On Feedback Linearization of Mobile Robots

Xiaoping Yun
University of Pennsylvania

Yoshio Yamamoto
University of Pennsylvania

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Abstract
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On Feedback Linearization of Mobile Robots

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Xiaoping Yun
Yoshio Yamamoto

University of Pennsylvania
School of Engineering and Applied Science
Computer and Information Science Department
Philadelphia, PA 19104-6389

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On Feedback Linearization of Mobile Robots

Xiaoping Yun and Yoshio Yamamoto
General Robotics and Active Sensory Perception (GRASP) Laboratory
University of Pennsylvania
3401 Walnut Street, Room 301C
Philadelphia, PA 19104-6228

ABSTRACT

A wheeled mobile robot is subject to both holonomic and nonholonomic constraints. Representing the motion and constraint equations in the state space, this paper studies the feedback linearization of the dynamic system of a wheeled mobile robot. The main results of the paper are: (1) It is shown that the system is not input-state linearizable. (2) If the coordinates of a point on the wheel axis are taken as the output equation, the system is not input-output linearizable by using a static state feedback; (3) but is input-output linearizable by using a dynamic state feedback. (4) If the coordinates of a reference point in front of the mobile robot are chosen as the output equation, the system is input-output linearizable by using a static state feedback. (5) The internal motion of the mobile robot when the reference point moves forward is asymptotically stable whereas the internal motion when the reference point moves backward is unstable. A nonlinear feedback is derived for each case where the feedback linearization is possible.

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1 Introduction

The feedback linearization of nonlinear systems has been extensively studied in the literature [1, 2, 3, 4, 5]. Broadly speaking, there are two types of linearization: input-state linearization and input-output linearization. Necessary and sufficient conditions have been established for each type of linearization [6, 7]. For a given nonlinear system, these conditions can be checked to determine if the system is linearizable. Two types of feedback are commonly employed for the purpose of linearization: static state feedback and dynamic state feedback. The dynamic state feedback is more general and includes the static state feedback as a special case. Consequently, the conditions for the dynamic state feedback are more complicated.

In this paper, we study the feedback linearization of a wheeled mobile robot. Due to the fact that the wheeled mobile robot is nonholonomically constrained, the wheeled mobile robot possesses a number of distinguishing properties as far as the feedback linearization is concerned. In particular, we will first show that the dynamic system of a wheeled mobile robot is not input-state linearizable. We then study the input-output linearization of the system for two types of output equations which are chosen for the trajectory tracking of the mobile robot. The first output takes the coordinates of the center point on the wheel axis, and the other output takes the coordinates of a reference point in front of the mobile robot. With the first output equation, we should that the system is not input-output linearizable by using a static state feedback but is input-output linearizable by using a dynamic state feedback. The dynamic feedback achieving the input-output linearization is constructed following the dynamic extension algorithm [7, 8]. With the second type of output equation, the system is input-output linearizable by simply using a static state feedback. Nevertheless, the internal dynamics of the system is not always stable. Specifically, when the reference point is controlled to move backward, the internal motion of the system is unstable.

Although motion planning of mobile robots have been an active topic in robotics in the past decade [9, 10, 11, 12, 13], the study on the feedback control of mobile robots is very recent [14, 15, 16]. The work which is most closely related to the present study is by d’Andrea-Novel et al. [17] who studied full linearization of wheeled mobile robots. Since they used a reduced model, the motions of mobile robots are not completely characterized. In particular, the nonlinear internal dynamics, which are a major topic of this study, are excluded from the motion equations. Bloch and McClamrock [18] showed that a nonholonomic system, including wheeled mobile robot systems, cannot be stabilized to a single equilibrium point by a smooth feedback. Walsh et al. [19] suggested a control law to stabilize the nonholonomic system about a trajectory, instead of a point. Other relevant work includes [20, 21] which proved that systems with nonholonomic constraints are small-time locally controllable.
2 Dynamics of a Wheeled Mobile Robot

2.1 Constraint Equations

In this section, we derive the motion equations and constraint equations of a wheeled mobile robot whose schematic top view is shown in Figure 1. We assume that the mobile robot is driven by two independent wheels and supported by four passive wheels at the corners (not shown in Figure 1). Before proceeding, let us fix some notations (see Figure 1).

- \( b \): the displacement from each of the driving wheels to the axis of symmetry.
- \( d \): the displacement from point \( P_0 \) to the mass center of the mobile robot, which is assumed to be on the axis of symmetry.
- \( r \): the radius of the driving wheels.
- \( c \): \( r/2b \).
- \( m_c \): the mass of the mobile robot without the driving wheels and the rotors of the motors.
- \( m_w \): the mass of each driving wheel plus the rotor of its motor.
- \( I_c \): the moment of inertia of the mobile robot without the driving wheels and the rotors of the motors about a vertical axis through the intersection of the axis of symmetry with the driving wheel axis.
- \( I_w \): the moment of inertia of each driving wheel and the motor rotor about the wheel axis.
- \( I_m \): the moment of inertia of each driving wheel and the motor rotor about a wheel diameter.

