

Taking Traction Control to Task

High-Adhesion-Point Tracking Based on a Disturbance Observer in Railway Vehicles



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Railway vehicles have been based on a wheel-rail system since the system's origination. Therefore, the adhesion-coupling behavior between wheel and rail is the fundamental element of railway traction [1]. However, low vehicle wheel-rail adhesion caused by high humidity, rain, snow, oil, or decomposing leaves is a common problem that can cause damage to some elements of the traction system and decrease traction performance [2]. It may even

give rise to safety problems and reduce comfort, not to mention delays and their corresponding economic impact. Good traction control to ensure the system is working at high adhesion point is therefore critical and mandatory for the traction control units (TCUs) [2]–[6].

In modern railway vehicles, a typical adhesion control system calculates the speeds of traction motors and trailer wheels, and the reference speed is set to a minimum speed among the components in the powering mode or the maximum speed in the braking mode. The TCU calculates the speed difference between

the motored truck and the reference speed, and the tractive effort of the traction motors is reduced as soon as the speed difference exceeds a certain value, until the slip or slide ceases. With this method, we can achieve a high tractive force.

The traction control of adhesion force is based on three fundamental pillars: speed measurement, traction motor control, and adhesion control. However, the number of pulses per rotation of the speed sensor is normally much smaller than that widely used in other industrial applications [7], so the level of accuracy of speed measurement using a basic pulse count is

Digital Object Identifier 10.1109/MIE.2016.2644699
Date of publication: 21 March 2017

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not usually sufficient. To improve the accuracy, a field-programmable gate array with a digital filter, a high-speed clock, and a parallel algorithm can be used in speed measurement.

Good motor control is essential to realizing good vehicle dynamic performance [8]; when slippage or sliding occurs, the tractive force is reduced by the motor controller automatically, the more the better. However, the acceleration of the vehicle will be reduced correspondingly, resulting in exceeding the running time constraints. Thus, it is the responsibility of the adhesion controller to maintain good tractive performance within the acceptable slip range.

To achieve this, we have to learn how the tractive force is transmitted in the traction system and what will happen on the wheel-rail contact surface. Innovative adhesion models considering the wheel-rail contact surface, which increase the control accuracy by reproducing degraded adhesion conditions in vehicle dynamics and railway systems, are described in [9]–[11]. However, such goals are not easy to achieve due to nonlinearities and uncertainties associated with wheel-rail surface conditions, train speed, and operating conditions [2].

In this article, a comparative study of adhesion control methods in different railways is made, including rubber-tire and steel-wheel types used in urban and mainline railways. The steel-wheel type of urban railway is chosen as a study case in this research, as the adhesion control methods of all these types are similar. The state-of-the-art of adhesion research is then discussed.

A control strategy using a first-order disturbance observer is proposed to realize better adhesion control performance on the basis of wheel-rail adhesion characteristics and a simplified equivalent single-axle model. An additional dynamic torque tuning function is used to adjust the torque slightly to maintain the robust performance of adhesion control. Finally, the torque reference given by the general vehicle logic controller is combined with the proposed adhesion control and used to control the traction motors on the basis of a field-oriented control (FOC) strategy. Simulations and experiments were carried out to validate the control strategy. The proposed adhesion method can also be extended to other types of railway vehicles, such as mainline railways or rubber-tire trains.

Adhesion Control Application and Research

We first compare the adhesion control application differences between rubber-tire and steel-wheel trains. Rubber-tire trains have advantages of greater adhesion and better acceleration in operation [12]; however, modern steel-wheel trains using distributed traction with a high proportion of powered axles have narrowed the adhesion performance gap compared to rubber-tire trains. Additionally, rubber-tire trains have similar traction logic and motor controls compared to those used in steel-wheel controls, which means that even though the adhesion characteristics are quite different, the adhesion control algorithm is similar [13]–[15].

Then we discuss the adhesion control differences between urban railways and mainline railways. In urban railways, the control is fulfilled through cooperation among the central control unit (CCU), the brake control unit (BCU), and the TCU. The general adhesion control diagram is shown in Figure 1. Generally, the TCU receives traction orders from the cab handle directly or from the CCU through the vehicle communication bus and then applies adhesion control to motor axles. The BCU regulates its mechanical braking force when sliding happens during the braking state. Of course, the BCU will always compensate the total braking force according to the actual electric braking force feedback from the TCU. The difference in mainline railways, especially electric multiple units using distributed traction, is that the BCU completes the braking slide algorithm for the motor axles instead and then sends the final reference of braking force to the TCU, which, by this point, just has to perform traction slip control. Another major difference in the operating pattern between urban railways and mainline railways is that urban railways normally require much more frequent starts and stops as the distances between stations are much shorter and the acceleration is much higher.

Therefore, the wheel slip/slide normally imposes a greater problem when compared with mainline services. The

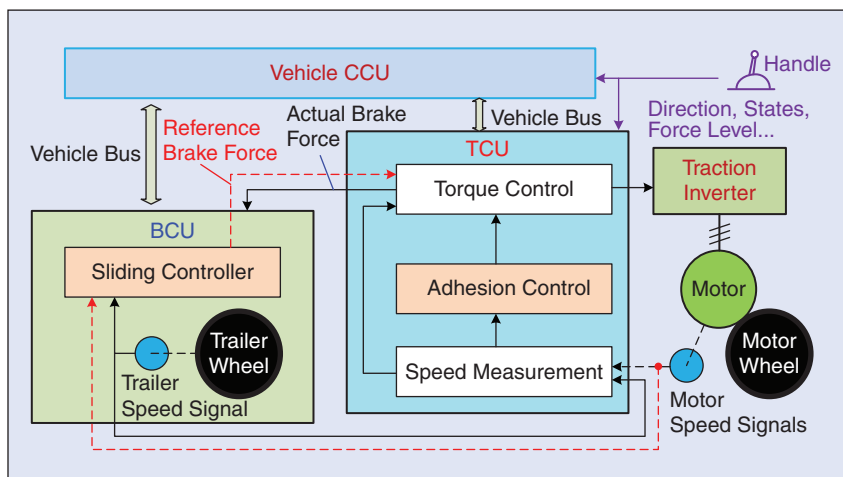


FIGURE 1 – A general diagram of adhesion control for traction systems.

adhesion control theory for urban and mainline railways remains quite similar, however, and it is very easy to transplant the algorithm from one to the other by changing the cooperation logic between the BCU and the TCU. Without loss of generality, this article mainly discusses adhesion control strategies of steel-wheel trains, especially those used in urban railways. Figure 2 summarizes different adhesion control approaches, each of which can be used to achieve a relatively ideal adhesion coefficient. The frequently proposed adhesion control schemes in the literature are based on conventional configurations with reference speed or on acceleration without a reference speed [1], [16]–[18], which are relatively simple and can achieve relatively high adhesion efficiency. According to [16], the actual adhesion coefficient can be improved, and the acceleration time can be shortened by 8% compared with an early control method. In [7], the derivative of motor current was utilized in place of the speed information for wheel slip detection of a multiple motor drive system, but additional current sensors are needed. The conventional control strategies given previously can be simply realized, and they have fast responses and low risk. When slippage or sliding occurs, however, the torque needs to be reduced quickly and then increased slowly to prevent slipping from occurring again. The adhesion point is often far away from the peak value, and adhesion efficiency is relatively low.

