Multi-Factor Modeling Method of the Load Sharing Ratio under Moving Train Loads

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In railway engineering, the load sharing ratio (LSR) is the ratio of the rail seat load (RSL) to the axle load, which is affected by many factors. The LSR can be used in the design and analysis of railway track structures as well as in the research of predicting the dynamic influence of railway tunnels and the environment. The “static loading method” commonly used to study the LSR does not conform to reality; using it, it is difficult to obtain a complete LSR curve, limiting its application. Besides, there is currently a lack of LSR prediction methods considering the impact of multiple factors. Therefore, this paper proposes a “moving loading method” for investigating the LSR under moving train excitation, verified to be rational by comparing with the experimental results. At the same time, a procedure for establishing the LSR multi-factor prediction model is put forward, namely, we (1) determine the LSR function form and the fitting algorithm; (2) perform parameter sensitivity analysis to determine the main influencing parameters of the LSR function; and (3) design a quadratic regression orthogonal test to obtain the prediction formula of the LSR function coefficients. Once establishing the prediction model for a type of train-track system, the LSR of similar systems can be calculated by adjusting the main parameters of the model. Shijiazhuang Metro Line 1 using the A-type vehicle and the monolithic trackbed is taken as a case study to develop a corresponding LSR multi-factor prediction model by the moving loading method and the procedure mentioned above. The results indicate that the proposed method performs well and can be adopted to enhance the accuracy of track design or tunnel and environmental vibration prediction.

1. Introduction

The rail seat load (RSL) is the load transferred from the rail to the underneath slabs via fastenings, rail seat plates, and sleepers. The load sharing ratio (LSR) is the ratio of the RSL to the axle load, reflecting the axle-load transmission law among wheelsets, rail, and fastenings. The maximum RSL, acting as the main parameter in the design and construction of railway tracks [1], needs to be calculated from the maximum axle load and the LSR. Moreover, recently increasing investigations have been devoted to reducing the model scale (for example, using 2.5D numerical methods) of the dynamic response prediction of railway tunnels [2] and the environment [3–7]. However, these models still include rail and fastenings, partly repeating with train-track coupling models. Omitting rail and fastenings and imposing the RSL time history calculated from the dynamic axle load and the LSR time history as the excitation can further simplify 2.5D numerical models.

Current practices in the analysis and design of railway track systems assume the axle load to be static [8]. A dynamic coefficient is generally adopted worldwide to modify the axle load to reflect the effect of the load dynamic properties [9]. However, the current dynamic coefficient formulas recommended are different both in form and main parameters [10–19]. Early methods for determining LSR are also based on the static assumption using the static load or the dynamic modified static load, mainly including the three adjacent sleepers method [11, 14, 20, 21], Australian formula...
2. Methodology for Obtaining the LSR under Moving Train Loads

2.1. The LSR under the Excitation of a Single Wheel. Existing studies generally use the “static loading method,” that is, applying a static load at a fixed position on the rail to calculate the LSR of surrounding fastening. When the axle load acts on a specific point, the load is distributed mainly by a total of six fastenings on both sides (five fastenings when the axle is directly above one fastening), as shown in Figure 1(a). This method is only useful when the wheel axle is located in a particular situation. Therefore, this article proposes a “moving loading method” as follows.

For a sure fastening, record the distance between the wheel axle and denote it as \(D\). As shown in Figure 1(b), the LSR is not zero only when \(D\) is less than \(3a\) (\(a\) is the distance between adjacent fastenings). When the train runs at a constant speed, for a specific train-track system, the LSR of a single fastening is merely related to \(D\). Regarding the metro track as a periodic structure, the LSR of each fastening varies with \(D\) in the same way. Therefore, as shown in Figure 2, the LSR of a single fastening time history includes all the static loading method results, and the complete LSR curve is obtained only by transforming the \(x\)-axis from time to \(D\).

2.2. The LSR under the Excitation of a Bogie. The “moving loading method” under a single wheel excitation regards the wheel as a point load and does not consider the real structure of the train and the complicated dynamic interaction between the wheel and the rail. The following assumptions are made to simplify the problem [28]: (1) when a single wheelset passes by, the LSR time-history curve is smooth and symmetrical; (2) in a certain train-track system, the LSR stimulated by each wheelset is exactly the same as the LSR stimulated by a single wheelset; and (3) all the LSR functions are superimposed assuming a linear superposition.

