

New Dynamics Model for Rail Vehicles and Optimizing Air Suspension Parameters Using GA

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Abstract. In this paper, a complete four axle rail vehicle model with 70 Degrees Of Freedom (DOFs) is addressed, which includes; a carbody, two bogies and four axles. In order to include track irregularity effects on vehicle behavior, a simplified track model for a straight line is proposed. As the performance of the suspension components, especially for air springs, has significant effects on rail-vehicle dynamics and the ride comfort of passengers, a complete nonlinear thermo-dynamical air spring model which is a combination of two different models is introduced and then implemented in the complete rail-vehicle dynamics. By implementing the Presthus formulation [1], the thermo-dynamical parameters of an air spring are estimated and then tuned, based on the experimental data. Effects of air reservoir volume and connecting pipe length and diameter on system performances are investigated. For improving passenger comfort during their trips, air suspension parameters of the modeled rail vehicle are tuned to minimize the Sperling ride comfort index. Results showed that by modification of air suspension parameters, passenger comfort is improved and the ride comfort index is reduced by about 10%. The Genetic Algorithm (GA) optimization method is also used to optimize air suspension parameters. Results showed that improved air suspension configurations are more practical, compared to optimized ones.

Keywords: Rail vehicle dynamics; Air spring model; Air suspension parameters; GA; Optimization.

INTRODUCTION

Nowadays, the speed-up in technology and its new features bring about higher speed, with reliable safety and better ride comfort, in rail transportation industries. Traffic jams in capital cities around the world, the waste of passenger time at air terminals, huge mass transportation and so on create a good opportunity for rail industries to attract more and more passengers and cargos to their services. In addition to safety, the other important issue for making passengers decide about transportation type is ride comfort during their trip. And that is why accessing better ride comfort during a trip for passengers is essential.

To serve good ride comfort to passengers, the secondary suspension of most new EMU and DMU rail-vehicles are equipped with air springs. Air springs are very important isolation components, which guarantee good ride comfort during the trip. In most published

rail-vehicle models, developed by research [2-12], the thermo-dynamical effects of air springs on rail-vehicle dynamics are ignored and secondary suspension is modeled by simple springs and dampers.

In the current work, the complete dynamics of one IRICo DMU trailer car with nonlinear component behavior is addressed. The dynamic behavior of all components is validated by some experimental results. In the proposed model, track rails are assumed to be rigid with a viscoelastic bed in vertical and lateral directions [13,14]. In order to consider rail irregularity effects on vehicle behavior, track data measured by an EM120 machine is used. In the model of a vehicle that moves along a straight line, effects of ballast and sleeper mass on vehicle behavior are ignored. Four contact parameters introduced by Shabana and Zaazaa [15,16] are used to define the contact point between the rail and wheel. In order to accelerate the computations, a Feed Forward Neural Network (FFNN) is trained and used to compute the contact parameters. Contact forces are calculated, based on the Polach theory [17]. Effects of air suspension parameters on system performances are investigated. Sperling ride comfort is selected as the fitness function and is minimized by engineering and Genetic Algorithm

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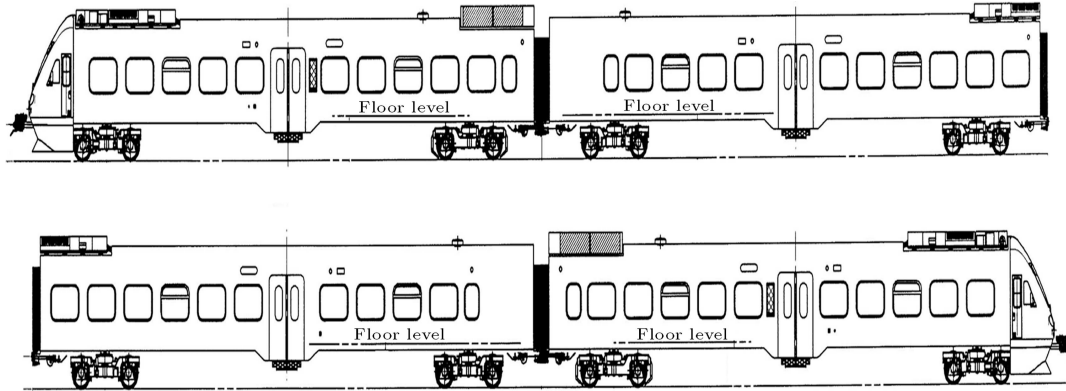


Figure 1. Composition of IRICo DMU.

optimization techniques. Results are summarized at the end.

VEHICLE DESCRIPTION

An IRICo DMU train is composed of four cars: two motor cars at each end and two trailer cars in the middle. The composition of one complete train is shown in Figure 1.

Each car is supported by two bogies. The side view of a two axle IRICo DMU trailer bogie is shown in Figure 2. To guarantee the stability of the vehicle and good ride comfort for the passenger, bogies

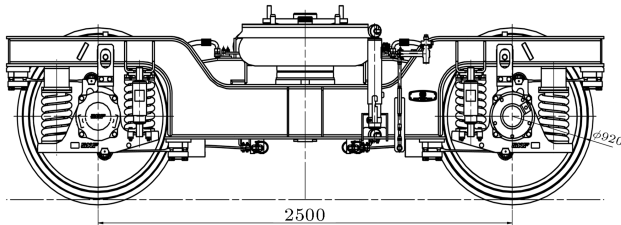


Figure 2. Side view of IRICo DMU trailer bogie.

are equipped with primary and secondary suspension systems. The secondary suspension has two air springs to suspend the vehicle body, four vertical and lateral dampers and two connection links to connect the bogie frame to the carbody. The primary suspension is made of two coil springs, two leaf springs and one vertical damper at each side of the wheel set.

DYNAMIC MODEL

Track Model

In order to include effects due to geometrical track irregularities in the rail-vehicle dynamics, a simplified version of the Jin and Wen [13,14] track model is used. In the proposed model, four track irregularities are included as bed disturbances. As the vehicle is modeled in a straight line, the effects of ballast and sleeper mass on the vehicle dynamics, and the coupling effects between the left and right ballast masses are ignored. The estimated parameters for track stiffness and damping rates are according to the work of Jin and Wen [13]. A schematic diagram of the track model containing these four irregularities is shown in Figure 3.

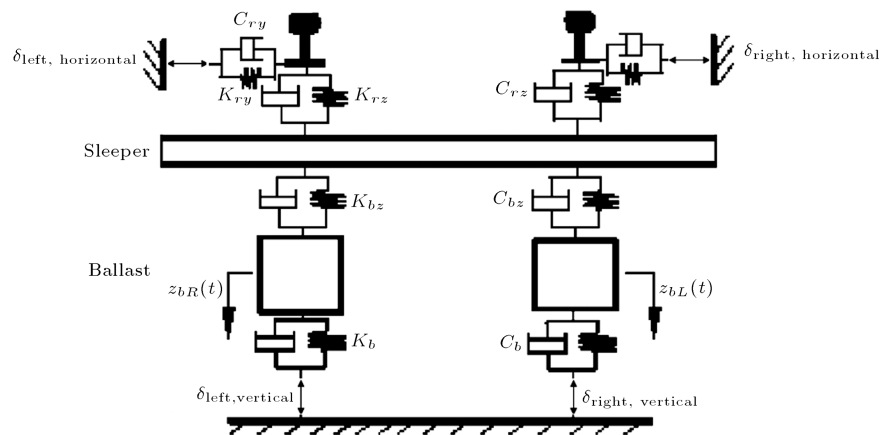


Figure 3. Track model - parameters are based on Jin & Wen work [13].

Contact Point and Contact Forces Model

By using four geometrical contact parameters, s_1^r , s_2^r , s_1^w and s_2^w , shown in Figure 4, and the method proposed by Shabana and Zaazaa [15], the contact point between the rail and wheel is determined. According to Shabana, s_1^r is found out by integrating the speed component of point Q along the rail shown in Figure 4; s_2^w is the angle between the Z -axis of the coordinate system fixed to the axle and vertical line. The remaining two contact parameters are calculated by a searching algorithm, which guarantees the perpendicularity of the rail reaction force and the tangent plane of the wheel at the contact point. To do this, four geometrical parameters, δ_L , δ_R , θ_A and ψ_A , shown in Figure 5 are investigated. By these assumptions, using four contact parameters and the normal reaction force of the rails, creep forces are calculated, based on the Polach theory [17].

For decreasing the simulation time, a feed forward neural network is suggested to compute these two contact parameters, as shown in Figure 6. This neural

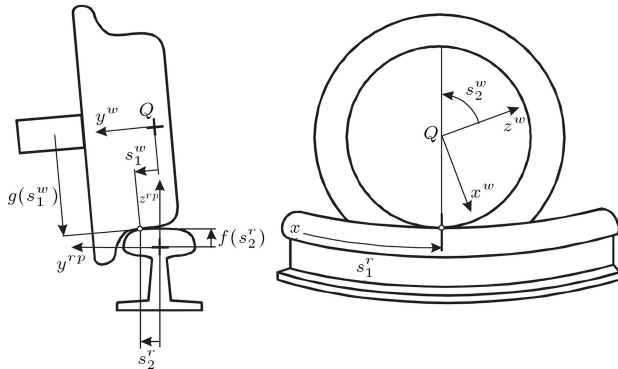


Figure 4. Four contact parameters introduced by Shabana and Zaazaa [15].

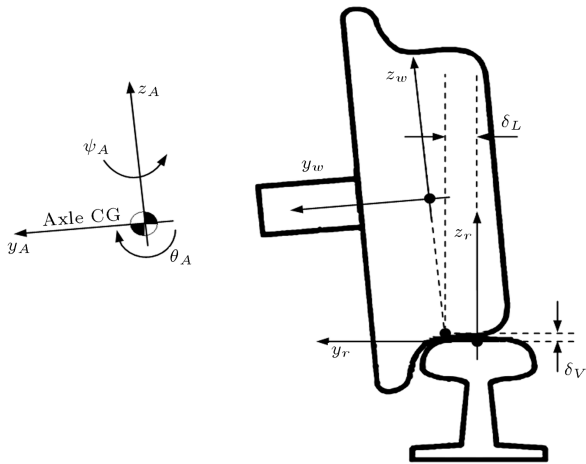


Figure 5. Effective parameters, δ_L , δ_V , ψ_A and θ_A for contact point extraction.

network has four inputs, δ_L , δ_R , θ_A and ψ_A , two outputs, s_2^r and s_1^w , and 14 neurons in two hidden layers. The activation function for hidden layers and the output layer is: tangent sigmoid and linear, respectively. The network is trained based on the Levenberg-Marquardt back-propagation method.

The proposed network is trained according to the searching algorithm results generated during sinusoidal variations of the axle states. The lateral and vertical axle states changed with 0.01 mm and 0.003 mm amplitude and 0.5 Hz and 0.25 Hz frequencies, respectively. At the same time, the yaw and roll angle of the axle varied with 0.4 deg. and 0.2 deg. amplitude and 0.334 Hz and 0.1667 Hz frequencies. The simulation time is selected so that each input signal has at least two periods in the simulation time. Results are shown in Figures 7 and 8. Good agreement between the network output and searching algorithm results guarantees the performance of the trained network. Synaptic weights of the trained network are presented in the Appendix.

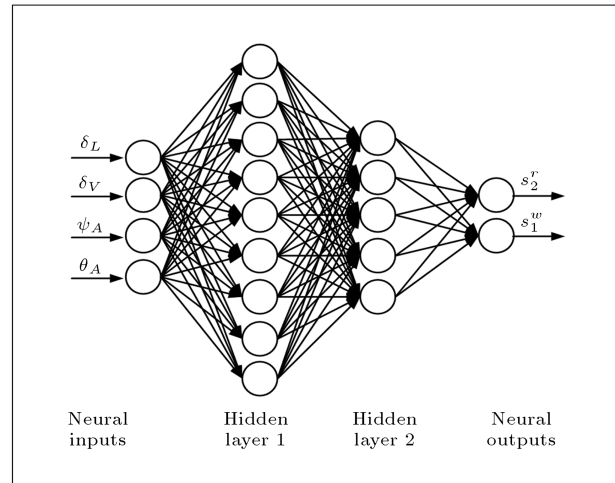


Figure 6. Feed-forward neural network.

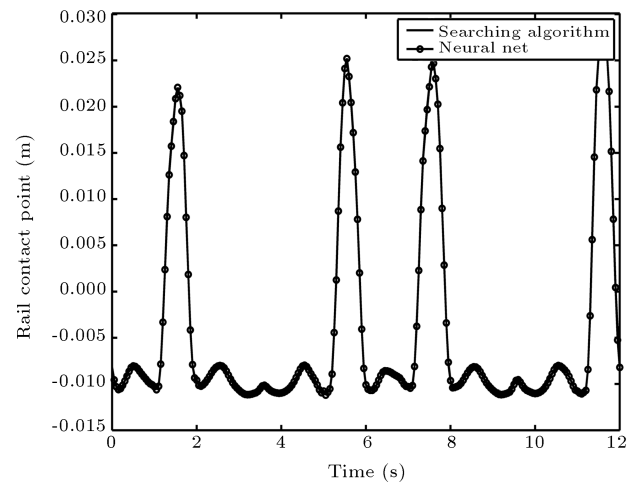


Figure 7. Contact point position in left rail.

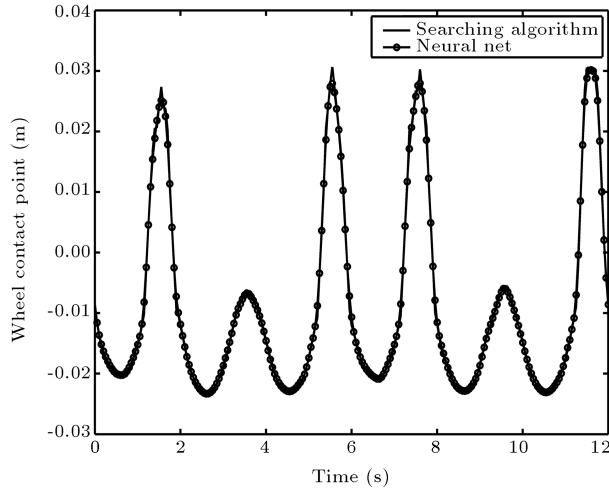


Figure 8. Contact point position in left wheel.

Vehicle Model

This work proposes a new method for studying the suspension components effects on the vehicle dynamic performances. Accordingly, real behaviors of all components will be described in this section. For this reason and due to the complexity of component behaviors, especially for air springs, describing complete vehicle dynamics with a single set of equations is almost impossible. So, the vehicle is modeled as a modular type and the internal forces of each component are calculated using a nonlinear description function and system states. A Newtonian approach is implemented for the dynamic modeling of different parts of the system and the interacting forces and moments between them are investigated. The vehicle model, with 42 Degrees Of Freedom (DOF), is shown in Figure 9

and some of its important specifications are listed in Table 1.

Air Spring Model

Air springs which are made of Carbon Black Filled Natural Rubber (CBFNR) have a long lifetime and isolate the vehicle body from unpredictable noise and vibration. Air spring behaviors are so complicated that they cannot be modeled simply by simple equations. Air spring response is independent of the excitation frequency [18], and it behaves like a stress relaxation function [19]. In addition, it has an asymmetric hysteresis loop, which is independent of the excitation frequency [20].

In this research work, the Berg model, which is validated by some experimental data up to 16 Hz frequencies [21], is used to simulate air spring dynamics. However, because of difficulties comprised from assigning previous turning point displacements to the variable, the frictional part of this model is replaced by the simplified viscoelastic model defined by Haupt and Sedlan. Schematic diagrams of the proposed modified models in vertical and lateral directions are shown in Figure 10. A complete explanation about this model is presented in [22].

According to Berg, and by taking into account the Haupt and Sedlan model, differential equations of air spring behavior in vertical, lateral and longitudinal directions are proposed as follows:

Vertical direction:

$$M\ddot{w}_s = K_{vz}(z - w_s) - C_{z\beta}|\dot{w}_s|^\beta \text{sign}(\dot{w}_s),$$

$$\beta = 1.8, \quad (1)$$

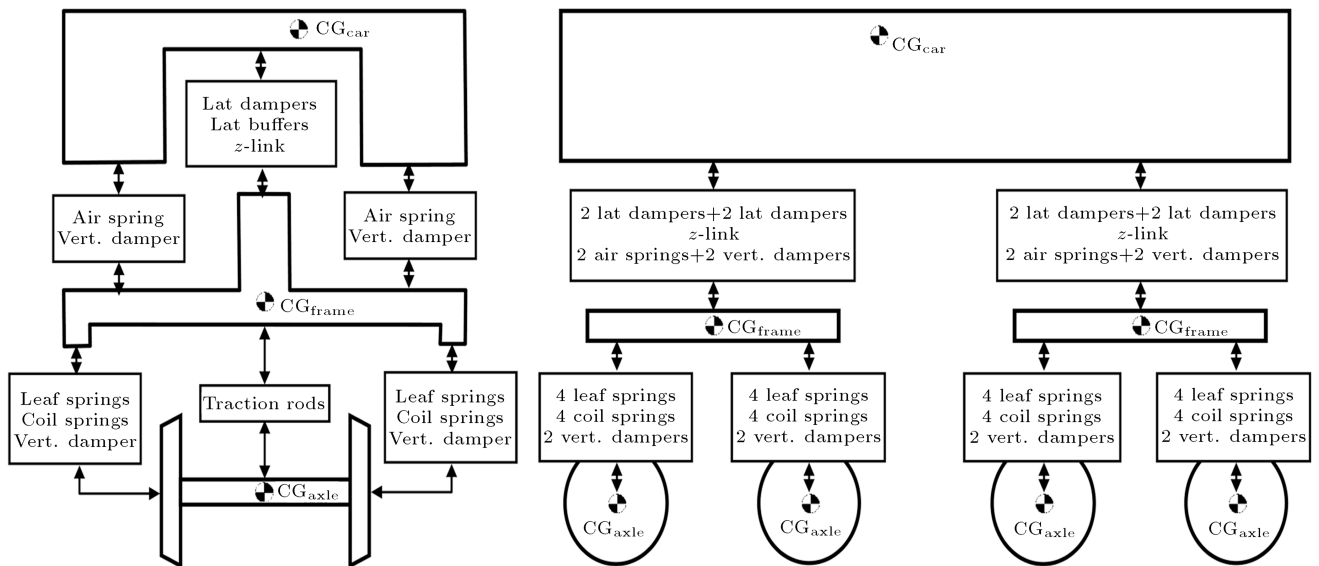
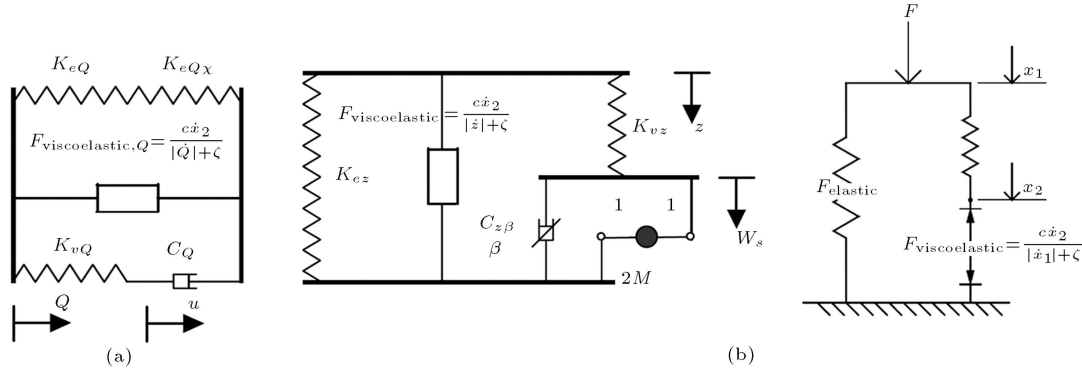


Figure 9. Vehicle model with 42 DOFs.

Table 1. Masses and inertia properties of the rail-vehicle.

Parameters	Values	Parameters	Values	Parameters	Values
m_{axle}	1747 kg	m_{frame}	2841.3 kg	m_{carbody}	33142 kg
$I_{xx,\text{axle}}$	1098 kgm ²	$I_{xx,\text{frame}}$	1030 kgm ²	$I_{xx,\text{carbody}}$	30000 kgm ²
$I_{yy,\text{axle}}$	191 kgm ²	$I_{yy,\text{frame}}$	1054 kgm ²	$I_{yy,\text{carbody}}$	687231 kgm ²
$I_{zz,\text{axle}}$	1098 kgm ²	$I_{zz,\text{frame}}$	2003 kgm ²	$I_{zz,\text{carbody}}$	687231 kgm ²

**Figure 10.** a) Proposed modified model for air spring; b) One-dimensional Haupt & Sedlan model.

$$F_z = (p_0 - p_a)A_e + K_{ez}z + K_{vz}(z - w_s) + \frac{c\dot{x}_2}{|\dot{z}| + \zeta}. \quad (2)$$

Lateral and longitudinal directions:

$$F_Q = K_{eQ}Q + K_{eQ\chi}\theta + F_{\text{visco-elastic},Q} + K_{vQ}(Q - u), \quad (3)$$

$$C_Q\dot{u} = K_{vQ}(Q - u), \quad (4)$$

and:

$$F_{\text{visco-elastic},Q} = \frac{c\dot{x}_2}{|\dot{Q}_1| + \zeta}, \quad Q = x, y. \quad (5)$$

The above parameters are estimated by the following equations [1]:

$$M = l_s A_s \rho \left(\frac{A_e}{A_s} \frac{V_{r0}}{V_{b0} + V_{r0}} \right)^2, \quad (6)$$

$$K_{ez} = \left(\frac{1}{\frac{p_0 A_e^2 n}{V_{b0} + V_{r0}} + p_g \frac{dA_e}{dz}} + \frac{1}{K_{\text{auxiliary}}} \right)^{-1}, \quad (7)$$

$$K_{vz} = \left(\frac{1}{\frac{p_0 A_e^2 n}{V_{b0}} + p_g \frac{dA_e}{dz}} + \frac{1}{K_{\text{auxiliary}}} \right)^{-1} - K_{ez}, \quad (8)$$

$$C_{z,\beta} = \frac{1}{2} \rho \cdot k_t \cdot A_s \left(\frac{A_e}{A_s} \frac{V_{r0}}{V_{b0} + V_{r0}} \right)^{1+\beta}, \quad (9)$$

$\beta = 2.$

The exact $C_{z,1.8}$ value is calculated based on the method introduced by Presthus, and the stiffness, $K_{eQ\chi}$, is approximated by $K_{eQ\chi} = 0.7(K_{eQ}h + \text{weight})$ [1]. The parameters in the above equations are described in Table 2.

To determine the exact values of the air spring parameters, a real test of the air spring was carried out for IRICo DMU air springs at the ContiTech Company in Germany. Figure 11 depicts the air spring test.

Based on the method introduced by Docquier et al. [23] and according to the real test data and simulation results, the loss angle, vertical dynamic stiffness and lateral hysteresis loop of the air spring are shown in Figures 12, 13 and 14. As can be concluded from these figures, a good agreement between the real

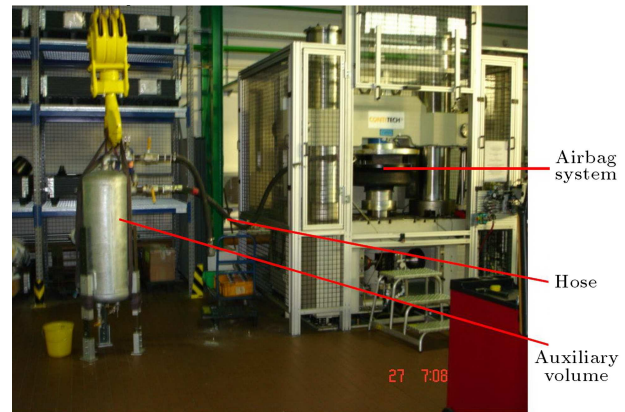
**Figure 11.** Test rig of air spring at ContiTech Company, Germany.

Table 2. Air spring parameters value.

Parameters	Description	Values
l_s	Connecting pipes' length	3.2 m
A_s	Connecting pipes' cross section	0.001359 m ²
A_e	Effective area of air spring	0.291 m ²
P_0	Initial absolute air spring pressure	3.806 bar
P_a	Air spring pressure	-
P_g	Gauge pressure	-
ρ	Density of air at P_0 pressure	4.523 kg/m ³
V_{r0}	Reservoir volume	0.04 m ³
V_{b0}	Air spring volume	0.064 m ³
k_t	Total loss coefficient of connection pipes	3.4727
$K_{\text{auxiliary}}$	Auxiliary spring stiffness in air spring	8234 kN/m
ζ	Constant coefficient	-
K_{eQ}, K_{eQx}	Elastic stiffness parameter	-
K_{vQ}	Viscous stiffness parameter	-
C_Q	Viscous linear damping parameter	-
θ	Vehicle roll angle	-
n	Polytropic index	-

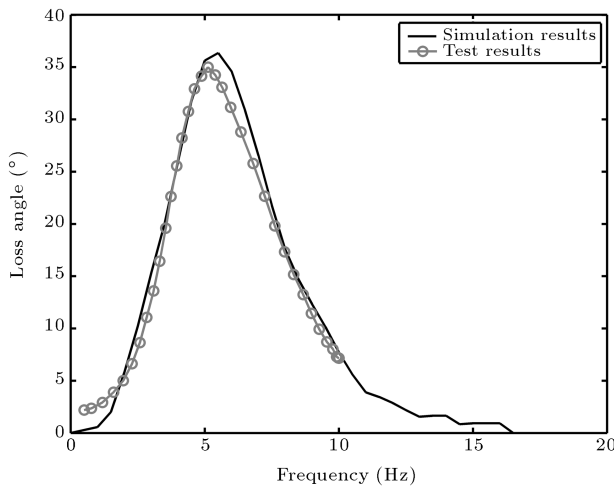


Figure 12. Loss angle of air spring, $M = 198.385$ kg, $K_{ez} = 461.629$ N/mm, $K_{vz} = 351.185$ N/mm, $C_{z,1.8} = 11.508$ kN(s/m)^{1.8}, $k_{\text{haupt}} = 96$ kN/m, $C_{\text{haupt}} = 1.52$ kNs/m, $\zeta = 0.00063$.

test and simulation results is achieved.

Dampers Model

Based on the experiments carried out by the SACHS Co. in Germany, the damping rates of the dampers are functions of velocity. The test results of IRICo DMU dampers are shown in Figure 15. The damping rate of each damper can be described by a bi-linear function; consequently:

$$F = a\dot{\Delta}_{\text{damper}} + b.$$

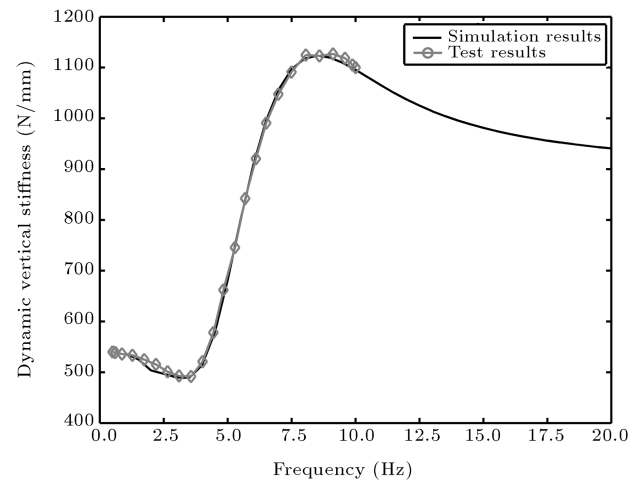


Figure 13. Dynamic vertical stiffness of the air spring, $M = 198.385$ kg, $K_{ez} = 461.629$ N/mm, $K_{vz} = 351.185$ N/mm, $C_{z,1.8} = 11.508$ kN(s/m)^{1.8}, $k_{\text{haupt}} = 96$ kN/m, $C_{\text{haupt}} = 1.52$ kNs/m, $\zeta = 0.00063$.

Parameters for different conditions and installation locations of dampers, based on the previous equation, are tabulated in Table 3.

Lateral Buffer Model

Lateral displacement of the carbody is restricted by four lateral buffers installed on each side of the bogie, as shown in Figure 16. Each lateral buffer has a primary compression force equal to 100 N. A force-displacement diagram of the lateral buffer, based on the experimental results, is shown in Figure 17. An air gap

Table 3. Dampers parameters.

