### AA274 (Winter 2017-18): Problem Set 1

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## 1 Optimal Control

(i) Let  $\mathbf{x} = (x, y, \theta)$  denote the robot state and  $\mathbf{u} = (V, \omega)$  be the robot control inputs. Our optimal control problem is

with initial and final conditions

$$x(0) = 0,$$
  $y(0) = 0,$   $\theta(0) = -\pi/2,$   
 $x(t_f) = 5,$   $y(t_f) = 5,$   $\theta(t_f) = -\pi/2.$ 

Here  $\lambda \in \mathbb{R}_+$  is a weighting factor and  $t_f$  is free. The Hamiltonian is

$$H(t) = \lambda + V(t)^{2} + \omega(t)^{2} + p_{1}(t)V(t)\cos(\theta(t)) + p_{2}(t)V(t)\sin(\theta(t)) + p_{3}(t)\omega(t)$$

where  $\mathbf{p} = (p_1, p_2, p_3)$  are the Lagrange multipliers associated with  $\dot{\mathbf{x}} = (\dot{x}, \dot{y}, \dot{\theta})$ . Our optimality conditions are

$$\begin{pmatrix}
\dot{x}^*(t) \\
\dot{y}^*(t) \\
\dot{\theta}^*(t)
\end{pmatrix} = \begin{pmatrix}
V^*(t)\cos(\theta^*(t)) \\
V^*(t)\sin(\theta^*(t)) \\
\omega^*(t)
\end{pmatrix}$$

$$\begin{pmatrix}
\dot{p}_1^*(t) \\
\dot{p}_2^*(t) \\
\dot{p}_3^*(t)
\end{pmatrix} = \begin{pmatrix}
0 \\
0 \\
p_1^*(t)V^*(t)\sin(\theta^*(t)) - p_2^*(t)V^*(t)\cos(\theta^*(t))
\end{pmatrix}$$
(2)

with the additional constraint (for the control inequality constraints)

$$V^{*}(t)^{2} + \omega^{*}(t)^{2} + p_{1}^{*}(t)V^{*}(t)\cos(\theta^{*}(t)) + p_{2}^{*}(t)V^{*}(t)\sin(\theta^{*}(t)) + p_{3}^{*}(t)\omega^{*}(t)$$

$$< V(t)^{2} + \omega(t)^{2} + p_{1}^{*}(t)V(t)\cos(\theta^{*}(t)) + p_{2}^{*}(t)V(t)\sin(\theta^{*}(t)) + p_{3}^{*}(t)\omega(t)$$
(3)

for all  $(V(t), \omega(t)) \in \mathbb{R}^2$ . Since  $t_f$  is free and  $\mathbf{x}(t_f)$  fixed, the boundary conditions amount to

$$x^*(0) = 0, \quad y^*(0) = 0, \quad \theta^*(0) = -\pi/2,$$

$$x^*(t_f) = 5, \quad y^*(t_f) = 5, \quad \theta^*(t_f) = -\pi/2,$$

$$\lambda + V^*(t_f)^2 + \omega^*(t_f)^2 + p_1^*(t_f)V^*(t_f)\cos(\theta^*(t_f)) + p_2^*(t_f)V^*(t_f)\sin(\theta^*(t_f)) + p_3^*(t_f)\omega^*(t_f)$$

$$= \lambda + V^*(t_f)^2 - p_2^*(t_f)V^*(t_f) + \omega^*(t_f)^2 + p_3^*(t_f)\omega^*(t_f) = 0.$$

We make two modifications in order to solve this problem. First, we ignore the control inequality constraints and manually vary  $\lambda$ . This means that (3) is replaced with

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 2V^*(t) + p_1^*(t)\cos(\theta^*(t)) + p_2^*(t)\sin(\theta^*(t)) \\ 2\omega^*(t) + p_3^*(t) \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} V^*(t) \\ \omega^*(t) \end{pmatrix} = -\frac{1}{2} \begin{pmatrix} p_1^*(t)\cos(\theta^*(t)) + p_2^*(t)\sin(\theta^*(t)) \\ p_3^*(t) \end{pmatrix}$$

Second, we use a change of variables to reformulate the BVP: we rescale the time to  $\tau = \frac{t}{t_f} \in [0,1]$ , so the derivatives become  $\frac{d}{d\tau} := t_f \frac{d}{dt}$ , then introduce a dummy state variable r that corresponds to  $t_f$  with dynamic  $\dot{r} = 0$  and replace all instances of  $t_f$  with r. Let  $\mathbf{z} = (x, y, \theta, p_1, p_2, p_3, r)$  denote the augmented state vector, then  $\frac{d\mathbf{z}}{dt} = \frac{1}{t_f} \frac{d\mathbf{z}}{d\tau} = \frac{1}{r} \frac{d\mathbf{z}}{d\tau}$  and our optimality conditions become

$$\frac{d\mathbf{z}}{d\tau} = z_7 \begin{pmatrix}
V \cos(z_3) \\
V \sin(z_3) \\
\omega \\
0 \\
V(z_4 \sin(z_3) - z_5 \cos(z_3)) \\
0
\end{pmatrix} \tag{4}$$

with boundary conditions

$$z_1(0) = 0, z_2(0) = 0, z_3(0) = -\pi/2$$
  

$$z_1(1) = 5, z_2(1) = 5, z_3(1) = -\pi/2$$
  

$$\lambda + V(1)^2 - z_5(1)V(1) + \omega(1)^2 + z_6(1)\omega(1) = 0$$
(5)

where we have defined

$$\begin{pmatrix} V \\ \omega \end{pmatrix} := -\frac{1}{2} \begin{pmatrix} z_4 \cos(z_3) + z_5 \sin(z_3) \\ z_6 \end{pmatrix} \tag{6}$$

(ii) See submitted code.

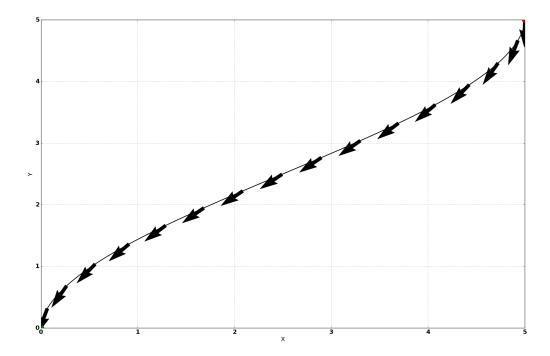
(iii) By choosing the largest feasible  $\lambda$ , we are solving the control problem with the smallest  $t_f$  feasible, i.e., our optimal controls drive the unicycle to its final waypoint in the shortest possible time. This can be seen by separating the objective into

$$J = \int_0^{t_f} [\lambda + V(t)^2 + \omega(t)^2] dt = \lambda t_f + \int_0^{t_f} [V(t)^2 + \omega(t)^2] dt.$$

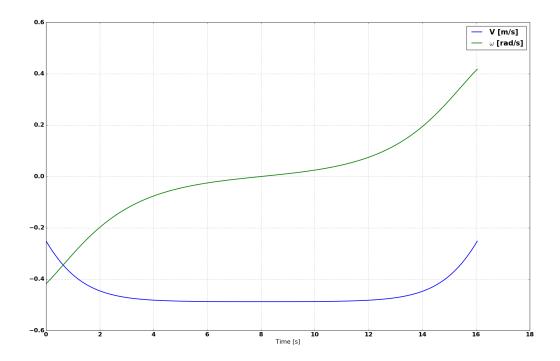
Here  $\lambda \geq 0$  acts like a regularization weight: a larger value of  $\lambda$  upweights  $t_f$  relative to the rest of the objective, which means (since we are minimizing J) that we place more importance on reducing  $t_f$  in the optimization. Selecting the largest feasible  $\lambda$  makes sense, since in general, we prefer to travel between the initial and final waypoints as quickly as possible.

(iv) See submitted code. With  $\lambda = 0.238$  and solution guess  $\mathbf{z}_0 = (0, 0, -\pi/2, 1, 1, 0, 10)$ , we get V < 0 and the robot backs up from (0, 0) to (5, 5). Its trajectory and control histories are shown below.

**Trajectory** 
$$(x(t), y(t))$$

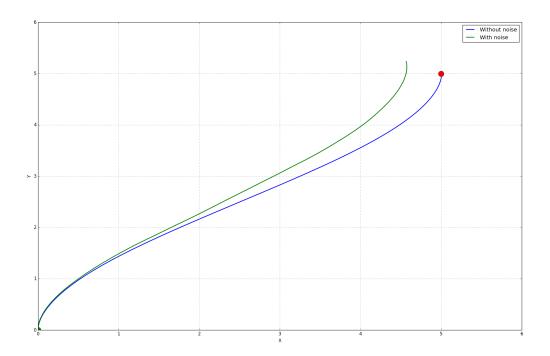


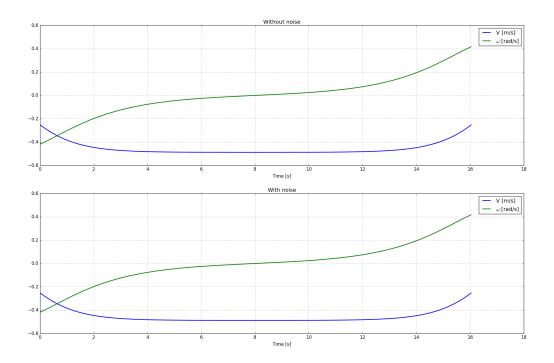
## History of V and $\boldsymbol{\omega}$



(v) See submitted code.

# Trajectory (x(t), y(t)))





### 2 Differential Flatness

(i) Let 
$$\psi_1(t) = 1, \psi_2(t) = t, \psi_3(t) = t^2$$
, and  $\psi_4(t) = t^3$ , then  $n = 4$  and

$$x(t) = \sum_{i=1}^{n} x_i \psi_i(t) = x_1 + x_2 t + x_3 t^2 + x_4 t^3$$

$$y(t) = \sum_{i=1}^{n} y_i \psi_i(t) = y_1 + y_2 t + y_3 t^2 + y_4 t^3$$

$$\dot{x}(t) = \sum_{i=1}^{n} x_i \dot{\psi}_i(t) = x_2 + 2x_3 t + 3x_4 t^2$$

$$\dot{y}(t) = \sum_{i=1}^{n} y_i \dot{\psi}_i(t) = y_2 + 2y_3 t + 3y_4 t^2$$

so the initial and final conditions can be expressed as

$$x(0) = 0 = x_1$$

$$y(0) = 0 = y_1$$

$$x(t_f) = 5 = x_1 + x_2 t_f + x_3 t_f^2 + x_4 t_f^3$$

$$y(t_f) = 5 = y_1 + y_2 t_f + y_3 t_f^2 + y_4 t_f^3$$

$$\dot{x}(0) = V(0)\cos(\theta(0)) = 0.5\cos(-\pi/2) = 0 = x_2$$

$$\dot{y}(0) = V(0)\sin(\theta(0)) = 0.5\sin(-\pi/2) = -1 = y_2$$

$$\dot{x}(t_f) = V(t_f)\cos(\theta(t_f)) = 0.5\cos(-\pi/2) = 0 = x_2 + 2x_3 t_f + 3x_4 t_f^2$$

$$\dot{y}(t_f) = V(t_f)\sin(\theta(t_f)) = 0.5\sin(-\pi/2) = -1 = y_2 + 2y_3 t_f + 3y_4 t_f^2$$

or more succinctly in matrix form,

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ 1 & t_f & t_f^2 & t_f^3 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 2t_f & 3t_f^2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 1 & 15 & 225 & 3375 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 30 & 675 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} 0 \\ 5 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} x(0) \\ x(t_f) \\ \dot{x}(0) \\ \dot{x}(t_f) \end{pmatrix}$$

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ 1 & t_f & t_f^2 & t_f^3 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 2t_f & 3t_f^2 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 1 & 15 & 225 & 3375 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 30 & 675 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix} = \begin{pmatrix} 0 \\ 5 \\ -1 \\ -1 \end{pmatrix} = \begin{pmatrix} y(0) \\ y(t_f) \\ \dot{y}(0) \\ \dot{y}(t_f) \end{pmatrix}$$

We cannot set  $V(t_f) = 0$  because then J would be singular at time  $t_f$ , and we could not recover the flat outputs from the virtual control inputs.

(ii) After solving for  $x_i, y_i$  for i = 1, ..., n, we can recover

$$\begin{pmatrix} x(t) \\ y(t) \\ \dot{x}(t) \\ \dot{y}(t) \\ \ddot{x}(t) \\ \ddot{y}(t) \end{pmatrix} = \begin{pmatrix} x_1 & x_2 & x_3 & x_4 \\ y_1 & y_2 & y_3 & y_4 \\ x_2 & 2x_3 & 3x_4 & 0 \\ y_2 & 2y_3 & 3y_4 & 0 \\ 2x_3 & 6x_4 & 0 & 0 \\ 2y_3 & 6y_4 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ t \\ t^2 \\ t^3 \end{pmatrix}$$

along with the alignment angle

$$\frac{\dot{y}(t)}{\dot{x}(t)} = \frac{V \sin(\theta(t))}{V \cos(\theta(t))} = \tan(\theta(t)) \Rightarrow \theta(t) = \arctan\left(\frac{\dot{y}(t)}{\dot{x}(t)}\right).$$

Since V > 0, we take the positive root of

$$\dot{x}(t)^2 + \dot{y}(t)^2 = V^2 \cos^2(\theta(t)) + V^2 \sin^2(\theta(t)) = V^2 \Rightarrow V(t) = \sqrt{\dot{x}(t)^2 + \dot{y}(t)^2},$$

then the matrix J is invertible, so we can calculate

$$J^{-1} = \begin{pmatrix} \cos(\theta) & -V\sin(\theta) \\ \sin(\theta) & V\cos(\theta) \end{pmatrix}^{-1} = \frac{1}{V} \begin{pmatrix} V\cos(\theta) & V\sin(\theta) \\ -\sin(\theta) & \cos(\theta) \end{pmatrix}$$
$$\begin{pmatrix} a \\ \omega \end{pmatrix} = J^{-1} \begin{pmatrix} \ddot{x}(t) \\ \ddot{y}(t) \end{pmatrix} = \begin{pmatrix} \ddot{x}(t)\cos(\theta) + \ddot{y}(t)\sin(\theta) \\ \frac{1}{V}(-\ddot{x}(t)\sin(\theta) + \ddot{y}(t)\cos(\theta)) \end{pmatrix}.$$

Putting this all together, our state-trajectory is

$$x(t) = x_1 + x_2t + x_3t^2 + x_4t^3$$

$$y(t) = y_1 + y_2t + y_3t^2 + y_4t^3$$

$$\theta(t) = \arctan\left(\frac{y_2 + 2y_3t + 3y_4t^2}{x_2 + 2x_3t + 3x_4t^2}\right)$$

and our control history is

$$V(t) = \sqrt{(x_2 + 2x_3t + 3x_4t^2)^2 + (y_2 + 2y_3t + 3y_4t^2)^2}$$
  

$$\omega(t) = \frac{1}{V(t)} \left( -(2x_3 + 6x_4t)\sin(\theta(t)) + (2y_3 + 6y_4t)\cos(\theta(t)) \right)$$

(iii) Given  $\dot{s}(t) = V(t)$  with s(0) = 0, we can integrate to get the path parameter

$$s(t) = \int_0^t V(t')dt' - V(0)$$

We wish to find an alternative velocity control  $\tilde{V}(s)$  that satisfies the control saturation constraints. Then, the corresponding angular velocity control and time history are

$$\tilde{\omega}(s) = \omega(s) \frac{\tilde{V}(s)}{V(s)}, \quad \tau(s) = \int_0^s \frac{ds'}{\tilde{V}(s')}.$$

Combining the constraint  $|\tilde{V}(s)| \le 0.5$  m/s with

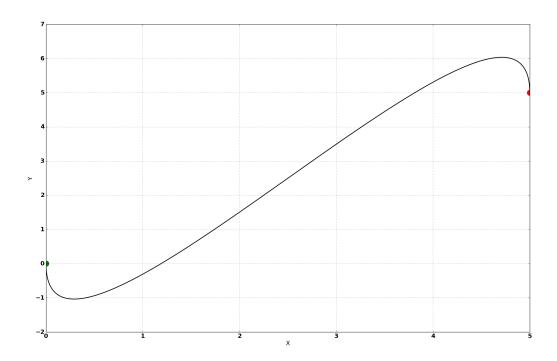
$$|\tilde{\omega}(s)| = \left|\omega(s)\frac{\tilde{V}(s)}{V(s)}\right| \le 1 \text{ rad/s} \Rightarrow |\tilde{V}(s)| \le \frac{V(s)}{|\omega(s)|} \text{ for } \omega(s) \ne 0$$

