


Numerical Methods for PDEs


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Outline


- Review midterm solutions
- Review basic material on numerical calculus
 - Expressions for derivatives, error and error order
- Numerical methods for the diffusion equation
 - Explicit and Implicit
 - First and second order time derivatives



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Numerical Analysis


- Want to express derivatives and integrals in terms of discrete data points
- Use different methods
 - Develop interpolation polynomial and integrate or differentiate this result
 - Use Taylor series to get expressions for derivatives
- Want expressions and measure of error with their use



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Finite Difference Grids

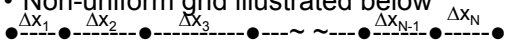
- Subdivide region into discrete points
- Spacing between the points may be uniform or non-uniform
- Example: grid for $x_{\min} \leq x \leq x_{\max}$ with N+1 nodes numbered from zero to N
- Initial node value, $x_0 = x_{\min}$
- Final grid node value, $x_N = x_{\max}$
- Nodal spacing $\Delta x_i = x_i - x_{i-1}$ ($i = 1, N$)
- Uniform spacing, $h = \Delta x_i = (x_{\min} - x_{\max})/N$
- N+1 nodes give N spaces



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Finite Difference Grids II

- Non-uniform grid illustrated below




$x_0 \quad x_1 \quad x_2 \quad x_3 \quad \dots \quad x_{N-2} \quad x_{N-1} \quad x_N$

- Two space dimensions require x and y grids (M+1 y nodes)

$x_0 = x_{\min} \quad x_N = x_{\max} \quad x_i - x_{i-1} = \Delta x_i$
 $y_0 = y_{\min} \quad y_M = y_{\max} \quad y_j - y_{j-1} = \Delta y_j$

- Most general case has three space dimensions (x, y, z, and time)




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Finite Difference Grids III

- Grid notation for four independent variables: x, y, z and t

$x_0 = x_{\min}$	$x_N = x_{\max}$	$x_i - x_{i-1} = \Delta x_i$
$y_0 = y_{\min}$	$y_M = y_{\max}$	$y_j - y_{j-1} = \Delta y_j$
$z_0 = z_{\min}$	$z_K = z_{\max}$	$z_k - z_{k-1} = \Delta z_k$
$t_0 = t_{\min}$	$t_L = t_{\max}$	$t_n - t_{n-1} = \Delta t_n$

- Dependent variable $u(x,y,z,t)$ at discrete points $u(x_i, y_j, z_k, t_n)$
- Use notation below for this value of u

$$u_{ijk}^n = u(x_i, y_j, z_k, t_n)$$


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Derivative Expressions

- Obtain from differentiating interpolation polynomials or from Taylor series
- Series expansion for $f(x)$ about $x = a$

$$f(x) = f(a) + \frac{df}{dx}\bigg|_{x=a} (x-a) + \frac{1}{2!} \frac{d^2f}{dx^2}\bigg|_{x=a} (x-a)^2 + \frac{1}{3!} \frac{d^3f}{dx^3}\bigg|_{x=a} (x-a)^3 + \dots$$

- Note: $d^0f/dx^0 = f$ and $0! = 1$
- What is error from truncating series?

$$f(x) = \sum_{n=0}^{\infty} \frac{1}{n!} \frac{d^n f}{dx^n}\bigg|_{x=a} (x-a)^n$$

Truncation Error

- If we truncate series after m terms

$$f(x) = \underbrace{\sum_{n=0}^m \frac{1}{n!} \frac{d^n f}{dx^n}\bigg|_{x=a} (x-a)^n}_{\text{Terms used}} + \underbrace{\sum_{n=m+1}^{\infty} \frac{1}{n!} \frac{d^n f}{dx^n}\bigg|_{x=a} (x-a)^n}_{\text{Truncation error, } \epsilon_m}$$

