60]: l_window = 4.

fig = plt.figure(figsize=[7,7])
    astar.plot_path(fig.number)
    plot_traj_smoothed(traj_smoothed)
    plot_traj_ol(states_ol)
    plot_traj_cl(states_cl)
    plot_traj_cl_sw(states_cl_sw)
    plt.legend(loc='upper center', bbox_to_anchor=(0.5, -0.03), fancybox=True, ncol=3)
```

plt.axis([x\_goal[0]-l\_window/2, x\_goal[0]+l\_window/2, x\_goal[1]-l\_window/2, x\_goal[1]plt.show()

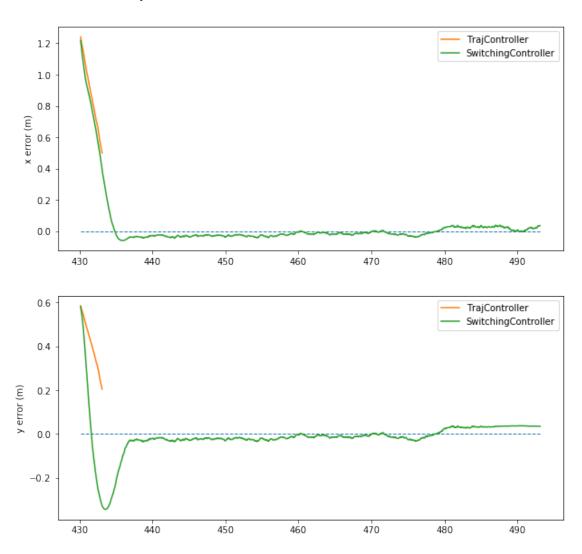


#### 0.3.7 Plot final sequence of states

To see just how well we're able to arrive at the target point (and to assist in choosing values for the pose stabilization controller gains  $k_1, k_2, k_3$ ), we plot the error in x and y for both the tracking controller and the switching controller at the end of the trajectory.

```
plt.plot(times_cl_extended[T:], states_cl_sw[T:,1] - x_goal[1], label='SwitchingControl
plt.legend()
plt.ylabel("y error (m)")
```

Out[61]: Text(0,0.5,'y error (m)')



In []: