

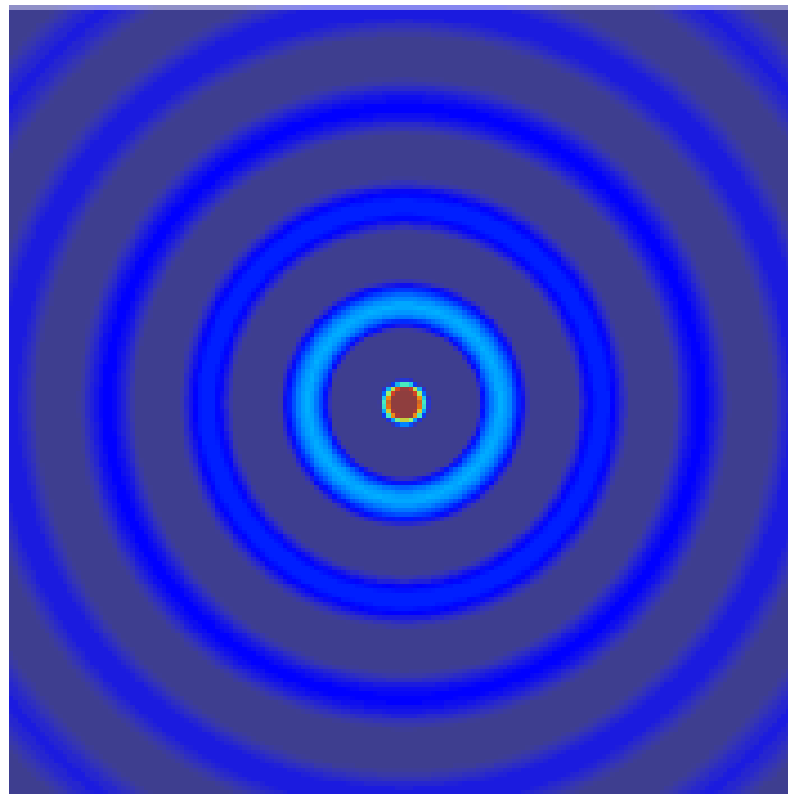
**MAP 2320 – MÉTODOS NUMÉRICOS EM EQUAÇÕES
DIFERENCIAIS II**

2º Semestre - 2020

Prof. Dr. Luis Carlos de Castro Santos

lsantos@ime.usp.br

$$\frac{\partial^2 u}{\partial t^2} = c^2 \nabla^2 u$$



Richard L. Burden
J. Douglas Faires

Numerical Analysis



Ninth Edition

Numerical Analysis

NINTH EDITION

Richard L. Burden

Youngstown State University

J. Douglas Faires

Youngstown State University

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12.3 Hyperbolic Partial Differential Equations

In this section, we consider the numerical solution to the **wave equation**, an example of a *hyperbolic* partial differential equation. The wave equation is given by the differential equation

$$\frac{\partial^2 u}{\partial t^2}(x, t) - \alpha^2 \frac{\partial^2 u}{\partial x^2}(x, t) = 0, \quad 0 < x < l, \quad t > 0, \quad (12.16)$$

subject to the conditions

$$u(0, t) = u(l, t) = 0, \quad \text{for } t > 0, \quad \leftarrow \text{extremidades fixas}$$

$$u(x, 0) = f(x), \quad \text{and} \quad \frac{\partial u}{\partial t}(x, 0) = g(x), \quad \text{for } 0 \leq x \leq l,$$

where α is a constant dependent on the physical conditions of the problem.

Select an integer $m > 0$ to define the x -axis grid points using $h = l/m$. In addition, select a time-step size $k > 0$. The mesh points (x_i, t_j) are defined by

$$x_i = ih \quad \text{and} \quad t_j = jk,$$

for each $i = 0, 1, \dots, m$ and $j = 0, 1, \dots$

At any interior mesh point (x_i, t_j) , the wave equation becomes

$$\frac{\partial^2 u}{\partial t^2}(x_i, t_j) - \alpha^2 \frac{\partial^2 u}{\partial x^2}(x_i, t_j) = 0. \quad (12.17)$$



$$h = \Delta x, \quad k = \Delta t$$

$$D^2 u(\bar{x}) = \frac{1}{h^2} [u(\bar{x} - h) - 2u(\bar{x}) + u(\bar{x} + h)]$$