There are three constraints. The first one is that the mobile robot can not move in lateral direction, i.e.,

\[
\dot{x}_2 \cos \phi - \dot{x}_1 \sin \phi = 0
\]  

(1)
where \((x_1, x_2)\) is the coordinates of point \(P_0\) in the fixed reference coordinated frame \(X_1-X_2\), and \(\phi\) is the heading angle of the mobile robot measured from \(x_1\)-axis. The other two constraints are that the two driving wheels roll and do not slip:

\[
\begin{align*}
\dot{x}_1 \cos \phi + \dot{x}_2 \sin \phi + b \dot{\phi} &= r \dot{\theta}_1 \\
\dot{x}_1 \cos \phi + \dot{x}_2 \sin \phi - b \dot{\phi} &= r \dot{\theta}_2
\end{align*}
\]

where \(\theta_1\) and \(\theta_2\) are the angular positions of the two driving wheels, respectively.

Let the generalized coordinates of the mobile robot be \(q = (x_1, x_2, \phi, \theta_1, \theta_2)\). The three constraints can be written as follows

\[
A(q) \dot{q} = 0
\]

where

\[
A(q) = \begin{bmatrix}
- \sin \phi & \cos \phi & 0 & 0 & 0 \\
- \cos \phi & - \sin \phi & -b & r & 0 \\
- \cos \phi & - \sin \phi & b & 0 & r
\end{bmatrix}
\]

We define a 5 x 2 dimensional matrix as follows

\[
S(q) = [s_1(q) \ s_2(q)] = \begin{bmatrix}
-rb \cos \phi & cb \cos \phi \\
-rc \sin \phi & cb \sin \phi \\
c & -c \\
1 & 0 \\
0 & 1
\end{bmatrix}
\]

The two independent columns of matrix \(S(q)\) are in the null space of matrix \(A(q)\), that is, \(A(q)S(q) = 0\). We define a distribution spanned by the columns of \(S(q)\)

\[
\Delta = \text{span}\{s_1(q), \ s_2(q)\}
\]

The involutivity of the distribution \(\Delta\) determines the number of holonomic or nonholonomic constraints [21]. If \(\Delta\) is involutive, from the Frobenius theorem [22], all the constraints are integrable (thus holonomic). If the smallest involutive distribution containing \(\Delta\) (denoted by \(\Delta^*\)) spans the entire 5-dimensional space, all the constraints are nonholonomic. If \(\text{dim}(\Delta^*) = 5 - k\), then \(k\) constraints are holonomic and the others are nonholonomic.

To verify the involutivity of \(\Delta\), we compute the Lie bracket of \(s_1(q)\) and \(s_2(q)\).

\[
s_3(q) = [s_1(q), s_2(q)] = \frac{\partial s_2}{\partial q} s_1 - \frac{\partial s_1}{\partial q} s_2 = \begin{bmatrix}
-rb \sin \phi \\
r \cos \phi \\
0 \\
0 \\
0
\end{bmatrix}
\]
which is not in the distribution $\Delta$ spanned by $s_1(q)$ and $s_2(q)$. Therefore, at least one of the constraints is nonholonomic. We continue to compute the Lie bracket of $s_1(q)$ and $s_3(q)$

$$s_4(q) = [s_1(q), s_3(q)] = \frac{\partial s_3}{\partial q} s_1 - \frac{\partial s_1}{\partial q} s_3 = \begin{bmatrix} -rc^2 \cos \phi \\ -rc^2 \sin \phi \\ 0 \\ 0 \end{bmatrix}$$

which is linearly independent of $s_1(q)$, $s_2(q)$, and $s_3(q)$. However, the distribution spanned by $s_1(q)$, $s_2(q)$, $s_3(q)$ and $s_4(q)$ is involutive. Therefore, we have

$$\Delta^* = \text{span}\{s_1(q), s_2(q), s_3(q), s_4(q)\}$$

It follows that, among the three constraints, two of them are nonholonomic and the third one is holonomic. To obtain the holonomic constraint, we subtract equation (2) from equation (3).

$$2b \dot{\phi} = r(\dot{\theta}_r - \dot{\theta}_l)$$

Integrating the above equation and properly choosing the initial condition of $\phi$, $\theta_r$, and $\theta_l$, we have

$$\phi = c(\theta_r - \theta_l)$$

which is clearly a holonomic constraint equation. Thus $\phi$ may be eliminated from the generalized coordinates. The new generalized coordinates are 4-dimensional, which will be denoted by $q$ again.

$$q = \begin{bmatrix} q_1 \\ q_2 \\ q_3 \\ q_4 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ \theta_1 \\ \theta_2 \end{bmatrix}$$

The two nonholonomic constraints are

$$\dot{x}_1 \sin \phi - \dot{x}_2 \cos \phi = 0$$

$$\dot{x}_1 \cos \phi + \dot{x}_2 \sin \phi = cb(\dot{\theta}_1 + \dot{\theta}_2)$$

where $cb = \frac{\pi}{2}$ as defined early. The second nonholonomic constraint equation in the above is obtained by adding equations (2) and (3). It is understood that $\phi$ is now a short-hand notation for $c(\theta_1 - \theta_2)$ rather than an independent variable. We write these two constraint equations in matrix form

$$A(q) \dot{q} = 0$$

where $q$ is now defined in equation (10) and $A(q)$ is given below

$$A(q) = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \end{bmatrix} = \begin{bmatrix} -\sin \phi & \cos \phi & 0 & 0 \\ -\cos \phi & -\sin \phi & cb & cb \end{bmatrix}$$
2.2 Dynamic Equations