To overcome these shortcomings, many scholars have proposed other adhesion control strategies [19]–[24]. For example, the slip velocity method in [20] is relatively simple and intuitive, but the precise measurement of slip velocity is not an easy task. The ordinary adhesion derivative method in [21] directly calculates the differential adhesion coefficient as a function of time, which causes noise to be amplified and increases the vulnerability of the control systems.

There are also some adhesion strategies based on intelligent control theories. The fuzzy logic–based control strategy described in [25] outputs the optimal slip velocity reference according to preestablished fuzzy rules and

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thus regulates the motor torque via a proportional integral (PI) controller. In [26], a wavelet denoising method is used to identify slip tendency, and the torque command is tuned by a cloud model. However, these intelligent rules require large amounts of data and computation, which makes robust software programming difficult and limits their application in rail transit, where high reliability, security, and real-time performance are needed. In contrast, the adhesion control strategy based on a model control law is simple and easy to apply because there is no need to calculate the slip velocity and the system is always working near the peak adhesion point by real-time adjustments of the motor torque; the only consideration is that the derivative of the current wheel-rail peak adhesion coefficient must be zero [27], [28].

Based on the model control, [29] proposes an antislip control system based on the ordinary disturbance observer, which is called a zero-order disturbance observer. It is simple, easy to realize, and has a good antislip response. However, the inputs and variables of a zero-order disturbance observer are susceptible to

interference by electromagnetic noise. Therefore, in the case of large variations of the adhesion force coefficient, it is sometimes difficult to keep the adhesion control reference near to the maximum adhesion force. To overcome this problem, [30] proposes an adhesion control method with a full-dimension state observer, which introduces a feedback correction path in the adjustment of the state variables. However, the zero-order disturbance observer and full-dimension state observer cannot directly obtain the load torque derivative values, which require additional derivative computing that will amplify noise and make the control system vulnerable to interference. In this study, we will use a first-order disturbance observer combined with a dynamic torque tuning function to compensate for the aforementioned shortcomings.

Vehicle Adhesion Characteristics and Dynamic Model

Wheel-Rail Adhesion Characteristics

The adhesion behavior of steel-wheel railways is determined by the forces arising between two surfaces in contact,

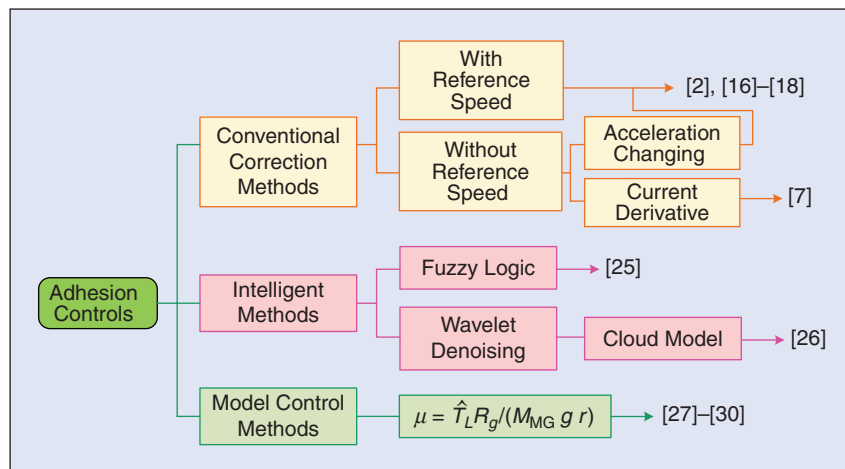


FIGURE 2 – A summary of the different adhesion control approaches.

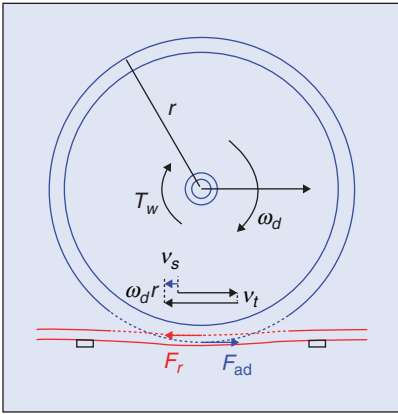


FIGURE 3 – An illustration of the creepage phenomenon.

where a phenomenon called *creepage* will occur. This phenomenon can be defined as a wheel-rail micro-elastic slide through relative deformation as shown in Figure 3 [1]: when a tractive

force (torque T_w) is applied on a railway wheel, a small difference arises between the forward velocity v_t and wheel circumference line velocity v_d (circumferential angular velocity ω_d). The relative difference is the creepage. It is resisted by the friction force F_{ad} on the wheel with radius r . The result slip velocity v_s can be defined as

$$\begin{cases} v_s = v_d - v_t \\ v_d = \omega_d r. \end{cases} \quad (1)$$

To design the adhesion control system, the adhesion characteristics must be considered as exactly as possible by using the adhesion coefficient μ , which is usually defined as the ratio between the wheel-rail tangential force (or adhesion force) F_{ad} and the normal force F_G on the wheel. The definition

shows that, when the load is constant, the available F_{ad} is proportional to μ

$$\mu = \frac{F_{ad}}{F_G}. \quad (2)$$

The fundamental relationship between the average value of μ and wheel slip under different rail conditions is shown in Figure 4 [17], [32]. It can be seen that a dry, clean wheel-rail surface condition will have a high μ , but a wet surface under rainy, snowy, foggy, or frosty weather conditions will have a much lower μ ; μ will be further reduced when the surface is oily. However, the different characteristic curves have their own peak adhesive point at a slip velocity v_{spot} . The adhesion system will be stable when the slip velocity is less than v_{spot} , and μ increases nearly linearly with the slip velocity. However, when the slip velocity is greater than v_{spot} , μ decreases rapidly when the slip velocity increases, and the system will be unstable.