The difference method is obtained using the centered-difference quotient for the second partial derivatives given by

$$\frac{\partial^2 u}{\partial t^2}(x_i, t_j) = \frac{u(x_i, t_{j+1}) - 2u(x_i, t_j) + u(x_i, t_{j-1}))}{k^2} - \frac{k^2}{12} \frac{\partial^4 u}{\partial t^4}(x_i, \mu_j),$$

where $\mu_j \in (t_{j-1}, t_{j+1})$, and

$$\frac{\partial^2 u}{\partial x^2}(x_i, t_j) = \frac{u(x_{i+1}, t_j) - 2u(x_i, t_j) + u(x_{i-1}, t_j))}{h^2} - \frac{h^2}{12} \frac{\partial^4 u}{\partial x^4}(\xi_i, t_j),$$

where $\xi_i \in (x_{i-1}, x_{i+1})$. Substituting these into Eq. (12.17) gives

$$\begin{aligned} & \frac{u(x_i, t_{j+1}) - 2u(x_i, t_j) + u(x_i, t_{j-1}))}{k^2} - \alpha^2 \frac{u(x_{i+1}, t_j) - 2u(x_i, t_j) + u(x_{i-1}, t_j))}{h^2} \\ &= \frac{1}{12} \left[k^2 \frac{\partial^4 u}{\partial t^4}(x_i, \mu_j) - \alpha^2 h^2 \frac{\partial^4 u}{\partial x^4}(\xi_i, t_j) \right]. \end{aligned}$$

Neglecting the error term

$$\tau_{ij} = \frac{1}{12} \left[k^2 \frac{\partial^4 u}{\partial t^4}(x_i, \mu_j) - \alpha^2 h^2 \frac{\partial^4 u}{\partial x^4}(\xi_i, t_j) \right], \quad (12.18)$$

leads to the difference equation

$$\frac{w_{i,j+1} - 2w_{i,j} + w_{i,j-1}}{k^2} - \alpha^2 \frac{w_{i+1,j} - 2w_{i,j} + w_{i-1,j}}{h^2} = 0.$$

Define $\lambda = \alpha k/h$. Then we can write the difference equation as

$$w_{i,j+1} - 2w_{i,j} + w_{i,j-1} - \lambda^2 w_{i+1,j} + 2\lambda^2 w_{i,j} - \lambda^2 w_{i-1,j} = 0$$

and solve for $w_{i,j+1}$, the most advanced time-step approximation, to obtain

$$w_{i,j+1} = 2(1 - \lambda^2)w_{i,j} + \lambda^2(w_{i+1,j} + w_{i-1,j}) - w_{i,j-1}. \quad (12.19)$$

This equation holds for each $i = 1, 2, \dots, m-1$ and $j = 1, 2, \dots$. The boundary conditions give

$$w_{0,j} = w_{m,j} = 0, \quad \text{for each } j = 1, 2, 3, \dots, \quad (12.20)$$

