Optimal Vibration Control of Vehicle Engine-Body System using Haar Functions

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Abstract: In this note a method of designing optimal vibration control based on Haar functions to control of bounce and pitch vibrations in engine-body vibration structure is presented. Utilizing properties of Haar functions, a computational method to find optimal vibration control for the engine-body system is developed. It is shown that the optimal state trajectories and optimal vibration control are calculated approximately by solving only algebraic equations instead of solving the Riccati differential equation. Simulation results are included to demonstrate the validity and applicability of the technique.

Keywords: Engine-body system, Haar function, optimal control, vibration control.

1. INTRODUCTION

Active control of sound and vibration has emerged as an important area of scientific and technological development in recent years. Developments in active control have allowed successful application of the concept in numerous industrial areas [1,2]. Recently, the noise and vibration of cars have become increasingly important. The predominant sources of interior noise in cars are engine and wheel vibrations, which propagate as structure-borne sound through the car body and finally radiate as airborne sound into the cabin [2,3-5]. A major comfort aspect is the transmission of engine-induced vibrations through powertrain mounts into the chassis (see Fig. 1). Engine and powertrain mounts are usually designed according to criteria that incorporate a trade-off between the isolation of the engine from the chassis and the restriction of engine movements. The engine mount is an efficient passive means to isolate the car chassis structure from the engine vibration. The passive means for isolation is efficient only in the high frequency range. However the vibration disturbance generated by the engine occurs mainly in the low frequency range [3,6-9]. These vibrations are result of the fuel explosion in the cylinder and the rotation of the different parts of the engine (see Fig. 2). The commercial use of engine and wheel mounts has been impeded so far by technical problems. Compact and robust combinations of conventional rubber mounts with electrodynamically driven hydraulics have been constructed as active hydromounts for a wide frequency range [10], but the stroke and power required for cars at low frequencies cannot yet be fulfilled by active hydromounts of reasonable size [2].

A variety of control techniques, such as PID or Lead-Lag compensation, LQG/$H_2$, $H_\infty$, $\mu$-synthesis and feedforward control have been used in active vibration systems [5,9,11-20]. The main characteristic of feedforward control is that information about the disturbance source is available and is usually realized with the Fx-LMS algorithms. However, the disturbance source is assumed to be unknown in feedback control, and then different strategies of feedback control for vibration attenuation of unknown disturbance exist ranging from classical methods to more advanced methods. Recently, the performance results obtained by feedback and feedforward controllers using Fx-LMS algorithms for vehicle engine-body vibration system were compared in [8,9].
On the other hand, in the field of dynamic systems and control, orthogonal functions-based techniques of analysis, identification and control have received considerable attention in the recent years. This is evident from the vast amount of literature published over the last two decades [21,22]. The various systems of orthogonal functions may be classified into two categories. The first is the so-called piecewise constant basis functions to which the orthogonal systems of Haar functions (HFs) [23-25], block pulse functions [26] and Walsh functions [27] belong. These functions are constant over different segments within their intervals of definition and the functions and solutions represented using this class as basis is always staircase-approximated. The main characteristic of the piecewise constant basis functions is that these problems are reduced to those of solving a system of algebraic equations for the solution of problems described by differential equations. Thus, the solution, identification and optimisation procedure are either greatly reduced or much simplified accordingly [25,28-32]. Despite this, orthogonal polynomials such as Legendre, Laguerre, Chebyshev, Jacobi, Hermite along with sine-cosine functions were extensively applied to many areas of systems and control in the last decade [33-36]. The problems considered so far for orthogonal functions-based solutions include response analysis, optimal control, parameter estimation, model reduction, controller design, and state estimation. They have been applied to linear time-invariant and time-varying systems, nonlinear and distributed parameter systems, which include scaled systems, stiff systems, delay systems, singular systems and multivariable systems [22].

In the sequel, we apply the HFs to the finite-time optimal control problem of the second-order vehicle engine-body vibration system. Mathematical model of the engine-body vibration structure is presented such the actuators and sensors used to investigate the optimal control are selected to be collocated. Moreover, the properties of HFs, Haar integral operational and Haar product operational matrices are given and are utilized to provide a systematic computational framework to find the optimal trajectory and finite-time optimal control of the vehicle engine-body vibration system approximately with respect to a quadratic cost function by solving only the linear algebraic equations instead of solving the differential equations. One of the main advantages is solving linear algebraic equations instead of solving nonlinear Riccati equation to optimize the control problem of the vehicle engine-body vibration system. Numerical results are presented to illustrate the applicability of the technique.