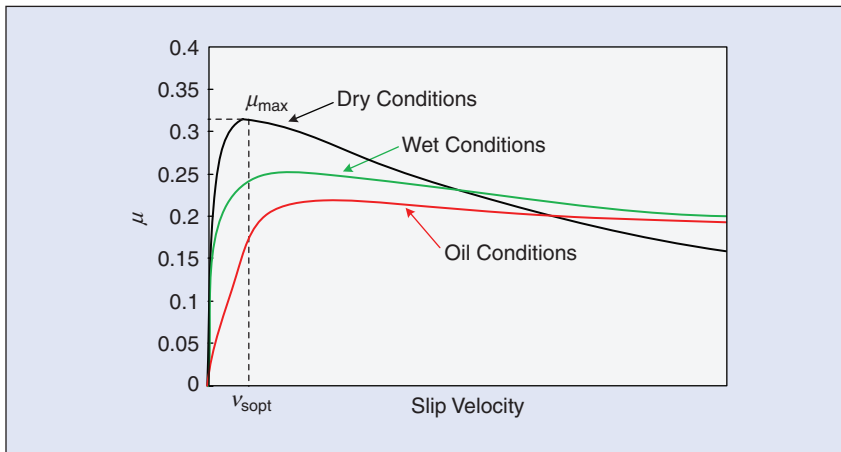


FIGURE 4 – Adhesion characteristic curves under different rail surface conditions.

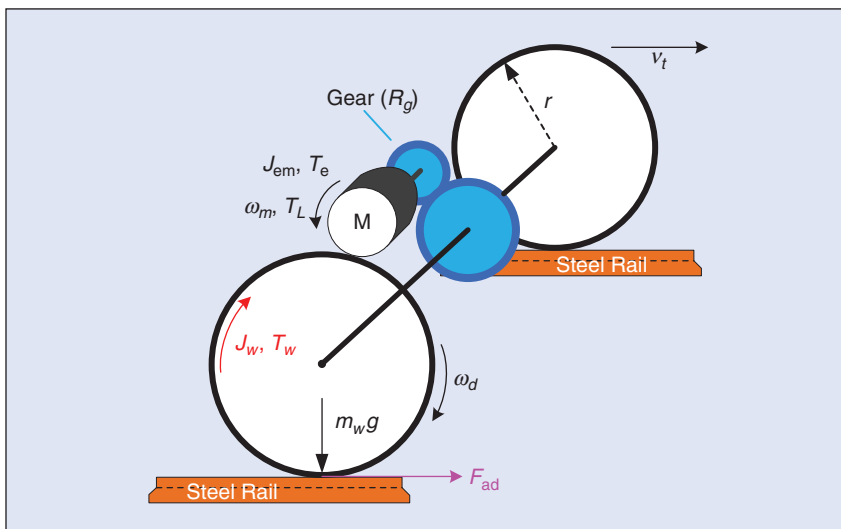


FIGURE 5 – The proposed single-axle model for one traction unit.

Vehicle Dynamic Model

To study the adhesion characteristics of the wheel-rail surface from the electric control perspective, it is necessary to establish a model that can reflect the relationship between electric control and adhesion characteristics. An accurate physical model considering all the dynamics of the entire vehicle, such as longitudinal, lateral, and vertical dynamics, would therefore be ideal. It is very difficult to capture all the dynamics, however, because of the complex and nonlinear behavior of the adhesion coefficient, as well as the external unknown contaminants that are present between the contact surfaces [9]. Additionally, most traction systems consist of one or several traction units, where the wheels in the same bogie or even different bogies are driven by motors connected in parallel and powered by the same traction converter. The differences of motor parameters, wheel diameters, weight on each axle, etc. will make an accurate model of a traction unit much more complicated.

We have focused primarily on the adhesion control itself, so many other factors will not be discussed in this article. A simplified one-dimensional single-axle model as shown in Figure 5 is proposed

in the analysis, which can be used to synthetically reflect the essential characteristics of one traction unit [29]. The single-axle model makes the following assumptions for its traction unit:

- The parallel motors have exactly the same characteristics and time-variant parameters. The diameter differences of the wheels are small enough to be neglected [33], [34].
- The mass is distributed equally on each motor axle. The transmission losses in the gears are also neglected.
- At every time interval, the rail condition for each wheel is the same, and the elastic deformation of the wheel-rail surfaces is quite similar.

In the model shown in Figure 5, the motor is powered by the traction converter, and the tractive force is transmitted from the motor to the wheels through a gearbox. The equivalent wheel and motor rotation equations of the model can be written as

$$\begin{cases} F_{ad} = \mu m_w g \\ J_w \frac{d\omega_d}{dt} = T_w - F_{ad} r \\ T_w = T_e R_g \\ T_L = F_{ad} r / R_g \\ J_{em} \frac{d\omega_m}{dt} = T_e - T_L \end{cases} \quad (3)$$

where g is the gravitational acceleration constant; R_g is the gear transmission ratio; ω_m , T_e , and T_L are the mechanical angular speed, total electromagnetic torque, and equivalent load torque of the traction motor, respectively; J_{em} and J_w are the total equivalent motor torque inertia and total equivalent wheel moment inertia, respectively; m_w is the total mass put on the motor axles of the traction unit; $m_w = (m_M + m_L) / N_c$; m_M and m_T are the dead mass of a powered carriage and a trailer carriage, respectively; m_L is the average load mass of each carriage; and N_c is the number of traction units for each powered carriage.

The motion equation of the whole train can be written as

$$\sum_{i=1}^{(N_c \times N_{mc})} F_{adi} - F_R = [N_{mc} (m_M + m_L) + N_{tc} (m_T + m_L)] \frac{dv_i}{dt}, \quad (4)$$

where F_{adi} represents the total adhesive force provided by number i traction units; F_R is the total resistance, including running resistance and slope resistance; and N_{mc} and N_{tc} denote the numbers of motored carriages and trailer carriages, respectively.

Adhesion Control Based on High Adhesion Utilization

The slope of the adhesion characteristic curve can be deduced from Figure 4 as

$$\frac{d\mu}{dv_s} = \frac{d\mu/dt}{dv_s/dt} = \frac{\dot{\mu}}{\dot{v}_s}. \quad (5)$$

The slope will be either greater than zero when the system's operating point is on the left side of the adhesion characteristic curve or less than zero when it is on the right side. Otherwise, it approximately equals zero in the vicinity of the peak point. Meanwhile, the slip velocity of the vehicle is always changing due to the ever-changing powering or braking conditions, i.e., its derivative value is not zero. Thus, a search for the adhesion coefficient's peak point can be directly performed using the condition that $\dot{\mu} = 0$.

Among uncertainties of adhesion control, the adhesion coefficient is one of the most difficult variables to predict or measure directly, especially under changing wheel-rail surface conditions when the train is running. Beam-and-bristle models were developed to investigate adhesion coefficient characteristics and verify them experimentally [35]. It is difficult to build such complicated test rigs, however, and there are also various inherent differences among simulations, labs, and reality. This study will use other parameters that can be detected by the TCU itself to indirectly calculate or be equivalent to μ .