and the initial condition implies that

$$w_{i,0} = f(x_i), \quad \text{for each } i = 1, 2, \dots, m-1. \quad (12.21)$$

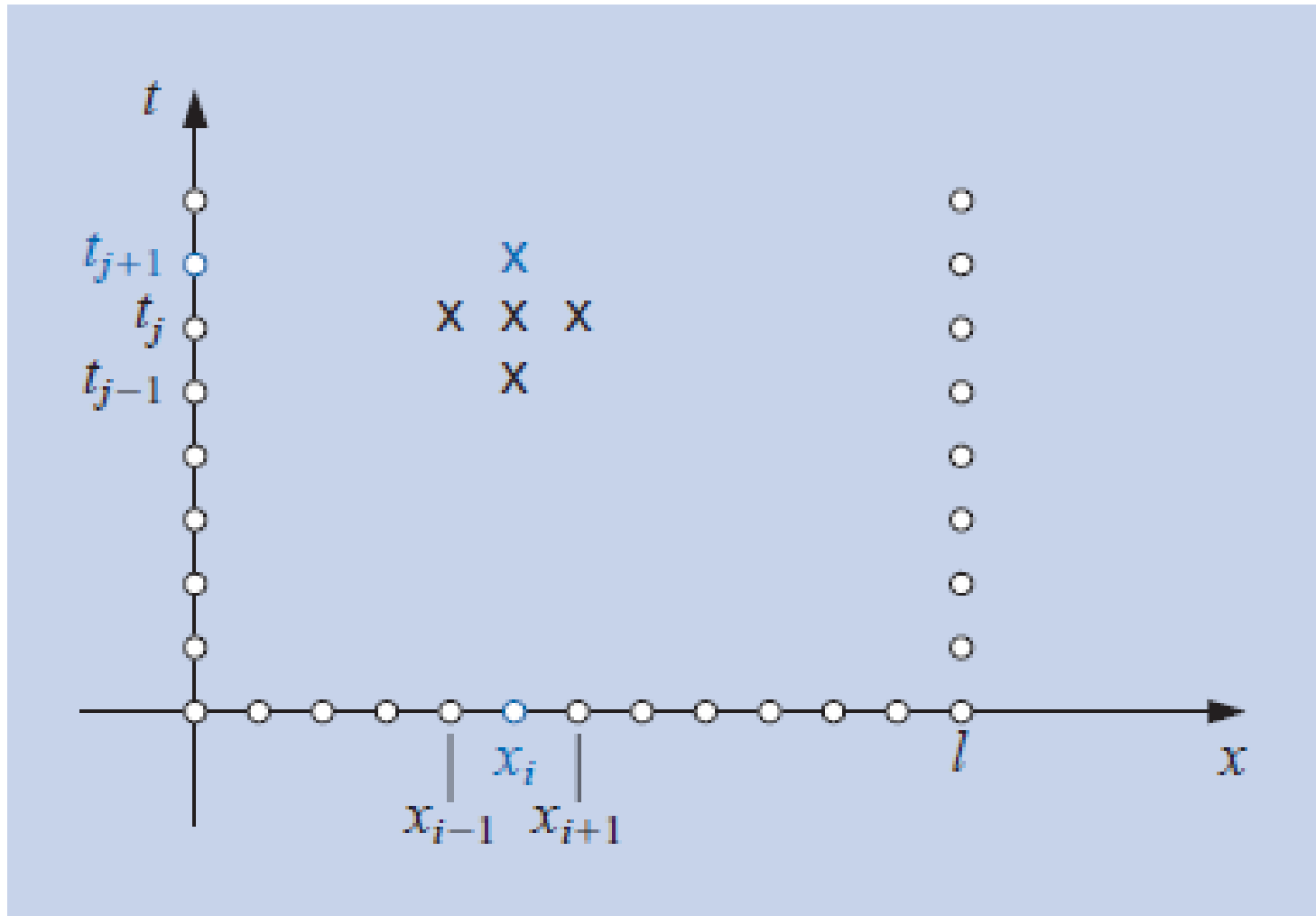
Writing this set of equations in matrix form gives

$$\begin{bmatrix} w_{1,j+1} \\ w_{2,j+1} \\ \vdots \\ w_{m-1,j+1} \end{bmatrix} = \begin{bmatrix} 2(1-\lambda^2) & \lambda^2 & 0 & \dots & 0 \\ \lambda^2 & 2(1-\lambda^2) & \lambda^2 & \dots & 0 \\ 0 & \dots & \dots & \dots & \lambda^2 \\ \vdots & \dots & \dots & \dots & \vdots \\ 0 & \dots & 0 & \dots & \lambda^2 & 2(1-\lambda^2) \end{bmatrix} \begin{bmatrix} w_{1,j} \\ w_{2,j} \\ \vdots \\ w_{m-1,j} \end{bmatrix} - \begin{bmatrix} w_{1,j-1} \\ w_{2,j-1} \\ \vdots \\ w_{m-1,j-1} \end{bmatrix}. \quad (12.22)$$