we see that a viable choice is

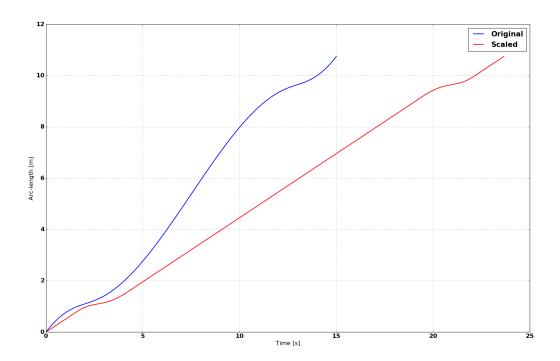
$$\tilde{V}(s) = \begin{cases} \min(V(s), 0.5) & \text{if } \omega(s) = 0\\ \min\left(V(s), \frac{V(s)}{|\omega(s)|}, 0.5\right) & \text{if } \omega(s) \neq 0 \end{cases}$$

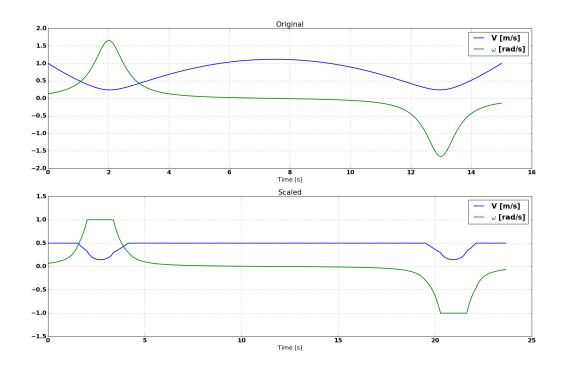
(iv) See submitted code.

# Trajectory (x(t), y(t))



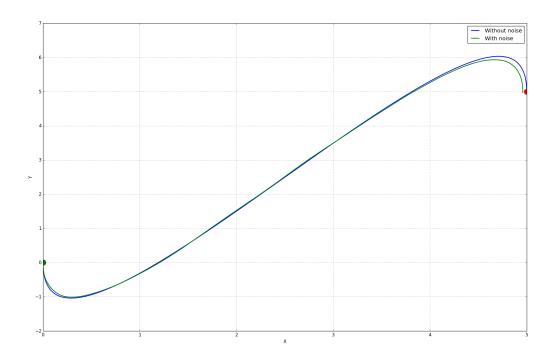
# $\mathbf{Arc\text{-length}}\ s(t)$

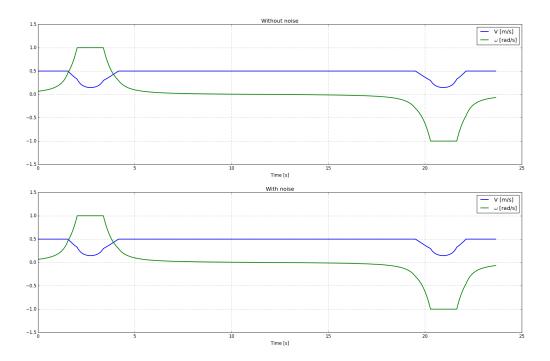




(v) See submitted code.

# **Trajectory** (x(t), y(t))





## 3 Closed-loop Control I

(i) We are given the control law

$$V = k_1 \rho \cos(\alpha)$$

$$\omega = k_2 \alpha + k_1 \frac{\sin(\alpha) \cos(\alpha)}{\alpha} (\alpha + k_3 \delta)$$

for the ODE system

$$\dot{\rho}(t) = -V(t)\cos(\alpha(t))$$

$$\dot{\alpha}(t) = V(t)\frac{\sin(\alpha(t))}{\rho(t)} - \omega(t)$$

$$\dot{\delta}(t) = V(t)\frac{\sin(\alpha(t))}{\rho(t)}$$

where  $k_1, k_2, k_3 > 0$  are constants. From Homework 1, Figure 2, we readily see that given a starting position  $(x, y, \theta)$  and final position  $(x_g, y_g, \theta_g)$ , the incremental change