We use the Lagrange formulation to establish equations of motion for the mobile robot. The total kinetic energy of the mobile base and the two wheels is

\[
K = \frac{1}{2}m(\dot{x}_1^2 + \dot{x}_2^2) + m_c d(\dot{\theta}_1 - \dot{\theta}_2)(\dot{x}_2 \cos \phi - \dot{x}_1 \sin \phi) + \frac{1}{2}I_w(\dot{\theta}_1^2 + \dot{\theta}_2^2) + \frac{1}{2}I_c^2(\dot{\theta}_1^2 - \dot{\theta}_2)^2 \quad (15)
\]

where

\[
m = m_c + 2m_w
\]
\[
I = I_c + 2m_w b^2 + 2I_m
\]

Lagrange equations of motion for the nonholonomic mobile robot system are governed by [23]

\[
\frac{d}{dt} \left( \frac{\partial K}{\partial \dot{q}_i} \right) - \frac{\partial K}{\partial q_i} = f_i - a_{i1}\lambda_1 - a_{i2}\lambda_2, \quad i = 1, \ldots, 4 \quad (16)
\]

where \(q_i\) is the generalized coordinate defined in equation (10), \(f_i\) is the generalized force, \(a_{ij}\) is from the constraint equation (14), and \(\lambda_1\) and \(\lambda_2\) are the Lagrange multipliers. Substituting the total kinetic energy (equation (15)) into equation (16), we obtain

\[
m\ddot{x}_1 - m_c d (\ddot{\phi} \sin \phi + \dot{\phi}^2 \cos \phi) = \lambda_1 \sin \phi + \lambda_2 \cos \phi \quad (17)
\]
\[
m\ddot{x}_2 + m_c d (\ddot{\phi} \cos \phi - \dot{\phi}^2 \sin \phi) = -\lambda_1 \cos \phi + \lambda_2 \sin \phi \quad (18)
\]
\[
m_c d(\ddot{x}_2 \cos \phi - \ddot{x}_1 \sin \phi) + (Ic^2 + I_w)\ddot{\theta}_1 - Ic^2 \ddot{\theta}_2 = \tau_1 - cb\lambda_2 \quad (19)
\]
\[
-m_c d(\ddot{x}_2 \cos \phi - \ddot{x}_1 \sin \phi) - Ic^2 \ddot{\theta}_1 + (Ic^2 + I_w)\ddot{\theta}_2 = \tau_2 - cb\lambda_2 \quad (20)
\]

where \(\tau_1\) and \(\tau_2\) are the torques acting on the two wheels. These equations can be written in the matrix form

\[
M(q)\ddot{q} + V(q, \dot{q}) = E(q)\tau - A^T(q)\lambda \quad (21)
\]

where \(A(q)\) is defined in equation (14) and

\[
M(q) = \begin{bmatrix}
m & 0 & -m_c d \sin \phi & m_c d \sin \phi \\
0 & m & m_c d \cos \phi & -m_c d \cos \phi \\
-m_c d \sin \phi & m_c d \cos \phi & Ic^2 + I_w & -Ic^2 \\
m_c d \sin \phi & -m_c d \cos \phi & -Ic^2 & Ic^2 + I_w
\end{bmatrix}
\]
\[
V(q, \dot{q}) = \begin{bmatrix}
-m_c d \dot{\phi}^2 \cos \phi \\
-m_c d \dot{\phi}^2 \sin \phi \\
0 \\
0
\end{bmatrix}
\]
\[
E(q) = \begin{bmatrix}
0 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 1 & 0 & 1
\end{bmatrix}
\]
\[
\tau = \begin{bmatrix}
\tau_1 \\
\tau_2
\end{bmatrix}
\]
\[
\lambda = \begin{bmatrix}
\lambda_1 \\
\lambda_2
\end{bmatrix}
\]

5
2.3 State Space Realization

In this subsection, we establish a state space realization of the motion equation (21) and constraint equation (13). Let \( S(q) \) be a \( 4 \times 2 \) matrix whose columns are in the null space of \( A(q) \) matrix in the constraint equation (13), i.e., \( A(q)S(q) = 0 \). From the constraint equation (13), the velocity \( \dot{q} \) must be in the null space of \( A(q) \). It follows that \( \dot{q} \in \text{span}\{s_1(q), \ s_2(q)\} \), and that there exists a smooth vector \( \eta = [\eta_1 \ \eta_2]^T \) such that

\[
\dot{q} = S(q)\eta \tag{23}
\]

and

\[
\ddot{q} = S(q)\dot{\eta} + \dot{S}(q)\eta \tag{24}
\]

For the specific choice of \( S(q) \) matrix in equation (22), we have \( \eta = \dot{\theta} \), where \( \dot{\theta} = [\dot{\theta}_1 \ \dot{\theta}_2]^T \).

