

## *Lesson 4*

# *Numerical Solution of PDE*

### **4.2 Parabolic PDE**

- *Semi-discretization*
- *Full finite difference discretization*
- ***Fourier Analysis***



# Numerical Methods for PDE: Parabolic PDE



Another useful way to analyze is to use Fourier analysis.

Recall that, on  $L(I_h)$ , we define the inner product

$$\langle u, v \rangle_h = h \sum_{k=1}^{N-1} u(kh)v(kh)$$

with the corresponding norm  $\|v\|_h$  and  $\varphi_m(x) = \sin \pi mx, m = 1, \dots, N - 1$ , form an orthogonal basis.

Also,

$$D_h^2 \varphi_m = -\lambda_m \varphi_m, \quad \lambda_m = \frac{2}{h^2} (\cos \pi mh - 1) = \frac{4}{h^2} \sin^2 \frac{\pi mh}{2},$$

where the eigenvalues satisfy

$$8 < \lambda_1 < \lambda_2 < \dots < \lambda_{N-1} < \frac{4}{h^2}.$$

We write the semi-discrete solution  $u_h(t, \cdot) = \sum_{m=1}^{N-1} a_m^h \varphi_m$ . Then,

$$\frac{\partial u_h}{\partial t} = \sum_{m=1}^{N-1} \frac{da_m^h}{dt} \varphi_m, \quad D_h^2 u_h = - \sum_{m=1}^{N-1} a_m^h \lambda_m \varphi_m.$$

Thus,

$$\frac{da_m^h}{dt} = -c a_m^h \lambda_m \quad \text{so} \quad a_m^h(t) = a_m^h(0) e^{-c \lambda_m t}.$$

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The numbers  $a_m^h(0)$  are the coefficients in the discrete Fourier sine transform of the initial data:

$$u_0(x) = \sum_{m=1}^{N-1} a_m^h(0) \varphi_m(x), \quad x \in \bar{I}_h.$$

Thus, the solution of the semi-discrete system may be written as

$$u(x, t) = \sum_{m=1}^{N-1} a_m^h(0) e^{-c \lambda_m t} \varphi_m(x), \quad x \in \bar{I}_h.$$

For the fully discrete forward-centered scheme, we write the solution as

$$u^j(x) = \sum_{m=1}^{N-1} A_m^j \varphi_m(x).$$

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For the fully discrete forward-centered scheme, we write the solution at time  $t = jk$  as

$$u^j(x) = \sum_{m=1}^{N-1} A_m^j \varphi_m(x).$$

The difference equation then gives

$$\sum_{m=1}^{N-1} A_m^{j+1} \varphi_m(x) = \sum_{m=1}^{N-1} A_m^j \varphi_m(x) - ck \sum_{m=1}^{N-1} A_m^j \lambda_m \varphi_m(x)$$

yielding

$$A_m^{j+1} = (1 - ck\lambda_m) A_m^j.$$

It follows that

$$A_m^j = (1 - ck\lambda_m)^j a_m^h(0), \quad u^j = \sum_{m=1}^{N-1} (1 - ck\lambda_m)^j a_m^h(0) \varphi_m.$$

If we assume  $ck/h^2 \leq 1/2$ , then  $ck\lambda_m \leq ck(4/h^2) \leq 2$  and hence  $|1 - ck\lambda_m| \leq 1$  for all  $m$  and the solution remains bounded. On the other hand, if  $|1 - ck\lambda_m| > 1$  for some  $m$ , the initial data will increase exponentially.

# Numerical Methods for PDE: Parabolic PDE



This idea can be used for rigorous stability analysis.

In the matrix form, we write

$$u^{j+1} = (I + ckD_h^2)u^j + kf^j, \quad j = 0, 1, \dots, M-1,$$

where  $(I + ckD_h^2)$  is a symmetric operator on  $L(I_h)$ . Thus,

$$\|(I + ckD_h^2)v\|_h \leq (\max_m |1 - ck\lambda_m|) \|v\|_h$$

and, therefore, we have

$$\|(I + ckD_h^2)\|_h \leq \max_m |1 - ck\lambda_m|.$$

Finally,

$$\|u^j\|_h \leq \max_m |1 - ck\lambda_m| \|u^{j-1}\|_h + k \|f^{j-1}\|_h \leq \left( \max_m |1 - ck\lambda_m| \right)^j \|u^0\|_h + Mk \max_j \|f^j\|_h.$$

Because of the condition  $ck/h^2 \leq 1/2$ , which we know is not only sufficient but necessary for stability, forward-centered difference method is called **conditionally stable**.