The rest of this paper is organized as follows. Section 2 introduces properties of the HFs. A dynamic model of the engine-body vibration structure is provided in Section 3. Algebraic solution of the engine-body system is given in Section 4 and development of optimal state trajectories and optimal vibration control by HFs are presented in Section 5. Simulation results of the vehicle engine-body vibration system are shown in Section 6 and finally the conclusion is discussed.

1.1. Notations

\[ A: r \times s \] matrix \( A \) with dimension \( r \times s \);

\[ I_r \] identity matrix with dimension \( r \times r \);

\[ 0_r \] zero matrix with dimension \( r \times r \);

\[ 0_{rs} \] zero matrix with dimension \( r \times s \);

\( \otimes \) Kronecker product;

\( \text{vec}(X) \) the vector obtained by putting matrix \( X \) into one column;

\( \text{tr}(A) \) trace of matrix \( A \).

2. HAAR FUNCTIONS

The original objective of the wavelet theory is to construct orthogonal bases of \( L_2(\mathbb{R}) \). These bases are constituted by translation and dilation of the same function \( \psi(.) \) and \( \phi(.) \), namely wavelet function and scaling function, respectively. These two functions generate a family of functions that can be used to break up or reconstruct a signal. To emphasis the ‘marriage’ involved in building this ‘family’ \( \phi(.) \) is sometimes called the ‘father wavelet’ and \( \psi(.) \), the ‘mother wavelet’ [21,37].

The oldest and most basic of the wavelet systems is named Haar wavelet, whose functions are given by

\[ \psi_0(t) = 1, \quad t \in [0, 1], \]
\[ \psi_i(t) = \begin{cases} 
1, & \text{for } t \in \left[0, \frac{1}{2}\right), \\
-1, & \text{for } t \in \left[\frac{1}{2}, 1\right), 
\end{cases} \]

where \( \phi(t) = \psi_0(t) \) and \( \psi_i(t) = \psi_i(2^j t - k) \) for \( i \geq 1 \) with \( i = 2^j + k \) for \( j \geq 0 \) and \( 0 \leq k < 2^j \). We can easily see that the \( \psi_0(t) \) and \( \psi_i(t) \) are compactly supported, they give a local description, at different scales \( j \), of the considered function [25].

The finite series representation of any square integrable function \( y(t) \) in terms of HFs in the interval \([0, 1)\), namely \( \hat{y}(t) \), is given by

\[
\hat{y}(t) = \sum_{i=0}^{m-1} a_i \psi_i(t) := a^T \Psi_m(t),
\]

where \( a := [a_0 \ a_1 \ \cdots \ a_{m-1}]^T \) and \( \Psi_m(t) := [\psi_0(t) \ \psi_1(t) \ \cdots \ \psi_{m-1}(t)]^T \) for \( m = 2^j \) and the Haar coefficients \( a_i \) are determined to minimize the mean integral square error \( e = \int_0^1 (y(t) - a^T \Psi_m(t))^2 \, dt \) and are given by

\[
a_i = 2^j \int_0^1 y(t) \, \psi_i(t) \, dt.
\]

Remark 1: The approximation error, \( \Xi_y(m) := y(t) - \hat{y}(t) \), is depending on the resolution \( m \) and is approaching zero by increasing parameter of the resolution.

The integration of the vector \( \Psi_m(t) \) can be approximated by

\[
\int_0^1 \psi_m(r) \, dr = P_m \Psi_m(t),
\]

where the matrix \( P_m \) represents the integral operator matrix for piecewise constant basis functions on the interval \([0, 1)\) at the resolution \( m \). For HFs, the square matrix \( P_m \) satisfies the following recursive formula [24]:

\[
P_m = \frac{1}{2m} \begin{bmatrix}
2m P_m & -H_m \\
H_m & 0
\end{bmatrix},
\]

with \( R_1 = \frac{1}{2} \) and \( H_m^{-1} = \frac{1}{m} H_m^T \text{diag}(r) \) where the vector \( r \) is represented by \( r := (1, 1, 2, 2, 4, 4, 4, 4, \cdots, \frac{m}{2}, \frac{m}{2}, \cdots, \frac{m}{2})^T \) for \( m > 2 \) and the matrix \( H_m \) for \( \frac{L}{m} \leq t_i < \frac{i+1}{m} \) is defined as

\[
H_m = \left[ \Psi_m(t_0), \Psi_m(t_1), \cdots, \Psi_m(t_{m-1}) \right].
\]

