```
# The autoreload extension will automatically load in new code as you
In [1]:
        edit 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, 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 siz
```

### Generate workspace, start and goal positions

```
In [2]: 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, max_size)
```

### **Solve A\* planning problem**

```
In [3]: astar = AStar((0, 0), (width, height), x_init, x_goal, occupancy)
if not astar.solve():
    print "No path found"
```

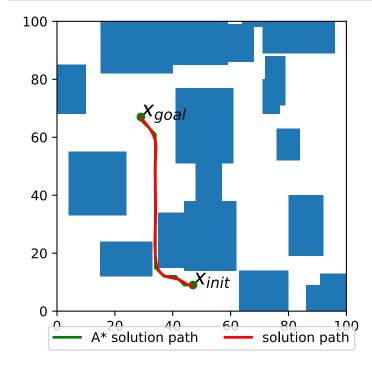
## **Smooth Trajectory Generation**

### **Trajectory parameters**

(Try changing these and see what happens)

```
In [4]: V_des = 0.3 # Nominal velocity
alpha = 3 # Smoothness parameter
dt = 0.05
```

### **Generate smoothed trajectory**



## **Control-Feasible Trajectory Generation and Tracking**

#### **Robot control limits**

```
In [6]: V_max = 0.5 # max speed
om_max = 1 # max rotational speed
```

### **Tracking control gains**

Tune these as needed to improve tracking performance.

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

### **Generate control-feasible trajectory**

### Create trajectory controller and load trajectory

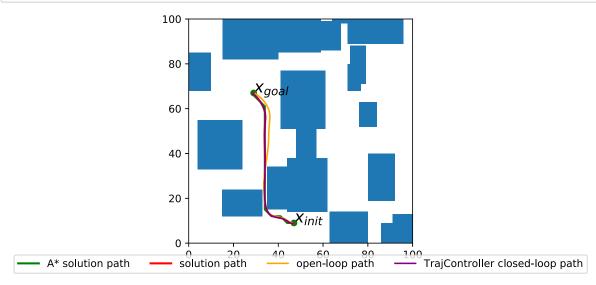
### Set simulation input noise

(Try changing this and see what happens)

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

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

```
In [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[
         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
         =1.5, label="open-loop path", zorder=10)
         def plot traj cl(states cl):
             plt.plot(states cl[:,0], states cl[:,1], color="purple", linewidt
         h=1.5, 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()
```

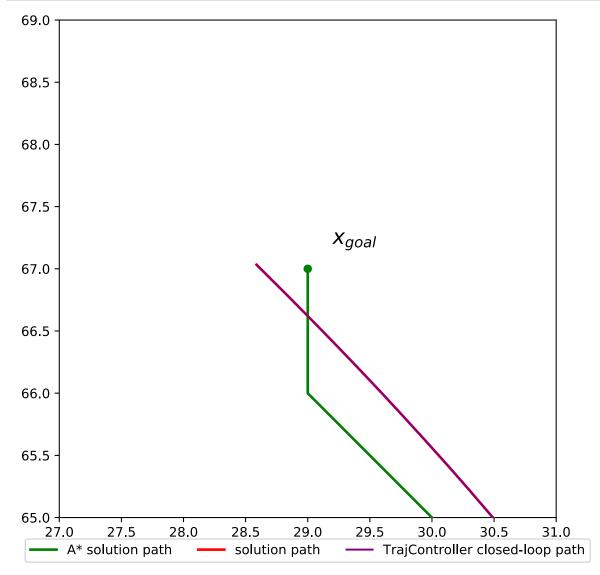


# Switching from Trajectory Tracking to Pose Stabilization Control

### Zoom in on final pose error

```
In [12]: 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]+l_window/2])
plt.show()
```



### Pose stabilization control gains

Tune these as needed to improve final pose stabilization.

```
In [13]: k1 = 1. k2 = 1. k3 = 1.
```

### Create pose controller and load goal pose

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

```
In [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])
```

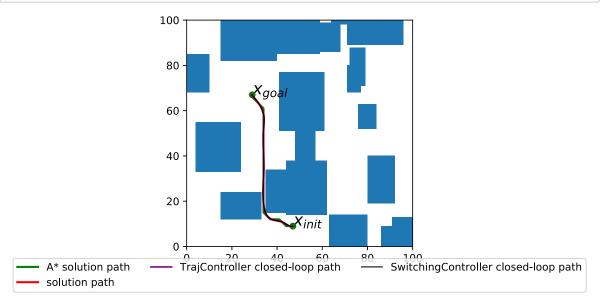
### Time before trajectory-tracking completion to switch to pose stabilization