- Can write truncation error as single term at unknown location (derivation based on the theorem of the mean)

$$\epsilon_m = \sum_{n=m+1}^{\infty} \frac{1}{n!} \frac{d^n f}{dx^n}\bigg|_{x=a} (x-a)^n = \frac{1}{(m+1)!} \frac{d^{m+1} f}{dx^{m+1}}\bigg|_{x=\xi} (x-a)^{m+1}$$

Derivative Expressions

- Look at finite-difference grid with **equal spacing**: $h = \Delta x$ so $x_i = x_0 + ih$
- Want Taylor series for $f_{i+k} = f(x_{i+k})$ in terms of $f_i = f(x_i)$ and derivatives at $x = x_i$

$$x_{i+k} - x_i = [x_0 + (i+k)h] - [x_0 + ih] = kh$$

$$f(x_i + kh) = f(x_i) + \frac{df}{dx}\bigg|_{x=x_i} kh + \frac{1}{2!} \frac{d^2f}{dx^2}\bigg|_{x=x_i} (kh)^2 + \frac{1}{3!} \frac{d^3f}{dx^3}\bigg|_{x=x_i} (kh)^3 + \dots$$

$$f_i' = \frac{df}{dx}\bigg|_{x=x_i} \quad f_i'' = \frac{d^2f}{dx^2}\bigg|_{x=x_i} \quad \dots \quad f_i^n = \frac{d^n f}{dx^n}\bigg|_{x=x_i}$$

Derivative Expressions II

- Combine all definitions for compact series notation

$$f(x_i + kh) = f(x_i) + \frac{df}{dx}\bigg|_{x=x_i} kh + \frac{1}{2!} \frac{d^2f}{dx^2}\bigg|_{x=x_i} (kh)^2 + \frac{1}{3!} \frac{d^3f}{dx^3}\bigg|_{x=x_i} (kh)^3 + \dots$$

$$f_{i+k} = f_i + f_i' kh + \frac{f_i'' (kh)^2}{2!} + \frac{f_i''' (kh)^3}{3!} + \dots$$

- Use this formula to get expansions for various grid locations about $x = x_i$ and use results to get derivative expressions

Derivative Expressions III

- Apply general equation for $k = 1$ and $k = -1$

$$f_{i+k} = f_i + f_i' kh + \frac{f_i'' (kh)^2}{2!} + \frac{f_i''' (kh)^3}{3!} + \dots$$

$$f_{i+1} = f_i + f_i' h + \frac{f_i'' h^2}{2!} + \frac{f_i''' h^3}{3!} + \dots$$

$$f_{i-1} = f_i - f_i' h + \frac{f_i'' h^2}{2!} - \frac{f_i''' h^3}{3!} + \dots$$

$$f_i' = \frac{f_{i+1} - f_i}{h} - \frac{f_i'' h}{2!} + \frac{f_i''' h^2}{3!} - \dots = \frac{f_{i+1} - f_i}{h} + Ah \quad \text{Forward}$$

$$f_i' = \frac{f_i - f_{i-1}}{h} + \frac{f_i'' h}{2!} - \frac{f_i''' h^2}{3!} + \dots = \frac{f_i - f_{i-1}}{h} + Ah \quad \text{Backward}$$

Derivative Expressions IV

- Subtract f_{i+1} and f_{i-1} expressions

$$f_{i+1} = f_i + f_i' h + \frac{f_i'' h^2}{2!} + \frac{f_i''' h^3}{3!} + \dots$$

$$f_{i-1} = f_i - f_i' h + \frac{f_i'' h^2}{2!} - \frac{f_i''' h^3}{3!} + \dots$$

$$f_{i+1} - f_{i-1} = 2f_i' h + \frac{2f_i''' h^3}{3!} + \frac{2f_i^{(5)} h^5}{5!} + \dots$$

$$f_i' = \frac{f_{i+1} - f_{i-1}}{h} + \frac{f_i''' h^2}{3!} + \frac{f_i^{(5)} h^4}{5!} - \dots = \frac{f_{i+1} - f_{i-1}}{2h} + Ah^2$$

