

Fast Evaluation of Nonviscously Damped System With A Kernel Independent Algorithm

Abstract

Keywords:

1. Nonviscous Damping With A Single Exponential Kernel

1.1. Nonviscous Damped System

Consider the equation of motion of a nonviscously damped inelastic multi-degree-of-freedom (MDOF) system,

$$\mathbf{Y}(\mathbf{u}, \mathbf{v}, \mathbf{a}) + \mathbf{F}(t) = \mathbf{P}(t), \quad (1)$$

where $\mathbf{u} = \mathbf{u}(t)$, $\mathbf{v} = \mathbf{v}(t) = \dot{\mathbf{u}}$ and $\mathbf{a} = \mathbf{a}(t) = \dot{\mathbf{v}}$ are the displacement, velocity and acceleration vectors, $\mathbf{Y} = \mathbf{Y}(\mathbf{u}, \mathbf{v}, \mathbf{a})$ is the resistance vector of the system, $\mathbf{P} = \mathbf{P}(t)$ is the external load vector, and \mathbf{F} is the nonviscous damping force which can be expressed in the form of the convolution of the kernel $f = f(t)$ and the vector \mathbf{w} , viz. $\mathbf{F}(t) = f * \mathbf{w}$.

Note here, \mathbf{w} can be either the exact velocity vector \mathbf{v} , or the subset of \mathbf{v} such that they share the same size but some velocity components in \mathbf{v} are replaced by zeros in \mathbf{w} on selected DoFs. This is beneficial when it comes to compositing flexible damping that will be discussed later in this work. Formally,

$$\mathbf{w} = \mathbf{T}\mathbf{v}, \quad (2)$$

where \mathbf{T} is a square diagonal matrix, the diagonal entries of which are either one or zero.

Since it is an inelastic system, the stiffness matrix \mathbf{K} , the viscous damping matrix \mathbf{C} and the mass matrix \mathbf{M} are

$$\frac{\partial \mathbf{Y}}{\partial \mathbf{u}} = \mathbf{K}, \quad \frac{\partial \mathbf{Y}}{\partial \mathbf{v}} = \mathbf{C}, \quad \frac{\partial \mathbf{Y}}{\partial \mathbf{a}} = \mathbf{M}. \quad (3)$$

20 The viscous damping matrix \mathbf{C} may not be trivial as the system may consist of viscous damping
 21 components (e.g., viscous damper devices). Using \mathbf{u} as the basic quantity, the effective stiffness
 22 matrix $\bar{\mathbf{K}}$

$$\bar{\mathbf{K}} = \frac{d\mathbf{Y}}{d\mathbf{u}} = \mathbf{K} + \mathbf{C} \frac{d\mathbf{v}}{d\mathbf{u}} + \mathbf{M} \frac{d\mathbf{a}}{d\mathbf{u}} \quad (4)$$

23 is the combination of the three, its specific form depends on the specific time integration method
 24 used.

25 1.2. A Single Exponential Kernel

26 For the moment, we focus on the scalar-valued exponential kernel function

$$f = f(t) = m \exp(-st), \quad (5)$$

27 where s is often denoted by the relaxation parameter μ , m is often denoted by $c\mu$ in which c is the
 28 damping constant. The convolution can be then expressed as

$$\mathbf{F}(t) = f * \mathbf{w} = \int_0^t f(t - \tau) \cdot \mathbf{w}(\tau) \, d\tau = \int_0^t m \exp(-s(t - \tau)) \cdot \mathbf{w}(\tau) \, d\tau. \quad (6)$$

29 Assuming trivial initial condition $\mathbf{v}(0) = \mathbf{0}$, Eq. (6) corresponds to the solution of the following
 30 ODE,

$$\mathbf{F}' = -s\mathbf{F} + m\mathbf{w}. \quad (7)$$

31 It can be validated by solving Eq. (7) with the assist of the integrating factor $\exp(st)$.

32 1.3. An Efficient Algorithm

33 Instead of directly integrating Eq. (6) using higher-order methods (such as the Runge–Kutta
 34 family), Eq. (7) can be combined with Eq. (1) to develop an efficient algorithm.

35 In the context of a discretised iterative solving schema, Eq. (7) can be rewritten as follows using
 36 the backward (implicit) Euler method,

$$\frac{\mathbf{F}_{n+1} - \mathbf{F}_n}{\Delta t} = -s\mathbf{F}_{n+1} + m\mathbf{w}_{n+1}, \quad (8)$$

in which subscripts $(\cdot)_{n+1}$ and $(\cdot)_n$ denote the corresponding quantity at t_n and $t_{n+1} = t_n + \Delta t$.
Rearranging Eq. (8) yields

$$(1 + s\Delta t) \mathbf{F}_{n+1} - \mathbf{F}_n - m\Delta t \mathbf{w}_{n+1}. \quad (9)$$

Assuming Eq. (1) is satisfied at t_{n+1} ¹, then, accounting for both Eq. (1) and Eq. (9), the
residual \mathbf{R} (with the subscript $(\cdot)_{n+1}$ dropped for brevity) is

$$\mathbf{R} = \begin{bmatrix} \mathbf{Q} \\ \mathbf{W} \end{bmatrix} = \begin{cases} \mathbf{Y} + \mathbf{F} - \mathbf{P}, \\ (1 + s\Delta t) \mathbf{F} - \mathbf{F}_n - m\Delta t \mathbf{w}. \end{cases} \quad (10)$$

The unknown quantity is $\mathbf{x} = \begin{bmatrix} \mathbf{u} & \mathbf{F} \end{bmatrix}^T$. Linearisation results in the following Jacobian.

$$\mathbf{J} = \begin{bmatrix} \frac{\partial \mathbf{Q}}{\partial \mathbf{u}} & \frac{\partial \mathbf{Q}}{\partial \mathbf{F}} \\ \frac{\partial \mathbf{W}}{\partial \mathbf{u}} & \frac{\partial \mathbf{W}}{\partial \mathbf{F}} \end{bmatrix} = \begin{bmatrix} \bar{\mathbf{K}} & \mathbf{I} \\ -m\Delta t \mathbf{T} \frac{d\mathbf{v}}{d\mathbf{u}} & (1 + s\Delta t) \mathbf{I} \end{bmatrix}. \quad (11)$$

Typically, $\frac{d\mathbf{v}}{d\mathbf{u}}$ reduces to a scalar constant (multiplied by an identity matrix), for example, in the
Newmark method, it is $\frac{\gamma}{\beta\Delta t}$.

Noting that $\frac{\partial \mathbf{W}}{\partial \mathbf{F}}$ is a diagonal matrix that can be easily inverted, there is no need to explicitly
formulate the Jacobian. Instead, one could perform static condensation such that, from the second
expression,

$$-m\Delta t \mathbf{T} \frac{d\mathbf{v}}{d\mathbf{u}} \delta \mathbf{u} + (1 + s\Delta t) \delta \mathbf{F} = \mathbf{W}, \quad (12)$$

the increment $\delta \mathbf{F}$ is

$$\delta \mathbf{F} = \frac{1}{1 + s\Delta t} \left(\mathbf{W} + m\Delta t \mathbf{T} \frac{d\mathbf{v}}{d\mathbf{u}} \delta \mathbf{u} \right), \quad (13)$$

¹This assumption is not always valid as some time integration methods establish the EOM elsewhere, see, for example, the generalised- α method, the GSSS method, the Bathe two-step method, the OALTS method, etc.

48 substituting it into the first expression yields

$$\bar{\mathbf{K}}\delta\mathbf{u} + \frac{1}{1+s\Delta t} \left(\mathbf{W} + m\Delta t \mathbf{T} \frac{d\mathbf{v}}{d\mathbf{u}} \delta\mathbf{u} \right) = \mathbf{Q}. \quad (14)$$

49 Rearranging gives

$$\left(\bar{\mathbf{K}} + \frac{m\Delta t}{1+s\Delta t} \mathbf{T} \frac{d\mathbf{v}}{d\mathbf{u}} \right) \delta\mathbf{u} = \mathbf{Q} - \frac{1}{1+s\Delta t} \mathbf{W}. \quad (15)$$

50 By denoting

$$\hat{\mathbf{K}} = \bar{\mathbf{K}} + \frac{m\Delta t}{1+s\Delta t} \mathbf{T} \frac{d\mathbf{v}}{d\mathbf{u}}, \quad \hat{\mathbf{Q}} = \mathbf{Q} - \frac{1}{1+s\Delta t} \mathbf{W}, \quad (16)$$

51 the system to be solved is simply

$$\hat{\mathbf{K}}\delta\mathbf{u} = \hat{\mathbf{Q}}. \quad (17)$$

52 The revised effective load vector $\hat{\mathbf{Q}}$ can be explicitly written as

$$\hat{\mathbf{Q}} = \mathbf{Y} - \mathbf{P} + \frac{1}{1+s\Delta t} \mathbf{F}_n + \frac{1}{1+s\Delta t} m\Delta t \mathbf{w}. \quad (18)$$

53 **References**