Now multiplying the both sides of equation (21) by \( S^T(q) \) and noticing that \( S^T(q)A^T(q) = 0 \) and \( S^T(q)E(q) = I_{2x2} \) (the \( 2 \times 2 \) identity matrix), we obtain

\[
S^T(q)M(q)\ddot{q} + S^T(q)V(q, \dot{q}) = S^T(q)E(q)\tau = \tau \tag{25}
\]

Substituting equation (24) into the above equation, we have

\[
S^T(q)M(q)(S(q)\dot{\eta} + \dot{S}(q)\eta) + S^T(q)V(q, \dot{q}) = \tau \tag{26}
\]

By choosing the following state variable

\[
x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ \theta_1 \\ \theta_2 \\ \eta_1 \\ \eta_2 \end{bmatrix} = \begin{bmatrix} q \\ \eta \end{bmatrix} \tag{27}
\]

we may represent the motion equation (26) in the state space form

\[
\dot{x} = f(x) + g(x)\tau \tag{28}
\]

where

\[
f(x) = \begin{bmatrix} S\eta \\ -(S^TMS)^{-1}(S^TMS\dot{\eta} + S^TV) \end{bmatrix}, \quad g(x) = \begin{bmatrix} 0 \\ (S^TMS)^{-1} \end{bmatrix}
\]

It is noted that the dependent variables for each term have been omitted in the above representation for clarity. All the terms are functions of the state variable \( x \) only. Since \( \dot{q} \) is not part of the state variable, it is replaced by \( S(q)\eta \).
3 Input-State Linearization

In this section, we study the input-state linearization of the control system (28) using smooth nonlinear feedbacks. To simplify the discussion, we first apply the following state feedback

\[ \tau = \alpha^1(x) + \beta^1(x) \mu \]

\[ = (S^T M \dot{\eta} + S^T V) + (S^T M S) S^T E \mu \]  

(29)

where \( \mu \) is the new input variable. The closed-loop system becomes

\[ \dot{x} = f^1(x) + g^1(x) \mu \]

(30)

where

\[ f^1(x) = \begin{bmatrix} S \eta \\ 0 \end{bmatrix} \quad g^1(x) = \begin{bmatrix} 0 \\ I_{2 \times 2} \end{bmatrix} \]

**Theorem 1** System (30) is not input-state linearizable by a smooth state feedback.

**Proof:** If the system is input-state linearizable, it has to satisfy two conditions: the strong accessibility condition and the involutivity condition [7, p. 179]. We will show that the system does not satisfy the involutivity condition.

Define a sequence of distributions

\[ D_j = \text{span}\{ L_{f^1} g^1 \mid i = 0,1,\ldots,j-1 \}, \quad j = 1,2,\ldots \]

Then the involutivity condition requires that the distributions \( D_1, D_2, \ldots, D_6 \) be all involutive, with 6 being the dimension of the system. \( D_1 = \text{span}\{ g^1 \} \) is involutive since \( g^1 \) is constant. Next we compute

\[ L_{f^1} g^1 = [f^1, g^1] = \frac{\partial g^1}{\partial x} f^1 - \frac{\partial f^1}{\partial x} g^1 = - \begin{bmatrix} S(q) \\ 0 \end{bmatrix} \]

It is easy to verify that the distribution spanned by the columns of \( S(q) \) is not involutive. (Actually, if the distribution were involutive, the two constraints (11) and (12) would be holonomic.) It follows that the distribution \( D_2 = \text{span}\{ g^1, L_{f^1} g^1 \} \) is not involutive. Therefore, the system is not input-state linearizable.

**Corollary 1** System (28) is not input-state linearizable by a smooth state feedback.

**Proof:** A proof similar to that of Theorem 1 can be carried out. Alternatively, system (30) can be regarded as a special case of system (28).
4 Input-Output Linearization and Decoupling

Although the dynamic system of a wheeled mobile robot is not input-state linearizable as shown in the previous section, it may be input-output linearizable. In this section, we study the input-output linearization of two types of outputs. First, the coordinates of the center point $P_0$ are chosen as the output equation. It will be shown that the input-output linearization is not possible by using static state feedback, but is possible by using a dynamic state feedback. Second, the coordinates of a reference point $P_r$ in front of the mobile robot is chosen as the output equation. In this case, the input-output linearization can be achieved by using a static state feedback. Nevertheless, the internal dynamics when the mobile robot moves backwards is unstable.