- Result called central difference expression

Order of the Error

- Forward and backward derivative have error term that is proportional to h
- Central difference error is proportional to h²
- Error proportional to hⁿ called nth order
- Reducing step size by a factor of a reduces nth order error by aⁿ

$$\varepsilon_2 \approx \varepsilon_1 \left(\frac{h_2}{h_1} \right)^n$$

Order of the Error Notation

- Write the error term for nth error term as O(hⁿ)
 - Big oh notation, O, denotes order
 - Recognizes that factor multiplying hⁿ may change slightly with h

First order forward $f'_i = \frac{f_{i+1} - f_i}{h} + O(h)$ First order backward $f'_i = \frac{f_i - f_{i-1}}{h} + O(h)$

Second order central $f'_i = \frac{f_{i+1} - f_{i-1}}{2h} + O(h^2)$

Other Derivatives

- Second-order, central-difference, second derivative

$$f''_i = \frac{f_{i+1} + f_{i-1} - 2f_i}{h^2} + \frac{f'''_i h^2}{3!} + \frac{f^{(4)}_i h^4}{5!} - \dots = \frac{f_{i+1} + f_{i-1} - 2f_i}{h^2} + O(h^2)$$

- Second-order, forward and backward difference, first derivatives

$$f'_i = \frac{-f_{i+2} + 4f_{i+1} - 3f_i}{2h} + f'''_i \frac{h^2}{3} + \dots$$

$$f'_i = \frac{f_{i-2} - 4f_{i-1} + 3f_i}{2h} + f'''_i \frac{h^2}{3} + \dots$$

Other Derivative Expressions

- Can derive various finite-difference expressions for derivatives
 - Derivative order, first, second, etc.
 - Order of the error (typically second although higher orders used)
 - Forward, backwards and central difference expressions (typically use central except at boundaries)
 - Derive by Taylor series manipulations
 - See results on page 271 of Hoffman

Order of Error Examples

- Table 1 in “introduction” notes shows first derivative error for e^x around x = 1
 - Using first- and second-order forward and second-order central differences
 - Step h = 0.4, 0.2, and 0.1
 - Error ratio for doubling step size
 - 4.01 to 4.02 for central differences
 - 2.07 to 2.15 for first-order forward differences
 - 4.32 to 4.69 for second-order forward

$$n \approx \frac{\log(\varepsilon_2 / \varepsilon_1)}{\log(h_2 / h_1)} = \frac{\log(\varepsilon_2) - \log(\varepsilon_1)}{\log(h_2) - \log(h_1)}$$

Roundoff Error

- Possible in derivative expressions from subtracting close differences
- Example f(x) = e^x: f'(x) ≈ (e^{x+h} - e^{x-h})/(2h) and error at x = 1 is (e^{1+h} - e^{1-h})/(2h) - e

$$E = \frac{3.004166 - 2.722815}{2(0.1)} - 2.718282 = 4.5 \times 10^{-3}$$

$$E = \frac{2.7185536702 - 2.7180100139}{2(0.0001)} - 2.718281828459 = 4.5 \times 10^{-9}$$

$$E = \frac{2.71828210028724 - 2.71828155660388}{2(0.0000001)} - 2.718281828 = 5.9 \times 10^{-9}$$

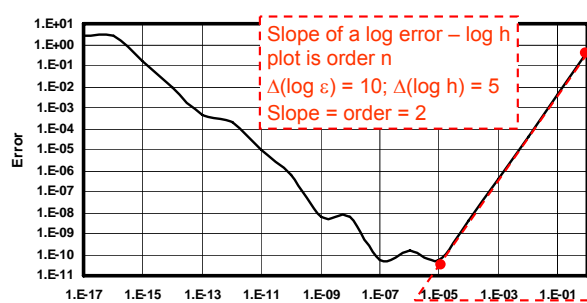
Error vs. Step Size Plot

- Plot the log of the error versus the log of the step size
- For regions where there is no roundoff error this will be a straight line whose slope is the order of the error

$$\varepsilon_2 \approx \varepsilon_1 \left(\frac{h_2}{h_1}\right)^n \Rightarrow \log \varepsilon_2 \approx \log \varepsilon_1 + n \log \left(\frac{h_2}{h_1}\right)$$

$$\log \varepsilon_2 - \log \varepsilon_1 \approx n(\log h_2 - \log h_1) \Rightarrow n \approx \frac{\log \varepsilon_2 - \log \varepsilon_1}{\log h_2 - \log h_1}$$