On the other hand, the product of two vectors \( \Psi_m(t) \) is also evaluated as

\[
R_m(t) := \Psi_m(t) \Psi_m^T(t),
\]

where \( R_m(t) \) satisfies the following recursive formula [24,28]

\[
R_m(t) = \frac{1}{2m} \begin{bmatrix}
R_m(t) & -H_m \text{diag}(\Psi_b(t)) \\
H_m^T \text{diag}(\Psi_b(t)) & \text{diag}(H_m^{-1} \Psi_a(t))
\end{bmatrix},
\]

with \( R_1(t) = \psi_0(t) \psi_0^T(t) \) and

\[
\Psi_a(t) := \left[ \psi_0(t), \psi_1(t), \cdots, \psi_{\frac{m}{2}}(t) \right]^T = \Psi_{\frac{m}{2}},
\]

\[
\Psi_b(t) := \left[ \psi_{\frac{m}{2}}(t), \psi_{\frac{m}{2}+1}(t), \cdots, \psi_{m-1}(t) \right]^T.
\]

3. THE VEHICLE ENGINE-BODY SYSTEM

In this section a dynamic formulation of the characteristics of the vehicle engine-body vibration system is provided for vibration control design. A schematic diagram of the vehicle engine-body vibration structure is shown in Fig. 3, where the engine with mass \( M_e \) and inertia moment \( I_e \) is mounted in the body by the engine mounts \( k_e, c_e \) and the vehicle body with mass \( M_b \) and inertia moment \( I_b \) is supported by front and rear tires, each of which is modeled as a system consisting of a spring \( k_b \) and a damping device \( c_b \). The front mount is the active mount, the output force of which can be controlled by an electric signal. The active mount consists of a main chamber where an oscillating mass (inertia mass) is moving up and down. The inertia mass is driven by an electro-magnetic force generated by a magnetic coil which is controlled by the input current.

In our study, only the bounce and pitch vibrations in the engine and body are considered. It is assumed that the actuator and sensor used to this control framework are selected to be collocated, since this arrangement is ideal to ensure the stability of the closed loop system for a slightly damped structure. Furthermore, the controller is tested for a single frequency signal, which is used to simulate the engine disturbance at particular frequency.
The derivation of the dynamic equations for a four degree-of-freedom vibration suspension model shown in Fig. 3 accordingly follows [9]:

\[
\begin{align*}
M_e \ddot{x}_1 + 2c_e \dot{x}_1 + 2k_e x_1 - 2c_e \dot{x}_2 - 2k_e x_2 - 2(L-l)c_e \dot{x}_4 - 2(L-l)k_e x_4 &= f + d_e, \\
M_b \ddot{x}_2 + 2(c_e + c_b) \dot{x}_2 + 2(k_e + k_b) x_2 - 2c_e \dot{x}_1 - 2k_e x_1 + 2(L-l)c_e \dot{x}_4 + 2(L-l)k_e x_4 &= -f, \\
I_e \ddot{x}_3 + 2L^2 c_e \dot{x}_3 + 2L^2 k_e x_3 - 2L^2 c_e \dot{x}_4 - 2L^2 k_e x_4 &= l f, \\
I_b \ddot{x}_4 + ((L^2 + (L-2l)^2)c_e + 2L^2 c_b) \dot{x}_4 + ((L^2 + (L-2l)^2)k_e + 2L^2 k_b) x_4 - 2L^2 c_e \dot{x}_3 - 2L^2 k_e x_3 - 2L^2 c_e \dot{x}_1 - 2L^2 k_e x_1 + 2(L-l)c_e \dot{x}_2 + 2(L-l)k_e \dot{x}_2 &= -l f,
\end{align*}
\]

(10)

where

- \( f(t) \): input force, which is used as the active force to compensate the vibration transmitted to vehicle body (or the chassis);
- \( d_e(t) \): engine disturbance, which can be the excitation, generated by the motion up/down of the different parts inside the engine;
- \( x_1(t), x_2(t), x_3(t), x_4(t) \): the bounces and pitches of the engine and body, respectively.

The state-space representation of the system (10) is

\[
M \ddot{x} + C \dot{x} + K x = B_f f(t) + B_d d_e(t),
\]

(11)

where the state-space matrices are defined as

\[
M = \begin{bmatrix}
M_e & 0 & 0 & 0 \\
0 & M_b & 0 & 0 \\
0 & 0 & I_e & 0 \\
0 & 0 & 0 & I_b
\end{bmatrix},
\]

\[
C = \begin{bmatrix}
2c_e & -2c_e & 0 & -2(L-l)c_e \\
-2c_e & 2(c_e + c_b) & 0 & 2(L-l)c_e \\
0 & 0 & 2L^2 c_e & -2L^2 c_e \\
-2L c_e & 2(L-l)c_e & -2L^2 c_e & 0
\end{bmatrix},
\]

\[
B_f = \begin{bmatrix}
1 \\
-1 \\
0 \\
-2\ell
\end{bmatrix}, \quad B_d = \begin{bmatrix}
1 \\
0 \\
0 \\
0
\end{bmatrix},
\]

\[
K = \begin{bmatrix}
2k_e & -2k_e & 0 & -2(L-l)k_e \\
-2k_e & 2(k_e + k_b) & 0 & 2(L-l)k_e \\
0 & 0 & 2L^2 k_e & -2L^2 k_e \\
-2\ell k_e & 2(L-l)k_e & -2L^2 k_e & \left(\frac{L^2 + (L-2\ell)^2}{2}\right)k_e
\end{bmatrix},
\]