Type of Damper	Condition	a	b
Primary vertical dampers	$0 \leq \dot{\Delta}_{\text{damper}} \leq 0.1$	40800	0
	$0.1 < \dot{\Delta}_{\text{damper}}$	19163	2163.7
	$\dot{\Delta}_{\text{damper}} < -0.1$	19025	-2177.5
Secondary vertical/lateral dampers	$0 \leq \dot{\Delta}_{\text{damper}} \leq 0.05$	61375	0
	$0.05 < \dot{\Delta}_{\text{damper}}$	18221	2157.7
	$\dot{\Delta}_{\text{damper}} < -0.05$	14143	-2361.6

of 17 mm between the carbody and the lateral buffer lets the carbody move freely in this range. Lateral displacement of the carbody over ± 40 mm is restricted by two stoppers installed on each bogie.

Based on the experimental results, the mechanical behavior of the lateral buffers is formulated as follows:

$$\begin{cases} F = 0 & 0 < x < 17 \\ F = 5 \times 10^{-5}(x - 17)^4 \\ \quad - 0.0019(x - 17)^3 \\ \quad + 0.0144(x - 17)^2 \\ \quad + 1.1525(x - 17) + 0.100 & 17 \leq x \leq 40 \end{cases} \quad (10)$$

Connection Link Model

Each center pivot is linked to the bogie frame by two connection links. Each link has a rubber bush at each end. Based on the experimental tests done by GMT Co. in Germany, the static stiffness of

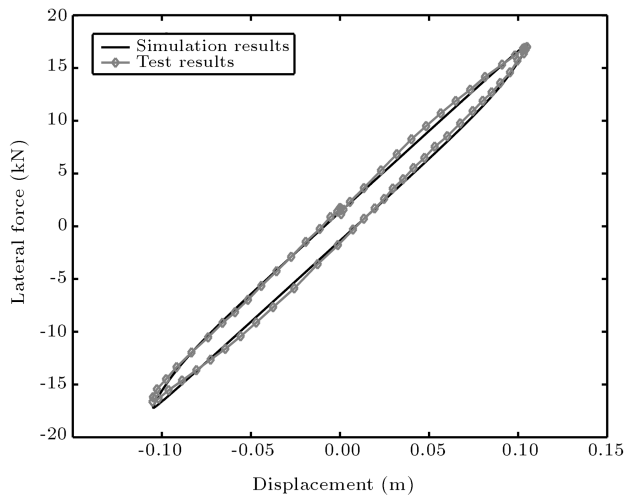


Figure 14. Lateral hysteresis loop of the air spring, $K_{ey} = 154$ kN/m, $K_{vy} = 82.26$ kN/m, $C_y = 1.109$ kNs/m, $k_{\text{haupt}} = 96$ kN/m, $C_{\text{haupt}} = 1.52$ kNs/m, $\zeta = 0.00063$.

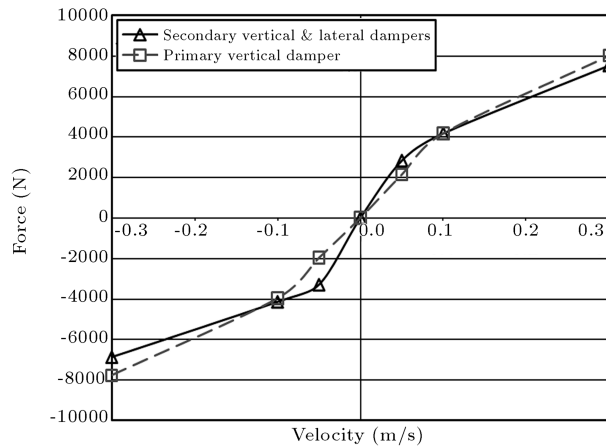


Figure 15. Test results of IRICo DMU dampers, done by SACHS, Germany.

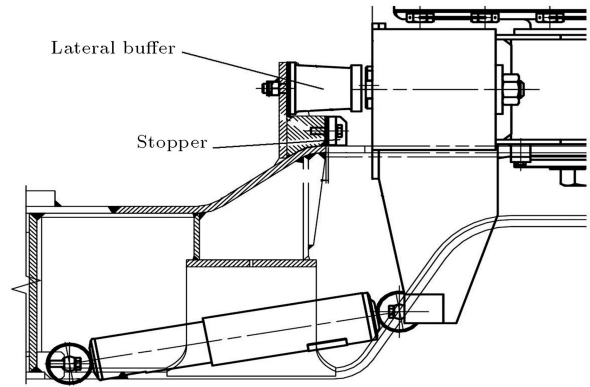


Figure 16. Lateral buffer position in bogie.

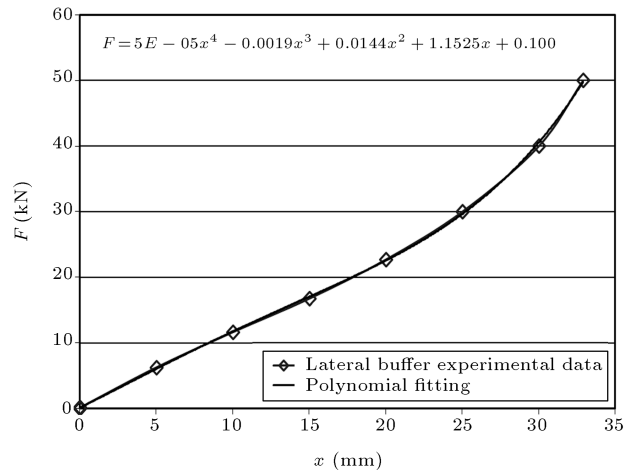


Figure 17. Lateral buffer stiffness.

the bush in radial, torsional and cardanic movements are calculated. Because these bushes are made from CBFNR rubber, for exact modeling of the connection link behavior in the vehicle dynamics, the simplified one-dimensional Haupt and Sedlan model is used for the radial movement of these bushes by the following function:

$$F_{\text{visco-elastic, radial}} = \frac{c\dot{x}_2}{|\dot{r}_1| + \zeta}. \quad (11)$$

The Haupt and Sedlan parameters defined in Table 4 are according to the Allen results [24].

Primary Suspension, Stiffness Coefficients Model

Four leaf springs in each axle direct the wheel-set during the curves. Each leaf spring has one rubber bush at each end. A schematic diagram of the leaf spring in the $x - z$ plane is shown in Figure 18. The same concept is developed for the other planes.

Considering the boundary condition of the leaf spring at two ends, the stiffness of the leaf spring in bending is determined as follows:

$$k_{\text{leaf Vertical}}^{-1} = \frac{1}{EI} \left[\frac{L^4 k_{ty, \text{bush}}}{4(Lk_{ty, \text{bush}} + EI)} - \frac{L^3}{3} \right]. \quad (12)$$

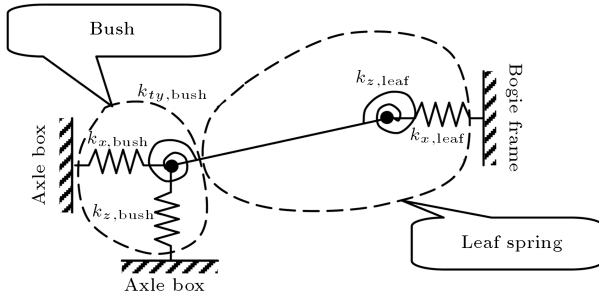


Figure 18. Leaf spring schematic diagram in $x - z$ plane.

Based on the manufacturer test reports, the coil spring stiffness coefficients are approximated linearly. By using the above equation, the stiffness coefficients of the primary suspension are determined as Table 5.

Complete Vehicle Model

All the components models used in the rail-vehicle dynamics with related internal DOFs are listed in Table 6. It can be seen that suspension components add 28 internal DOFs to the model. Whereas the vehicle masses have totally 42 DOFs, the complete rail-vehicle will be a model of 70 DOFs.

MODEL SIMULATION AND TEST RESULTS

A dynamic test of the IRICo DMU was performed on the Tehran-Ghazvin route to validate the performance of the proposed dynamic model. The VBOX III GPS, manufactured by Racelogic in the UK, is used to measure the exact position and speed of the train. The measured speed is monitored and recorded by one of the 16 channels of the CRONOS PL data acquisition system manufactured by IMC in Australia, which was connected to a portable computer. The accelerations of the trailer car masses are recorded by 15 remaining channels by means of accelerometers with the specifications presented in Table 7. These accelerometers are installed on two axles, two bogie frames and on the floor of the carbody inside the car according to Figures 19 and 20. As the track data and geometrical irregularities recorded by the EM120 machine are not up-to-date and were recorded 25 months before the dynamic test, by using track data a poor agreement between the test and simulation results is achieved. For this reason, the vehicle model

Table 4. Haupt and Sedlan model parameters and stiffness coefficients of bushes.

Parameters	Values	Parameters	Values
k_{haupt}	9 kN/mm	k_{radial}	2222.2 N/mm
C_{haupt}	3.8 kNs/m	k_{torsion}	31.25 Nm/°
ζ	0.00063	k_{cardanic}	45.525 Nm/°

Table 5. Primary suspension system, mechanical specifications.