Equations (12.18) and (12.19) imply that the $(j + 1)$ st time step requires values from the j th and $(j - 1)$ st time steps. (See Figure 12.12.) This produces a minor starting problem because values for $j = 0$ are given by Eq. (12.20), but values for $j = 1$, which are needed in Eq. (12.18) to compute $w_{i,2}$, must be obtained from the initial-velocity condition

$$\frac{\partial u}{\partial t}(x, 0) = g(x), \quad 0 \leq x \leq l.$$

Figure 12.12



One approach is to replace $\partial u/\partial t$ by a forward-difference approximation,

$$\frac{\partial u}{\partial t}(x_i, 0) = \frac{u(x_i, t_1) - u(x_i, 0)}{k} - \frac{k}{2} \frac{\partial^2 u}{\partial t^2}(x_i, \tilde{\mu}_i), \quad (12.23)$$

for some $\tilde{\mu}_i$ in $(0, t_1)$. Solving for $u(x_i, t_1)$ in the equation gives

$$\begin{aligned} u(x_i, t_1) &= u(x_i, 0) + k \frac{\partial u}{\partial t}(x_i, 0) + \frac{k^2}{2} \frac{\partial^2 u}{\partial t^2}(x_i, \tilde{\mu}_i) \\ &= u(x_i, 0) + kg(x_i) + \frac{k^2}{2} \frac{\partial^2 u}{\partial t^2}(x_i, \tilde{\mu}_i). \end{aligned}$$

Deleting the truncation term gives the approximation,

$$w_{i,1} = w_{i,0} + kg(x_i), \quad \text{for each } i = 1, \dots, m-1. \quad (12.24)$$

However, this approximation has truncation error of only $O(k)$ whereas the truncation error in Eq. (12.19) is $O(k^2)$.

To obtain a better approximation to $u(x_i, 0)$, expand $u(x_i, t_1)$ in a second Maclaurin polynomial in t . Then

$$u(x_i, t_1) = u(x_i, 0) + k \frac{\partial u}{\partial t}(x_i, 0) + \frac{k^2}{2} \frac{\partial^2 u}{\partial t^2}(x_i, 0) + \frac{k^3}{6} \frac{\partial^3 u}{\partial t^3}(x_i, \hat{\mu}_i),$$

for some $\hat{\mu}_i$ in $(0, t_1)$. If f'' exists, then

$$\frac{\partial^2 u}{\partial t^2}(x_i, 0) = \alpha^2 \frac{\partial^2 u}{\partial x^2}(x_i, 0) = \alpha^2 \frac{d^2 f}{dx^2}(x_i) = \alpha^2 f''(x_i)$$

and

$$u(x_i, t_1) = u(x_i, 0) + kg(x_i) + \frac{\alpha^2 k^2}{2} f''(x_i) + \frac{k^3}{6} \frac{\partial^3 u}{\partial t^3}(x_i, \hat{\mu}_i).$$

This produces an approximation with error $O(k^3)$:

$$w_{i1} = w_{i0} + kg(x_i) + \frac{\alpha^2 k^2}{2} f''(x_i).$$

If $f \in C^4[0, 1]$ but $f''(x_i)$ is not readily available, we can use the difference equation in Eq. (4.9) to write

$$f''(x_i) = \frac{f(x_{i+1}) - 2f(x_i) + f(x_{i-1}))}{h^2} - \frac{h^2}{12} f^{(4)}(\tilde{\xi}_i),$$

for some $\tilde{\xi}_i$ in (x_{i-1}, x_{i+1}) . This implies that

$$u(x_i, t_1) = u(x_i, 0) + kg(x_i) + \frac{k^2\alpha^2}{2h^2}[f(x_{i+1}) - 2f(x_i) + f(x_{i-1}))] + O(k^3 + h^2k^2).$$

Because $\lambda = k\alpha/h$, we can write this as

$$\begin{aligned} u(x_i, t_1) &= u(x_i, 0) + kg(x_i) + \frac{\lambda^2}{2}[f(x_{i+1}) - 2f(x_i) + f(x_{i-1}))] + O(k^3 + h^2k^2) \\ &= (1 - \lambda^2)f(x_i) + \frac{\lambda^2}{2}f(x_{i+1}) + \frac{\lambda^2}{2}f(x_{i-1}) + kg(x_i) + O(k^3 + h^2k^2). \end{aligned}$$

Thus, the difference equation,

$$w_{i,1} = (1 - \lambda^2)f(x_i) + \frac{\lambda^2}{2}f(x_{i+1}) + \frac{\lambda^2}{2}f(x_{i-1}) + kg(x_i), \quad (12.25)$$

can be used to find $w_{i,1}$, for each $i = 1, 2, \dots, m - 1$. To determine subsequent approximates we use the system in (12.22).