Figure 2-1. Effect of Step Size on Error



Numerical PDE Solutions

- Define a finite-difference grid in the independent variables (x, y, z, t)
- Place grid points on region boundary whose values are found from boundary conditions for the problem
- At some grid location convert differential equation into a finite difference equation
 - Observe truncation error in process
 - Neglect truncation error to get set of algebraic equations to solve

Diffusion Equation

- Apply difference formulas derived for ordinary derivatives to partial derivatives
- Use notation to consider different coordinate directions $\left[\frac{\partial u}{\partial t} = \alpha \frac{\partial^2 u}{\partial x^2}\right]_i$
- Apply to diffusion equation
- Grids $x_i = x_0 + i\Delta x$ and $t_n = t_0 + n\Delta t$
- Try finite difference expressions below to get simple finite-difference equation

$$\frac{\partial u}{\partial t} = \frac{u_i^{n+1} - u_i^n}{\Delta t} + O(\Delta t) \quad \text{and} \quad \frac{\partial^2 u}{\partial x^2} = \frac{u_{i+1}^n + u_{i-1}^n - 2u_i^n}{(\Delta x)^2} + O[(\Delta x)^2]$$

Diffusion Equation II

- Substitute finite difference expressions into differential equation

$$\frac{u_i^{n+1} - u_i^n}{\Delta t} = \alpha \frac{u_{i+1}^n + u_{i-1}^n - 2u_i^n}{(\Delta x)^2} + O[\Delta t, (\Delta x)^2]$$

- Ignore truncation error, solve for u_i^{n+1}

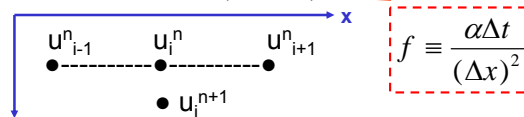
$$u_i^{n+1} = \frac{\alpha \Delta t}{(\Delta x)^2} (u_{i+1}^n + u_{i-1}^n) + \left(1 - \frac{2\alpha \Delta t}{(\Delta x)^2}\right) u_i^n$$

- Obtain potential at $x = x_i$ and $t = t_{n+1}$ in terms of u values at old time step

Explicit (FTCS) Method

- Method just derived is called explicit method; can solve one equation at a time

$$u_i^{n+1} = \frac{\alpha \Delta t}{(\Delta x)^2} (u_{i+1}^n + u_{i-1}^n) + \left(1 - \frac{2\alpha \Delta t}{(\Delta x)^2}\right) u_i^n = f(u_{i+1}^n + u_{i-1}^n) + (1 - 2f)u_i^n$$



- u_i^{n+1} does not depend on other u values at the new time step (n+1)

Explicit Method Example

- Pick $\alpha = 1$, $\Delta x = 0.25$, $N_x = 4$, $\Delta t = 0.01$
- $f = \alpha \Delta t / (\Delta x)^2 = 1(.01) / (.25)^2 = 0.16$
- Pick initial $u_i^0 = 1000$ and boundaries, $u_0^n = u_4^n = 0$ for time > 0 ($n \geq 0$)

$$\text{Apply } u_i^{n+1} = f(u_{i+1}^n + u_{i-1}^n) + (1 - 2f)u_i^n$$

$$u_1^1 = f[u_0^0 + u_2^0] + (1 - 2f)u_1^0 = 0.16[0 + 1000] + 0.68[1000] = 840$$

$$u_2^1 = f[u_1^0 + u_3^0] + (1 - 2f)u_2^0 = 0.16[1000 + 1000] + 0.68[1000] = 1000$$