Parameters	Values	Parameters	Values
$k_{x, \text{bush}}, k_{z, \text{bush}}$	45000 N/mm	$k_{ty, \text{bush}}$	32 Nm/°
$k_{y, \text{bush}}$	2118 N/mm	$k_{tx, \text{bush}}, k_{tz, \text{bush}}$	1659 Nm/°
$k_{x, \text{leaf spring}}$	44145 N/mm	$k_{x, \text{coil spring}}$	365.69 N/mm
$k_{y, \text{leaf spring}}$	2118 N/mm	$k_{y, \text{coil spring}}$	365.69 N/mm
$k_{z, \text{leaf spring}}$	71.9618 N/mm	$k_{z, \text{coil spring}}$	682.77 N/mm

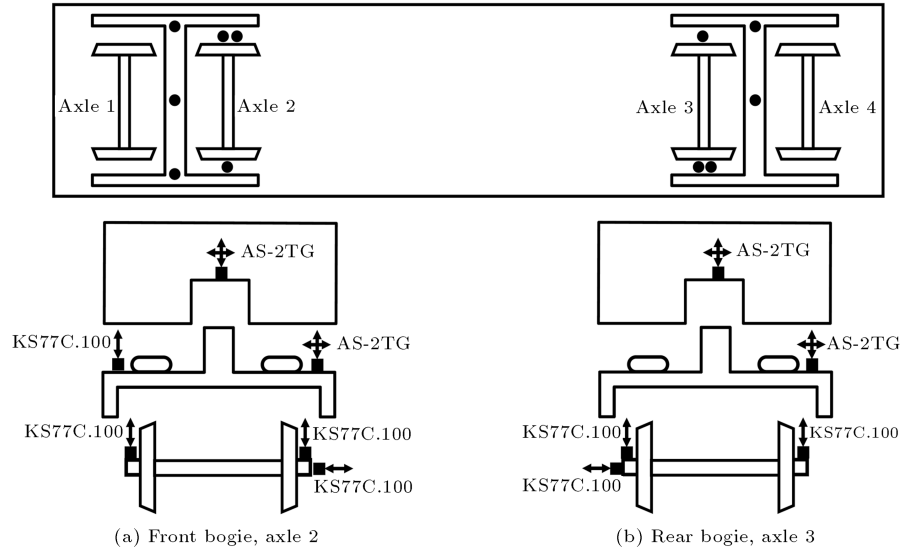


Figure 19. Accelerometers installation position.

Table 6. Suspension components of rail vehicle model (28 internal DOFs).

Components	No.	Identifier Equations		Internal DOF
Secondary vert. damper	2 per bogie	Bi-linear function		–
Secondary lat. damper	2 per bogie			
Primary vert. damper	2 per bogie			
Air spring	4 per car	Vertical	Nonlinear, Berg & Sedlan Model	x_2, w_s
		Lateral	Nonlinear, Berg & Sedlan Model	x_2, u
		Longitudinal	Nonlinear, Berg & Sedlan Model	x_2, u
Lateral buffer	4 per car	Nonlinear-polynomial, order 4		–
Bush-link	2 pairs per bogie	Nonlinear, Sedlan model		x_2
coil spring	4 per bogie	Linear in space, based on experimental data		–
Leaf spring	4 per bogie	Linear in space, based on analytical calculations		–

Table 7. Accelerometers specifications.

	KS77C.100	AS-2TG
Manufacturer	Manfred Weber, Germany	Kyowa, Japan
Type	ICP ® compatible	Strain gauge
Range (g: 9.81 m/s ²)	± 60 g	± 2 g

is studied in two different phases. In the first phase and to validate the track model, the model was executed by considering old track irregularities, and axle accelerations are investigated. In the second phase, measured axle accelerations and their integrals in the time domain are used as reference states to validate the complete vehicle model.

A complete explanation about model validation and test results are presented in [25]. The vehicle test results and the proposed vehicle model

have almost the same behavior in the frequency domain.

AIR SUSPENSION PARAMETER INFLUENCES

In this section, the influence of rail-vehicle secondary suspension parameters on the performance of the air spring including the air reservoir volume and connecting pipe lengths and diameters will be discussed.

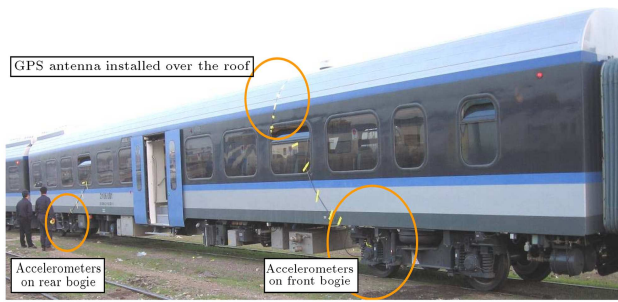


Figure 20. Sensors installation position on the trailer car.

Air Reservoir Volume

The behavior of air springs is highly affected by air reservoir volume. For studying the influences of air reservoir volume on the air suspension performances, the dynamic stiffness and loss angle of the air spring reaction force for various air reservoir volumes are studied within a 0 ~ 20 Hz frequency, as shown in Figures 21 and 22. As can be seen, by adjusting the air reservoir volume to zero, the air spring behaves as a simple stiffness component and the loss angle of reaction force is zero for all excitation frequencies. This fact indicates that no damping characteristics exist in the system when the air reservoir volume is zero.

Increasing air reservoir volume divides dynamic stiffness to low, transient and high frequency sections and appends damping to the system by creating a loss angle within a specific frequency range. By increasing air reservoir volume, maximum loss angle is occurred at lower frequencies with higher amplitude, as increasing the air reservoir volume from 5 to 80 liters, increases the maximum loss angle from 15.7° at 14.5 Hz to 48.1° at 4 Hz frequency.

By increasing air reservoir volume, the dynamic stiffness of the air suspension system is decreased at low

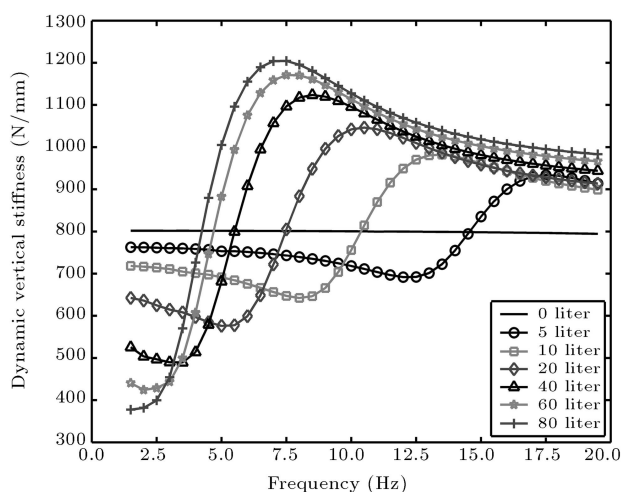


Figure 21. Dynamic vertical stiffness of air spring for various air reservoir volumes.

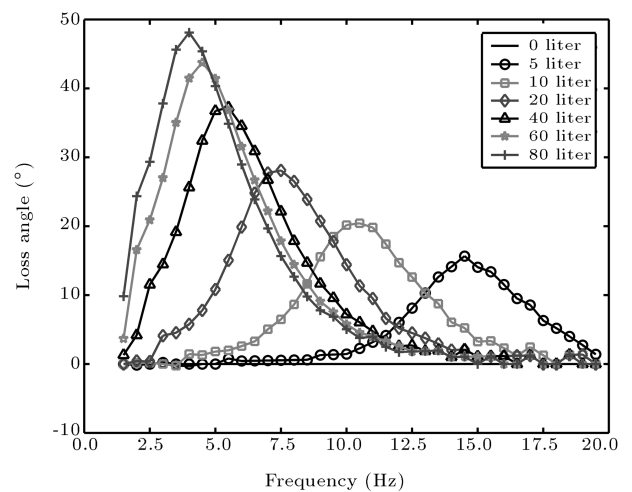


Figure 22. Loss angle of air spring for various air reservoir volumes.

frequencies and increased at high frequencies. In the same way, and as can be seen in Figure 22, a maximum loss angle is also occurred at lower frequencies by increasing air reservoir volume. So, by identifying the frequency resonance of the rail-vehicle, the transient section of the air spring response can be adjusted so that vehicle responses are restricted.

Connecting Pipe Diameter

As can be seen in Figure 23, by increasing the connecting pipe diameter, the transient section of the air spring dynamic stiffness is shifted to higher frequencies. The same concept is valid for the loss angle, as shown in Figure 24. By increasing the connecting pipe diameter, the maximum loss angle is shifted to higher frequencies with higher amplitude.

Whereas increasing air reservoir volume shifts the graphs to lower frequencies, increasing the connecting

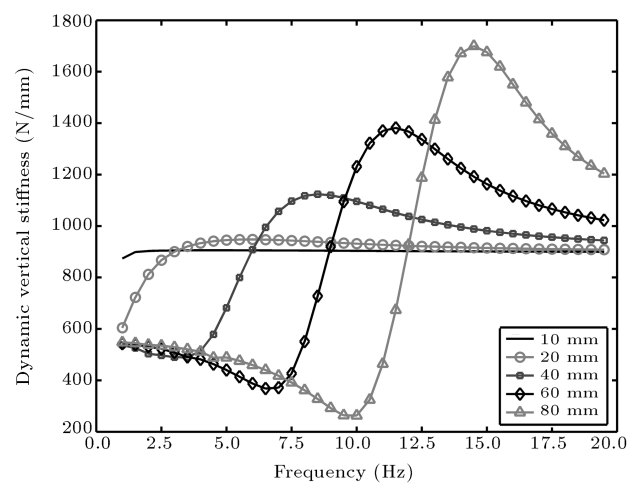


Figure 23. Dynamic vertical stiffness of air spring for various pipe diameters.

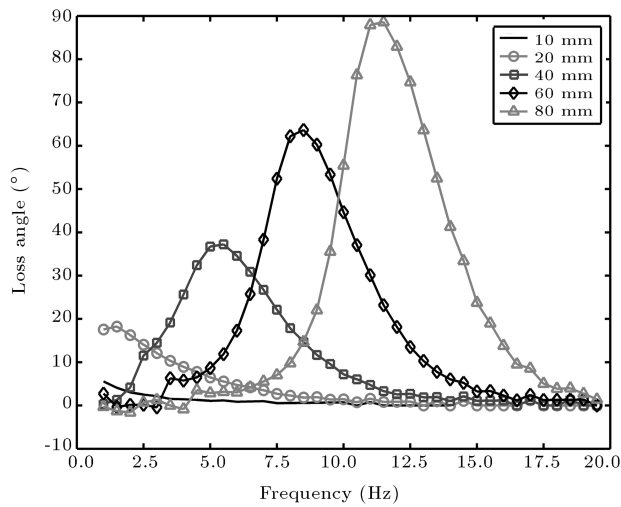


Figure 24. Loss angle of air spring for various pipe diameters.

pipe diameter shifts the graphs to the higher frequencies. Comparing reservoir volume and connecting pipe diameter influences, shows both factors add a damping characteristic to the system.