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From stability, we obtain the convergence result in the same way as earlier. Let  $e_n^j = u_n^j - u(nh, jk)$ , we have

$$\begin{aligned} \frac{e_n^{j+1} - e_n^j}{k} &= c \frac{e_{n+1}^j - 2e_n^j + e_{n-1}^j}{h^2} - \ell_n^j, \quad 0 < n < N, j = 0, 1, \dots, M-1, \\ e_0^j &= e_N^j = 0, \quad j = 0, 1, \dots, M-1, \\ e_n^0 &= 0, \quad 0 < n < N. \end{aligned}$$

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The stability result then yields

$$\|e^j\|_h \leq \left( \max_m |1 - ck\lambda_m| \right)^j \|e^0\|_h + Mk \max_j \|\ell^j\|_h \leq T \max_j \|\ell^j\|_h.$$

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# *Numerical Solution of PDE*

### ***4.2 Parabolic PDE***

- *Semi-discretization*
- *Full finite difference discretization*
- *Fourier Analysis*
- ***Unconditional Stability***



# Numerical Methods for PDE: Parabolic PDE



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# Numerical Methods for PDE: Parabolic PDE



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The backward-centered difference method for the heat equation leads to

$$\frac{u_n^{j+1} - u_n^j}{k} = c \frac{u_{n+1}^{j+1} - 2u_n^{j+1} + u_{n-1}^{j+1}}{h^2} + f_n^{j+1}, \quad 0 < n < N, j = 0, 1, \dots, M-1,$$
$$u_0^j = u_N^j = 0, \quad j = 0, 1, \dots, M,$$
$$u_n^0 = u_0(nh), \quad 0 < n < N.$$

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Thus,  $u_n^{j+1}$  can be obtained by solving a tridiagonal system

$$-\lambda u_{n+1}^{j+1} + (1 + 2\lambda)u_n^{j+1} - \lambda u_{n-1}^{j+1} = u_n^j + kf_n^{j+1}, \quad u_0^{j+1} = u_N^{j+1} = 0.$$

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The operator  $I - ckD_h^2$  has eigenvalues  $1 + ck\lambda_m$  which are all greater than one, so  $\|(I - ckD_h^2)^{-1}v\|_h \leq \|v\|_h$ .



We, therefore, have

$$\|u^{j+1}\|_h = \|(I - ckD_h^2)^{-1}(u^j + kf^{j+1})\|_h \leq \|u^j\|_h + k\|f^j\|_h, \quad j = 0, 1, \dots, M-1.$$



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Combining the stability estimate with the local truncation error, we obtain the error estimate

$$\max_{1 \leq j \leq M} \|e^j\|_h = O(k + h^2)$$

for the method.

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If we use trapezoidal rule to discretize in time, we get the **Crank-Nicolson method**:

$$\frac{u^{j+1} - u^j}{k} = \frac{c}{2} (D_h^2 u^j + D_h^2 u^{j+1}) + \frac{1}{2} (f^j + f^{j+1}).$$

# Numerical Methods for PDE: Parabolic PDE



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Using the Taylor expansion about the point  $(nh, (j + 1/2)k)$ , it is straightforward to show that the local truncation error is  $O(k^2 + h^2)$ , so the Crank-Nicholson method is second order in both space and time.

# Numerical Methods for PDE: Parabolic PDE



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Thus

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The eigenvalues of  $\left( I - \frac{1}{2} ck D_h^2 \right)^{-1} \left( I + \frac{1}{2} ck D_h^2 \right)$  is  $(1 - ck\lambda_m/2)/(1 + ck\lambda_m/2)$  which is less than 1. Therefore, we get unconditional stability. The Crank-Nicolson method converges with  $O(k^2 + h^2)$ .

