sim_traj_planning

October 7, 2020

```
[1]: # The autoreload extension will automatically load in new code as you edit
     \hookrightarrow files,
     # so you don't need to restart the kernel every time
     %load_ext autoreload
     %autoreload 2
     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,_u
     →SwitchingController
     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
```

0.0.1 Generate workspace, start and goal positions

0.0.2 Solve A* planning problem

```
[3]: 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

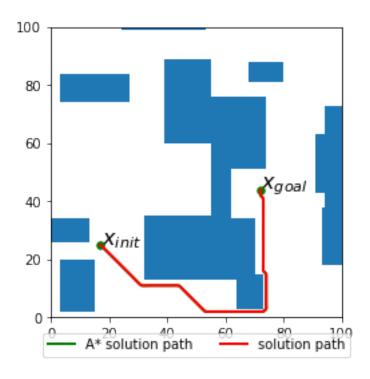
(Try changing these and see what happens)

```
[4]: V_des = 0.3 # Nominal velocity
alpha = 0.1 # Smoothness parameter
dt = 0.05
```

0.1.2 Generate smoothed trajectory

/Users/Elju/.conda/envs/aa274_python27/lib/python2.7/sitepackages/matplotlib/cbook/deprecation.py:107: MatplotlibDeprecationWarning:
Adding an axes using the same arguments as a previous axes currently reuses the earlier instance. In a future version, a new instance will always be created and returned. Meanwhile, this warning can be suppressed, and the future behavior ensured, by passing a unique label to each axes instance.

warnings.warn(message, mplDeprecation, stacklevel=1)



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.

```
[7]: kpx = 2
kpy = 2
kdx = 2
kdy = 2
```

0.2.3 Generate control-feasible trajectory

```
[8]: t_new, V_smooth_scaled, om_smooth_scaled, traj_smooth_scaled = → modify_traj_with_limits(traj_smoothed, t_smoothed, V_max, om_max, dt)
```

0.2.4 Create trajectory controller and load trajectory

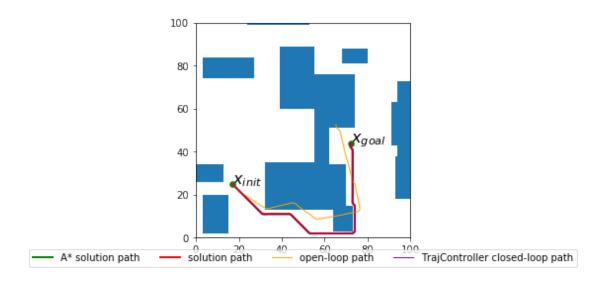
0.2.5 Set simulation input noise

(Try changing this and see what happens)

```
[10]: noise_scale = 0.05
```

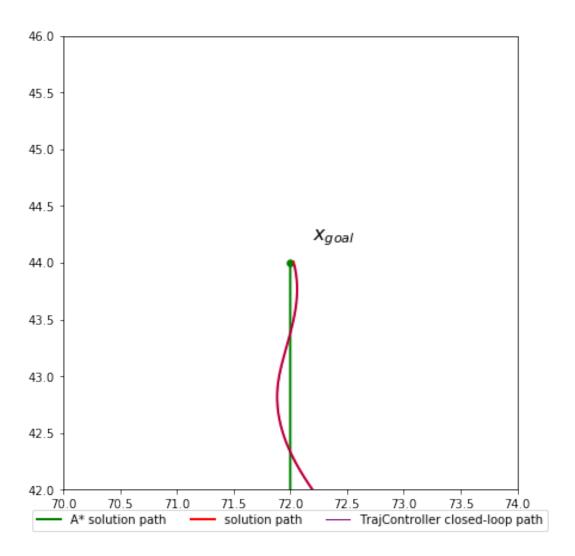
0.2.6 Simulate closed-loop tracking of smoothed trajectory, compare to open-loop

```
[11]: tf actual = t new[-1]
      times_cl = np.arange(0, tf_actual, dt)
      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_ol, noise_scale=noise_scale)
      states_cl, ctrl_cl = simulate_car_dyn(s_0.x, s_0.y, s_0.th, times_cl,_
      →controller=traj_controller, noise_scale=noise_scale)
      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-loop path", zorder=10)
      def plot_traj_cl(states_cl):
          plt.plot(states_cl[:,0], states_cl[:,1], color="purple", linewidth=1,_
      ⇒label="TrajController closed-loop path", zorder=10)
      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



0.3.2 Pose stabilization control gains

Tune these as needed to improve final pose stabilization.