Design of a First-Order Disturbance Observer

According to (3), the adhesion coefficient μ is proportional to the equivalent motor load torque, T_L . However, the traction control system cannot detect the load torque of the motor in real time, i.e., during the actual running state. Therefore, a state observer needs to be used by reconstructing the current equivalent motor load torque according to input variables or variables that can be measured directly [26].

Traction motors drive rail vehicles by transmitting torque to the heavy wheels through gears, and the mechanical inertia of the whole vehicle is enormous. Hence, the equivalent load torque of the motors can be considered to be changing at a constant speed during a short calculation period, i.e., \dot{T}_L is constant. Using T_L and its derivative as state variables, a first-order disturbance observer of the linear state space is described in (6) in the box at the bottom of the page.

In (6), A, B, and C are coefficient matrices and x is the state matrix of the model. The equivalent motor electromagnetic torque T_e is the input variable, and $y = \omega_m$ is the output variable, which can be directly measured.

Define $\tilde{x}_1 = [\omega_m]$, $\tilde{x}_2 = [\dot{T}_L \ T_L]^T$; then, (6) can be rewritten as

$$\begin{cases} \dot{\tilde{x}}_1 \\ \dot{\tilde{x}}_2 \end{cases} = \begin{bmatrix} \tilde{A}_{11} & \tilde{A}_{12} \\ \tilde{A}_{21} & \tilde{A}_{22} \end{bmatrix} \begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \end{bmatrix} + \begin{bmatrix} \tilde{B}_1 \\ \tilde{B}_2 \end{bmatrix} T_e, \quad (7)$$

where the coefficient matrices with \sim symbols are derived from A or B in (6) separated by dotted lines.

If a state space for variable \tilde{x}_2 is defined as

$$\begin{cases} \dot{\tilde{x}}_2 = \tilde{A}_{22} \tilde{x}_2 + \tilde{u} \\ \tilde{y} = \tilde{A}_{12} \tilde{x}_2, \end{cases} \quad (8)$$

then combining (7) and (8) yields

$$\begin{cases} \tilde{u} = \tilde{A}_{21} \tilde{x}_1 + \tilde{B}_2 T_e \\ \tilde{y} = \tilde{x}_1 - \tilde{A}_{11} \tilde{x}_1 - \tilde{B}_1 T_e. \end{cases} \quad (9)$$

$$\begin{cases} \begin{bmatrix} \dot{\omega}_m \\ \dot{T}_L \\ \dot{T}_L \end{bmatrix} = \begin{bmatrix} 0 & 0 & -1/J_{em} \\ 0 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \omega_m \\ T_L \\ T_L \end{bmatrix} + \begin{bmatrix} 1 + J_{em} \\ 0 \\ 0 \end{bmatrix} T_e \\ y = \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} \omega_m \\ T_L \\ T_L \end{bmatrix} \end{cases} \quad (6)$$

TABLE 1 – VEHICLE AND CONTROL PARAMETERS.

SYMBOL	VALUE	SYMBOL	VALUE
R_g	6.37	m_T	34 t
r	805 mm	m_M	38 t
J_{em}	5 kg.m ²	m_L	25 t
μ_{cal}	0.16	T_{ad-C1}	0.15 T_{ref}
l_1, l_2	-150	T_{ad-C2}	0.25 T_{ref}

From (6) and (7), it is obvious that the eigenvalues of the matrix \tilde{A}_{22} are $\lambda_1 = \lambda_2 = 0$; thus, the linear system given in (8) is unstable. Therefore, use L as the feedback gain matrix to construct a closed-loop state observer as

$$\begin{cases} \dot{\hat{x}}_2 = \tilde{A}_{22}\hat{x}_2 + \tilde{u} + L(\tilde{y} - \tilde{A}_{12}\hat{x}_2) \\ = (\tilde{A}_{22} - L\tilde{A}_{12})\hat{x}_2 + L\tilde{u} + \tilde{u} \\ L = [l_1 \quad l_2]^T, \end{cases} \quad (10)$$

where $\hat{\cdot}$ indicates the observed quantities.

Considering the tradeoff between different performance objectives, such as dynamic response and noise sensitivity, a reasonable choice of the values of the feedback gain matrix coefficients, l_1, l_2 , can be selected to ensure the observer's stability, where the real parts of the eigenvalues for the system matrix $\tilde{A}_{22} - L\tilde{A}_{12}$ are negative.

However, (9) shows that the calculation of \tilde{y} includes a derivative operation of $y = \dot{x}_1 = \omega_m$, which will amplify high-frequency noise in \tilde{y} . It therefore leads to a direct impact on the accuracy of the observation results and the stability of the control system. To avoid this, define an intermediate state variable

$$\begin{aligned} z &= [z_1 \quad z_2]^T \text{ and} \\ z &= \hat{x}_2 - Ly. \end{aligned} \quad (11)$$

Taking the derivative of both sides of (11), and substituting (9) and (10) yields

$$\begin{aligned} \dot{z} &= \dot{\hat{x}}_2 - L\dot{x}_1 \\ &= (\tilde{A}_{22} - L\tilde{A}_{12})\hat{x}_2 + L\tilde{y} \\ &\quad + \tilde{u} - L(\tilde{y} + \tilde{A}_{11}y + \tilde{B}_1 T_e) \\ &= (\tilde{A}_{22} - L\tilde{A}_{12})(z + Ly) \\ &\quad + (\tilde{A}_{21} - L\tilde{A}_{11})y + (\tilde{B}_2 - L\tilde{B}_1)T_e \\ &= (\tilde{A}_{22} - L\tilde{A}_{12})z + [(\tilde{A}_{21} - L\tilde{A}_{11}) \\ &\quad + (\tilde{A}_{22} - L\tilde{A}_{12})L]y + (\tilde{B}_2 - L\tilde{B}_1)T_e. \end{aligned} \quad (12)$$

Substituting the coefficient matrices in (6) and (10) into (12), the final model of the first-order disturbance observer can be derived as shown in (13) at the bottom of the page.

As shown in (3), adhesion coefficient μ can be directly calculated using the observed value \hat{T}_L by $\mu = \hat{T}_L R_g / (m_w g r)$, i.e., μ can be derived from (13). To assess the observer's performance with different values of the feedback gain matrix coefficients, simulations of changing wheel-rail surface conditions are carried out with different sets of l_1, l_2 and other parameters given in Table 1.