Taking displacement of the chassis \((x_2(t))\) as an output then a comparison of the displacement

Fig. 3. The sketch of engine-body vibration system [9].

Fig. 4. Displacement of the chassis respect to \(f(t)\) (a) and \(d_e(t)\) (b).
response respect to the input force \( f(t) \) and the external disturbance \( d_e(t) \) in the frequency range up to 1 KHz is depicted in Fig. 4(a) and 4(b). Three relevant modes occur around the frequencies 1, 5 and 9 Hz, respectively, which represent the dynamics of the main degrees of freedom (DOFs) of the system.

4. ALGEBRAIC SOLUTION OF SYSTEM EQUATIONS

The problem of solving the second-order differential equations of the engine-body system (10) in terms of the input control and exogenous disturbance is investigated using HFs and an appropriate algebraic equation is developed.

Based on definition of HFs on the time interval \([0, 1]\), we need to rescale the finite time interval \([0, T_f] \) into \([0, 1]\) by considering \( t = T_f \sigma; \) normalizing the system (11) with the time scale would be as follows

\[
M\ddot{x}(\sigma) + C\dot{x}(\sigma) + Kx(\sigma) = B_f f(\sigma) + B_d d_e(\sigma).
\]

Now by integrating the system above in an interval \([0, \sigma]\), we obtain

\[
M(\dot{x}(\sigma) - \dot{x}(0)) + T_f C(x(\sigma) - x(0)) + T_f^2 J K \int_0^\sigma x(\tau)d\tau = T_f^2 B_f \int_0^\sigma f(\tau)d\tau + T_f^2 B_d \int_0^\sigma d_e(\tau)d\tau.
\]

To avoid the differentiation of HFs, we take again the integration of (13) in the interval \([0, \sigma]\) as follows:

\[
M(x(\sigma) - x(0)) + T_f C \int_0^\sigma x(\tau)d\tau + T_f^2 J K \int_0^\sigma x(\tau)d\tau = T_f^2 B_f \int_0^\sigma f(\tau)d\tau + T_f^2 B_d \int_0^\sigma d_e(\tau)d\tau + \int_0^\sigma (M\dot{x}(0) + T_f Cx(0))d\tau.
\]

By using the HF expansion (2), we express in the following solution \( x(\sigma) \), input force \( f(\sigma) \) and engine disturbance \( d_e(\sigma) \) in terms of HFs

\[
x(\sigma) = X_0 \Psi_m(\sigma), \quad f(\sigma) = F \Psi_m(\sigma), \quad d_e(\sigma) = D_e \Psi_m(\sigma),
\]

where \( X : 4 \times m \), \( F : 1 \times m \) and \( D_e : 1 \times m \) denote the wavelet coefficients of \( x(\sigma) \), \( f(\sigma) \) and \( D_e(\sigma) \), respectively. The initial conditions of \( x(0) \) and \( \dot{x}(0) \) are represented by \( x(0) = X_0 \Psi_m(\sigma) \) and \( \dot{x}(0) = \dot{X}_0 \Psi_m(\sigma) \), where the matrices \( X_0 : 4 \times m \) and \( \dot{X}_0 : 4 \times m \) are defined as

\[
X_0 := \begin{bmatrix} x(0) & 0_{4 \times d} & \ldots & 0_{4 \times d} \end{bmatrix}, \quad \dot{X}_0 := \begin{bmatrix} \dot{x}(0) & 0_{4 \times d} & \ldots & 0_{4 \times d} \end{bmatrix}.
\]

Therefore, using the HF expansions (15), the relation (14) becomes

\[
M(X - X_0) \Psi_m(\sigma) + T_f CX_0 \int_0^\sigma \Psi_m(\tau)d\tau + T_f^2 J KX
\]

\[
\times \int_0^\sigma \Psi_m(\tau)d\tau = T_f^2 B_f \int_0^\sigma \Psi_m(\tau)d\tau + T_f^2 B_d D_e
\]

\[
\times \int_0^\sigma \Psi_m(\tau)d\tau + (M\dot{X}_0 + T_f Cx(0)) \int_0^\sigma \Psi_m(\tau)d\tau.
\]

