The Development of Self-Balancing Controller for One-Wheeled Vehicles

Chung-Neng Huang
Graduate Institute of Mechatronic System Engineering, National University of Tainan, Taiwan, China
E-mail: kosono@mail.nutn.edu.tw
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Abstract
The purpose of this study is to develop a self-balancing controller (SBC) for one-wheeled vehicles (OWVs). The composition of the OWV system includes: a DSP motion card, a wheel motor, and its driver. In addition, a tilt and a gyro, for sensing the angle and angular velocity of the body slope, are used to realize self-balancing controls. OWV, a kind of unicycle robot, can be dealt with as a mobile-inverted-pendulum system for its instability. However, for its possible applications in mobile carriers or robots, it is worth being further developed. In this study, first, the OWV system model will be derived. Next, through the simulations based on the mathematical model, the analysis of system stability and controllability can be evaluated. Last, a concise and realizable method, through system pole-placement and linear quadratic regulator (LQR), will be proposed to design the SBC. The effectiveness, reliability, and feasibility of the proposal will be confirmed through simulation studies and experimenting on a physical OWV.

Keywords: Self-balancing Controller, One-wheeled Vehicle, Mobile-inverted-pendulum, Pole-placement, Linear Quadratic Regulator (LQR)

1. Introduction
In recent years, because of the surging consciousness of global pollution and energy-shortage crises, automobiles and motorcycles are no longer the best for transportation. In order to fit the daily required and improve above problems, exploring new energy or developing lighter and innovative mobile carriers are beginning to be known as new trends. The earliest two-wheeled balancing robot was published in 1987 by Prof. Yamafuji [1]. From then on, the concerning researches with this topic have been increasing [2-6] and have even been a commercialized product. For example, the Human Transporter, was developed by Segway Co., U.S.A., which is a very famous two-wheeled balancing vehicle [7,8]. In addition, NASA’s Robonaut, Segway platform puts robots in motion, is now aim in the military projects [9].

However, the balancing mechanism of a two-wheeled system is rather complicated. Whereas, for the problems, the two-wheeled synchronization and body balancing should be considered simultaneously, making a lot of sensors a requirement. It does not only complicate the system, but also increases the cost. For the reasons above, how to simplify the two-wheeled system has become one of the studying motives of OWV. By this motive, not only can it maintain the advantages of a two-wheeled system, but it can also simplify the system mechanism further, decreasing the cost.

Nowadays, although a lot of references engaging in the studies of two-wheeled balancing carriers can be found [2-6], the ones for OWV studies are still insufficient [10,11]. Trevor Blackwell [12], an American engineer, has proposed an OWV design on his personal website, and the concept OWV, EMBRIO, is published by BRP [13], a motorcycle manufacturer in Canada. Besides, [10] is the newest one based on fuzzy controls for balancing.

According to the above studies, one can find that the proposed OWV is rather new and is a worthy topic in studies on modern robots and carriers. However, most of the studies on robot-balancing subjects are often adopting the fuzzy control theory to handle those balancing control problems [4,14,15], but carrying out system modeling and analysis, and confirming the effectiveness and feasibility through experiment studies [14,15]. For fuzzy control, using linguistic information, can model complex systems without employing precise quantitative analyses, particularly for the controlled plant with in-
complete knowledge [4,14]. However, it has not been viewed as rigorous due to lack of formal synthesis technologies, which guarantee the basic requirements for control systems such as global stability [15]. That is, it is not only difficult to precisely know the mutual influence between each system state but also might be difficult and expensive to design the controller for its complexity highly depending on the number of fuzzy rules [16].

Consequently, finishing system modeling and analysis for proposed SBC is set as the first step of this study. Next, by linearizing the system model, pole-placement and state-feedback controls can be found and designed. Finally, through human-computer interaction, a DSP card featured with motion control is adopted to handle the real-time computation of digital signals.

However, from above studies found that only using the pole-placement and state-feedback controls, the feedback gains can not be adjusted easily. It is difficult to lead the system in a better stability. Here, LQR is used to improve the above problems. For comparison, both the numerical and experiment studies on SBC, basing on pole-placement and LQR controls, respectively, are made. The studies have proved that the proposed SBC is able to perform higher controllability and stability.

2. Modeling for OWV

The concept of the OWV prototype is simply constructed by a wheel motor and two sensors as a tilt and gyro. Here, the wheel motor is adopted as the driving wheel of the OWV that uses the feedback signals from sensors to balance the system body. Figure 1 shows the outward appearance of OWV prototype where the box on the upper side of wheel motor can be treated as the balancing weight of inverted-pendulum, and in which the motor driver, sensors are set. Figure 2 shows the free body diagram of OWV in Figure 1.