The results are shown in Figure 6, in which the load torque and its derivative values have been changed to μ and its derivative, $\dot{\mu}$. Figure 6(a) with $l_1 = l_2 = -30$ shows that the observed values have small steady-state errors but the dynamic response is slow, which makes the adhesion response not fast enough when the wheel-rail surface condition changes suddenly. But in Figure 6(b) with $l_1 = l_2 = -150$, the dynamic response becomes faster, which can meet the requirement of dynamic response.

By comparing Figure 6(a) and (b), we see that small absolute values of the feedback gain matrix coefficients will cause large steady-state error and slow response. Larger values will also cause the observer to be more sensitive to interfering signals. The appropriate values are those that result in small steady-state errors,

$$\begin{cases} \begin{bmatrix} \dot{z}_1 \\ \dot{z}_2 \end{bmatrix} = \begin{bmatrix} 0 & l_1/J_{em} \\ 1 & l_2/J_{em} \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} + \begin{bmatrix} l_1 l_2 / J_{em} \\ l_1 + l_2^2 / J_{em} \end{bmatrix} \omega_m + \begin{bmatrix} -l_1 / J_{em} \\ -l_2 / J_{em} \end{bmatrix} T_e \\ \begin{bmatrix} \hat{T}_L \\ \dot{\hat{T}}_L \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} + \begin{bmatrix} l_1 \\ l_2 \end{bmatrix} \omega_m. \end{cases} \quad (13)$$

which help to improve the dynamic response of observation and reduce noise sensitivity. Accordingly, the values of the parameter l_1 and l_2 are chosen to be -150.

Dynamic Torque Tuning Function

Using the output of the first-order disturbance observer, combined with the knowledge that the derivative of the adhesion coefficient of the peak point is zero, a PI regulator is normally used. Under different rail conditions, however, unsuitable PI parameters can lead to skidding or slipping. It is difficult to find parameters that meet all rail conditions.

To solve this problem and improve the system's robustness, we used an additional dynamic torque tuning function ΔT_C , which is combined with reasonable selection of the parameters of the PI regulator. The regulation method of the tuning function is shown in Figure 7 and is described as follows:

- First, define a reference slip velocity $v_{s,ref}$ on the basis of the related adhesion test criteria and adhesion control experiences.
- Second, set two equidistant sidebands of slip velocity Δv_s , then the actual slip velocity v_s must be controlled within the range of $[v_{s,ref} - \Delta v_s, v_{s,ref} + \Delta v_s]$. At time t1, if v_s is lower than the range and $\dot{\mu}$ is less than zero, the operating point will move far away from the peak point. Though the operating point is in the stable region, the adhesion efficiency is too small. So ΔT_C increases at a preset rate until it reaches its upper limit T_{ad-C1} to achieve higher adhesion efficiency. Otherwise, at time t4, if v_s is larger than the range and $\dot{\mu}$ is less than zero, the operating point has gone through the peak point into the unstable region. For this occasion, ΔT_C has to decrease at a preset rate until it reaches its lower limit

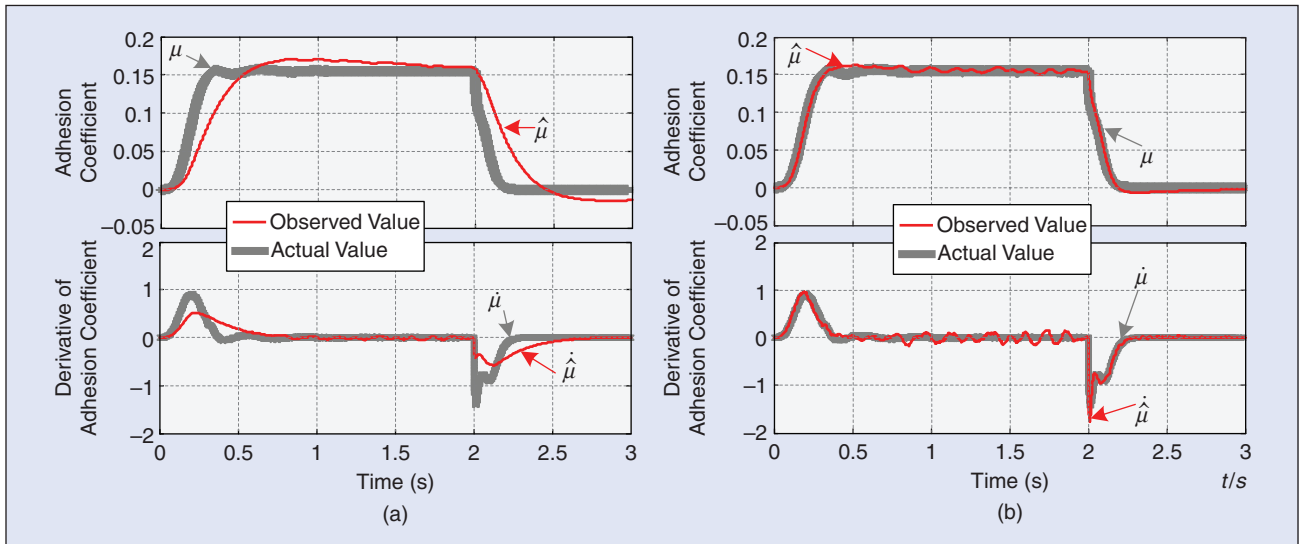


FIGURE 6 – The simulation results of first-order disturbance observer with (a) $I_1 = I_2 = -30$ and (b) $I_1 = I_2 = -150$.

$-T_{ad-C2}$. For both occasions, ΔT_C will recover to zero with another slope when v_s is back to the range.

- Third, μ will increase when the operation point moves toward the peak point due to the characteristics of the adhesion curve in Figure 4, i.e., $\dot{\mu}$ will be greater than zero during the movement. This characteristic can be used to achieve a real-time correction of v_{s_ref} . For example, at time $[t_2, t_3]$, when ΔT_C reaches its upper limit T_{ad-C1} and $\dot{\mu}$ is larger than zero, there is still more space for larger v_{s_ref} , and it could be increased to approach higher adhesion efficiency.

The auxiliary tuning torque ΔT_C can help the PI regulator track much higher adhesion efficiency under uncertain rail conditions, which also prevents the operating point from back adjustment or moving to the unstable region and effectively improves the system's robustness.