Moreover, using the Haar integral operational matrix \( P_m \) in (4) and omitting \( \Psi_m(\sigma) \) in both sides of (17), we have

\[
M(X - X_0) + T_f CX_0 P_m + T_f^2 KXP_m = T_f^2 B_f F P_m^2
\]

\[
+ T_f^2 B_d D_e P_m + (M\dot{X}_0 + T_f Cx(0)) P_m.
\]

For calculating the matrix \( X \), we apply the operator \( vec(\cdot) \) to (18) and according to the property of the Kronecker product in the Appendix A1, the following algebraic relation is obtained

\[
(I_m \otimes M)(vec(X) - vec(X_0)) = T_f(P_m^T \otimes C)vec(X)
\]

\[
+ T_f^2 (P_m^T \otimes K)vec(X) = T_f(P_m^T \otimes B_f)vec(F)
\]

and

\[
T_f^2 (P_m^T \otimes B_d)vec(D_e) + T_f(P_m^T \otimes C)vec(X_0) + (P_m^T \otimes M)vec(\dot{X}_0).
\]

Solving (19) for \( vec(X) \) leads to

\[
vec(X) = \Delta_1 vec(F) + \Delta_2 vec(D_e) + \Delta_3 vec(X_0) + \Delta_4 vec(\dot{X}_0),
\]

where the matrices \( \Delta_1 : 4m \times m, \quad \Delta_2 : 4m \times m, \quad \Delta_3 : 4m \times 4m, \quad \text{and} \quad \Delta_4 : 4m \times 4m \) are defined as

\[
\Delta_1 = T_f^2 (T_f(P_m^T \otimes C)
\]

\[
+ T_f^2 (P_m^T \otimes K) + I_m \otimes M)^{-1}(P_m^T \otimes B_f),
\]

\[
\Delta_2 = T_f^2 (T_f(P_m^T \otimes C)
\]

\[
+ T_f^2 (P_m^T \otimes K) + I_m \otimes M)^{-1}(P_m^T \otimes B_d),
\]

\[
\Delta_3 = T_f^2 (P_m^T \otimes C)
\]

\[
+ T_f^2 (P_m^T \otimes K) + I_m \otimes M)^{-1}(I_m \otimes M),
\]

\[
\times (I_m \otimes M + T_f P_m^T \otimes C),
\]
\[ \Delta_4 = (T_f (P_m^T \otimes C) + T_f^2 (P_m^T \otimes K) + I_m \otimes M)^{-1} (P_m^T \otimes M). \]

Consequently, from (20), (21) and the properties of the Kronecker product, the solution of the system (11) is approximately
\[ x(\sigma) = (\Psi_m^T (\sigma) \otimes I_4) \text{vec}(X). \] (22)

5. OPTIMAL VIBRATION CONTROL DESIGN

The control objective is to find the optimal control \( f(t) \) with respect to a quadratic cost functional approximately such acts as the active force to compensate the vibration transmitted to vehicle body (or the chassis). The quadratic cost functional weights the states and their derivatives with respect to time in the cost function as follows:
\[ J = \frac{1}{2} \int x^T(T_f) S_1 x(T_f) + \frac{1}{2} \hat{x}^T(T_f) S_2 \hat{x}(T_f) \]
\[ + \frac{1}{2} \int S_1 \left( \frac{1}{T_f} x^T(t) \right) + \hat{x}^T(t) \hat{Q}_2 \hat{x}(t) + Rf(t)^2 \] dt,
where \( S_1 : 4 \times 4, S_2 : 4 \times 4, Q_1 : 4 \times 4 \) and \( Q_2 : 4 \times 4 \) are positive-definite matrices and \( R \) is a positive scalar. We can rewrite the cost function (23) as follows:
\[ J = \frac{1}{2} \left[ x^T(T_f) \right] \hat{S} \left[ \frac{x(T_f)}{\hat{x}(T_f)} \right] \]
\[ + \frac{1}{2} \int S_1 \left( \frac{1}{T_f} x^T(t) \right) \frac{x(t)}{\hat{x}(t)} + \hat{x}^T(t) \hat{Q}_2 \hat{x}(t) + Rf(t)^2 \] dt,
where \( \hat{S} = \text{diag}(S_1, S_2) \) and \( \hat{Q} = \text{diag}(Q_1, Q_2) \).

Normalizing (24) with the time scale \( t = T_f \sigma \) yields
\[ J = \frac{1}{2} \left[ x^T(l) \right] \frac{T_f}{\hat{S}} \left[ \frac{x(l)}{\hat{x}(l)} \right] + \frac{T_f}{2} \]
\[ \times \int \left( \frac{1}{T_f} x^T(\sigma) \right) \frac{T_f}{\hat{S}} \left[ \frac{x(\sigma)}{\hat{x}(\sigma)} \right] + Rf(\sigma)^2 \] d\sigma.