Algorithm 12.4 uses Eq. (12.25) to approximate $w_{i,1}$, although Eq. (12.24) could also be used. It is assumed that there is an upper bound for the value of t to be used in the stopping technique, and that $k = T/N$, where N is also given.

ALGORITHM

12.4

Wave Equation Finite-Difference

To approximate the solution to the wave equation

$$\frac{\partial^2 u}{\partial t^2}(x, t) - \alpha^2 \frac{\partial^2 u}{\partial x^2}(x, t) = 0, \quad 0 < x < l, \quad 0 < t < T,$$

subject to the boundary conditions

$$u(0, t) = u(l, t) = 0, \quad 0 < t < T,$$

and the initial conditions

$$u(x, 0) = f(x), \quad \text{and} \quad \frac{\partial u}{\partial t}(x, 0) = g(x), \quad \text{for} \quad 0 \leq x \leq l,$$

INPUT endpoint l ; maximum time T ; constant α ; integers $m \geq 2, N \geq 2$.

OUTPUT approximations $w_{i,j}$ to $u(x_i, t_j)$ for each $i = 0, \dots, m$ and $j = 0, \dots, N$.

Step 1 Set $h = l/m$;
 $k = T/N$;
 $\lambda = k\alpha/h$.

Step 2 For $j = 1, \dots, N$ set $w_{0,j} = 0$;
 $w_{m,j} = 0$;

Step 3 Set $w_{0,0} = f(0)$;
 $w_{m,0} = f(l)$.

Step 4 For $i = 1, \dots, m - 1$ (*Initialize for $t = 0$ and $t = k$.*)
 set $w_{i,0} = f(ih)$;

$$w_{i,1} = (1 - \lambda^2)f(ih) + \frac{\lambda^2}{2}[f((i+1)h) + f((i-1)h)] + kg(ih).$$

Step 5 For $j = 1, \dots, N - 1$ (*Perform matrix multiplication.*)
 for $i = 1, \dots, m - 1$
 set $w_{i,j+1} = 2(1 - \lambda^2)w_{i,j} + \lambda^2(w_{i+1,j} + w_{i-1,j}) - w_{i,j-1}$.

Step 6 For $j = 0, \dots, N$
 set $t = jk$;
 for $i = 0, \dots, m$
 set $x = ih$;
OUTPUT $(x, t, w_{i,j})$.

Step 7 STOP. (*The procedure is complete.*)

Example 1

Approximate the solution to the hyperbolic problem

$$\frac{\partial^2 u}{\partial t^2}(x, t) - 4 \frac{\partial^2 u}{\partial x^2}(x, t) = 0, \quad 0 < x < 1, \quad 0 < t,$$

with boundary conditions

$$u(0, t) = u(1, t) = 0, \quad \text{for } 0 < t,$$

and initial conditions

$$u(x, 0) = \sin(\pi x), \quad 0 \leq x \leq 1, \quad \text{and} \quad \frac{\partial u}{\partial t}(x, 0) = 0, \quad 0 \leq x \leq 1,$$

using $h = 0.1$ and $k = 0.05$. Compare the results with the exact solution

$$u(x, t) = \sin \pi x \cos 2\pi t.$$

Example 1

Solution Choosing $h = 0.1$ and $k = 0.05$ gives $\lambda = 1$, $m = 10$, and $N = 20$. We will choose a maximum time $T = 1$ and apply the Finite-Difference Algorithm 12.4. This produces the approximations $w_{i,N}$ to $u(0.1i, 1)$ for $i = 0, 1, \dots, 10$. These results are shown in Table 12.6 and are correct to the places given. ■

