Development of Oil Damper Test Equipment Capable of Simulating the Actual Conditions of Railway Vehicles

1. Introduction

Railway vehicles are equipped with multiple oil dampers, whose characteristics have a great effect on riding comfort for passengers as well as on the safety of vehicles. The technical specifications of these dampers are therefore essential in vehicle design. We have developed test equipment to allow estimation of oil damper characteristics without the need to perform field tests, making it possible to operate an oil damper three-dimensionally to simulate the actual running conditions of railway vehicles. This paper gives an outline and describes the specifications of the test equipment and the automatic identification of oil damper characteristics.

**Keywords:** damper, test equipment, HILS, iteration, neural network

Table 1 Damper specifications applicable to the damper test equipment

<table>
<thead>
<tr>
<th>Damper Type</th>
<th>Max. force (kN)</th>
<th>Max. speed (m/s)</th>
<th>Max. test frequency (Hz)</th>
<th>Stroke (mm)</th>
<th>Direction</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lateral damper</td>
<td>15</td>
<td>0.2</td>
<td>15</td>
<td>20 200 80</td>
<td>Y</td>
</tr>
<tr>
<td>Axle damper</td>
<td>8</td>
<td>0.3</td>
<td>30</td>
<td>20 10 100</td>
<td>Z</td>
</tr>
<tr>
<td>Anti-yaw damper between cars</td>
<td>22</td>
<td>0.2</td>
<td>10</td>
<td>200 80</td>
<td>X</td>
</tr>
<tr>
<td>Anti-yaw damper</td>
<td>15</td>
<td>0.2</td>
<td>30</td>
<td>100 200 80</td>
<td>X</td>
</tr>
<tr>
<td>Anti-rolling damper</td>
<td>10</td>
<td>0.1</td>
<td>10</td>
<td>200 160</td>
<td>Z</td>
</tr>
<tr>
<td>Anti-tilting damper</td>
<td>5</td>
<td>0.3</td>
<td>2</td>
<td>300 20</td>
<td>Y</td>
</tr>
</tbody>
</table>
2. Outline of the damper test equipment

The specifications of this damper equipment were set to allow installation of any damper applied to railway vehicles. Table 1 shows the damper specifications applicable to the equipment, which allows installation of test dampers in two directions, i.e. horizontally and vertically. There are three acting directions in actual dampers, longitudinal (X), lateral (Y) and vertical (Z). Accordingly, axis transform may be necessary in test execution. Figure 1 shows the damper equipment developed.

2.2 Measurement device

To measure three-dimensional forces and moment, we have chosen a six-component force transducer making use of a strain gage, which is installed between a damper bracket and the equipment frame. The rated capacity in the principal direction is up to 20 kN.

2.3 Controller

The equipment controller performs the two roles of motion order to the motion system and data acquisition from the six-component force transducer. To cooperate...
with other computer simulators in HILS, it can be connected to an optical real-time network.

3. Performance improvement of the test equipment

3.1 Introduction

It is difficult for actual motion movement to be realized by the motion order without compensation in each time step. The motion system controls the top board by adjusting the actuator length, and some dynamical error exists between the motion order and the actual movement because of the motion transfer characteristics. In addition, static error may also exist because of the tolerance manufacturing and setting stage (Fig. 4).

Real-time simulation is executed in collaboration with the test equipment during HILS testing. It may be a critical issue from the standpoint of simulation accuracy and stability for the system to control the damper position properly in each time step without delay, otherwise the simulation may receive an erroneous response (in terms of forces) from the transducer.

The performance improvement test was executed as follows:

1. Six actuators’ length of the motion system was measured at the static movement order. The top-board movement without static error was then estimated from these lengths (referred to below as the estimated movement). If it agrees with the motion order, we can also evaluate the estimated movement dynamically in each time step.

2. The top-board movement including static error was measured using theodolites at the static movement order (referred to below as the actual movement), and the relationship between the actual movement and the estimated movement (equal to the motion order from step (1)) was acquired. We can statically adjust the actual movement using this relationship.

3. The characteristics of motion were acquired by measuring the actuators’ length during dynamic operation, and the compensator was considered. The actual movement can also be adjusted dynamically together with step (2) using a compensator of appropriate design.

3.2 Estimation of top-board movement from the actuators’ length

The motion system controls the top-board movement by adjusting the length of each actuator that is set without tolerance as described above.