## *Lesson 4*

# *Numerical Solution of PDE*

### ***4.3 Hyperbolic PDE***



# Numerical Methods for PDE: Hyperbolic PDE



For the first example, we take the advection equation

$$\frac{\partial u}{\partial t} + c \frac{\partial u}{\partial x} = 0,$$

where  $c$  is a constant. We consider an initial value problem so that the function  $u = u(x, t)$  is given when  $t = 0$  and is to be found for  $t > 0$ .

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The spatial domain can be an interval in which case we will need to impose boundary conditions to obtain a well posed initial value-boundary value problem, or the entire real line (no boundaries) for which the pure initial value problem is well-posed.

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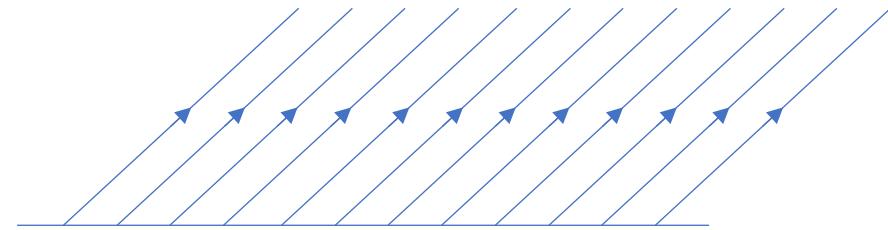
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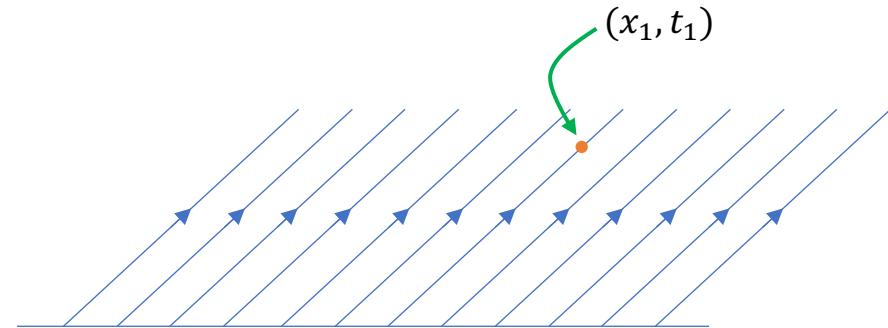
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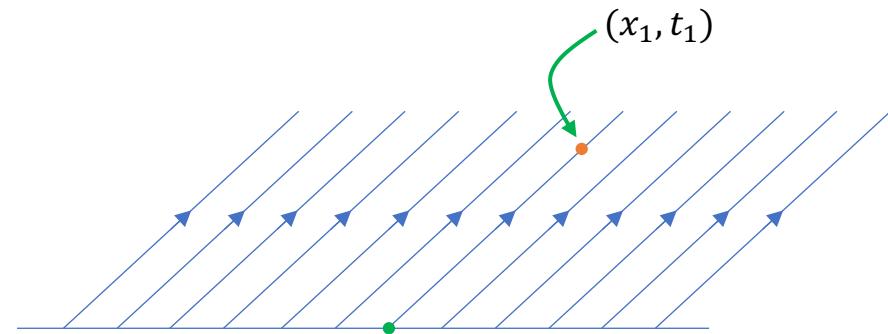
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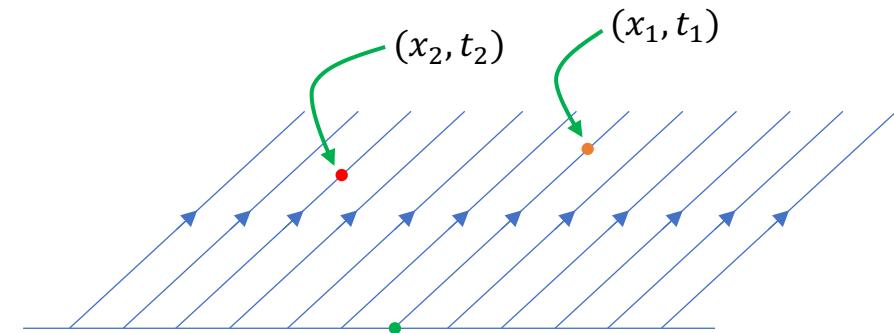
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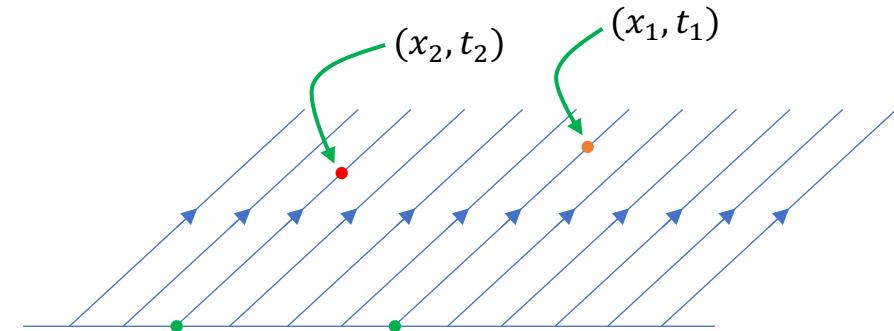
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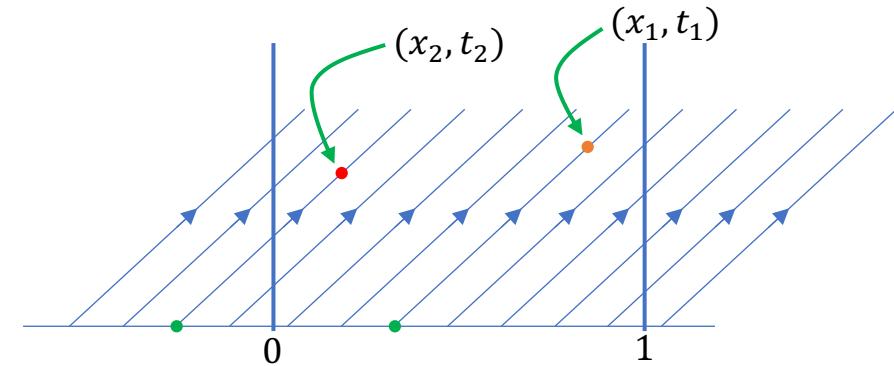


