

Chapter 4

An Introduction to

Numerical Methods for Parabolic Equations

First Session Contents:

- 1) Introduction
- 2) Finite Difference Methods
- 3) Explicit methods for Parabolic Equations
- 4) Truncation Error (T.E.)
- 5) Consistency
- 6) Stability

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Parabolic PDE

In this chapter, we will discuss about the numerical solution of one-dimensional parabolic PDE as given below:

$$\frac{\partial u}{\partial t} = \frac{\partial^2 u}{\partial x^2} \quad a < x < b$$

B.C.'s $\begin{cases} u(a, t) = u_0(t) \\ u(b, t) = u_0(t) \end{cases}$

I.C. $u(x, 0) = f(x)$

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Computational Domain for Parabolic PDEs

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Computational Domain for Parabolic PDEs

Grid Points or Mesh Points

$$x_j = (j - 1)\Delta x = (j - 1)h \quad j = 1, 2, 3, \dots, imax$$

$$t_n = (n - 1)\Delta t = (n - 1)k \quad n = 1, 2, 3, \dots$$

$k = \Delta t$ **Time Step**

$h = \Delta x$ **Grid Spacing**

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Discretization

General PDE $\frac{\partial u}{\partial t} = \frac{\partial^2 u}{\partial x^2}$

Transient Term $\left. \frac{\partial u}{\partial t} \right|_i^n = \frac{u_i^{n+1} - u_i^n}{k} + O(k)$ (Forward Difference)

Diffusion Term $\left. \frac{\partial^2 u}{\partial x^2} \right|_i^n = \frac{u_{i+1}^n - 2u_i^n + u_{i-1}^n}{h^2} + O(h^2)$ (Central Difference)

Substituting transient and diffusion terms in the PDE, we have

$$\frac{u_i^{n+1} - u_i^n}{k} = \frac{u_{i+1}^n - 2u_i^n + u_{i-1}^n}{h^2} + O(k, h^2)$$

$$u_i^{n+1} = u_i^n + r(u_{i+1}^n - 2u_i^n + u_{i-1}^n) \quad \text{where } r = \frac{k}{h^2}$$

$$u_i^{n+1} = u_i^n + r\delta_x^2 u_i^n = (1 + r\delta_x^2)u_i^n \quad \text{where } \delta_x \text{ is the central difference operator}$$

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Explicit Methods for Parabolic PDEs

$\frac{\partial u}{\partial t} = \frac{\partial^2 u}{\partial x^2}$ 1-D transient Heat Conduction Problem

Using the following Eq., the temperature at time step $n+1$ can be obtained from the temperature at time step n .

$$u_i^{n+1} = ru_{i+1}^n + (1 - 2r)u_i^n + ru_{i-1}^n \quad , \quad r = \frac{k}{h^2} \quad , \quad \text{T.E.} = O(k, h^2)$$

At time step $n=1$, temperature is known as initial condition. Therefore, Temperature at $n+1, n+2, \dots$ can be obtained. $[u^{n+1}] = [A][u^n]$

A method which calculates the state of a system at a later time from the state of the system at the current time is called **Explicit Method**

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Explicit Method (FTCS)

Transient Term (Forward Difference) } **FTCS**
 Diffusion Term (Central Difference) } **Forward Time Central Space**

$$u_i^{n+1} = u_i^n + r(u_{i+1}^n - 2u_i^n + u_{i-1}^n)$$

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Example (FTCS)

$\frac{\partial u}{\partial t} = \frac{\partial^2 u}{\partial x^2} \quad 0 \leq x \leq 1$

Initial Condition $u(x, 0) = \begin{cases} 2x & 0 \leq x \leq \frac{1}{2} \\ 2(1-x) & \frac{1}{2} \leq x \leq 1 \end{cases}$

Boundary Conditions $\begin{cases} u(0, t) = 0 \\ u(1, t) = 0 \end{cases}$

$h = \frac{1}{10} \quad k = \frac{1}{1000} \quad r = \frac{k}{h^2} = \frac{1}{10}$

↓

$$u_i^{n+1} = \frac{1}{10}(u_{i-1}^n + 8u_i^n + u_{i+1}^n)$$

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Example (FTCS)

$$u_i^{n+1} = \frac{1}{10}(u_{i-1}^n + 8u_i^n + u_{i+1}^n)$$

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Example (FTCS)

Numerical Solution

$$u_i^{n+1} = \frac{1}{10}(u_{i-1}^n + 8u_i^n + u_{i+1}^n)$$

$$u_6^2 = \frac{1}{10}(u_5^1 + 8 \times u_6^1 + u_7^1) = \frac{1}{10}[0.8 + (8 \times 1) + 0.8] = 0.9600$$

$$u_5^3 = \frac{1}{10}(u_4^2 + 8 \times u_5^2 + u_6^2) = \frac{1}{10}[0.6 + (8 \times 0.8) + 0.96] = 0.7960$$