If we assume that the coordinates of the upper connecting point of the \( i \)-th actuator are given as:

\[
(x_{Ci}(t), y_{Ci}(t), z_{Ci}(t))
\]

Then, the following equation can be obtained:

\[
P_{Ci}(t) = P_{Bi}(t) + (x(t), y(t), z(t))
\]

where: \( P_{Bi}(t) \) = estimated movement, translation and roll-pitch-yaw angle.

The squared length of the \( i \)-th actuator \( L_{i}(t) \) is obtained by:

\[
L_{i}(t) = (x_{B_i} - x_{A_i})^2 + (y_{B_i} - y_{A_i})^2 + (z_{B_i} - z_{A_i})^2
\]

where: \( x_{B_i}, y_{B_i}, z_{B_i} \) are the coordinates at the fixed bottom connecting point of the \( i \)-th actuator.

3.3 Measurement of actual top-board movement using theodolites

Theoretically, the coordinates of the fixed measured points on the top board can be calculated by measuring the angles of direction and elevation of these points using two theodolites set at a known distance from each other. In reality though, the accuracy of such calculation results is not high because the measured angles introduce a certain level of error. The coordinates are therefore calculated using the following method:

1. The coordinates at the foot of a perpendicular line from the measured point to the horizontal plane (including the first theodolite) are calculated using the angles of direction and the distance between the two theodolites. This foot point is referred to as \( H \).

2. The coordinates of the fixed measured point are calculated on the local coordinate system (whose origin is the first theodolite) from the angle of elevation of the first theodolite and the distance between the first theodolite and \( H \).

3. The coordinates of the second theodolite are calculated on the same local coordinate system as above from the angle of elevation of the second theodolite and the distance between the second theodolite and \( H \).

Measurement errors appear as a dispersion of the second theodolite’s coordinates, so we can check the dispersion of the results to confirm the validity of measurement.

Figure 5 shows actual movement error at static motion orders up to \( \pm 150 \) mm (in the X, Y and Z directions). These errors change linearly against the motion order. The actual movement can therefore be thought of as statically equal to the motion order with adjustment using these regression curves.
3.4 Acquisition of motion characteristics and consideration of the compensator

3.4.1 Transfer function of the motion system

The transfer function of the motion system is calculated from the ratio of the input-output cross spectrum to the input spectrum. Figure 6 shows the transfer function of the motion system calculated from random wave inputs along the X-axis direction, where the half amplitude is up to 15 mm and the frequency is up to 3 Hz.

3.4.2 Consideration of the compensator

There are two points of difficulty in designing a compensator applicable to HILS testing: the compensator cannot use future inputs or preparative test results, because it receives an order at the very moment when the real-time simulator finishes calculating at each time step in HILS testing.

We designed the compensator to flatten the transfer characteristic of the motion system using the Yule-Walker technique, which is an IIR filter design method. The Yule-Walker technique compensates only the gain characteristic, and does not always compensate the phase characteristic. However, the phase characteristic can generally be made flat by using another appropriate compensator with it. Basically, the compensator is represented by the inverse transfer function of the motion system. At this time, the calculated transfer function is filtered because...
of noise from the system, and gain characteristic beyond 30 Hz is intentionally cut off. Figure 7 shows the compensating function and the expected transfer function of the motion system after being compensated.

Figure 8 shows a time history of the motion results by compensated orders. Improvement of the phase characteristic over that of the gain characteristic is not seen, indicating time delays in the software that controls the motion system. Improvement of this software is therefore needed.

4. Automatic identification of oil damper characteristics

4.1 Aim of automatic identification

Although a virtual field-test environment enables actual devices to be combined with computer simulations for hybrid tests, it is not a realistic test method to install all devices in test equipment on board a vehicle. Most of these devices are therefore represented by computer simulation models.

Conventional modeling methodology does not provide highly accurate models in a wide range of simulation conditions for strongly non-linear devices such as dampers. This is because the methodology generates more characteristic-comprehensive models with fewer parameters using linearization and approximation in restricted coverage.

In automatic identification of damper characteristics, automatic generation of simulation models aims at a three-step development. The first is automatic generation of the actuating pattern for the damper test equipment from parameters such as stroke and designed damping force diagrams, which are shown in drawings. The second step involves data acquisition during the test with this pattern, and the final one consists of automatic generation of a simulation model. We have focused only on the relationship between input (damper displacement and velocity) and output (damper force) without paying attention to modeling a detailed damper interior structure.