Connecting Pipe Length

Dynamic stiffness and the loss angle of the air spring for different connecting pipe lengths are shown in Figures 25 and 26. By decreasing the connecting pipe length, dynamic stiffness and loss angle graphs are widened. As can be seen, for all connecting pipe lengths, the low frequency section almost remained unchanged. Furthermore, the transient section is widened and the high frequency section is shifted to higher frequencies.

So, by adjusting the connecting pipe length, the damping characteristic of the air spring can cover a

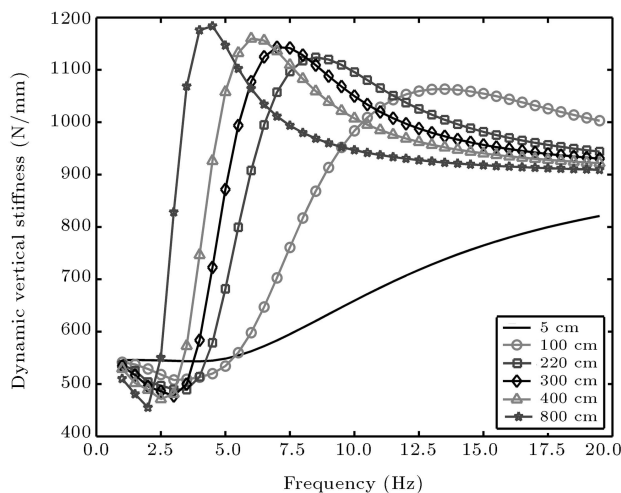


Figure 25. Dynamic vertical stiffness of air spring for various pipe lengths.

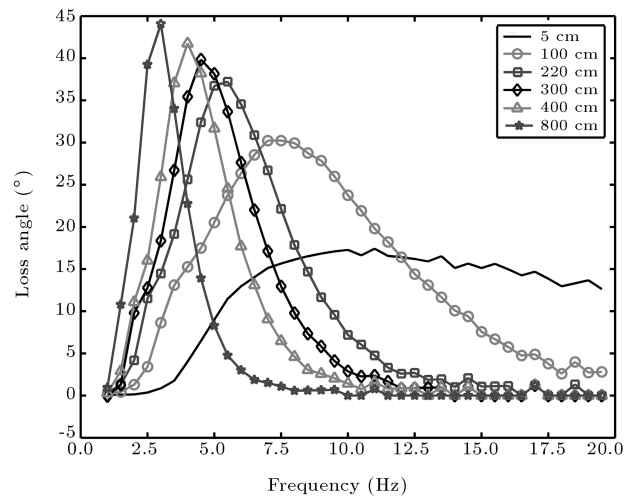


Figure 26. Loss angle of air spring for various pipe lengths.

wider range of excitation frequencies. By choosing a proper value for these three air suspension parameters, and based on the design requirements in rail-vehicles, obtaining satisfactory behavior from the system is possible.

PASSENGER COMFORT IMPROVEMENT

Engineering Approach

In this section, secondary suspension parameters of rail vehicles are tuned so that good ride comfort for passengers is guaranteed during their trip. Accordingly, performances of the air suspension system in rail vehicles are investigated. The Power Spectral Density (PSD) of the first axle, the front bogie frame and carbody vertical accelerations for real air suspension parameters are shown in Figure 27. As can be seen from this figure,

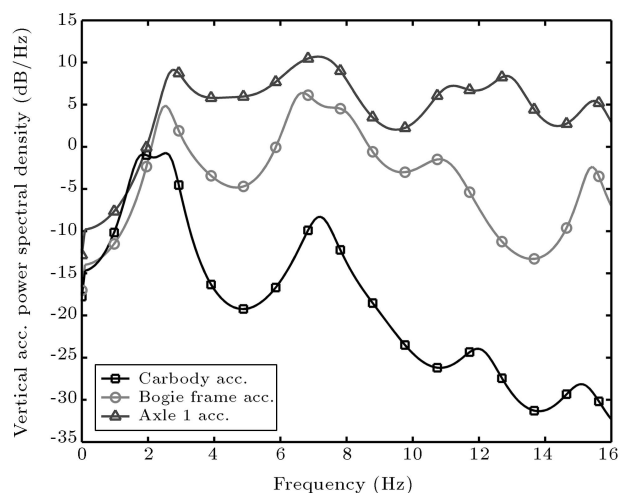


Figure 27. PSD diagram of first axle, front bogie frame and carbody vertical accelerations.

the carbody is excited strongly between 1.5 and 3 Hz frequencies.

For real air suspension parameters, the maximum loss angle appears around 6 Hz frequencies, as shown in Figure 28. Whereas secondary air suspension damps vertical excitations around 6 Hz frequencies, the carbody vertical acceleration has no significant excitation within this range. So, the capabilities of the air suspension system for damping carbody vertical vibrations in the modeled rail vehicle are not well employed. The carbody is excited with maximum amplitude within 1.5 and 3 Hz frequencies, as shown in Figure 27, which should be damped with a secondary air suspension system.

To reduce transmitted vertical vibrations to the carbody, the air suspension parameters of the modeled rail vehicle are tuned so that maximum loss angle occurs at lower frequencies, as shown in Figure 29. Based on the presented graphs in section 5, by increasing the air reservoir volume from 40 liters to 100 liters, increasing the connection pipe length from 2.2 m to 3.2 m and decreasing the connecting pipe diameter from 40 mm to 30 mm, the maximum loss angle is shifted to the lower frequencies. In this way, the maximum damping of the system occurs at lower frequencies where the carbody is excited with strong vibrations.

PSD diagrams of the first axle, the front bogie frame and the carbody vertical accelerations for two sets of air suspension parameters are shown in Figure 30. As can be seen from this figure, adjusting air suspension parameters reduced the transmitted vertical low frequency vibrations to the carbody.

For an evaluation of achieved improvement in the system performances, the Sperling ride comfort index [26] which is defined by the following equation is implemented:

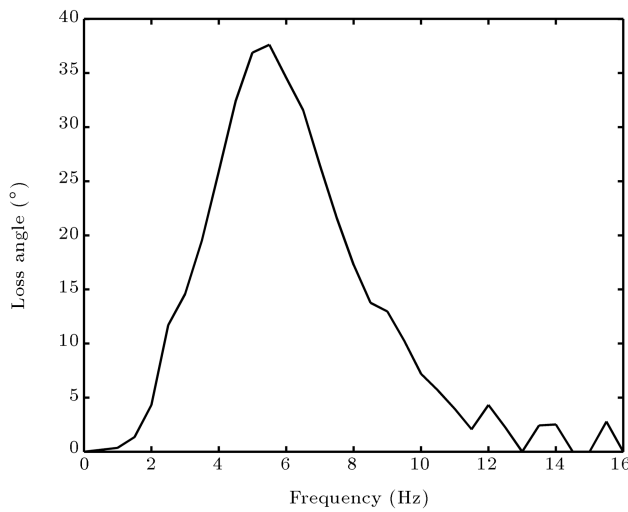


Figure 28. Loss angle of air spring response. $V_{\text{reservoir}} = 40 \text{ Lit}$, $L_{\text{pipe}} = 2200 \text{ mm}$, $D_{\text{pipe}} = 40 \text{ mm}$.

$$Wz = \left(\int_{0.5}^{30} a^3 B^3 df \right)^{1/10}. \quad (13)$$

In the above equation, f is frequency, a is acceleration in the frequency domain (cm/s^2) and B is defined by the following equations:

$$A = (1 - 0.277f^2)^2 + (1.563f - 0.0368f^3)^2, \quad (14)$$

$$B = 0.588 \left[\frac{1.911f^2 + (0.25f^2)^2}{A} \right]^{1/2}.$$

The Sperling ride comfort index is expressed according to Table 8.

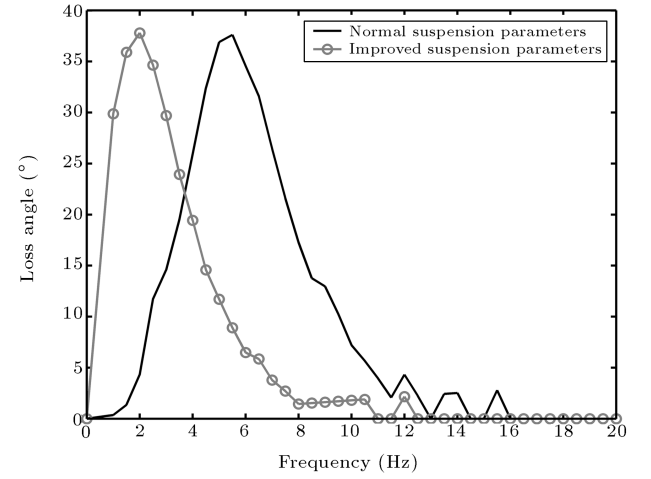


Figure 29. Loss angle of air spring response. $V_{\text{normal}} = 40 \text{ Lit}$, $L_{\text{normal}} = 2200 \text{ mm}$, $D_{\text{normal}} = 40 \text{ mm}$, $V_{\text{improved}} = 100 \text{ Lit}$, $L_{\text{improved}} = 3200 \text{ mm}$, $D_{\text{improved}} = 30 \text{ mm}$.

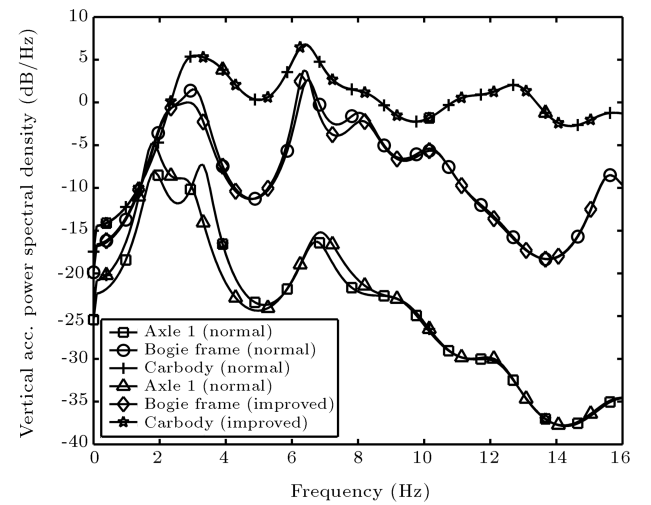


Figure 30. PSD diagram. $V_{\text{normal}} = 40 \text{ Lit}$, $L_{\text{normal}} = 2200 \text{ mm}$, $D_{\text{normal}} = 40 \text{ mm}$, $V_{\text{improved}} = 100 \text{ Lit}$, $L_{\text{improved}} = 3200 \text{ mm}$, $D_{\text{improved}} = 30 \text{ mm}$.

Table 8. Sperling ride comfort index (W_z).

Subjective Ride Comfort	W_z
Just noticeable	1
Clearly noticeable	2
More pronounced but not unpleasant	2.5
Strong, irregular but still tolerable	3
Very irregular	3.25
Extremely irregular, unpleasant, annoying, prolonged exposure intolerable	3.5
Extremely unpleasant, prolonged exposure harmful	4

Simulation results for two sets of air suspension parameters are shown in Figure 31. As can be seen, by simply modifying air suspension parameters, passengers comfort is improved and the ride comfort index is reduced about 10%.

Genetic Algorithm Optimization Approach

Whereas only the ride comfort index is available for evaluating suspension performances, based on the complexity of rail vehicle dynamic equations, the Genetic Algorithm approach is utilized to optimize air suspension parameters and improve the ride comfort of passengers [27,28]. To do this, a searching range, as defined in Table 9, is defined for suspension parameters.

Based on the above parameter variations, one string with 20 bits can cover the searching range very well. Dedicated bits for each parameter in the individuals are defined in Figure 32. In this way, the searching range is divided into 127 sections for the

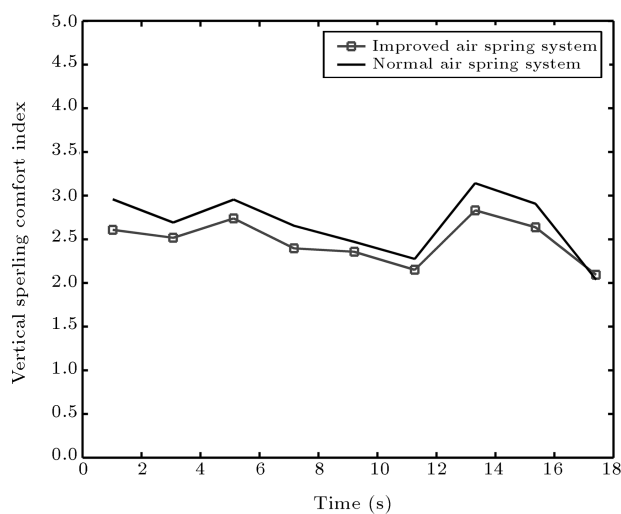


Figure 31. Sperling ride comfort index variation in Tehran-Ghazvin route (straight section) with 80 km/h speed. $V_{\text{normal}} = 40$ Lit, $L_{\text{normal}} = 2200$ mm, $D_{\text{normal}} = 40$ mm, $V_{\text{improved}} = 100$ Lit, $L_{\text{improved}} = 3200$ mm, $D_{\text{improved}} = 30$ mm.

connecting pipe lengths and reservoir volume and into 63 sections for the connecting pipe diameters.

The equivalent binary code for each section and the real value of parameters are related to each other, according to the following equations:

$$d_s = 20 + n_{ds} \times 60/63, \quad (15)$$

$$l_s = 40 + n_{ls} \times 760/127, \quad (16)$$

$$V_{r0} = 10 + n_{V_{ro}} \times 190/127. \quad (17)$$

As reaching to a ride comfort index equal to 1.5 in trains with a passive suspension is almost impossible, and in order to highlight the lower ride comfort indexes, a fitness function which should be maximized is introduced based on the following equation. By finding the best configuration of suspension parameters where the following function has its maximum value, passenger comfort will be improved during their trip.

$$\text{Fitness Function} = \frac{1}{W_z - 1.5}. \quad (18)$$

Because three different variants are defined by a single individual and for speeding up the convergence of GA and improving optimization performances, the following statements are assumed [29]:

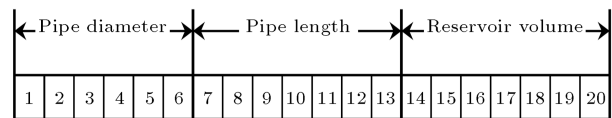


Figure 32. Dedicated bits for each parameter in individuals.

Table 9. Searching space.

Parameters	Values	Variation Range
d_s -pipe diameter (mm)	$20 \leq d_s \leq 80$	60
l_s -pipe length (mm)	$40 \leq l_s \leq 800$	760
V_{r0} -reservoir volume (liter)	$10 \leq V_{r0} \leq 200$	190

1. Two points cross over and bitwise inversion mutation methods are used.
2. The observed best-fitted individual is selected for the next generation without any random experiment to avoid the best-fitted individual dying (elitism).

Six populations of the first generation containing as-built suspension parameters of an IRICo rail vehicle are shown in Table 10.

The mean value of the fitness function and the Sperling ride comfort index of the best individuals for

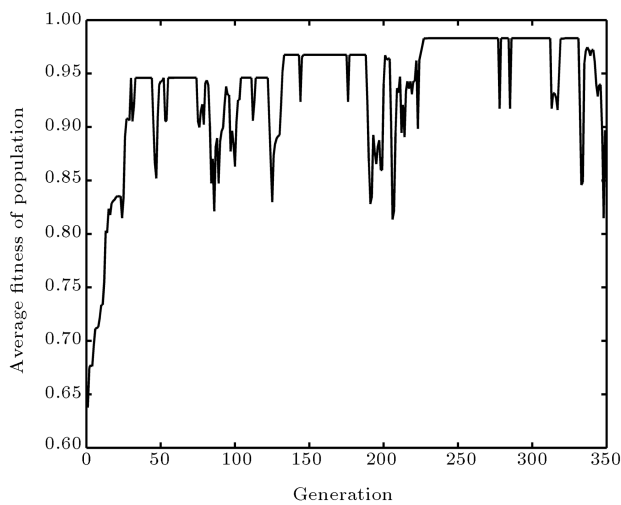


Figure 33. Mean value of fitness function during generation.

400 generations with a defined algorithm are shown in Figures 33 and 34. Drops in Figure 31 are related to the times when the mutation operator acts on the population. As can be seen from Figure 32, after 400 generations, the ride comfort index is reduced from $W_z = 3.131$ to $W_z = 2.517$. 400th generation individuals are presented in Table 11.

PSD diagrams of the front bogie frame and carbody vertical accelerations for three different air suspension configurations including initial, improved (engineering approach) and optimized are shown in Figure 35. As is clear, the transmitted vibration to

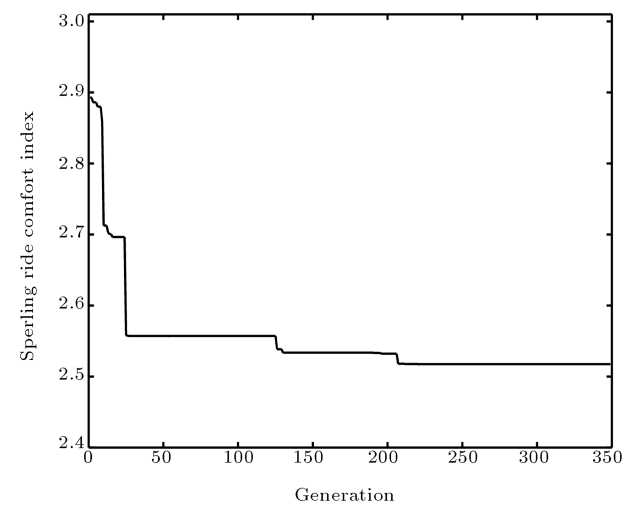


Figure 34. Sperling ride comfort index of best individuals during 400 generations.

Table 10. Initial population.

No	Pipe Diameter (mm)	Pipe Length (cm)	Reservoir Volume (liter)	Binary Individual	Fitness
1	40	220	40	010101—0011110—0100001	0.6131
2	68	648	55	110010—1100110—0011110	0.6653
3	20	200	80	000000—0011011—0101111	0.6209
4	56	344	30	100110—0110011—0001101	0.5760
5	32	192	48	001101—0011001—0011001	0.6414
6	80	40	10	111111—0000000—0000000	0.5041

Table 11. 400th generation individuals.