Exact Solution

$$u = \frac{8}{\pi^2} \sum_{n=1}^{\infty} \frac{1}{n^2} (\sin \frac{n\pi}{2}) (\sin n\pi x) \exp(-n^2 \pi^2 t)$$

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Example (FTCS)

$r = \frac{k}{h^2} = 0.10$

Time	$i = 1$ $x = 0$	$i = 2$ $x = 0.1$	$i = 3$ $x = 0.2$	$i = 4$ $x = 0.3$	$i = 5$ $x = 0.4$	$i = 6$ $x = 0.5$	$i = 7$ $x = 0.6$
0.000	0	0.2000	0.4000	0.6000	0.8000	1.0000	0.8000
0.001	0	0.2000	0.4000	0.6000	0.8000	0.9600	0.8000
0.002	0	0.2000	0.4000	0.6000	0.7960	0.9280	0.7960
0.003	0	0.2000	0.4000	0.5996	0.7896	0.9016	0.7896
0.004	0	0.2000	0.4000	0.5986	0.7818	0.8792	0.7818
0.005	0	0.2000	0.3999	0.5971	0.7732	0.8597	0.7732
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
0.01	0	0.1996	0.3968	0.5822	0.7281	0.7867	0.7281
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
0.02	0	0.1938	0.3781	0.5373	0.6486	0.6891	0.6486

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Example (FTCS)

$r = \frac{k}{h^2} = 0.10$

Time	$i = 1$ $x = 0$	$i = 2$ $x = 0.1$	$i = 3$ $x = 0.2$	$i = 4$ $x = 0.3$	$i = 5$ $x = 0.4$	$i = 6$ $x = 0.5$	$i = 7$ $x = 0.6$
0.000	0	0.2000	0.4000	0.6000	0.8000	1.0000	0.8000
0.001	0	0.2000	0.4000	0.6000	0.8000	0.9600	0.8000
0.002	0	0.2000	0.4000	0.6000	0.7960	0.9280	0.7960
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Example (FTCS)

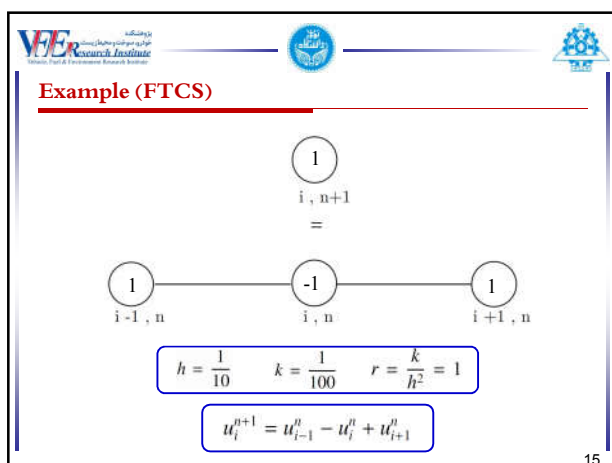
	Time	Numerical Solution (x = 0.3)	Exact Solution (x = 0.3)	Absolute Error	Relative Error %
x = 0.30	t = 0.005	0.5971	0.5966	0.0005	0.08
	t = 0.01	0.5822	0.5799	0.0023	0.4
	t = 0.02	0.5373	0.5334	0.0039	0.7
	t = 0.1	0.2472	0.2444	0.0028	1.1
	Time	Numerical Solution (x = 0.5)	Exact Solution (x = 0.5)	Absolute Error	Relative Error %
x = 0.50	t = 0.005	0.8597	0.8404	0.0193	2.3
	t = 0.01	0.7867	0.7743	0.0124	1.6
	t = 0.02	0.6891	0.6809	0.0082	1.2
	t = 0.1	0.3056	0.3021	0.0035	1.2

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Example (FTCS)