Console

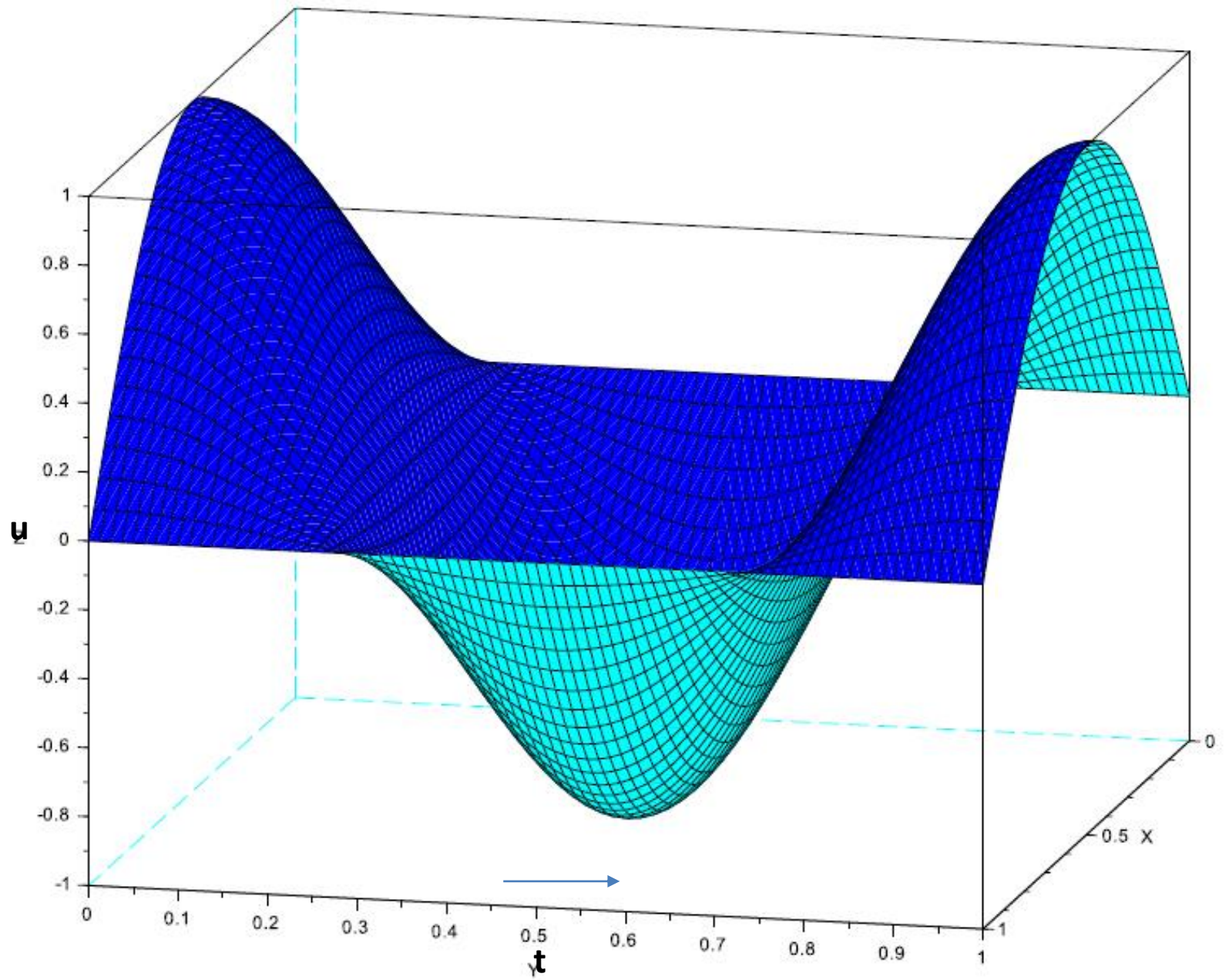
```
-->u(:,21)
ans =

    0.
    0.3090170
    0.5877853
    0.8090170
    0.9510565
    1.
    0.9510565
    0.8090170
    0.5877853
    0.3090170
    0.
```

Table 12.6

| x_i | $w_{i,20}$ |
|-------|--------------|
| 0.0 | 0.0000000000 |
| 0.1 | 0.3090169944 |
| 0.2 | 0.5877852523 |
| 0.3 | 0.8090169944 |
| 0.4 | 0.9510565163 |
| 0.5 | 1.0000000000 |
| 0.6 | 0.9510565163 |
| 0.7 | 0.8090169944 |
| 0.8 | 0.5877852523 |
| 0.9 | 0.3090169944 |
| 1.0 | 0.0000000000 |