No	Binary Individual	Pipe Diameter (mm)	Pipe Length (cm)	Reservoir Volume (liter)	Fitness
1	011010—1100110—1101111	44.8	650.4	176.1	0.9833
2	011010—1100110—1101111	44.8	650.4	176.1	0.9833
3	100101—0011000—1100000	55.2	183.6	153.6	0.8331
4	011010—1100111—0011111	44.8	656.4	54.4	0.7073
5	011010—1100110—1101111	44.8	650.4	176.1	0.9833
6	011010—1100110—1101111	44.8	650.4	176.1	0.9833

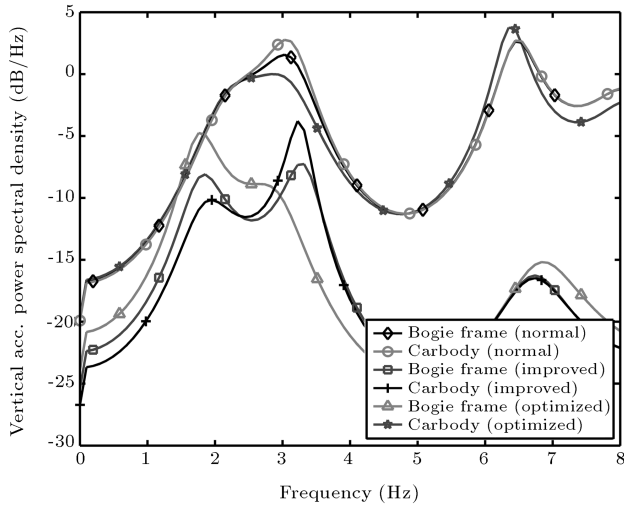


Figure 35. PSD diagrams of front bogie frame and carbody vertical accelerations. $V_{\text{normal}} = 40$ Lit, $L_{\text{normal}} = 2200$ mm, $D_{\text{normal}} = 40$ mm, $V_{\text{improved}} = 100$ Lit, $L_{\text{improved}} = 3200$ mm, $D_{\text{improved}} = 30$ mm, $V_{\text{optimized}} = 176.1$ Lit, $L_{\text{optimized}} = 6504$ mm, $D_{\text{optimized}} = 44.8$ mm.

the carbody is reduced around 2 Hz frequencies and is increased around 3 Hz frequencies to improve the ride comfort of passengers during their trip.

Sperling ride comfort index variations for the three above mentioned air suspension configurations are shown in Figure 36. As can be seen and by comparing with the improved one, an optimized suspension configuration slightly improves the ride comfort of passengers. Loss angle diagrams of the

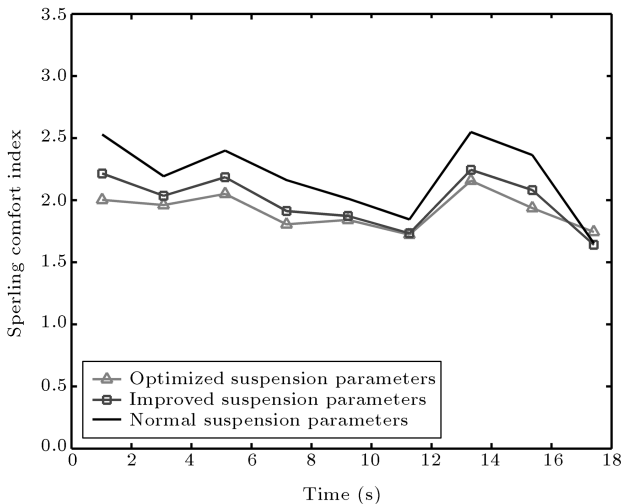


Figure 36. Sperling ride comfort index for Tehran-Ghazvin route (straight section) with 80 km/h speed. $V_{\text{normal}} = 40$ Lit, $L_{\text{normal}} = 2200$ mm, $D_{\text{normal}} = 40$ mm, $V_{\text{improved}} = 100$ Lit, $L_{\text{improved}} = 3200$ mm, $D_{\text{improved}} = 30$ mm, $V_{\text{optimized}} = 176.1$ Lit, $L_{\text{optimized}} = 6504$ mm, $D_{\text{optimized}} = 44.8$ mm.

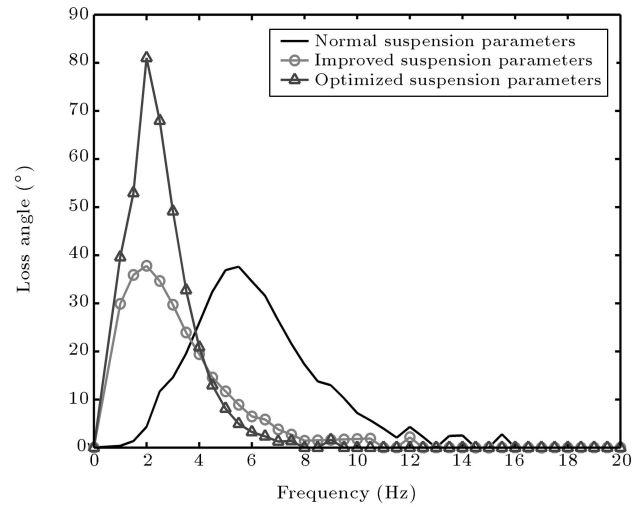


Figure 37. Loss angle of air spring response. $V_{\text{normal}} = 40$ Lit, $L_{\text{normal}} = 2200$ mm, $D_{\text{normal}} = 40$ mm, $V_{\text{improved}} = 100$ Lit, $L_{\text{improved}} = 3200$ mm, $D_{\text{improved}} = 30$ mm, $V_{\text{optimized}} = 176.1$ Lit, $L_{\text{optimized}} = 6504$ mm, $D_{\text{optimized}} = 44.8$ mm.

air suspension system for three different suspension configurations are shown in Figure 37. As is clear, and without any forcing policy to the optimization algorithm, a 2 Hz frequency is achieved for best suspension performances, which is in agreement with engineering technique results. Both improved and optimized configurations focused on a 2 Hz frequency as the objective value where transmitted vibrations to the carbody should be reduced. Meanwhile, for improving ride comfort and for better vibration isolating, the optimized configuration tried to increase the loss angle by around 2 Hz frequencies.

CONCLUSION

This research work proposes a new model for studying the influences of suspension component behavior on rail vehicle dynamics. To do this, a complete track-vehicle model with 70 degrees of freedom is addressed as a modular type. As the air springs are a very important isolating component in rail vehicles, a complete nonlinear air spring model, taking into consideration thermo-dynamical effects, is developed. The model coefficients are tuned, based on real test data, and can be easily used in the dynamic modeling of air springs. The effects of air reservoir volume and connecting pipe length and diameter on system performances are investigated. By using the achieved illustrative graphs about air suspension parameter influences on system performances, the parameters are tuned so that transmitted vibration to the passengers is reduced. To optimize air suspension behavior, a GA optimization approach is investigated. By using optimized suspension parameters and comparing them with

an engineering approach, suspension performances are slightly improved. Simulation results showed that in an optimization approach, the objective frequency range is well matched with the engineering one, which approves the engineering technique results. Whereas the ride comfort index in an optimized configuration is slightly improved with respect to the engineering one, and by comparing connecting pipe lengths and reservoir volume in engineering and optimization technique configurations, it is concluded that, because of space limitation under the carbody, engineering approach results are more practical.

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APPENDIX

Neural network activation functions and synaptic weights of the trained network for the contact parameters are as follows:

$$\mathbf{y} = f(\mathbf{W} \cdot \mathbf{x} + \mathbf{b}),$$

$$f_{\text{hidden layers}}(x) = \frac{2}{1 + e^{-2x}} - 1,$$

$$f_{\text{Output layer}}(x) = x,$$

$$W_1 = \begin{bmatrix} 0.4879 & -143.29 & 153.58 & 296.96 \\ 440.24 & -1618.3 & -922.66 & -1334.3 \\ 31.414 & 61.038 & 58.806 & 286.63 \\ -24.646 & -3.7075 & -68.557 & 168.59 \\ -90.293 & 204.3 & 141.32 & 326.46 \\ -25.573 & -141.03 & -86.023 & -164.47 \\ -74.14 & 5.2036 & -304.34 & -42.909 \\ 19.799 & 90.972 & 249.49 & -199.13 \\ -29.727 & -80.828 & -67.982 & -241.81 \end{bmatrix},$$

$$b_1 = \begin{bmatrix} -1.7844 \\ -343.02 \\ -21.85 \\ 18.173 \\ 71.443 \\ 16.336 \\ 58.029 \\ -11.211 \\ 20.332 \end{bmatrix},$$

$$W_2 = \begin{bmatrix} 3.8078 & 0.34866 & -81.023 & 13.621 \\ -23.264 & -6.9369 & -340.7 & -27.582 \\ 3.8083 & 0.34666 & -80.939 & 13.612 \\ -1.464 & -0.20806 & -51.506 & 3.1561 \\ -0.76366 & -0.01884 & -9.4441 & -0.49465 \end{bmatrix}$$

$$\begin{bmatrix} -0.1302 & 28.59 & -1.6923 & 10.078 & -107.8 \\ 6.3036 & 210.98 & 10.203 & -1.4993 & -556.53 \\ -0.12951 & 28.561 & -1.6881 & 10.077 & -107.69 \\ -2.0031 & 23.906 & -3.4124 & 3.1543 & -76.946 \\ -0.0555 & 6.2466 & -0.25575 & 0.15104 & -16.544 \end{bmatrix},$$

$$b_2 = \begin{bmatrix} -4.2292 \\ -2.5291 \\ -4.2332 \\ -1.0352 \\ 0.5185 \end{bmatrix},$$

$$W_3 = \begin{bmatrix} 7.73 & 0.003244 & -7.7259 \\ 7.5261 & 0.003135 & -7.5231 \\ 0.022746 & -0.00677 \\ 0.021734 & 0.020275 \end{bmatrix},$$

$$b_3 = \begin{bmatrix} 0.014172 \\ -0.00269 \end{bmatrix}.$$