	Time	Numerical Solution (x = 0.3)	Exact Solution (x = 0.3)	Absolute Error	Relative Error %
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Example (FTCS)

r = 0.10								r = 1							
t	i=1	i=2	i=3	i=4	i=5	i=6	i=7	t	i=1	i=2	i=3	i=4	i=5	i=6	i=7
x=0	x=0.1	x=0.2	x=0.3	x=0.4	x=0.5	x=0.6		x=0	x=0.1	x=0.2	x=0.3	x=0.4	x=0.5	x=0.6	
0.000	0	0.2000	0.4000	0.6000	0.8000	1.0000	0.8000	0.000	0	0.2	0.4	0.6	0.8	1.0	0.8
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0.002	0	0.2000	0.4000	0.6000	0.7960	0.9280	0.7960	0.002	0	0.2	0.4	0.6	0.4	1.0	0.4
0.003	0	0.2000	0.4000	0.5996	0.7896	0.9016	0.7896	0.003	0	0.2	0.4	0.2	1.2	-0.2	1.2
0.004	0	0.2000	0.4000	0.5986	0.7818	0.8792	0.7818	0.004	0	0.2	0.0	1.4	-1.2	2.6	-1.2
0.005	0	0.2000	0.3999	0.5971	0.7732	0.8597	0.7732	0.005	0	0.2					
...
0.01	0	0.1996	0.3968	0.5822	0.7281	0.7867	0.7281	0.01	0						
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Example (FTCS)

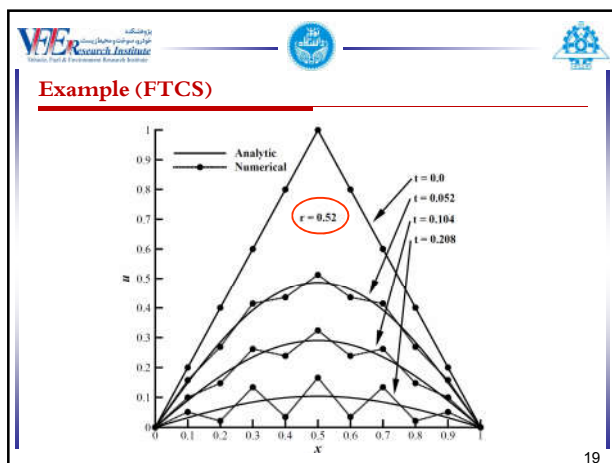
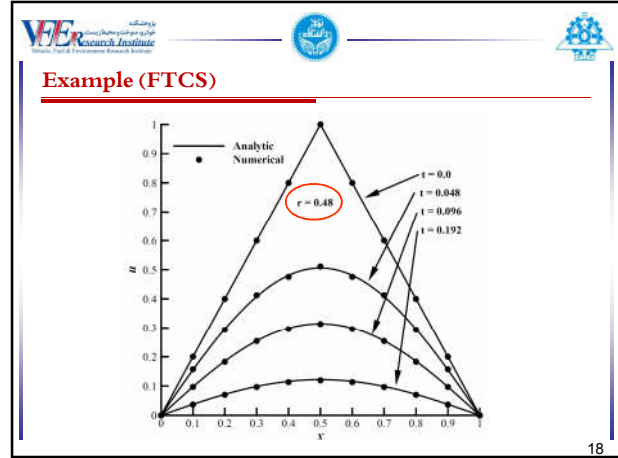
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$r = 1$

t	i=1	i=2	i=3	i=4	i=5	i=6	i=7
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0.01	0	0.2	0.4	0.6	0.8	0.6	0.8
0.02	0	0.2	0.4	0.6	0.4	1.0	0.4
0.03	0	0.2	0.4	0.2	1.2	-0.2	1.2
0.04	0	0.2	0.0	1.4	-1.2	2.6	-1.2

The value of r plays an important role in explicit methods

$$0 < r \leq \frac{1}{2}$$


Propagation Speed of a Disturbance

Consider a 1-D parabolic PDE with all initial conditions are zero except at point A

$$u_i^{n+1} = u_i^n + r(u_{i+1}^n - 2u_i^n + u_{i-1}^n)$$

Propagation Speed of a Disturbance

Slope of Characteristics lines:
 $\frac{dt}{dx} = \frac{\Delta t}{\Delta x} \neq 0$

The propagation speed of a disturbance in parabolic PDEs for given time is **infinity**

?

The propagation speed of a disturbance in finite-difference form of the Parabolic PDE for given time is $\Delta x / \Delta t$

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Propagation Speed of a Disturbance

Slope of Characteristics lines:
 $\frac{dt}{dx} = \frac{\Delta t}{\Delta x} \neq 0$

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The propagation speed of a disturbance in finite-difference form of the Parabolic PDE for given time is $\Delta x / \Delta t$

Solution: $\frac{\Delta x}{\Delta t} = \frac{\Delta x}{r(\Delta x)^2} = \frac{1}{r\Delta x} \left\{ \begin{array}{l} \Delta x \rightarrow 0 \\ r = \text{Const.} \end{array} \right\} \Rightarrow \frac{dt}{dx} \rightarrow 0$

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Truncation Error (T.E.)