According to the free body diagram in Figure 2, the kinetic energy $E$ and potential energy $U$ can be expressed as follows, respectively.

$$E = \frac{1}{2} M_v \dot{X}_v^2 + \frac{1}{2} m(\dot{X}_v + \dot{\theta} l \cos \theta)^2$$

$$+ \frac{1}{2} m(-\dot{\theta} l \sin \theta)^2 + \frac{1}{2} J_\theta \dot{\theta}^2 + \frac{1}{2} J_\omega (\frac{\dot{X}_w}{R})^2$$

$$U = -mgl + mgl \cos \theta$$  \hspace{1cm} (1)  

Basing on Lagrange’s equation as well as the coefficient transformation for deriving mathematic model of OWV, the Lagrange equation can be written as:

$$\frac{d}{dt} \left( \frac{\partial L}{\partial \dot{X}_v} \right) - \frac{\partial L}{\partial X_v} = T_v \frac{R}{R}$$

where $L = E - U$

From (3), the accelerations of body movement and included angle between vertical axis and body can be found as follows.

$$\ddot{X}_v = \frac{T_v + ml \sin \theta \dot{\theta}^2 - ml l \cos \theta}{m l^2 + J_\omega}$$

$$\ddot{\theta} = \frac{(-ml \cos \theta(T_v + ml \sin \theta \dot{\theta}^2) - m l^2 \cos \theta)}{M_v + m + J_\omega} + \frac{m l^2 + J_\omega}{M_v + m + J_\omega}$$  \hspace{1cm} (4)  

Since (4) and (5) are nonlinear equations, for simplifying and modeling, let them be represented on the balanced point $\theta = 0$ and given $M_v = M_v + m + \frac{J_\omega}{R^2}$, then
the equations can be rewritten as
\[
\dot{X} = \begin{bmatrix} -m_l^2 l g \theta \\ M_r (m_l^2 + J_u) - m_l^2 l^2 \end{bmatrix} + \begin{bmatrix} (ml^2 + J_u)T_u \\ M_r (ml^2 + J_u) - m_l^2 l^2 \end{bmatrix}
\]

\[
\ddot{\theta} = \frac{ml^2 g}{M_r (ml^2 + J_u) - m_l^2 l^2} + \frac{-mlT_u}{M_r (ml^2 + J_u) - m_l^2 l^2}
\]  

By using (6) and (7), the block diagram for OWV control can be illustrated as shown in Figure 3. Where,
\[
J_s = \frac{ml^2 + J_u}{R [M_r (ml^2 + J_u) - ml^2 l^2]}, \quad K_s = \frac{-ml}{M_r (ml^2 + J_u) - ml^2 l^2}
\]

From Figure 3, it can be found that OWV is a single input and multiple output system (SIMO). Besides, OWV is similar to the inverted-pendulum vehicle, belonging to the non-minimum phase system \[3,17\].

3. Analysis and Design for State Variables

3.1. Performance of System Instability

Since a system can be represented in state space by following equations:
\[
\begin{align*}
\dot{X} &= AX + Bu \\
y &= CX + Du
\end{align*}
\]  

for \( t \geq t_0 \) and initial conditions with respect to the equilibrium point, \( X(t_0) = (0,0,0,0) \), where
\[
X: \text{state vector} \\
\dot{X}: \text{derivative of the state vector with respect to time} \\
y: \text{output vector} \\
u: \text{input or control vector}
\]

\( A \): system matrix  
\( B \): input matrix  
\( C \): output matrix  
\( D \): feed forward matrix

(8) is called the state equation, and the vector \( X \), the state vector, contains the state variables. It can be solved for the state variables. Besides, (9) is called the output equation. This equation is used to calculate any other state variables.

For the linear, time-invariant, second-order system as OWV, its system dynamics can be transformed and expressed by state equations; the state space of OWV can be taken on the following form by (6) and (7).

\[
\begin{bmatrix}
\dot{X}_w \\
\dot{Y}_w \\
\end{bmatrix} =
\begin{bmatrix}
0 & 1 & 0 & 0 \\
0 & 0 & -m_l^2 l^2 & 0 \\
0 & 0 & 0 & M_{mg} / 1 \\
0 & 0 & M_r (ml^2 + J_u) - ml^2 l^2 & 0
\end{bmatrix}
\begin{bmatrix}
X_w \\
Y_w \\
\end{bmatrix}
\]

(8')

\[
\begin{bmatrix}
\dot{X}_w \\
\dot{Y}_w \\
\end{bmatrix}
=
\begin{bmatrix}
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0
\end{bmatrix}
\begin{bmatrix}
X_w \\
Y_w \\
\end{bmatrix}
\]

(9')

In order to confirm the effectiveness of the derived model, through the simulation by mechanical design software Solidworks, or CATIA, system analysis and estimation of system coefficients are done.