Try changing this!

```
In [15]: t_before_switch = 5.0
```

### Create switching controller and compare performance

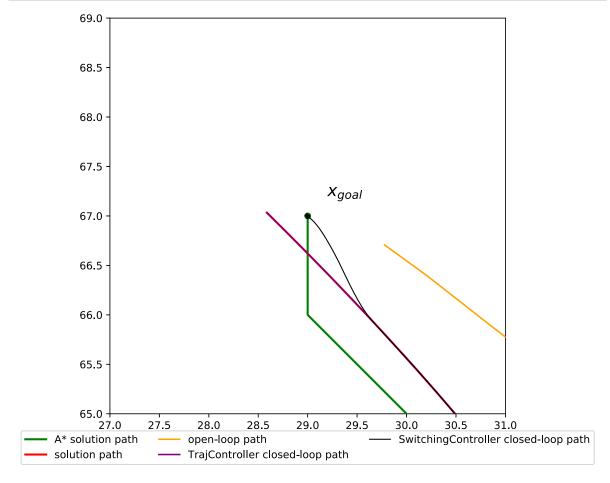
```
switching controller = SwitchingController(traj controller, pose cont
In [16]:
         roller, t before switch)
         t extend = 60.0 # Extra time to simulate after the end of the nominal
         trajectory
         times cl extended = np.arange(0, tf actual+t extend, dt)
         states cl sw, ctrl cl sw = simulate car dyn(s 0.x, s 0.y, s 0.th, tim
         es cl extended, controller=switching controller, noise scale=noise sc
         ale)
         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", lin
         ewidth=1, 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, ncol=3)
         plt.show()
```



### Zoom in on final pose

```
In [17]: l_window = 4.

fig = plt.figure(figsize=[7,7])
    astar.plot_path(fig.number)
    plot_traj_smoothed(traj_smoothed)
    plot_traj_cl(states_cl)
    plot_traj_cl_sw(states_cl_sw)
    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]+l_window/2])
    plt.show()
```

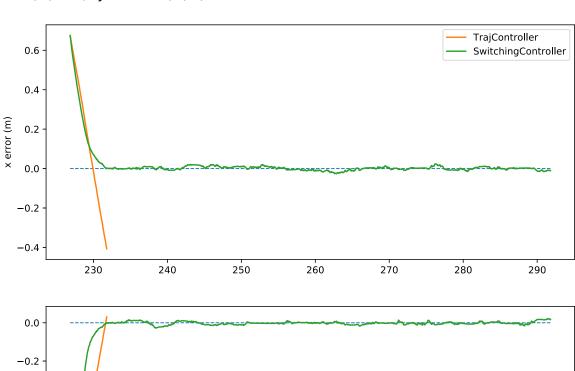


### Plot final sequence of states

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.

```
T = len(times_cl) - int(t_before_switch/dt)
In [18]:
         fig = plt.figure(figsize=[10,10])
         plt.subplot(2,1,1)
         plt.plot([times cl extended[T], times cl extended[-1]], [0,0], linest
         yle='--', linewidth=1)
         plt.plot(times_cl[T:], states_cl[T:,0] - x_goal[0], label='TrajContro
         ller')
         plt.plot(times cl_extended[T:], states_cl_sw[T:,0] - x_goal[0], label
         ='SwitchingController')
         plt.legend()
         plt.ylabel("x error (m)")
         plt.subplot(2,1,2)
         plt.plot([times_cl_extended[T], times_cl_extended[-1]], [0,0], linest
         vle='--', linewidth=1)
         plt.plot(times_cl[T:], states_cl[T:,1] - x_goal[1], label='TrajContro
         ller')
         plt.plot(times cl extended[T:], states cl sw[T:,1] - x goal[1], label
         ='SwitchingController')
         plt.legend()
         plt.ylabel("y error (m)")
```

#### Out[18]: Text(0,0.5,'y error (m)')



-0.4

-0.6

-0.8

-1.0

TrajController

In [ ]: