

## 9

Algorithms for Hyperbolic  
and Parabolic-Hyperbolic Problems**9.1 ONE-STEP ALGORITHMS FOR THE SEMIDISCRETE  
EQUATION OF MOTION****9.1.1 The Newmark Method**

Recall from Chapter 7 that the semidiscrete equation of motion is written as

$$M\ddot{\mathbf{d}} + C\dot{\mathbf{d}} + K\mathbf{d} = \mathbf{F} \quad (9.1.1)$$

where  $M$  is the mass matrix,  $C$  is the viscous damping matrix,  $K$  is the stiffness matrix,  $F$  is the vector of applied forces, and  $\mathbf{d}$ ,  $\dot{\mathbf{d}}$ , and  $\ddot{\mathbf{d}}$  are the displacement, velocity and acceleration vectors, respectively. We take  $M$ ,  $C$ , and  $K$  to be symmetric;  $M$  is positive-definite, and  $C$  and  $K$  are positive-semidefinite.

The initial-value problem for (9.1.1) consists of finding a displacement,  $\mathbf{d} = \mathbf{d}(t)$ , satisfying (9.1.1) and the given initial data:

$$\mathbf{d}(0) = \mathbf{d}_0 \quad (9.1.2)$$

$$\dot{\mathbf{d}}(0) = \mathbf{v}_0 \quad (9.1.3)$$

Perhaps the most widely used family of direct methods for solving (9.1.1) to (9.1.3) is the *Newmark family* [1], which consists of the following equations:

$$M\mathbf{a}_{n+1} + C\mathbf{v}_{n+1} + K\mathbf{d}_{n+1} = \mathbf{F}_{n+1} \quad (9.1.4)$$

$$\mathbf{d}_{n+1} = \mathbf{d}_n + \Delta t \mathbf{v}_n + \frac{\Delta t^2}{2} [(1 - 2\beta)\mathbf{a}_n + 2\beta\mathbf{a}_{n+1}] \quad (9.1.5)$$

$$v_{n+1} = v_n + \Delta t[(1 - \gamma)a_n + \gamma a_{n+1}] \quad (9.1.6)$$

where  $d_n$ ,  $v_n$ , and  $a_n$  are the approximations of  $d(t_n)$ ,  $\dot{d}(t_n)$ , and  $\ddot{d}(t_n)$ , respectively. Equation (9.1.4) is simply the equation of motion in terms of the approximate solution, and (9.1.5) and (9.1.6) are finite difference formulas describing the evolution of the approximate solution. The parameters  $\beta$  and  $\gamma$  determine the stability and accuracy characteristics of the algorithm under consideration. Equations (9.1.4) to (9.1.6) may be thought of as three equations for determining the three unknowns  $d_{n+1}$ ,  $v_{n+1}$ , and  $a_{n+1}$ , it being assumed that  $d_n$ ,  $v_n$ , and  $a_n$  are known from the previous step's calculations. The Newmark family contains as special cases many well-known and widely used methods.

**Implementation: a-form**

In our case  $\beta = 0$  (purely explicit scheme),  $\gamma = 1/2$ , and matrix  $C$  is zero in the elastic case.

There are several possible implementations. We will sketch one, but we leave further details until Sec. 9.4, which deals with operator and mesh partitions. The results in Sec. 9.4 include the Newmark method as a special case. Define *predictors*:

$$\tilde{d}_{n+1} = d_n + \Delta t v_n + \frac{\Delta t^2}{2}(1 - \text{X})a_n \quad (9.1.7)$$

$$\tilde{v}_{n+1} = v_n + (1 - \gamma)\Delta t a_n \quad (9.1.8)$$

Equations (9.1.5) and (9.1.6) may then be written as

$$d_{n+1} = \tilde{d}_{n+1} + \beta \text{X} a_{n+1} \quad (9.1.9)$$

$$v_{n+1} = \tilde{v}_{n+1} + \gamma \Delta t a_{n+1} \quad (9.1.10)$$

The recursion relation determines  $a_{n+1}$ :

$$(M + \gamma \text{X} C + \beta \text{X} K)a_{n+1} = F_{n+1} - \text{X} F_n - K \tilde{d}_{n+1} \quad (9.1.12)$$

Equations (9.1.9) and (9.1.10) may then be used to calculate  $d_{n+1}$  and  $v_{n+1}$ , respectively.

This form of implementation is convenient for generalization to algorithms that employ “mesh partitions” (see Sec. 9.4) but is not the most efficient implementation.

$$v_{n+1} = v_n + \Delta t[(1 - \gamma)a_n + \gamma a_{n+1}] \quad (9.1.6)$$

where  $d_n$ ,  $v_n$ , and  $a_n$  are the approximations of  $d(t_n)$ ,  $\dot{d}(t_n)$ , and  $\ddot{d}(t_n)$ , respectively. Equation (9.1.4) is simply the equation of motion in terms of the approximate solution, and (9.1.5) and (9.1.6) are finite difference formulas describing the evolution of the approximate solution. The parameters  $\beta$  and  $\gamma$  determine the stability and accuracy characteristics of the algorithm under consideration. Equations (9.1.4) to (9.1.6) may be thought of as three equations for determining the three unknowns  $d_{n+1}$ ,  $v_{n+1}$ , and  $a_{n+1}$ , it being assumed that  $d_n$ ,  $v_n$ , and  $a_n$  are known from the previous step's calculations. The Newmark family contains as special cases many well-known and widely used methods.

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$$d_{n+1} = \tilde{d}_{n+1} + \beta\Delta t^2 a_{n+1} \quad (9.1.9)$$

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To start the process,  $a_0$  may be calculated from

$$Ma_0 = F - Cv_0 - Kd_0 \quad (9.1.11)$$

or specified directly. The recursion relation determines  $a_{n+1}$ :

$$(M + \gamma\Delta t C + \beta\Delta t^2 K)a_{n+1} = F_{n+1} - C\tilde{v}_{n+1} - K\tilde{d}_{n+1} \quad (9.1.12)$$

Equations (9.1.9) and (9.1.10) may then be used to calculate  $d_{n+1}$  and  $v_{n+1}$ , respectively.

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