When loads from two wheels in a bogie are superimposed on a fastening, the corresponding M-shaped LSR time-history curve is shown in Figure 3(a). Due to the superposition of adjacent axle loads, the LSR curve of a single wheel is incomplete, but considering the symmetry, a complete curve can be obtained by half of the curve, as illustrated in Figure 3(b). The LSR under a whole train can be obtained by adding the LSR under the excitation of each wheel.

2.3. Verification. To investigate the effect of the moving loading method, a calculation is carried out using the static loading model test parameters of the literature [28], including the train speed \(v = 100\) km/h, the fastening stiffness \(k = 28 \cdot 5\) MN/m, and the fastening spacing \(a = 0.63\) m. Figure 4 shows that the closer the fastener to the loading point, the greater the difference in LSR between the two curves. However, the maximum relative error between the two methods is about 18%, and the maximum absolute error is merely 1.7%, indicating that the results are very close. Both approaches suggest that the axle load is mainly borne by the surrounding five fastenings when the load is above the fastening and six when it is in the middle of two adjacent fastenings. Considering the negligible error between the two methods, the prediction model of the moving loading method can cover all the static loading method results.

3. Methodology for Establishing an LSR Prediction Model

A procedure for establishing the LSR multi-factor prediction model is put forward, which includes the following:

1. Determination of the LSR function form and fitting algorithm.
2. Parameter sensitivity analysis for determining the main parameters of the LSR function.
3. Quadratic regression orthogonal tests for obtaining the prediction formula of the LSR function coefficients.

3.1. The LSR Function. The LSR curve is approximately a Gaussian curve [27]. Therefore, a Gaussian-like function is used to fit the LSR curve as follows:
\[ y(x) = e^{(A - B x^2)} \times 1\%, \]  
\[ (1) \]

where \( y(x) \) is the LSR; \( x \) is the ratio of \( D \) to \( a \) (\( a \) is fastening spacing); and \( A \) and \( B \) are constants determined by actual conditions. The Levenberg–Marquardt optimization algorithm is used for fitting.

3.2. Parameter Sensitivity Analysis. The LSR is affected by many factors, including train speed, axle load, fastening stiffness, fastening damping, fastening spacing, trackbed elastic modulus, and subsoil stiffness [30]. To achieve quantitative analysis, a parameter sensitivity analysis method is employed to find the main parameters. The parameter sensitivity in the form of the ratio of the objective function to the relative change rate of the influencing parameter is adopted as follows:

\[ S_{ij} = \frac{\Delta y_j / y_j^0}{\Delta x_j / x_j^0}, \]  
\[ (2) \]

where \( S_{ij} \) is the sensitivity of the LSR \( y_i \) to the \( j \)th parameter \( x_j \) at the \( i \)th position of the load; \( \Delta x_j \) is the parameter variation; \( \Delta y_j \) is the LSR variation with parameter \( x_j \); \( y_j^0 \) is the initial value of the LSR at the \( i \)th position; \( x_j^0 \) is the initial value of the \( j \)th parameter \( x_j \); \( i = 1, 2, \ldots, n \); \( j = 1, 2, \ldots, m \); and \( n \) and \( m \) are the maximum numbers of positions and parameters, respectively.

The LSR function sensitivity to different parameters is still a function related to variable \( D \), bringing difficulties in selecting the main parameters. Regarding the coefficients \( A \) and \( B \) as target parameters, equation (2) is transformed into the following form:

\[ S_A^j = \frac{\Delta A_j / \Delta x_j / x_j^0}{\Delta A_0 / \Delta x_j / x_j^0}, \]
\[ S_B^j = \frac{\Delta B_j / \Delta x_j / x_j^0}{\Delta B_0 / \Delta x_j / x_j^0}, \]  
\[ (3) \]

where \( S_A^j \) and \( S_B^j \) are separately the sensitivity factors of \( A \) and \( B \) to the \( j \)th parameter \( x_j \); \( \Delta A_j \) and \( \Delta B_j \) are separately the variations of \( A \) and \( B \) with \( x_j \); and \( A_0 \) and \( B_0 \) are individually the initial values of \( A \) and \( B \).