4.1 Controlling the Center Point $P_0$

Since the mobile robot has two inputs, we may choose an output equation with two independent components. A natural choice for the output equation is the coordinates of the center point $P_0$, i.e.,

$$y = h(x) = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}.$$  \hfill (31)

Together with this output equation, we will consider the state equation (30), assuming that the nonlinear feedback (29) is applied to cancel the dynamic nonlinearity. To verify if the system is input-output linearizable, we compute the time derivatives of $y$.

$$\dot{y} = \frac{\partial h}{\partial x} \dot{x} = \frac{\partial h}{\partial x} \left( f^1(x) + g^1(x)\mu \right) = S_1(x)\eta$$

where

$$S_1(x) = \begin{bmatrix} cb \cos \phi & cb \cos \phi \\ cb \sin \phi & cb \sin \phi \end{bmatrix}.$$  

Since $\dot{y}$ is not a function of the input $\mu$, we differentiate once more.

$$\ddot{y} = S_1(x)\ddot{\eta} + \ddot{S}_1(x)\eta = S_1(x)\mu + \dot{S}_1(x)\eta$$  \hfill (32)

where the second term on the right-hand side is evaluated to be

$$\dot{S}_1(x)\eta = c^2 b(\eta_1^2 - \eta_2^2) \begin{bmatrix} -\sin \phi \\ \cos \phi \end{bmatrix}.$$  

Now that $\ddot{y}$ is a function of the input $\mu$, the decoupling matrix of the system is $S_1(x)$. Since $S_1(x)$ is singular, the system is not input-output linearizable and the output can not be decoupled by using any static state feedback [6, 14, 15].
4.2 Dynamic Feedback Control

As shown above, the mobile robot under the output equation (31) is not input-output linearizable with any static feedback of the form

\[ \mu = \alpha(x) + \beta(x)u \]  \hspace{1cm} (33)

Nevertheless the input-output linearization may be achieved by using a dynamic feedback of the form [7, 24, 25, 26, 81]

\begin{align*}
\dot{\xi} &= f(x, \xi) + g(x, \xi)u \\
\mu &= \alpha(x, \xi) + \beta(x, \xi)u
\end{align*}

We follow the dynamic extension algorithm [7, pp.258-269] to derive \( f(\cdot, \cdot), g(\cdot, \cdot), \alpha(\cdot, \cdot), \) and \( \beta(\cdot, \cdot) \) if they exist at all. We divide the algorithm in three steps.

**Step 1:** Since the rank of the decoupling matrix \( S_1(x) \) in equation (32) is one, we first apply a static feedback to linearize and decouple one output from the others. For the mobile robot, there are two outputs \( y = [y_1 \ y_2]^T \). We choose to linearize \( y_1 \) and decouple it from \( y_2 \). Substituting the following static feedback into equation (32)

\[ \mu = \alpha^2(x) + \beta^2(x)u = \begin{bmatrix} c(\eta_1^2 - \eta_2^2) \tan \phi \\ 0 \end{bmatrix} + \begin{bmatrix} 1 \\ -1 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} \]  \hspace{1cm} (36)

the closed-loop input-output map is then

\[ \ddot{y} = \begin{bmatrix} c^2b(\eta_1^2 - \eta_2^2) \frac{1}{\cos \phi} \\ \tan \phi \end{bmatrix} + \begin{bmatrix} 1 \\ \tan \phi \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} \]  \hspace{1cm} (37)

It is clear that \( \ddot{y}_1 = u_1 \), that is, the first output \( y_1 \) is linearized and controlled only by \( u_1 \). Thus \( u_1 \) can be designed to achieve the performance requirements for \( y_1 \). On the other hand, \( y_2 \) is still nonlinear. Further, it is also driven by \( u_1 \).

**Step 2:** We substitute the static feedback (36) into equation (30) to obtain the new state equation

\[ \dot{x} = f^1(x) + g^1(x)\mu = f^1(x) + g^1(x) \left( \alpha^2(x) + \beta^2(x)u \right) \]  \hspace{1cm} (38)

\[ = \begin{bmatrix} c(\eta_1^2 - \eta_2^2) \tan \phi \\ S_1 \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{1}{\cos \phi} \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = f^2(x) + g^2(x)u \]  \hspace{1cm} (39)
We now differentiate the second output with respect to the new state equation \( \dot{x} = f^2(x) + g^2(x)u \), hoping that \( u_2 \) will appear in the derivative of \( y_2 \). In the following differentiation, \( u_1 \) is treated as a (time-varying) parameter.

\[
\begin{align*}
\dot{y}_2 &= cb(\eta_1 + \eta_2) \sin \phi \\
\ddot{y}_2 &= c^2b(\eta_1^2 - \eta_2^2) \frac{1}{\cos \phi} + \tan \phi u_1 \\
y_2^{(3)} &= c^3b(\eta_1^2 - \eta_2^2)(\eta_1 - \eta_2) \frac{\sin \phi}{\cos^2 \phi} + c(\eta_1 - \eta_2) \frac{u_1}{\cos^2 \phi} \\
&\quad + \frac{2c^2b\eta_1}{\cos \phi} \left( c(\eta_1^2 - \eta_2^2) \tan \phi + \frac{u_1}{cb \cos \phi} \right) \\
&\quad + \tan \phi \dot{u}_1 + \frac{2c^2b(\eta_1 + \eta_2)}{\cos \phi} u_2
\end{align*}
\]

It is seen that \( u_2 \) appears in the third-order derivative of \( y_2 \). We note that \( y_2^{(3)} \) has the following structure

\[
y_2^{(3)} = Q_1(x) + Q_2(x)u_1 + Q_3\dot{u}_1 + Q_4u_2
\]

where \( Q_i(x) \) can be easily identified.