Traction Control Using the Proposed Adhesion Controller

The traction control system combined with the proposed adhesion controller is shown in Figure 8, where the reference vehicle velocity v_t can be calculated from (14). In the figure, the final torque reference input to the FOC strategy is calculated through two parts: general vehicle logic control and the proposed adhesion control

$$\begin{cases} v_t = \min(v_{d1} \dots v_{d4}, v_T) & \text{when powering} \\ v_t = \max(v_{d1} \dots v_{d4}, v_T) & \text{when braking} \\ \omega_m = \text{ave}(v_{d1} \dots v_{d4})R_g/r \\ T_{ref} = \min(T_{limit}, T_{ad}). \end{cases} \quad (14)$$

The general vehicle logic control is used to produce the effective torque command value T_{limit} , which is first derived from the handle and traction characteristics. Then the states of the entire traction system, such as the temperature coefficients of the heat-sink, motors, filter, and air, the wheel diameter differences, and the weight change, will also limit the amplitude of the reference torque. After that, the jerk control, which is used to control passenger comfort, will regulate the change rate of the torque.

Meanwhile, in the proposed adhesion control, the observed motor torque from the FOC algorithm and the equivalent wheel speed ω_m in (14) are used as the input signal for the first-order disturbance observer, which is proposed to obtain the derivative value, \dot{T}_L , of the current traction condition. Then the PI regulator combined with auxiliary torque tuning, ΔT_C , outputs the final available torque reference T_{ad} .

In a general traction calculation of the traction system design, the adopted adhesion coefficient is normally less than the available maximum value. Therefore, under good wheel-rail surface conditions, T_{ad} is usually greater than T_{limit} , and T_{limit} is chosen to meet the traction demand. When the rail conditions worsen, however, the available adhesion coefficient

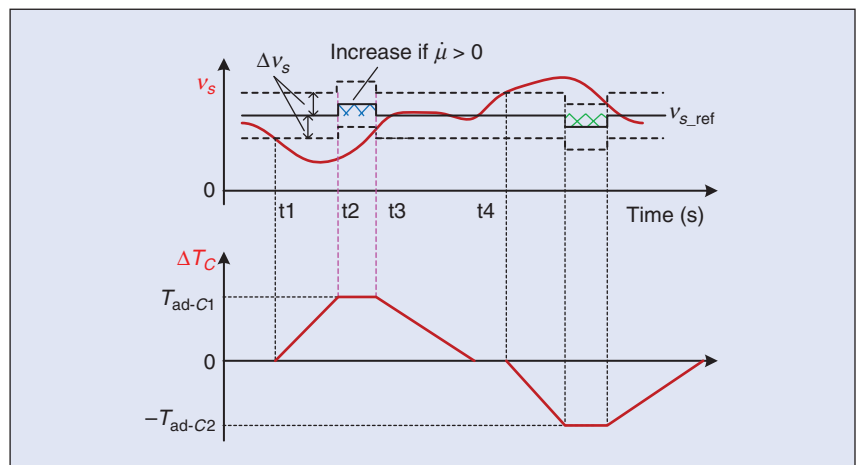


FIGURE 7 – An illustration of auxiliary dynamic torque tuning.

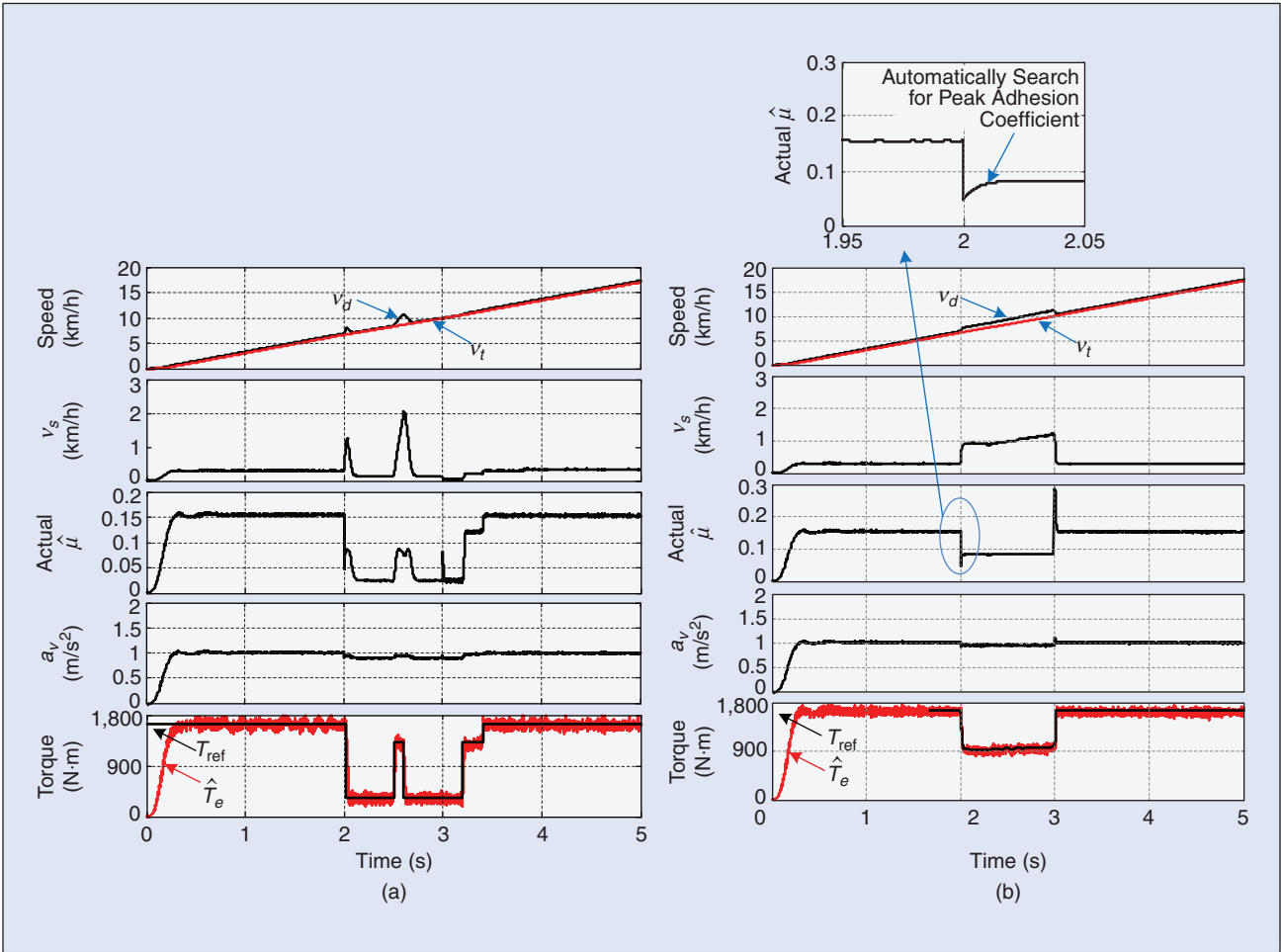


FIGURE 9 – Adhesion control simulations of the acceleration process with (a) the conventional correction method and (b) the proposed method.