From (15) and the relation \( \hat{x}(\sigma) = \hat{X} \Psi_m(\sigma) \), where \( \hat{X} : 4 \times m \) denotes the wavelet coefficients of \( \hat{x}(\sigma) \) after its expansion in terms of HFs, we read
\[ \begin{bmatrix} x(\sigma) \\ T_f^{-1} \hat{x}(\sigma) \end{bmatrix} = \begin{bmatrix} X \\ T_f^{-1} \hat{X} \end{bmatrix} \Psi_m(\sigma) = X_{\text{aug}} \Psi_m(\sigma), \] (26)
where \( X_{\text{aug}} = \begin{bmatrix} X \\ T_f^{-1} \hat{X} \end{bmatrix} \) and
\[ \text{vec}(X_{\text{aug}}) = \begin{bmatrix} \text{vec}(X) \\ T_f^{-1} \text{vec}(\hat{X}) \end{bmatrix}^T. \] (27)

Remark 2: By substituting \( \hat{x}(\sigma) = \hat{X} \Psi_m(\sigma) \) into \( x(\sigma) - x(0) = \int_0^\sigma \hat{x}(t) dt \), we have:
\[ X \Psi_m(\sigma) - X_0 \Psi_m(\sigma) = \int_0^\sigma \hat{X} \Psi_m(\tau) d\tau. \] (28)

and using (4), we read \( X - X_0 = \hat{X} P_m \). Then, by applying the operator of \( \text{vec}(\cdot) \) and according to the properties of Kronecker product in Appendix A1, we obtain
\[ \text{vec}(X) - \text{vec}(X_0) = (P_m^T \otimes I_n) \text{vec}(\hat{X}). \] (29)

By substituting the definition (26) in (29) and using the properties of the operator \( \text{tr}(\cdot) \) in Appendix A1, the cost function (25) is given by
\[ J = \frac{1}{2} \left( \text{tr}(M_f X_{\text{aug}}^T \tilde{S} X_{\text{aug}}) \right) + \frac{T_f}{2} \]
\[ \times \text{tr}(M X_{\text{aug}} \tilde{Q} X_{\text{aug}} + \text{tr}(M F^T F)), \]
where the matrices \( M_m : m \times m \) and \( M_{mf} : m \times m \) are defined as \( M_m := \int_0^1 \Psi_m(\sigma) \Psi_m^T(\sigma) d\sigma \) and \( M_{mf} := \Psi_m(l) \Psi_m^T(l) \), respectively. Using the properties of the Kronecker product in Appendix A1, we can write (30) as
\[ J = \frac{1}{2} \left( \text{vec}^T(X_{\text{aug}} M_{mf}) (I_m \otimes \tilde{S}) \text{vec}(X_{\text{aug}}) \right) \]
\[ + \frac{T_f}{2} \left( \text{vec}^T(X_{\text{aug}} M_m) (I_m \otimes \tilde{Q}) \text{vec}(X_{\text{aug}}) \right) \]
\[ + R \text{vec}^T(F) M_m \text{vec}(F)) \]
or
\[ J = \frac{1}{2} \left( \text{vec}^T(X_{\text{aug}}) \Pi_m \text{vec}(X_{\text{aug}}) \right) \]
\[ + \text{vec}^T(F) \Pi_{m2} \text{vec}(F)), \] (32)
where the matrices \( \Pi_{m1} : 8m \times 8m \) and \( \Pi_{m2} : m \times m \) are defined as \( \Pi_{m1} = (M_f^T \otimes \hat{S} + T_f (M^T \otimes \hat{Q})) \) and \( \Pi_{m2} = RT_f M_m \), respectively.

It is clear that the cost function of \( J(\cdot) \) is a
function of $\frac{i}{m} \leq \sigma_i < \frac{i+1}{m}$, then for finding the optimal control law, which minimizes the cost functional $J(\cdot)$, the following necessary condition should be satisfied

$$\frac{\partial J}{\partial \text{vec}(F)} = 0.$$  

(33)

By considering $\text{vec}(X_{aug})$, which is a function of $\text{vec}(F)$, and using the properties of derivatives of inner product of Kronecker product in Appendix A2, we find

$$\frac{\partial J}{\partial \text{vec}(F)} = \frac{1}{2} \left( \frac{\partial }{\partial \text{vec}(F)} \text{vec}(X_{aug}) \right) (\text{vec}(X_{aug}))^T \times \Pi_m \text{vec}(X_{aug})$$

$$+ \frac{\partial }{\partial \text{vec}(F)} (\text{vec}(F) \Pi_{m2} \text{vec}(F))).$$  