**Step 3:** Noting equation (40), \( y_2 \) will be linearized if we apply the following feedback

\[
u_2 = Q_4^{-1}(x)(v_2 - Q_1(x) - Q_2(x)u_1 - Q_3(x)\dot{u}_1)
\]

with \( v \) being the reference input. However, this feedback depends on \( \dot{u}_1 \), which can be eliminated by introducing an integrator on the first input channel. Formally, we utilize the
following dynamic feedback

\[
\begin{align*}
\dot{x} &= \alpha^4(x, \xi) + \beta^4(x, \xi)v \\
u &= \alpha^3(x, \xi) + \beta^3(x, \xi)v
\end{align*}
\] (42) (43)

where \( \xi \) is one-dimensional and

\[
\begin{align*}
\alpha^4(x, \xi) &= 0 \\
\beta^4(x, \xi) &= [1 \ 0] \\
\alpha^3(x, \xi) &= \begin{bmatrix} \xi \\ -Q_4^{-1}(x)(Q_1(x) + Q_2\xi) \end{bmatrix} \\
\beta^3(x, \xi) &= \begin{bmatrix} 0 & 0 \\ -Q_4^{-1}(x)Q_3(x) & Q_4^{-1}(x) \end{bmatrix}
\end{align*}
\]

After applying the above dynamic feedback, we finally obtain two linearized and decoupled subsystems:

\[
\begin{align*}
y_1^{(3)} &= v_1 \\
y_2^{(3)} &= v_2
\end{align*}
\] (44) (45)

It is noted that the first subsystem is now of third order due to the introduction of the integrator on its input channel. This concludes the dynamic extension algorithm. The resulting extended system hence is decouplable with static state feedback.

The overall dynamic feedback control of the mobile robot is depicted in Figure 2. The first feedback (29) is to cancel the dynamic nonlinearity in order to simplify the subsequent discussion. The second feedback (36) is to linearize \( y_1 \) and also decouple it from \( y_2 \). The third feedback represented by equations (42) and (43) is to linearize \( y_2 \).

Finally we comment on the invertibility of the system [27, 28, 29]. Since the differential output rank \( \rho^* \) of this particular system is computed by [8]

\[
\rho^* = rank \left( \frac{\partial g^{(3)}}{\partial v} \right) = 2
\]

which is equal to the number of outputs, the system is right-invertible [27]. This guarantees the success of the above dynamic extension algorithm since a right-invertible system can always be locally decoupled via a dynamic state feedback [27]. Furthermore, since the differential output rank is equal to the number of inputs, the system is also left-invertible [28, 29, 30].

4.3 Look-Ahead Control

In Section 4.1, we showed that the center point \( P_o \) of the mobile robot cannot be controlled by using a static feedback. A dynamic feedback is necessary. In this section, we present
an alternative control method. The method is motivated from vehicle maneuvering. When operating a vehicle, a driver looks at a point or an area in front of the vehicle. We define a reference point \( P_r \) which is \( L \) distance (called look-ahead distance) from \( P_o \) (see Figure 1). We take the coordinates of \( P_r \) in the fixed coordinate frame as the output equation, \( \text{i.e.}, \)

\[
y = h(x) = \begin{bmatrix} x_1 + L \cos \phi \\ x_2 + L \sin \phi \end{bmatrix}
\] (46)

To verify if the system is input-output linearizable with this output equation, we compute the derivatives of \( y \).

\[
\dot{y} = \frac{\partial h}{\partial x} \dot{x} = \frac{\partial h}{\partial x} \begin{bmatrix} f^1(x) + g^1(x) \mu \\ \eta \end{bmatrix}
\]

\[
= \begin{bmatrix} cb \cos \phi - cL \sin \phi & cb \cos \phi + cL \sin \phi \\ cb \sin \phi + cL \cos \phi & cb \sin \phi - cL \cos \phi \end{bmatrix} \begin{bmatrix} \eta_1 \\ \eta_2 \end{bmatrix} = \Phi(x) \eta
\]

Since \( \dot{y} \) is not a function of the input \( \mu \), we differentiate it once more.

\[
\ddot{y} = \Phi(x) \ddot{\eta} + \dot{\Phi}(x) \eta = \Phi(x) \mu + \dot{\Phi}(x) \eta
\]

The input \( \mu \) shows up in the second order derivative of \( y \). Clearly, the decoupling matrix in this case is \( \Phi(x) \). Since the determinant of \( \Phi(x) \) is \((-2c^2bL)\), it is nonsingular as long as the look-ahead distance \( L \) is not zero. It follows that the system can be input-output linearized and decoupled [6]. The nonlinear feedback for achieving the input-output linearization and decoupling is

\[
\mu = \Phi^{-1}(x) \left( u - \dot{\Phi}(x) \eta \right)
\] (47)

Applying this nonlinear feedback, we obtain

\[
\ddot{y}_1 = u_1
\] (48)

\[
\ddot{y}_2 = u_2
\] (49)

Therefore, the mobile robot can be controlled so that the reference point \( P_r \) tracks a desired trajectory. The motion of the mobile robot itself, particularly the motion of the center point \( P_o \), is determined by the internal dynamics of the system which is the topic of the next section. We note that the look-ahead control method degenerates to the control of the center point if \( L = 0 \).