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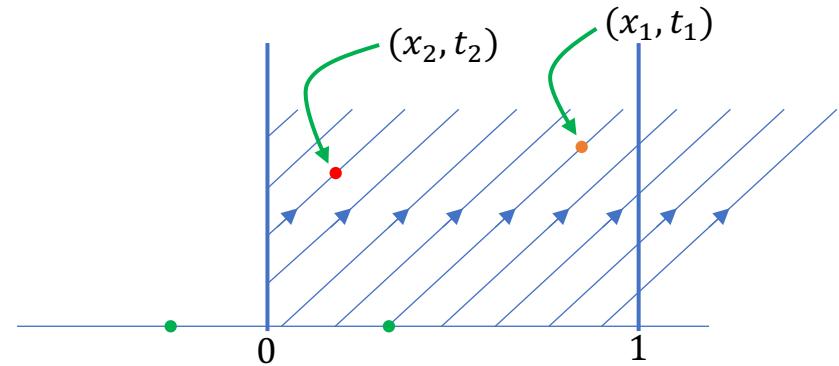


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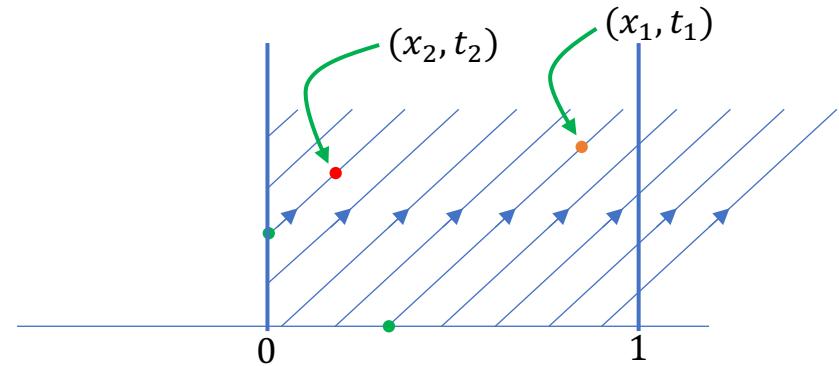