(34)

To further investigate the relation (34), from Appendix A3 we obtain

$$\frac{\partial \text{vec}^T(X_{aug})}{\partial \text{vec}(F)} = \left[ \text{vec}^T(U)T_f \Delta_f T_f \left( P_m^{-1} \otimes I_4 \right) \right] \Pi_m \text{vec}(X_{aug})$$

$$+ \Pi_{m2} \text{vec}(F).$$  

(35)

$$\text{where}$$

$$(*) = \text{vec}^T(U)\Delta_f^T + \text{vec}^T(D_e)\Delta_2^T$$

$$+ \text{vec}^T(X_0)\Delta_3^T + \text{vec}^T(\bar{X}_0)\Delta_4^T,$$

$$(**) = T_f^{-1}(P_m^{-1} \otimes I_4) (\Delta_3 \text{vec}(F) + \Delta_2 \text{vec}(D_e))$$

$$+ \Delta_3 \text{vec}(X_0) + \Delta_4 \text{vec}(\bar{X}_0) - \text{vec}(X_0)).$$

Therefore, we find

$$\frac{\partial J}{\partial \text{vec}(F)} = \left[ \Delta_1^T T_f^{-1} \Delta_1^T \left( P_m^{-1} \otimes I_4 \right) \right] \Pi_m \text{vec}(X_{aug})$$

$$+ \Pi_{m2} \text{vec}(F).$$  

(36)

Then the wavelet coefficients of the optimal control law will be in vector form as

$$\text{vec}(F) = -\Pi_{m2} \left[ \Delta_1^T T_f^{-1} \Delta_1^T \left( P_m^{-1} \otimes I_4 \right) \right] \Pi_m \text{vec}(X_{aug})$$

$$\times \text{vec}(X_{aug}).$$  

(37)

Consequently, from (20), (27), (29), and (37) the optimal vectors of $\text{vec}(X)$ and $\text{vec}(F)$ are found, respectively, in the following forms

$$\text{vec}(X) = (I_{4m} + \Delta_1 \Pi_{m2} \Delta_1^T T_f^{-1} \left( P_m^{-1} \otimes I_4 \right) \Pi_m) \text{vec}(X_{aug})$$

$$\times \text{vec}(X_{aug}).$$

(38)

and

$$\text{vec}(F) = -\Pi_{m2} \Delta_1^T T_f^{-1} \Delta_1^T \left( P_m^{-1} \otimes I_4 \right) \Pi_m \text{vec}(X_{aug})$$

$$\times \text{vec}(X_{aug}).$$  

(39)

Finally, the Haar function-based optimal trajectories and optimal control are obtained approximately from (22) and $f(t) = \Psi_m^T(t) \text{vec}(F)$.

**Remark 3:** According to the properties of HFs and Haar product operational matrix in the Section 2, the matrix $M_m$ can be calculated from the following recursive formula:

$$M_m = \frac{1}{2m} \left[ M_m \frac{1}{2} \left( H_m \text{diag}(\Psi_m(t)) \right)^T \text{diag}(H_m^{-1} \Psi_m(t)) \right]$$

(40)

with $M_1(t) = 1$ and

$$\Psi_m(t) := \left[ e_1 P_m \Psi_m(1), e_2 P_m \Psi_m(1), \ldots, e_m P_m \Psi_m(1) \right]^T,$$

$$\Psi_m(t) := \left[ e_1 P_m \Psi_m(1), e_2 P_m \Psi_m(1), \ldots, e_m P_m \Psi_m(1) \right]^T,$$

(41)

where $e_i = \lfloor 0_d(i-1), 1, 0_d(m-i) \rfloor$ for $i = 1, 2, \ldots, m$.

**Remark 4:** Since the vector $\Psi_m(\sigma)$ is constant within each of the $m$ time intervals, the approximated optimal trajectories (38) and optimal control (39) can be expressed as
\( x(t) = \sum_{i=1}^{m} G_i \vec{v}(X_0) + \sum_{i=1}^{m} \vec{G}_i \vec{v}(X_0) + \sum_{i=1}^{m} \vec{G}_i \vec{v}(D_c), \)

(42)

\( f(t) = -i \left( \sum_{i=1}^{m} F_i \vec{v}(X_0) + \sum_{i=1}^{m} \vec{F}_i \vec{v}(X_0) + \sum_{i=1}^{m} \vec{F}_i \vec{v}(D_c) \right) \)

(43)

with constant matrices \( G_i : 4 \times 4, \vec{G}_i : 4 \times 4, \vec{G}_i : 4 \times m, F_i : 1 \times 4, \vec{F}_i : 1 \times 4 \) and \( \vec{F}_i : 1 \times m \) within each of time intervals \( \frac{i}{m} \leq \sigma_i < \frac{i+1}{m} \) for \( i = 0, 1, \ldots, (m - 1) \).