4.4 Internal Dynamics

The previous section addresses the input-output properties of the mobile robot with the look-ahead control output equation (46). In this section, we proceed to study the behavior of the internal dynamics including the zero dynamics of the system. For a general discussion of internal dynamics and zero dynamics, see Chapter 5 of [31].

We first construct a diffeomorphism by which the overall system can be represented in the norm form of nonlinear systems [31]. Since the relative degree of each output is two,
we may construct four components of the needed diffeomorphism from the two outputs and its Lie derivative, i.e., \( h_1(x) \), \( L_fh_1(x) \), \( h_2(x) \) and \( L_fh_2(x) \). Since the state variable \( x \) is six dimensional, we need two more components. We choose the two components to be \( \theta_1 \) and \( \theta_2 \). Thus the proposed diffeomorphic transformation would be

\[
z = T(x) = \begin{bmatrix}
z_1 \\
z_2 \\
z_3 \\
z_4 \\
z_5 \\
z_6 \\
\end{bmatrix} = \begin{bmatrix}
h_1(x) \\
L_fh_1(x) \\
h_2(x) \\
L_fh_2(x) \\
\theta_1 \\
\theta_2 \\
\end{bmatrix}
\]  

(50)

To verify that \( T(x) \) is indeed a diffeomorphism, we compute its Jacobian.

\[
\frac{\partial T}{\partial x} = \begin{bmatrix}
1 & 0 & -cL \sin \phi & cL \sin \phi & 0 & 0 \\
0 & 0 & * & * & cb \cos \phi - cL \sin \phi & cb \cos \phi + cL \sin \phi \\
0 & 1 & cL \cos \phi & -cL \cos \phi & 0 & 0 \\
0 & 0 & * & * & cb \sin \phi + cL \cos \phi & cb \sin \phi - cL \cos \phi \\
0 & 0 & 1 & 0 & 0 & 0 \\
0 & 0 & 0 & 1 & 0 & 0 \\
\end{bmatrix}
\]

It is easy to check that \( \frac{\partial T}{\partial x} \) has full rank\(^1\). Thus \( T(x) \) is a valid state space transformation. The inverse transformation \( T^{-1}(z) \) is given by

\[
\begin{bmatrix}
x_1 \\
x_2 \\
\theta_1 \\
\theta_2 \\
\eta_1 \\
\eta_2 \\
\end{bmatrix} = \Phi^{-1} \begin{bmatrix}
z_2 \\
z_4 \\
\end{bmatrix}
\]

We partition the state variable \( z \) into two blocks

\[
z^1 = \begin{bmatrix}
z_1 & z_2 & z_3 & z_4 \\
\end{bmatrix}^T \\
z^2 = \begin{bmatrix}
z_5 & z_6 \\
\end{bmatrix}^T
\]

After applying the feedback (47), the system of the mobile robot is represented in the following normal form.

\[
\begin{align*}
\dot{z}^1 &= Az^1 + Bu \\
\dot{z}^2 &= w(z^1, z^2) \\
y &= Cz^1
\end{align*}
\]  

\(^1\)The terms denoted by * do not affect the computation of the rank.
where

$$A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}, \quad B = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \quad C = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

$$w(z^1, z^2) = \Phi^{-1}(z) \begin{bmatrix} z_2 \\ z_4 \end{bmatrix} = -\frac{1}{2c^2bL} \begin{bmatrix} cb \sin \phi - cL \cos \phi & -cb \cos \phi - cL \sin \phi \\ -cb \sin \phi - cL \cos \phi & cb \cos \phi - cL \sin \phi \end{bmatrix} \begin{bmatrix} z_2 \\ z_4 \end{bmatrix}$$

It is understood that $\phi$ in the expression of $w(z^1, z^2)$ is a short-hand notation for $c(z_5 - z_6)$. Together, the linear state equation (51) and the linear output equation (53) are an equivalent representation of the input-output map (equations (48) and (49)). Equation (52) represents the unobservable internal dynamics of the mobile robot under the look-ahead control.

The zero dynamics of a control system is defined as the dynamics of the system when the outputs are identically zero (i.e., $y = 0, \dot{y} = 0, \ddot{y} = 0, \ldots$). If the outputs are identically zero, it implies that $z^1 = 0$, and the zero dynamics is

$$\dot{z}^2 = w(0, z^2) = 0 \quad (54)$$

Thus, $z^2$ remains constant while the outputs are identically zero. The zero dynamics is stable but not asymptotically stable. In other words, if the reference point $P_r$ remains still, so does the mobile robot (or more specifically, the wheels do not move).