[13]:
$$k1 = 1$$
. $k2 = 1$. $k3 = 1$.

0.3.3 Create pose controller and load goal pose

Note we use the last value of the smoothed trajectory as the goal heading θ

```
[14]: pose_controller = PoseController(k1, k2, k3, V_max, om_max) pose_controller.load_goal(x_goal[0], x_goal[1], traj_smooth_scaled[-1,2])
```

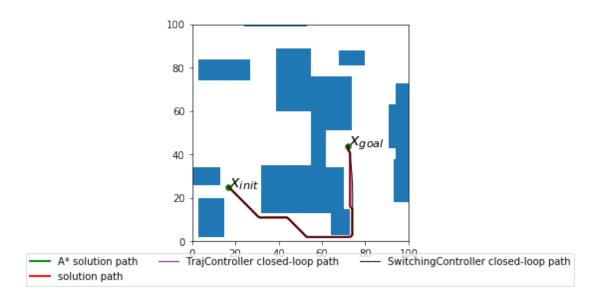
0.3.4 Time before trajectory-tracking completion to switch to pose stabilization

Try changing this!

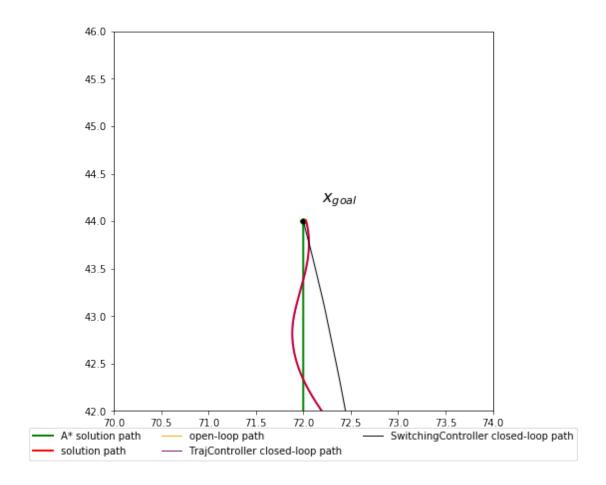
```
[35]: t_before_switch = 250
```

0.3.5 Create switching controller and compare performance

```
[36]: switching_controller = SwitchingController(traj_controller, pose_controller,__
       →t_before_switch)
      t_extend = 60.0 # Extra time to simulate after the end of the nominal trajectory
      times_cl_extended = np.arange(0, tf_actual+t_extend, dt)
      states_cl_sw, ctrl_cl_sw = simulate_car_dyn(s_0.x, s_0.y, s_0.th,_
       -times_cl_extended, controller=switching_controller, noise_scale=noise_scale)
      fig = plt.figure()
      astar.plot_path(fig.number)
      plot_traj_smoothed(traj_smoothed)
      plot_traj_cl(states_cl)
      def plot_traj_cl_sw(states_cl_sw):
          plt.plot(states_cl_sw[:,0], states_cl_sw[:,1], color="black", linewidth=1,_u
      →label="SwitchingController closed-loop path", zorder=10)
      plot_traj_cl_sw(states_cl_sw)
      plt.legend(loc='upper center', bbox_to_anchor=(0.5, -0.03), fancybox=True,__
       \rightarrowncol=3)
      plt.show()
```



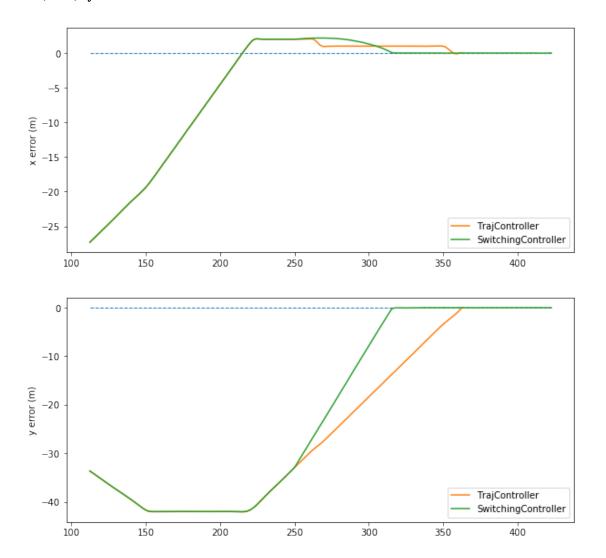
0.3.6 Zoom in on final pose



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.

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



```
[]:
```