Table 1 and Figure 4 show the estimated system coefficients of OWV by Solidworks' analysis.

Based on above, system coefficients and system stability, the analysis finds that for OWV it is not a stable system. Initially, it would fall down by an outside disturbance. This phenomenon can be confirmed by the root locus as shown in Figure 4. Where, for one of the poles locating in the right of s-plane, it can be judged that it is an unstable system. Besides, this result also can be confirmed by finding out the eigenvalues of system matrix in (8') as \( s = 0,0.4116, -4.4116 \).

Table 1. System coefficients of OWV.

| \( m \) | 4.31 kg |
| \( J_u \) | 0.1984 kg-m^2 |
| \( l \) | 0.1 m |
| \( R \) | 0.07 m |
| \( M_w \) | 5.1819 kg |
| \( J_w \) | 0.0123 kg-m^2 |
3.2. Analysis for the Controllability of OWV

Accepting to discuss the system stability in state-space analysis, the controllability of system is also an important index. That is, before designing state-feedback controllers, the analysis for controllability should be done in advance to make sure the system is able to achieve the desired responses through state-feedback control and observing.

Pole-placement is a viable design technique only for systems that are controllable. In order to be able to determine controllability or, alternatively, to design state feedback for a plant under any representation or choice of state variables, a matrix can be derived that must have a particular property if all state variables are to be controlled by the plant input, \( u \). For an \( n \)-th order plant whose state equation is (8), which is completely controllable if the matrix

\[
Q_c = \begin{bmatrix} B & AB & A^2B & \ldots & A^{n-1}B \end{bmatrix}
\]

is of rank \( n \), where \( Q_c \) is called the controllability matrix.

Now, by substituting the coefficients of Table 1 to the system and inputting matrices in (8\(^\prime\)) and using (10), then \( rank(Q) = 4 \) can be found. That is, OWV is completely controllable on the balancing point. Its closed-loop system’s poles can be placed at desired locations on the s-plane through using the state feedback \( u(t) = -Kx(t) \).

3.3 State-Feedback Control

In order to improve the system instability, the state feedback as well as the pole-placement design is adopted to make the system stable and balanced. By which, the block diagram of the system state feedback is shown as Figure 5. Where, the feedback is given as \( u = -Kx(t) \).

Thus, the transfer function of the closed-loop system

\[
det \left[ sI - (A-BK) \right] \neq 0
\]

is

\[
\begin{bmatrix}
0 & -1 & 0 & 0 \\
1.2717k_1 & 1.2717k_2 & 0.6987+1.2717k_3 & 1.2717k_4 \\
0 & 0 & 0 & -1 \\
-2.2696k_1 & -2.2696k_2 & -19.4625-2.2696k_3 & -2.2696k_4
\end{bmatrix}
\]

Here, assume an example for placing the system’s poles at \( s= 0, -4.5, -4 \), since the system transfer function should satisfy the condition \( s^2(s+4.5)(s+4) = 0 \), the feedback gains of system state can be found as \( K = [k_1, k_2, k_3, k_4] = [0, 0, -16.5062, -3.7452] \). In order to confirm the effectiveness of pole-placement, a simulation study on OWV by using the results (feedback gains) of the above example is done. Figure 6 shows that through system feedback control, all system’s poles of OWV have located in the left of s-plane. In addition, the feedback system, being disturbed by a continuous impulse...
3.4 LQR Control

In above section, it shows that pole-placement and state-feedback controls used for SBC are available and realizable. However, if the unstable pole could only be placed in the left of s-plane by using assignments (experienced studies), then adjusting the feedback gains to a better stability for OWV would be difficult.

In order to improve above problem, LQR, the continuous time infinite horizon linear quadratic regular with control constraints, is adopted to design SBC. Deterministic LQR theory was first introduced by Kalman, and has been playing a central role in modern control theory as well as various engineering practices; the reader can be referred to the well-known book [18], as the following problem:

Minimize quadratic performance index

$$ J = \int_{0}^{\infty} \left[ X^T(t)QX(t) + u^T(t)Ru(t) \right] dt $$

subject to linear dynamics (8), (9) and a constraint on the input $u(t) \in U$, for all $t \in [0, \infty)$.