$L[u] = f \quad (x, t) \in \Omega$ General PDE

$D[U_i^n] = f_i^n \quad (x, t) \in \Omega$ Discretized PDE

$U_i^n \approx u_i^n = u(x_i, t_n)$ If h and k are small enough then T.E. approaches to zero and the discretized form is a good approximation of general PDE

Truncation Error is Defined as $T_i^n = D[u_i^n] - f_i^n$ as $h, k \rightarrow 0$ independently

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Truncation Error (T.E.)

$$\frac{\partial u}{\partial t} - \frac{\partial^2 u}{\partial x^2} = \underbrace{\frac{u_i^{n+1} - u_i^n}{\Delta t} - \frac{1}{(\Delta x)^2}(u_{i+1}^n - 2u_i^n + u_{i-1}^n)}_{\text{FDE}}$$

$$+ \underbrace{\left[-\frac{\partial^2 u}{\partial t^2} \Big|_i \frac{\Delta t}{2} + \frac{\partial^4 u}{\partial x^4} \Big|_i \frac{(\Delta x)^2}{12} + \dots \right]}_{\text{T.E.}}$$

T.E. is obtained by writing the Taylor series of each term around point (x_i, t_n)

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Truncation Error (T.E.)

$$\underbrace{\frac{\partial u}{\partial t} - \frac{\partial^2 u}{\partial x^2}}_{\text{PDE}} = \underbrace{\frac{u_i^{n+1} - u_i^n}{\Delta t} - \frac{1}{(\Delta x)^2}(u_{i+1}^n - 2u_i^n + u_{i-1}^n)}_{\text{FDE}} + \underbrace{\left[-\frac{\partial^2 u}{\partial t^2} \Big|_i \frac{\Delta t}{2} + \frac{\partial^4 u}{\partial x^4} \Big|_i \frac{(\Delta x)^2}{12} + \dots \right]}_{\text{T.E.}}$$

T.E. is obtained by writing the Taylor series of each term around point (x_i, t_n)

- ❑ Is the finite difference form of PDE acceptable?
- ❑ Does the marching method give a good approximation of PDE?

The finite-difference form of PDE should satisfy both **Stability** and **Consistency** conditions

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Consistency

Finite-difference form of a PDE is consistent if:

$$\lim_{h, \Delta t \rightarrow 0} \text{T.E.}_i^n = 0$$

In some finite-difference methods the T.E. is $\frac{O(\Delta x)}{O(\Delta t)}$

These methods are consistent if $\frac{\Delta x}{\Delta t} \rightarrow 0$

DuFort-Frankel method $\frac{u_i^{n+1} - u_i^{n-1}}{2\Delta t} = \frac{1}{(\Delta x)^2} [u_{i+1}^n - (u_i^{n+1} + u_i^{n-1}) + u_{i-1}^n]$

T.E. $\leftarrow \left[\frac{1}{12} \frac{\partial^4 u}{\partial x^4} \Big|_i (\Delta x)^2 - \frac{\partial^2 u}{\partial t^2} \Big|_i \left(\frac{\Delta t}{2} \right)^2 - \frac{1}{6} \frac{\partial^3 u}{\partial t^3} \Big|_i (\Delta t)^3 \right]$

The DuFort-Frankel method is consistent if $\lim_{\Delta x, \Delta t \rightarrow 0} \left(\frac{\Delta x}{\Delta t} \right) = 0$

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Stability

- ❖ Concept of stability occurs in marching problems
- ❖ A numerical method is **stable** if the errors of any kind (Round of error and Truncation error) will not grow (increasing unconditionally) during time marching.
- ❖ Generally, analysis of Consistency in a numerical method is more easier than analysis of Stability
- ❖ An explicit method is stable if $r = 1 - \frac{\Delta t}{(\Delta x)^2} \leq \frac{1}{2}$

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Example

Forward Time Central Space


$u_i^{n+1} = u_i^n + r(u_{i+1}^n - 2u_i^n + u_{i-1}^n)$

$r = 1$


$u_i^{n+1} = u_{i-1}^n - u_i^n + u_{i+1}^n$

$u^{n+1} = 100 - 0 + 100 = 200$


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CFD

Convergence for Marching Problems

A numerical method **converges** if its global discretization error approaches zero as the mesh is refined

$$\lim_{h, k \rightarrow 0} |U_i^n - u_i^n| = 0 \quad (x_i, t_n) \in \Omega$$

Generally, a numerical method **converges** if it is both **stable** and **consistence**

Lax's Equivalence Theorem
Given a well-posed initial value problem and a finite-difference approximation to it that satisfies the consistency condition, stability is the necessary and sufficient condition for convergence

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