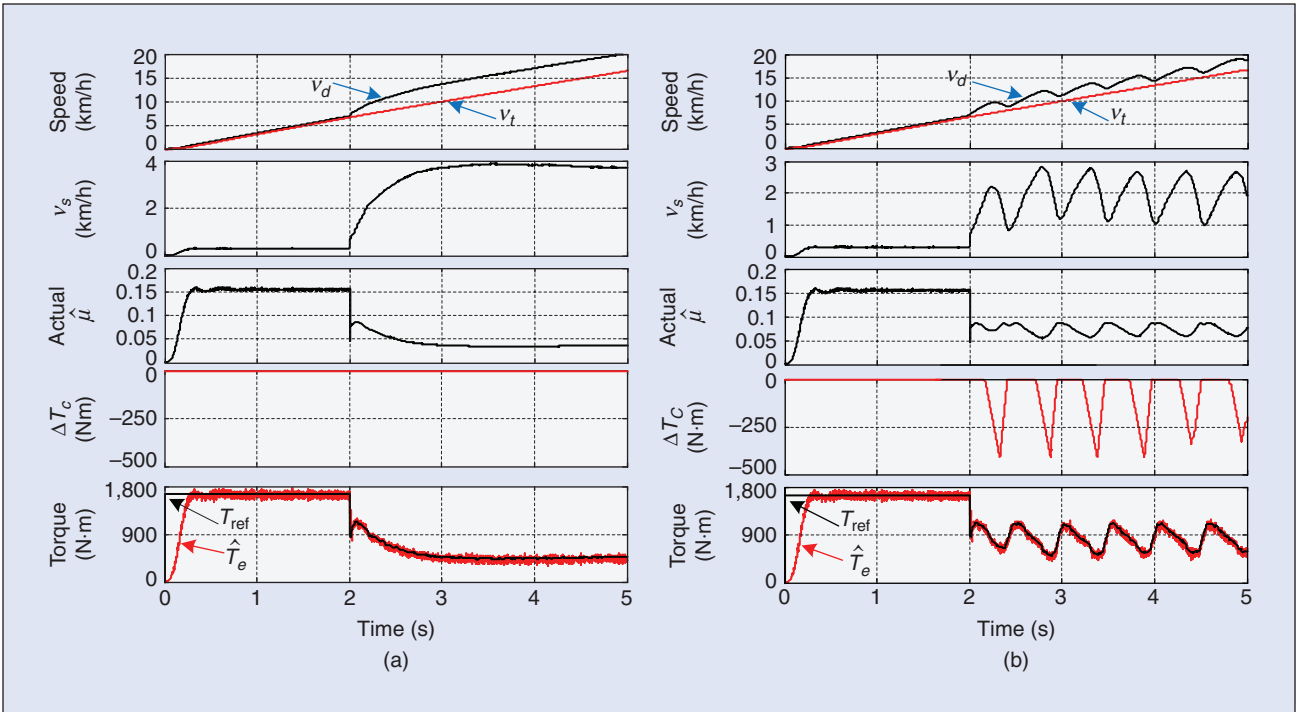


FIGURE 10 – Simulation results of adhesion control by (a) PI regulator only and (b) PI regulator combined with dynamic torque tuning.

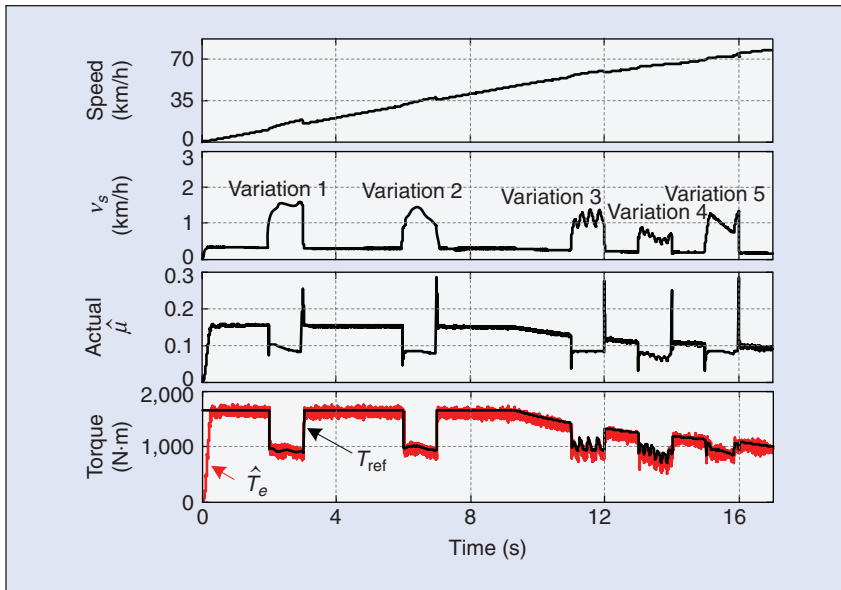


FIGURE 11 – Simulation results of a whole speed range under variable conditions.

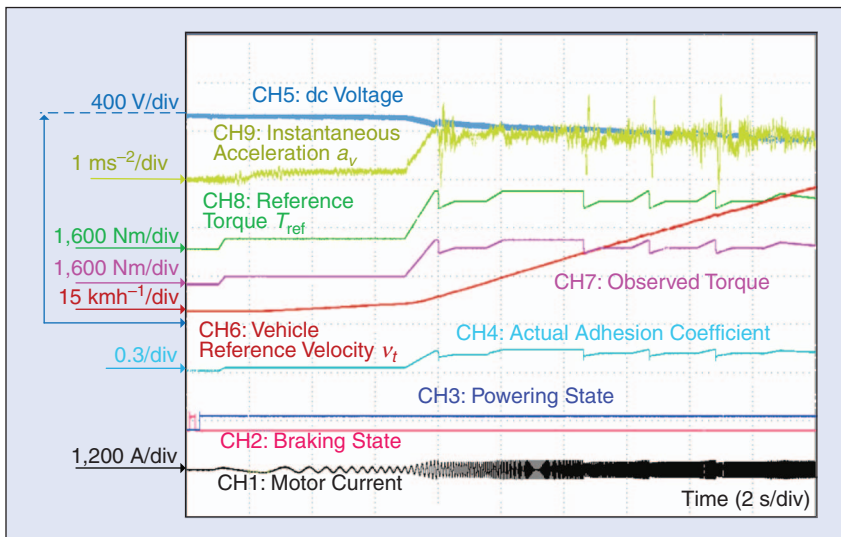


FIGURE 12 – Adhesion test results during slippage when operating in the powering state.