3.3. Determining the Coefficient Formulas. To generalize equation (1) into a universal empirical formula, it is necessary to further explore the valuing method for \( A \) and \( B \). After determining the main parameters of the LSR function, a quadratic regression orthogonal test is designed, and the
value functions of A and B under multiple parameters are established, respectively. To simplify the treatment process of data, the actual value of each test factor $x_j$ is firstly linearly transformed into the factor level code $z_j$ as follows:

$$z_j = \frac{x_j - x_{0j}}{\Delta_j},$$

$$\Delta_j = \frac{x_{2j} - x_{0j}}{c},$$

where $x_{0j}$ is the average of the upper and lower bounds of the $i$th experimental factor $x_i$; $x_{2j}$ is the upper bound of $x_j$; $\Delta_j$ is the variation range of $x_j$; and $c$ is the asterisk arm which can be calculated as follows:

$$c = \sqrt{\frac{(m_c + 2m + m_0)m_c - m_c}{2}},$$

where $m$ is the number of test factors; $m_0$ is the number of zero-level tests; and $m_c$ is the number of two-level trials.

Quadratic regression orthogonal analysis needs to consider the influence of a single factor $z_j$ and the impact between two factors $z_k z_j$; hence, the quadratic regression equations of parameters A and B are shown as follows:

$$\hat{A} = a_0 + \sum_{j=1}^{n} a_j z_j + \sum_{j=1}^{n-1} \sum_{k=j+1}^{n} a_{jk} z_j z_k + \sum_{j=1}^{n} a_{jj} z_j^2,$$

$$\hat{B} = b_0 + \sum_{j=1}^{n} b_j z_j + \sum_{j=1}^{n-1} \sum_{k=j+1}^{n} b_{jk} z_j z_k + \sum_{j=1}^{n} b_{jj} z_j^2,$$
where \( \tilde{A} \) and \( \tilde{B} \) are the estimated values of parameters \( A \) and \( B \); \( n \) is the number of test factors; and \( a_0, a_j, a_{kj}, b_0, b_j, b_{kj}, \) and \( b_{jj} \) are all regression coefficients.

4. Case Study

The concrete monolithic trackbed has the characteristics of good integrity and high rigidity and is a commonly used track form in metro underground lines [31]. Shijiazhuang Metro Line 1 using the A-type train and the monolithic trackbed was taken as an example. The LSR data were obtained employing a numerical model based on the moving loading method, and a multi-factor prediction model of LSR was developed.

4.1. Numerical Model. The instance is an A-type metro train with six cars including four motor cars and two trailer cars. The train size is shown in Figure 5, and the main parameters are shown in Table 1. Each vehicle subsystem had several rigid bodies including one carbody, two frames, four wheelsets, and eight axle boxes, and each body had six degrees of freedom in the longitudinal, lateral (transverse), vertical (up and down), side roll, pitch, and yaw. Therefore, each vehicle subsystem had 90 degrees of freedom. The train model established by the MB software named Universal Mechanism (UM) is shown in Figure 6(a). The axle load of the fully loaded train was about 14.8 t, and the vehicle runs at a constant speed of 80 km/h.

The rails were modeled as Timoshenko beams, and the fastenings were modeled as a series of spring-damper pairs [32] with a fastening spacing of \( a = 0.6 \) m. The monolithic trackbed and the circular cross section tunnel lining were modeled as a whole FE system without creating separate sleeper models. The geometry and properties of the track-tunnel-soil system are separately shown in Figure 7(a) and Table 2. Based on the linear viscoelastic constitutive model, a FE model of the trackbed-tunnel bottom was established using the element type of SOLID 45 in the software ANSYS, as shown in Figure 7(b). The longitudinal length of the model was 120 m, and the element size was 0.11–0.3 m. The Craig–Bampton method was adopted to couple the FE model and the MB model in the software UM. Spring-damper pairs were used to simulate the subsoil, whose stiffness and damping coefficients were determined separately according to the literatures [33, 34].

There are many wheel-rail interaction models. In this paper, the Kik–Piotrowski multi-point contact algorithm [35] was used to simulate the contact between wheels and rail. Figure 7(b) shows the train-track coupling model.

Track irregularity is the main reason for the dynamic excitation of trains [36]. The environmental vibration response caused by the metro operation is mainly vertical [37]. Hence, it is generally considered that the vertical wheel-rail force is the primary excitation source, so only vertical track irregularities were considered [38]. At present, the American Class 6 track irregularity spectrum and the Sato track irregularity spectrum are commonly used to simulate medium and long wave and short wave irregularities in metro tracks, respectively [36]. However, some studies have shown that the American Class 6 track irregularity spectrum differs significantly from the actual metro track irregularities [39–41]. This research used the Shanghai Metro track spectrum [39] and Sato track irregularity spectrum [42] to simulate vertical track irregularities. Figure 8 shows the sample data.