Minha implementação em Scilab



Minha implementação em Scilab, $m = 40$

The results of the example were very accurate, more so than the truncation error $O(k^2 + h^2)$ would lead us to believe. This is because the true solution to the equation is infinitely differentiable. When this is the case, Taylor series gives

$$\begin{aligned} & \frac{u(x_{i+1}, t_j) - 2u(x_i, t_j) + u(x_{i-1}, t_j))}{h^2} \\ &= \frac{\partial^2 u}{\partial x^2}(x_i, t_j) + 2 \left[\frac{h^2}{4!} \frac{\partial^4 u}{\partial x^4}(x_i, t_j) + \frac{h^4}{6!} \frac{\partial^6 u}{\partial x^6}(x_i, t_j) + \dots \right] \end{aligned}$$

and

$$\begin{aligned} & \frac{u(x_i, t_{j+1}) - 2u(x_i, t_j) + u(x_i, t_{j-1}))}{k^2} \\ &= \frac{\partial^2 u}{\partial t^2}(x_i, t_j) + 2 \left[\frac{k^2}{4!} \frac{\partial^4 u}{\partial t^4}(x_i, t_j) + \frac{h^4}{6!} \frac{\partial^6 u}{\partial t^6}(x_i, t_j) + \dots \right]. \end{aligned}$$

Since $u(x, t)$ satisfies the partial differential equation,

$$\begin{aligned} & \frac{u(x_i, t_{j+1}) - 2u(x_i, t_j) + u(x_i, t_{j-1}))}{k^2} - \alpha^2 \frac{u(x_{i+1}, t_j) - 2u(x_i, t_j) + u(x_{i-1}, t_j))}{h^2} \\ &= 2 \left[\frac{1}{4!} \left(k^2 \frac{\partial^4 u}{\partial t^4}(x_i, t_j) - \alpha^2 h^2 \frac{\partial^4 u}{\partial x^4}(x_i, t_j) \right) \right. \\ & \quad \left. + \frac{1}{6!} \left(k^4 \frac{\partial^6 u}{\partial t^6}(x_i, t_j) - \alpha^2 h^4 \frac{\partial^6 u}{\partial x^6}(x_i, t_j) \right) + \dots \right]. \end{aligned} \tag{12.26}$$

However, differentiating the wave equation gives

$$\begin{aligned} k^2 \frac{\partial^4 u}{\partial t^4}(x_i, t_j) &= k^2 \frac{\partial^2}{\partial t^2} \left[\alpha^2 \frac{\partial^2 u}{\partial x^2}(x_i, t_j) \right] = \alpha^2 k^2 \frac{\partial^2}{\partial x^2} \left[\frac{\partial^2 u}{\partial t^2}(x_i, t_j) \right] \\ &= \alpha^2 k^2 \frac{\partial^2}{\partial x^2} \left[\alpha^2 \frac{\partial^2 u}{\partial x^2}(x_i, t_j) \right] = \alpha^4 k^2 \frac{\partial^4 u}{\partial x^4}(x_i, t_j), \end{aligned}$$

and we see that since $\lambda^2 = (\alpha^2 k^2 / h^2) = 1$, we have

$$\frac{1}{4!} \left[k^2 \frac{\partial^4 u}{\partial t^4}(x_i, t_j) - \alpha^2 h^2 \frac{\partial^4 u}{\partial x^4}(x_i, t_j) \right] = \frac{\alpha^2}{4!} [\alpha^2 k^2 - h^2] \frac{\partial^4 u}{\partial x^4}(x_i, t_j) = 0.$$

Continuing in this manner, all the terms on the right-hand side of (12.26) are 0, implying that the local truncation error is 0. The only errors in Example 1 are those due to the approximation of $w_{i,1}$ and to round-off.

As in the case of the Forward-Difference method for the heat equation, the Explicit Finite-Difference method for the wave equation has stability problems. In fact, it is necessary that $\lambda = \alpha k / h \leq 1$ for the method to be stable. (See [IK], p. 489.) The explicit method given in Algorithm 12.4, with $\lambda \leq 1$, is $O(h^2 + k^2)$ convergent if f and g are sufficiently differentiable. For verification of this, see [IK], p. 491.

Although we will not discuss them, there are implicit methods that are unconditionally stable. A discussion of these methods can be found in [Am], p. 199, [Mi], or [Sm,G].

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2º Semestre - 2020

Roteiro do curso

- Introdução
- Séries de Fourier
- **Método de Diferenças Finitas**
- Equação do calor transiente (parabólica)
- Equação de Poisson (elíptica)
- **Equação da onda (hiperbólica)**