in polar coordinates is

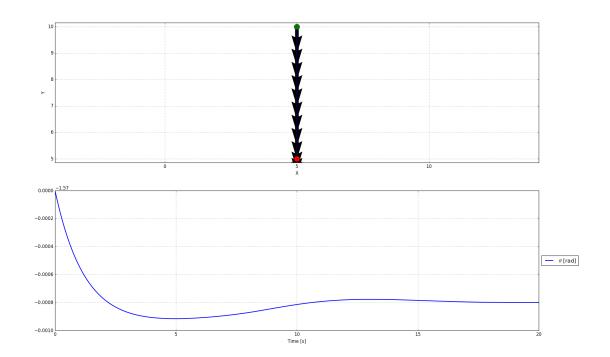
$$\rho = \sqrt{(x_g - x)^2 + (y_g - y)^2}$$
$$\tan(\alpha + \theta) = \frac{y_g - y}{x_g - x} \Rightarrow \alpha = \arctan\left(\frac{y_g - y}{x_g - x}\right) - \theta$$
$$\alpha + \theta = \delta + \theta_g \Rightarrow \delta = \alpha + \theta - \theta_g$$

(ii) See submitted code. The parameters used were

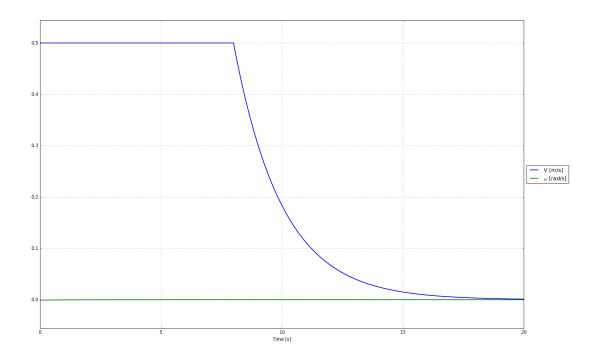
$$(x_0, y_0, \theta_0, t_f) = \begin{cases} (5, 10, -\pi/2, 20) & \text{for forward parking} \\ (5, 0, -\pi/2, 12) & \text{for reverse parking} \\ (0, 5, -\pi/2, 18) & \text{for parallel parking} \end{cases}$$

$$(k_1, k_2, k_3) = \begin{cases} (0.5, 0.5, 1.2) & \text{for forward parking} \\ (1.45, 0.01, 1.45) & \text{for reverse parking} \\ (0.5, 0.5, 1.2) & \text{for parallel parking} \end{cases}$$