4.2. Results. \( S_A^4 \) and \( S_B^4 \) obtained from equation (3) do not change with \( D \), facilitating an intuitive comparison of the influence of various parameters. The reference values of parameters were determined according to the actual situation of Shijiazhuang Metro Line 1, and the change rate of each parameter took the same small amount (10%). The calculation conditions are shown in Table 3, where condition 3 increases the mass of each rigid body in the train model by 10%. In fact, the mass of carbody, bogie, and wheelset has different influence on LSR, but here the vehicle is taken as a whole, and only the influence of total vehicle mass on LSR is considered to reduce the number of influencing factors. In the process of changing the mass of each body, the change of vehicle structure is not expected. Therefore, assuming that the influence of mass change as small as 10% on vehicle structural characteristics can be ignored, the mass of each rigid body is increased by 10%.

According to the conditions in Table 3, the vehicle-rail coupling calculation is performed to obtain the corresponding LSR data. The calculated discrete data cannot fully reflect the LSR when the wheel is at any position, so the Levenberg–Marquardt optimization algorithm was used to fit each data group according to equation (1) (see Figure 9).

Figures 10(a) and 10(b) show the sensitivity curves of LSR and the coefficients \( A \) and \( B \) to each parameter calculated by equations (2) and (3) individually. As shown in Figure 10(a), factors have different effects on the LSR of different locations, resulting in different parameter sensitivity distribution laws. Besides, the impact of axle load varies with \( D \) in the opposite trend with other factors. Figure 10(b) implies that the coefficients \( A \) and \( B \) have considerable absolute values of the parameter sensitivity of the fastening stiffness and the fastening spacing, so they were regarded as the main influencing parameters.

The fastening stiffness change range was taken to be 20–50 MN m\(^{-3} \), and the fastening spacing change range was assumed to be 0.5–0.65 m. A two-factor quadratic regression orthogonal test was designed, and the factor level coding is shown in Table 4. The number of test factors was \( m = 2 \), and the length of the asterisk arm was \( y = 1.078 \). The number of two-level tests was \( m_r = 2^m = 4 \); the number of asterisks was \( m_s = 2m = 4 \); and the number of zero-level tests was \( m_b = 2 \). Therefore, the total number of test groups was \( N = 10 \).
Table 1: Key parameters of the vehicle model.

<table>
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<tr>
<th>Location</th>
<th>Notation</th>
<th>Item</th>
<th>Value</th>
<th>Unit</th>
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<td>Carbody</td>
<td>$M_c$</td>
<td>Mass of fully loaded carbody</td>
<td>$4.81 \times 10^4$</td>
<td>kg</td>
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<td>$I_{cx}$</td>
<td>Side roll inertia moment of carbody</td>
<td>$4.38 \times 10^4$</td>
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<td>$I_{cy}$</td>
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<tr>
<td></td>
<td>$I_{cz}$</td>
<td>Yaw inertia moment of carbody</td>
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<td>Frame</td>
<td>$M_f$</td>
<td>Mass of frame</td>
<td>$1.65 \times 10^3$</td>
<td>kg</td>
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<td></td>
<td>$I_{fx}$</td>
<td>Side roll inertia moment of frame</td>
<td>$1.09 \times 10^3$</td>
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<td>$I_{fy}$</td>
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<td>$1.78 \times 10^3$</td>
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<td>$I_{fz}$</td>
<td>Yaw inertia moment of frame</td>
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<td>Wheelsets</td>
<td>$M_w$</td>
<td>Mass of wheelset</td>
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<td>$I_{wx}$</td>
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<td>$I_{wy}$</td>
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<td>$I_{wz}$</td>
<td>Yaw inertia moment of wheelset</td>
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<td>Mass of axle box</td>
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<td>$I_{bx}$</td>
<td>Side roll inertia moment of axle box</td>
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<td>$I_{by}$</td>
<td>Pitch inertia moment of axle box</td>
<td>3.9</td>
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<td>$I_{bz}$</td>
<td>Yaw inertia moment of axle box</td>
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<td>kg·m²</td>
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<td>Suspension</td>
<td>$k_1$</td>
<td>Vertical stiffness of primary suspension</td>
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<td>N/m</td>
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<td>$c_1$</td>
<td>Vertical damping of primary suspension</td>
<td>$2.00 \times 10^4$</td>
<td>N·s/m</td>
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<td></td>
<td>$k_2$</td>
<td>Vertical stiffness of secondary suspension</td>
<td>$2.04 \times 10^5$</td>
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<td>$c_2$</td>
<td>Vertical damping of secondary suspension</td>
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<td>N·s/m</td>
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Figure 6: Train model and train-track coupling model. (a) Multi-rigid body train model. (b) Train-track coupling model.