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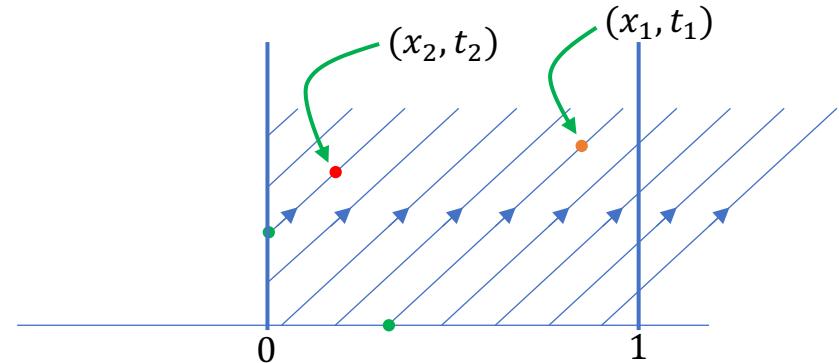
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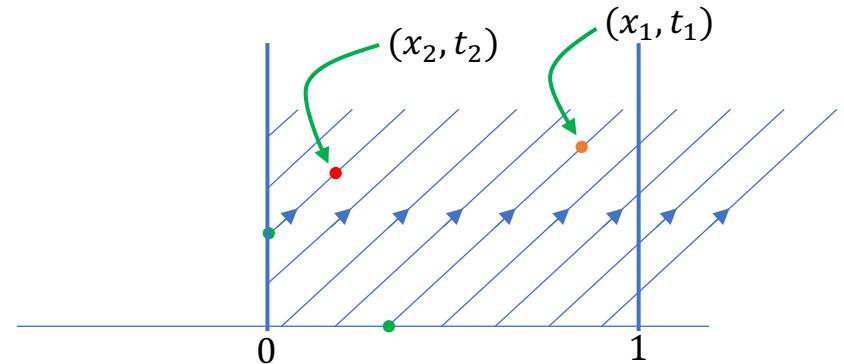
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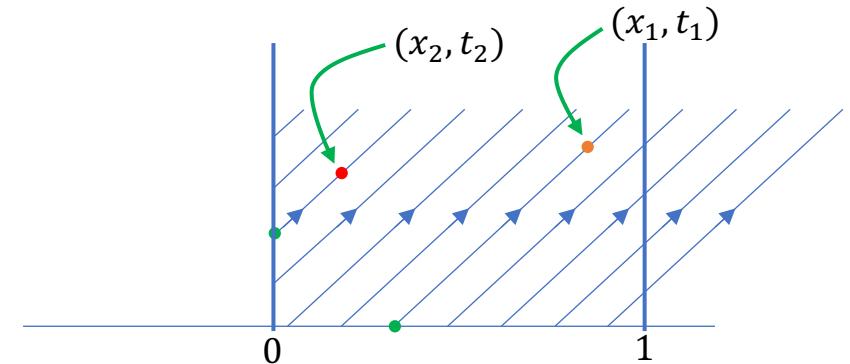
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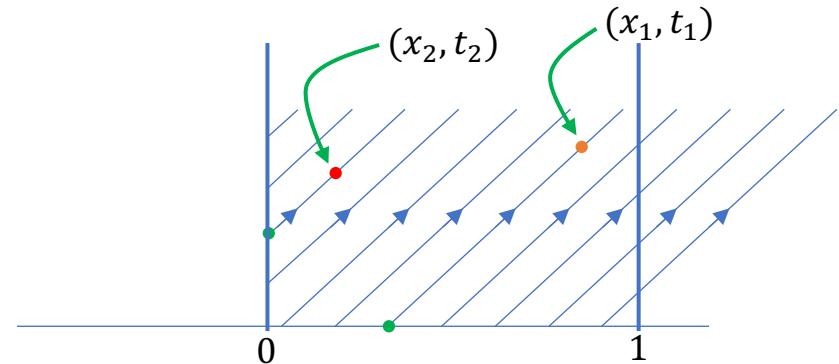
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