4.2 Neural network

A neural network [2, 3] is a non-parametrical identification method, and is applied to automatic identification of damper specifications. Such networks constitute a parallel-distributed processing model developed by simulating simplified neural processing in the brain. The network is represented by connections between artificial neurons (Fig. 9), each of which has input weights and bias as its parameters. Its output to the adjacent neurons is calculated from the weighted sum of inputs with its own predefined function, and the final output on the output layer is the network output. A neural network automatically adjusts input weights and biases to minimize the error of the desired input-output pattern presented (known as the training sample). One of the most popular techniques for adjustment is back propagation, and this process is referred to as training.

Two types of neural network (the multi-layer feedforward type and the multi-layer recurrent type) are considered for application to identification. A recurrent neural network contains feedback loops, and enables training to be established with fewer neurons in time-series-dependent systems. However, the feedback loops may cause undesirable output to continue once it occurs, and in the worst case diffused output may cause a fatal result. In contrast, a feedforward neural network needs many neurons for training in time-series-dependent systems, and avoids the propagation of undesirable output if it occurs.

We have established automatic identification with a multi-layer feedforward neural network using the damper displacement and speed as the network inputs and the force of the damper as the output. To improve the performance of the feedforward network, we have also used a simplified dynamics model that calculates the strain of the elastic devices installed at both ends of the damper as pre-work for preparing the network input (Fig. 10).

4.3 Identification of the principal direction

4.3.1 Identification experiment procedure

To confirm the validity of the identification method described above, we used a lateral damper installed on a Shinkansen bogie in an identification experiment of the principal direction. Two different types of sample were prepared (i.e. training and verifying samples), and the damper test equipment was used to certify that output from the trained neural network was equal to that of the verifying samples.

4.3.2 Actuating data

The trained neural network gives highly accurate output if the input data are within the training input range. However, if the data are outside this range, the

\[
y = f \left( \sum_{i} w_i x_i + b \right)
\]

\[
y : \text{Output} \\
f : \text{Transfer function} \\
w_i : \text{Input weight} \\
x_i : \text{Input} \\
b : \text{Bias}
\]
network often outputs very low-accuracy values because it recognizes these inputs as unknown patterns. It is therefore important that the range of training input data includes the data ranges used by the trained network. For this reason, actuating data to make training samples were prepared with a five-fold amplitude of the damper stroke obtained in a field test, and a sine wave (with a frequency of 0.05 Hz and a peak amplitude of 10 mm) was applied to the data. The displacement and velocity distribution of the actuating data includes that of a field test on a Shinkansen line, as shown in Fig. 11.

The actuating data for verification are non-correlative with those for training, and consist of measurements taken in field testing on a meter-gauged line. These are different from both the training data and the data measured in a field test on a Shinkansen line, as shown in Fig. 12.

The actuating data for verification are non-correlative with those for training, and consist of measurements taken in field testing on a meter-gauged line. These are different from both the training data and the data measured in a field test on a Shinkansen line, as shown in Fig. 12.

4.3.3 Identification results

We confirmed the extent to which the calculated outputs of the trained network (Fig. 10) will coincide with the measured force of the actual damper using test equipment and verifying data.

Figure 13 shows the results of the calculated outputs of the trained network. To confirm its effect, the calculated outputs of an ARX model (a popular parametrical identification method) are also shown in Fig. 13. These outputs are different from the measured data, especially at the peaks. However, the calculated outputs of the neural network model are overall the same as the measured data, irrespective of the damper force amplitude.

5. Conclusions

(1) We have developed damper test equipment capable of simulating actual conditions. It can actuate any damper installed on vehicles three-dimensionally, and is capable of acting as a component of HILS to cooperate with real-time simulation.

(2) The relationship between the motion order and the actual movement has been clarified through performance improvement.

(3) We have acquired the characteristics of the motion system, and have designed a compensator to flatten the gain characteristic.

(4) We have executed an automatic characteristics identification experiment using a lateral damper. The proposed neural network model is effective.

We plan to improve the controller of this equipment to execute HILS testing, and to develop an automatic identification method for the characteristics of air suspension.

Acknowledgment

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References