Figure 7: Geometry and the simplified model adopted for modeling the track-tunnel-soil system. (a) Geometry of the track-tunnel-soil system. (b) The simplified model of the track-tunnel-soil system.
Table 2: Properties of the track-tunnel-soil system.

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<th>Unit</th>
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<td>Track and tunnel</td>
<td>$m_r$</td>
<td>Mass per meter of rail</td>
<td>60.64</td>
<td>kg/m</td>
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<td></td>
<td>$k_f$</td>
<td>Vertical stiffness of fastening</td>
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<td>$c_f$</td>
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<td>N·s/m</td>
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<td></td>
<td>$\rho_t$</td>
<td>Density of trackbed and tunnel</td>
<td>$2.50 \times 10^3$</td>
<td>kg/m$^3$</td>
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<td></td>
<td>$E_t$</td>
<td>Young’s modulus of trackbed and tunnel</td>
<td>$3.55 \times 10^{10}$</td>
<td>Pa</td>
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<td></td>
<td>$\mu_t$</td>
<td>Poison ratio of trackbed and tunnel</td>
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<td></td>
<td>$\zeta_t$</td>
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<td>Soil</td>
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<td>Young’s modulus of soil</td>
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<td></td>
<td>$\mu_s$</td>
<td>Poisson’s ratio of soil</td>
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<td></td>
<td>$C_s$</td>
<td>Shear wave velocity of soil</td>
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<td>$C_p$</td>
<td>Compression shear wave velocity of soil</td>
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<td>m/s</td>
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Figure 8: Sample data of vertical track irregularities. (a) Short wave irregularities. (b) Medium and long wave irregularities. (c) Superimposed wave irregularities.

Table 3: Parameter sensitivity analysis conditions.

<table>
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<tr>
<th>Condition</th>
<th>$v$ (km·h$^{-1}$)</th>
<th>$m_a$ ($t$)</th>
<th>$k_f$ (MN·m$^{-1}$)</th>
<th>$c_f$ (kN·s·m$^{-1}$)</th>
<th>$a$ (m)</th>
<th>$E_t$ (MPa)</th>
<th>$k_s$ (MN·m$^{-1}$)</th>
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<td>1</td>
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<td>14.80</td>
<td>40.73</td>
<td>9.90</td>
<td>0.60</td>
<td>35 500</td>
<td>862.11</td>
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<tr>
<td>2</td>
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<td>14.80</td>
<td>40.73</td>
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<tr>
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<tr>
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<td>14.80</td>
<td>40.73</td>
<td>10.89</td>
<td>0.60</td>
<td>35 500</td>
<td>862.11</td>
</tr>
<tr>
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<td>14.80</td>
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<td>35 500</td>
<td>862.11</td>
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<tr>
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<td>14.80</td>
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<td>39 050</td>
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<tr>
<td>8</td>
<td>80</td>
<td>14.80</td>
<td>40.73</td>
<td>9.90</td>
<td>0.60</td>
<td>35 500</td>
<td>948.32</td>
</tr>
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</table>

$v$ is the train speed; $m_a$ is the axle load; $k_f$ is the fastening stiffness; $c_f$ is the fastening damping; $a$ is the fastening spacing; $E_t$ is the trackbed elastic modulus; and $k_s$ is the subsoil stiffness, that is, the stiffness of the equivalent spring of subsoil.
Figure 9: The fitting curves of the LSRs under different conditions. (a) Condition 1 (initial). (b) Condition 2 (v is modified). (c) Condition 3 (m_a is modified). (d) Condition 4 (k_f is modified). (e) Condition 5 (c_f is modified). (f) Condition 6 (a is modified). (g) Condition 7 (E_b is modified). (h) Condition 8 (k_s is modified). The values of coefficients A and B are shown in the figure, and R^2 is the goodness of fit.
Table 5 shows the values of test factor levels and the corresponding coefficients of fitting functions. By carrying out regression analysis according to the orthogonal experiment results, the regression equations were obtained as follows:

\[
\hat{A} = 3 \cdot 544 + 0 \cdot 072z_1 + 0 \cdot 124z_2 - 0 \cdot 074z_1z_2 + 0 \cdot 023z_1^2 + 0 \cdot 018z_2^2, \quad R^2 = 0.928, \tag{7a}
\]