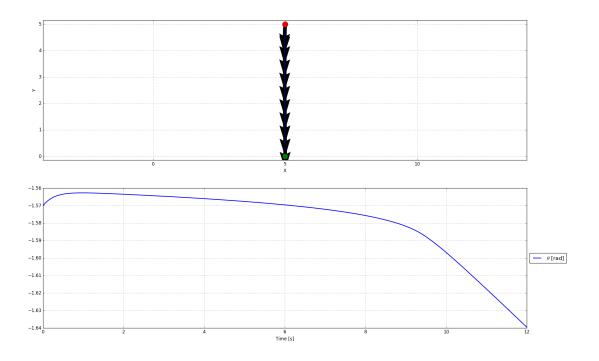
#### Forward Trajectory



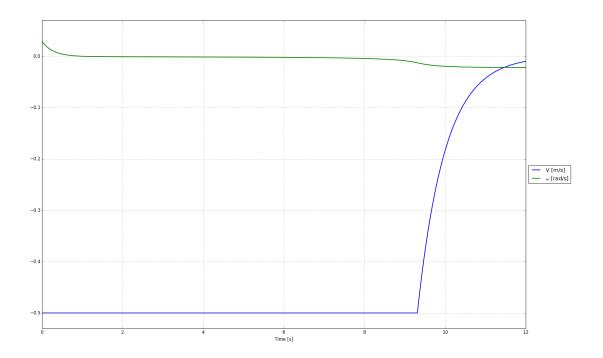
### Forward Control



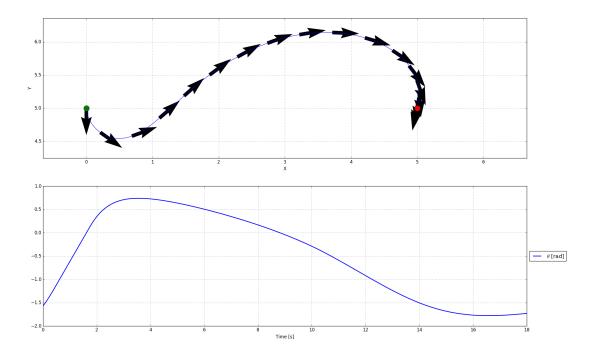
## Reverse Trajectory



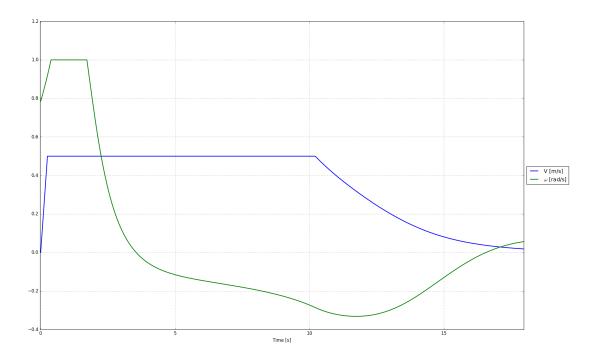
### Reverse Control



## Parallel Trajectory



#### Parallel Control



## 4 Closed-loop Control II

(i) In Problem 2, part (ii), we showed that for a given state  $\mathbf{x} = (x, y, \theta)$ ,

$$\begin{pmatrix} a \\ \omega \end{pmatrix} = \begin{pmatrix} \dot{V} \\ \omega \end{pmatrix} = \begin{pmatrix} u_1 \cos(\theta) + u_2 \sin(\theta) \\ \frac{1}{V}(-u_1 \sin(\theta) + u_2 \cos(\theta)) \end{pmatrix}$$
 (7)

where the virtual controls are  $(u_1, u_2) = (\ddot{x}, \ddot{y})$ .

(ii) See submitted code. We assume the control law is implemented at a high enough frequency that the state variables  $\mathbf{x} = (x, y, \theta)$  can be treated as constant. Furthermore, at each step t, we consider the current velocity to be that commanded at t - 1, i.e.,

$$\dot{x}_t = V_{t-1}\cos(\theta), \quad \dot{y}_t = V_{t-1}\sin(\theta),$$

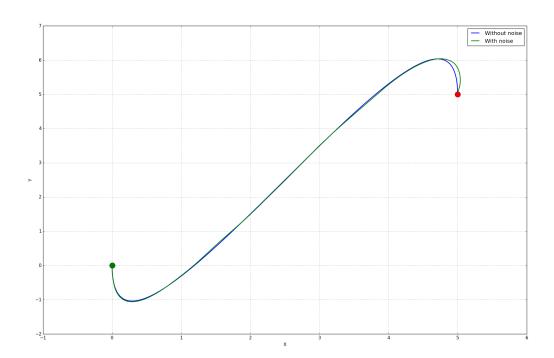
Using this, we can calculate  $(u_1, u_2)$  and plug into (7) to get the current acceleration,  $\dot{V}_t$ . We then apply the Euler method to update

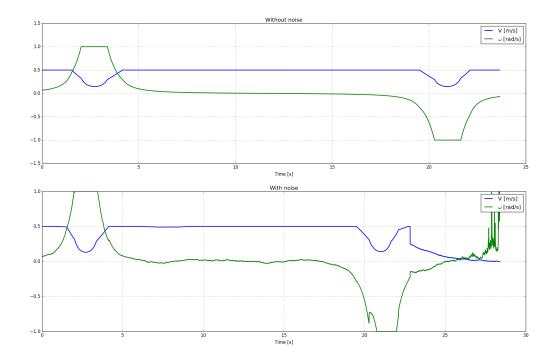
$$V_t = V_{t-1} + \delta \dot{V}_t$$
 where  $\delta$  is the timestep.

If  $V_t \approx 0$ , we reset it to the nominal (desired) velocity  $V_d = \sqrt{\dot{x}_d^2 + \dot{y}_d^2}$  according to the robot kinematic model. Finally, we calculate  $\omega$ .

- (iii) See submitted code.
- (iv) See submitted code. For this simulation, we set  $(k_{px}, k_{py}, k_{dx}, k_{dy}) = (1, 1, 0.5, 0.5)$  and reset to the nominal velocity whenever  $|V_t| \le \epsilon = 10^{-5}$ .

**Trajectory** (x(t), y(t))





# 5 Robot Operating System

- (i) See submitted code.
- (ii) rosbag play filename.bag
- (iii) See submitted code.