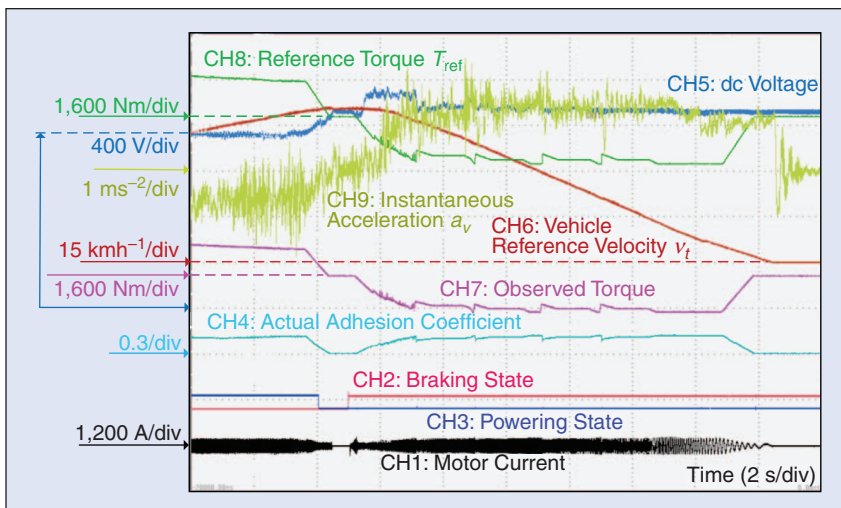


FIGURE 13 – Adhesion test results during sliding when operating in the braking state.

efficiency will be better when the slope is somewhat large under poor adhesion conditions, but the torque will fluctuate if there is no sideband and the slope is too large. This study used a slope of 1,250 Nm/s.

By using the given dynamic vehicle model, Figure 11 shows the control performance over the entire speed range under varying rail conditions, where the traction motor changes from the constant torque zone, to the constant power zone, and to the natural zone. The simulation waveforms demonstrate that the adhesion controller using the proposed strategies will be activated to prevent heavy slippage and achieve high adhesion efficiency.

Line Experimental Verification

To verify the adhesion control performance in practice, an onsite test was carried out on the Guangzhou Subway Line 1. According to the metro user's demands for a traction system-type test, the adhesion test must be done during the horizontal straight line test under no-load conditions with 100% traction and braking force, and the maximum speed level must be not higher than 50 km/h on the test line. To simulate the slippage conditions on the dry rails, a liquid mixture of detergent and water was sprayed on the wheel-rail surface in front of the motor bogie wheel. The test waveforms under powering and braking states are shown in Figure 12 and Figure 13, respectively. Both results have been converted to one traction motor.

When slippage or sliding occurred, the adhesion controller reduced the torque reference T_{ref} (CH8), and the observed electromagnetic torque \hat{T}_e (CH7) tracked with the reference value very closely, which indicates good performance of FOC algorithm. The proposed first-order disturbance observer estimated the load torque and calculated the actual adhesion coefficient $\hat{\mu}$ (CH4) of the actual wheel-rail surface condition. Then the torque and the adhesion coefficient recovered to the maximum reference values quickly when slippage or sliding stopped. Meanwhile, the measured instantaneous acceleration a_v (CH9)

changed dramatically when the adhesion problem happened, but the system remained stable overall. Meanwhile, the motor current (CH1), dc voltage (CH5), and vehicle speed (CH6) remained stable and did not exceed any of the protection limits of the traction motor drive. The controller achieved good traction and high dynamic adhesion performance.

According to the relevant International Union of Railway (UIC) standard, during slip tests, the instantaneous slide value for the wheelsets of the brake system shall not exceed 30 km/h for more than 3 s [36]. Until now, there has not been a specific standard for slip protection for traction converters, and the criterion typically adopted for the traction converter is that the slip speed must be below 5 km/h during slip tests. Additionally, the adhesion efficiency η_{ad} calculated by the following equation can also be adopted to indirectly evaluate adhesion performance, where the control efficiency must not fall below 80%:

$$\eta_{ad} = 100\% \times \frac{\int_{t1}^{t2} a_{measured}}{\int_{t1}^{t2} a_{fictitious}}, \quad (15)$$

where $a_{measured}$ is the measured acceleration within the slide or slip period [t1, t2] and $a_{fictitious}$ is a fictitious acceleration, which is calculated from the total reference force divided by the total mass of the vehicle.

The adhesion efficiencies of the simulation results in Figure 9 and the test results can be calculated by using (15); these are shown and compared in Table 2, where the onsite line test control efficiency of the conventional adhesion method is obtained from our previous test. From the table, it can be seen that the adhesion efficiency of the proposed method has improved significantly when compared with the conventional method.

The simulation waveforms demonstrate that the adhesion controller using the proposed strategies will be activated to prevent heavy slippage and achieve high adhesion efficiency.

From the simulation and experiments, we have validated the good performance of the proposed adhesion control method; the merits of this method include high dynamic response to varying rail surface conditions, immunity to interference, and easy modeling and engineering usage. The single-axle model, however, makes several simplifying assumptions. There will also be some limitations when parallel motors are used, e.g., the observed torque of the parallel motors may not be distributed equally among motor axles when the difference in wheel diameters reaches a large value, because not all of the motor axles can reach the highest adhesion utilization. In the one-converter/one-motor driving mode, however, such as in an independent-wheel tram or in rolling stock, this issue does not arise.

Conclusions

The main focus of this article is on an adhesion control strategy for urban railway vehicles utilizing a disturbance observer, combined with an additional dynamic torque tuning function, which was adopted to adjust the motor torque command after the PI controller. This tuning function helps compensate for the drawback of the PI controller and improves the control robustness under different rail conditions. This strategy prevents heavy wheel slippage and achieves high adhesion efficiency. Simulation and line experimental results are carried out

by using practical vehicle and traction converters, and the results show good slippage regulation and robustness under different rail conditions. The proposed adhesion control method has a good dynamic response under varying rail surface conditions, it is immune to interference, and it is easy for modeling and implementation. The method can also be easily extended to other railway vehicles, such as mainline railways and rubber-tire trains.

Acknowledgment

This work was supported in part by the National Natural Science Foundation of China (U1134204) and the China National Science and Technology Support Program (2013BAG21Q00 and 2015BAG13B01). The authors would like to thank the engineers of Guangzhou Metro and Beijing Qiansiyu Electric Co., Ltd., for their valuable cooperation during onsite testing. The authors would also like to thank the China Scholarship Council.

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TABLE 2 – ADHESION CONTROL EFFICIENCY COMPARISON.

SIMULATION ADHESION EFFICIENCY		EXPERIMENT ADHESION EFFICIENCY		
Figure 9(a)	Figure 9(b)	Conventional method	Traction Figure 12	Brake Figure 13
85.2%	92.6%	86.2%	91.7%	92.9%

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