\[
\hat{B} = 0 \cdot 339 + 0 \cdot 038z_1 + 0 \cdot 061z_2 - 0 \cdot 044z_1z_2 + 0 \cdot 036z_1^2 + 0 \cdot 049z_2^2, \quad R^2 = 0.863. \tag{7b}
\]
According to equations (4) and (6), equations (7a) and (7b) are transformed into functions represented by the actual levels as follows:

\[
A = 2.175 + 0.041a + 0.181k_f - 0.076ak_f + 0.0001a^2 + 0.018k_f^2, \quad R^2 = 0.928, \quad (8a)
\]

\[
B = 2.399 + 0.016a - 9.174k_f - 0.045ak_f + 0.0002a^2 + 10.123k_f^2, \quad R^2 = 0.863, \quad (8b)
\]

where \( k_f \) is the fastening stiffness in MN·m\(^{-1} \) and \( a \) is the fastening spacing in m.

Equations (1), (8a), and (8b) are the final LSR formulas. Through this group of functions, the LSR can be predicted from the fastening stiffness and the fastening spacing.

4.3 Error Analysis. The estimated values of coefficients \( A \) and \( B \) were compared with the calculated values, and the relative errors \( \delta \) were calculated individually, as shown in Figure 11(a). The maximum prediction errors of coefficients \( A \) and \( B \) appear in the seventh and the sixth groups separately, and the errors are 5.72% and 5.21% individually, both less than 6%. Hence, using equations (8a) and (8b) for prediction is reliable.

Under the sixth group of test parameters, the LSR curve was generated utilizing the prediction model composed of equations (1), (8a), and (8b). The LSRs obtained by the prediction model and numerical calculation are compared in Figure 11(b). The absolute error is always less than 3%, which is acceptable.

The LSR curve under the excitation of a bogie was obtained by superimposing two LSR curves under the
excitation of a single wheel. Figure 11(c) shows the curves of the prediction result, numerical result, and the absolute error. The prediction error is within 5%, indicating that the prediction effect is satisfactory.

5. Conclusions

The load sharing ratio (LSR) is a significant factor for designing railway tracks, but current LSR formulas or calculation methods are too rough to satisfy the requirements, and the widely used “static loading method” for obtaining the LSR is inconsistent with reality.

This research has led to an improvement of conventional approaches in the calculation of the LSR by putting forward a new procedure of establishing multi-factor formulas of LSR under moving train excitation. Addressing the limitation of the “static loading method,” a “moving loading method” was proposed to obtain the LSR under moving train excitation. On this basis, a procedure for establishing LSR multi-factor formulas was presented, containing the LSR fitting algorithm, parameter sensitivity analysis, and quadratic regression orthogonal tests. To investigate the performance of this procedure, a case study of Shijiazhuang Metro Line 1 was taken. The main conclusions of this article are as follows:

(1) When the axle is located directly above a fastening, the axle load is mainly borne by the surrounding five fastenings; when the wheel is located in the middle of adjacent fastenings, the axle load is primarily distributed by the surrounding six fastenings. The results of moving loading and static loading methods are almost the same, but the latter results are only part of the former.

(2) For the metro using the A-type vehicle and the monolithic trackbed, the fastening spacing and the fastening stiffness are the primary factors of the LSR function. At the same time, the fastening damping, the train speed, and the axle load have little influence, while the subsoil stiffness and the trackbed elastic modulus have almost no effect.

(3) For the metro using the A-type vehicle and the monolithic trackbed, the error between the LSR formulas and the vehicle-rail coupling simulation result is satisfactory.

For similar trains and track types, only one prediction model needs to be established according to the proposed method; then, the LSR prediction results can be obtained by adjusting the main parameters. Replacing current LSR calculation approaches with those recommended in this research will considerably improve the accuracy of the LSR prediction model.

It is worth noting that the LSR prediction model can only consider the influence of quantifiable factors, while non-quantitative factors (including train type and track type) need to be reflected by establishing a new LSR prediction model. For example, cars of type A, type B, or type C may be used in metros, and passenger cars or freight cars may be used in railways. Track types include ballastless tracks and ballasted tracks, and track vibration reduction measures such as vibration damping fasteners may be adopted. The influence of different train and track types on the main parameters of LSR and the distribution of LSR requires to be further investigated in future research.

Data Availability

The data that support the findings of this study are available from the corresponding author upon reasonable request.

Disclosure

The opinions expressed in this paper are those of the authors.

Conflicts of Interest

The authors declare that there are no conflicts of interest regarding the publication of this paper.

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