**Remark 5:** This fact of constant coefficient matrices is a consequence of using piecewise constant basis functions like HFs or Walsh functions, and cannot be achieved with smooth function sets like Legendre or Laguerre polynomials. Compared to Walsh functions, the HFs have additional advantages in computational effort. Of course, more detailed investigations on using other basis functions than HFs would be of interest.

### 6. NUMERICAL RESULTS

The parameters of the vehicle engine-body vibration model, used for the design and simulation are given in Tables 1 and 2. The objective is to find the optimal states and optimal input force approximately using HFs at the finite time interval \([0, 3]\). Moreover, the matrices \( S_1 : 4 \times 4, S_2 : 4 \times 4, Q_1 : 4 \times 4 \) and \( Q_2 : 4 \times 4 \) and scalar \( R \) in the cost function (23) are chosen as \( S_1 = S_2 = 0_4, Q_1 = diag(1, 2, 1, 1), Q_2 = diag(0.1, 0.2, 0.1, 0.1) \) and \( R = 1 \).

![Fig. 5. Comparison of state trajectories found by HFs at resolution level \( j = 5 \) (solid line) and by analytic solution (dashed line).](image1)

![Fig. 6. Comparison of input force found by HFs at resolution level \( j = 5 \) (solid line) and by analytic solution (dashed line).](image2)

![Fig. 7. Comparison of displacement of the chassis found by HFs at resolution level \( j = 5 \) (solid line), \( j = 4 \) (dashed line) and by analytic solution (dashed-dot line).](image3)
the comparison of states $x(\sigma)$ and optimal vibration control $f(\sigma)$ found by HFs at resolution level $j=5$ and the analytic solution found by solving the differential Riccati equation in Appendix A4, respectively. Figures show that the HFs can construct the vibration signals as well. Moreover, by increasing frequency of the external disturbance to 0.5KHz, the approximation displacement of the chassis $x(t)$ at the resolution levels $j=4$ and $5$ are plotted and compared to the analytic solution in Fig. 7. It is clear that by increasing the resolution level $j$ the accuracy of the approximation can be improved as well.

Different from the analytic solution using the nonlinear Riccati equation, the approximate solutions (38) and (39) deliver both, control $f(t)$ and state trajectory $x(t)$ in one step, while mean square error can easily be improved by increasing the resolution level $j$.

7. CONCLUSION

This paper presented a method of designing optimal vibration control based on Haar functions (HFs) to control of bounce and pitch vibrations in engine-body vibration structure. Utilizing properties of HFs, a computational method to find optimal vibration control for the engine-body system was developed. It was shown that the optimal state trajectories and optimal vibration control are calculated approximately by solving only algebraic equations instead of solving the Riccati differential equation. The simulation results were included to illustrate the validity and applicability of the proposed technique.

APPENDIX A

A.1. Some properties of Kronecker product [38]

Let $A: p \times q$, $B: q \times r$, $C: r \times s$ and $D: q \times t$ be fixed matrices, then we have:

$$\text{vec}(ABC) = (C^T \otimes A) \text{vec}(B),$$

$$\text{tr}(ABC) = \text{vec}^T(A^T)(I_p \otimes B) \text{vec}(C),$$

$$\text{tr}(ABC) = \text{vec}^T(A^T)(I_p \otimes B) \text{vec}(C),$$

$$(A \otimes C)(D \otimes B) = AD \otimes CB.$$

A.2. Derivatives of inner products of Kronecker product [38]

Let $A: n \times n$ be fixed constants and $x: n \times 1$ be a vector of variables. Then, the following results can be established:

$$\frac{\partial(x^T Ax)}{\partial x} = Ax + A^T x,$$

A.3. Chain rule for matrix derivatives using Kronecker product [38]

Let $Z$ be a $p \times q$ matrix whose entries are a matrix function of the elements of $Y: s \times t$, where $Y$ is a function of matrix $X: m \times n$. That is, $Z = H_1(Y)$, where $Y = H_2(X)$. The matrix of derivatives of $Z$ with respect to $X$ is given by

$$\frac{\partial Z}{\partial X} = \left\{ \frac{\partial \text{vec}(Y)}{\partial X} \otimes I_p \right\} \left\{ I_n \otimes \frac{\partial Z}{\partial \text{vec}(Y)} \right\}.$$


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