We now look at the internal dynamics while the reference point is in motion. More specifically, we are interested in the internal motion of the mobile robot when it moves straight forward or backward. Let the mobile robot be initially headed in the positive $X_1$ direction. We assume that the reference point is controlled to move in the negative $X_1$ direction. The velocity of the reference point is then

$$\begin{bmatrix} \dot{y}_1 \\ \dot{y}_2 \end{bmatrix} = \begin{bmatrix} z_2 \\ z_4 \end{bmatrix} = \begin{bmatrix} -c(t) \\ 0 \end{bmatrix}$$

where $c(t) > 0$. Substituting this into the internal dynamics (52), we obtain

$$\begin{bmatrix} \dot{z}_5 \\ \dot{z}_6 \end{bmatrix} = \frac{c(t)}{2c^2bL} \begin{bmatrix} cb \sin \phi - cL \cos \phi \\ -cb \sin \phi - cL \cos \phi \end{bmatrix}$$

A solution of this internal dynamics is

$$z_5^* = -\frac{1}{r}t + c_1 \quad (55)$$

$$z_6^* = -\frac{1}{r}t + c_1 \quad (56)$$

where $c_1$ is a constant. That is, the two wheels rotate at exactly the same angular velocity and the mobile platform moves straight in the negative $X_1$ direction.
We now study the stability of the internal motion described by equations (55) and (56). We first change the state variable so that the stability of the internal motion in $z^2$ can be formulated as the stability of equilibrium points in $\zeta$.

$$\zeta_1 = z_5 - z_5^*$$
$$\zeta_2 = z_6 - z_6^*$$

We may express the internal dynamics in terms of $\zeta = \begin{bmatrix} \zeta_1 \\ \zeta_2 \end{bmatrix}^T$.

$$\dot{\zeta} = \begin{bmatrix} \dot{\zeta}_1 \\ \dot{\zeta}_2 \end{bmatrix} = \frac{\epsilon(t)}{2c^2bL} \begin{bmatrix} cb\sin(c\zeta_1 - c\zeta_2) - cL\cos(c\zeta_1 - c\zeta_2) + \frac{1}{r} \\ -cb\sin(c\zeta_1 - c\zeta_2) - cL\cos(c\zeta_1 - c\zeta_2) + \frac{1}{r} \end{bmatrix}$$

This system has an equilibrium subspace characterized by

$$E_\zeta = \{ \zeta \mid \zeta_1 = \zeta_2 \}$$

We may not draw any conclusion based on the linear approximation of the internal dynamics which has an eigenvalue at the origin. We will utilize the Liapunov method to establish the stability condition. Consider the following candidate for a Liapunov function

$$V(\zeta) = 1 - \cos(c\zeta_1 - c\zeta_2)$$

In a neighborhood of $E_\zeta$, $V(\zeta) = 0$ if $\zeta \in E_\zeta$, and $V(\zeta) > 0$ if $\zeta \notin E_\zeta$. Thus $V(\zeta)$ is positive definite with respect to $E_\zeta$, and may serve as a Liapunov function for testing the stability of $E_\zeta$. We compute the derivative of $V(\zeta)$ with respect to the time

$$\dot{V}(\zeta) = \frac{\partial V}{\partial \zeta} \dot{\zeta} = \frac{\epsilon(t)}{L} \sin^2(c\zeta_1 - c\zeta_2)$$

Since $\epsilon(t) > 0$, $\dot{V}(\zeta)$ is also positive definite with respect to $E_\zeta$. Therefore the equilibrium subspace $E_\zeta$ is not stable.

On the other hand, if the reference point is controlled to move in the positive $X_1$ direction, the velocity of the reference point is

$$\begin{bmatrix} \dot{y}_1 \\ \dot{y}_2 \end{bmatrix} = \begin{bmatrix} \dot{z}_2 \\ \dot{z}_4 \end{bmatrix} = \begin{bmatrix} \epsilon(t) \\ 0 \end{bmatrix}$$

where $\epsilon(t) > 0$. Using the same Liapunov function, we can similarly show that

$$\dot{V}(\zeta) = -\frac{\epsilon(t)}{L} \sin^2(c\zeta_1 - c\zeta_2)$$

along the forward internal motion. Therefore, the forward internal motion is stable. Intuitively, if the mobile platform is "pushed" at the reference point, the internal motion is not stable. If it is "pulled" or "dragged" at the reference point, the internal motion is stable.
5 Conclusion

We presented a number of interesting results on the feedback linearization of the dynamic system of a wheeled mobile robot. The first result reveals that the system is not input-state linearizable. The proof of this result is based on the fact a wheeled mobile robot is nonholonomically constrained. The other results are on the input-output linearization and decoupling of the system. Two types of outputs have been addressed. In the first type of output, the center point of the mobile robot on the wheel axis is intended to be controlled. It has been known that the point on the wheel axis cannot be controlled using a static feedback \[14, 15\]. We show that the center point can be controlled to track a trajectory by using a dynamic nonlinear feedback. The dynamic feedback for achieving the input-output linearization and decoupling has been developed through a three-step algorithm. The second output takes the coordinates of a reference point in front of the mobile robot. The input-output linearization of the system under this output is possible by simply using a static nonlinear feedback. The last part of the paper investigates the behavior of the internal dynamics of the system with the second type of output. We showed that the internal motion of the system is asymptotically stable when the reference point is controlled to move forward, but is unstable when it is controlled to move backward. These results, together with the results on controllability and feedback stabilization \[18, 20, 14, 15, 16\] provide a theoretical foundation for feedback control of wheeled mobile robots.

References