Here, the state vector $X: [0, \infty) \rightarrow IR^n$ is locally absolutely continuous, and the minimization is carried out over all locally integrable controls $u: [0, \infty) \rightarrow IR^n$. Throughout the note, the standing assumptions are

1) $Q$ and $R$, the weighting matrices of $X(t)$ and $u(t)$, are symmetric and with positive semi-definite and positive definite, respectively.

2) The pair $(A, B)$ is controllable. The pair $(A, C)$ is observable.

3) The set $U$ is closed, convex, and $0 \in \text{int}U$.

For the unconstrained problem (8), (9), and (12), the optimal solution of the problem is the following state-feedback control:

$$ u(t) = -K(t)X(t) $$

where, $K(t) = R^{-1}B^TP(t)$, and $P$, the unique symmetric and positive definite matrix, is the solution of the Riccati differential equation as

$$ \dot{P}(t) = -P(t)A + A^TP(t) + P(t)BR^{-1}B^TP(t) - Q $$

Since there is only stable state considered for OWV system, that is $\dot{P}(t) = 0$, (14) can be rewritten as

$$ PA + A^TP - PBR^{-1}B^TP = -Q $$

By substituting (15) into (13) with system coefficients, $\hat{K}(t)$, the optimal feedback gains, can be found as $K = [k_1 \ k_2 \ k_3 \ k_4]$

$$ = [-3.1623 \ -3.9851 \ -32.5896 \ -7.5687] $$, by giving matrices $R = [1]$, and $Q = \text{diag}([10 \ 1 \ 1 \ 1])$. The $R$ and $Q$ are provided to have constraint of control action satisfied.

Figure 8 shows the comparison of stability responses by using pole-placement and LQR, respectively. Where, SBC, basing on LQR control, let OWV have a better system stability is confirmed.

4. Physical Demonstration of OWV

For further confirming above numerical studies and the feasibility of OWV, the testing setup is accomplished and pictured in Figure 9.

A motor driver and sensors (SSY0090 tilt and CRS03-02 gyro) are set in the box and a 24-V 250-W wheel motor (SA176-6) [19] is used as the driving wheel. All the algorithms for pole-placement, feedback-gain calculation, and LQR etc. are developed with C language and implemented in an ITRI PMC32-6000 DSP board with a clock frequency of 40MHz (Figure 10).

Figure 11 illustrates that when the instability of body is detected by those sensors, the angular signals, through A/D transformation, transfer to DSP for real-time digital
Figure 8. Comparison of stability responses.

Figure 9. OWV setup.

Figure 10. DSP board.

Figure 11. Control block diagram of OWV.

terrupting algorithm is executed by every 1 ms.

Figure 12 shows the voltage outputs of motor control, angular signals of OWV’s body from gyro and tilt while SBC bases on pole-placement and LQR controls, respectively. Here, by examining angular velocity (Figure 12(b)), since it, the derivatives of body tilt angle can be taken as the estimation of next-state body falls, the
(a) voltage outputs

(b) angular velocity

(c) tilt angle

Figure 12. Control signals.

voltage signal (Figure 12(a)) is approximately inverse to angular velocity for falling-preventive control. Besides, in Figure 12(c), even through some tiny oscillations within ±2 degree occur when the body of OWV is starting to swing up by the wheel-motor driving a horizontal force to move OWV back and forth, after 3.75 seconds it can return to the desired balancing state. Consequently, it proves that SBC is able to be realized by both above controls. Moreover, through the comparisons of the signals in Figure 12, finds that the magnitude of the state signals by LQR control is smaller than those by pole-placement control. It is not only shows LQR with better system stability performance but also corresponding to the objective of LQR control in (12).

5. Conclusions

This study has proposed a SBC, based on concise construction and control theories for OWV. Only based on state-space modeling and real-time sensing, the SBC is attractive for its conciseness and feasibility. Studied results show that the real-time SBC is not only can be realized but also can let OWV with optimal stability by LQR control.

By further developing OWV technologies, such as SBC etc., the benefits not only can serve the handling problems, existing in present electrical mobile-robots, welfare carriers, or two-wheel entertainment vehicles as large steering radius (angle), differential gear, or synchronization control etc., but also with advantages on cost, setup, light quantity, and saving energy, etc.

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Appendix A

Nomenclature

- \( m \): pendulum payload mass
- \( J_m \): pendulum moment of inertia
- \( l \): Pendulum length
- \( R \): wheel radius
- \( M_w \): single wheel mass
- \( J_w \): single wheel moment of inertia
- \( X_w \): center cart trajectory
- \( \theta \): pendulum tilt angle
- \( T_w \): driving torque

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