$$u_3^1 = f[u_2^0 + u_4^0] + (1 - 2f)u_3^0 = 0.16[1000 + 0] + 0.68[1000] = 840$$

- Repeat for subsequent time steps

Explicit Method Results $f = 0.16$

		i = 0	i = 1	i = 2	i = 3	i = 4
		x = 0.00	x = 0.25	x = 0.50	x = 0.75	x = 1.00
	t = 0	1000	1000	1000	1000	1000
n = 0	t = 0+	0	1000	1000	1000	0
n = 1	t = 0.01	0	840	1000	840	0
n = 2	t = 0.02	0	731.2	948.8	731.2	0
n = 3	t = 0.03	0	649	879.2	649	0
n = 4	t = 0.04	0	582	805.5	582	0
n = 5	t = 0.05	0	524.6	734	524.6	0
n = 6	t = 0.06	0	474.2	667	474.2	0
n = 7	t = 0.07	0	429.2	605.3	429.2	0
n = 8	t = 0.08	0	388.7	548.9	388.7	0

Explicit Method Results $f = 0.16$

		i = 0	i = 1	i = 2	i = 3	i = 4
		x = 0.00	x = 0.25	x = 0.50	x = 0.75	x = 1.00
n = 12	t = 0.12	0	262	370.5	262	0
n = 13	t = 0.13	0	237.5	335.8	237.5	0
n = 14	t = 0.14	0	215.2	304.4	215.2	0
n = 15	t = 0.15	0	195	275.8	195	0
n = 16	t = 0.16	0	176.8	250	176.8	0
n = 17	t = 0.17	0	160.2	226.5	160.2	0
n = 18	t = 0.18	0	145.2	205.3	145.2	0
n = 19	t = 0.19	0	131.6	186.1	131.6	0
n = 20	t = 0.20	0	119.2	168.6	119.2	0
Exact	t = 0.20	0	125.1	176.9	125.1	0
Error	t = 0.20	0	5.8	8.2	5.8	0

Explicit Results $f = 0.32$

		i = 0	i = 1	i = 2	i = 3	i = 4
		x = 0.00	x = 0.25	x = 0.50	x = 0.75	x = 1.00
	t = 0	1000	1000	1000	1000	1000
n = 0	t = 0+	0	1000	1000	1000	0
n = 1	t = 0.02	0	680	1000	680	0
n = 2	t = 0.04	0	564.8	795.2	564.8	0
n = 3	t = 0.06	0	457.9	647.7	457.9	0
n = 8	t = 0.16	0	162.2	229.4	162.2	0
n = 9	t = 0.18	0	131.8	186.4	131.8	0
n = 10	t = 0.20	0	107.1	151.4	107.1	0
Exact	t = 0.20	0	125.1	176.9	125.1	0
Error	t = 0.20	0	18	25.4	18	0

Explicit Results $f = 0.64$

		i = 0	i = 1	i = 2	i = 3	i = 4
		x = 0.00	x = 0.25	x = 0.50	x = 0.75	x = 1.00
	t = 0	1000	1000	1000	1000	1000
n = 0	t = 0+	0	1000	1000	1000	0
n = 1	t = 0.04	0	360	1000	360	0
n = 2	t = 0.08	0	539.2	180.8	539.2	0
n = 3	t = 0.12	0	-35.3	639.6	-35.3	0
n = 4	t = 0.16	0	419.2	-224.2	419.2	0
n = 5	t = 0.20	0	-260.9	599.3	-260.9	0
Exact	t = 0.20	0	125.1	176.9	125.1	0
Error	t = 0.20	0	385.9	422.5	385.9	0

What Happened?

- We are seeing effects of instability
- Difference equations may not converge
 - Unstable equations grow without bound
 - May have stable equations that produce incorrect results
 - Conditional stability requires step size less than that needed for accuracy
 - Goal of absolute stability not always possible
 - Discussions of stability complex, can sometimes use physical arguments

Stability of Explicit Method

- If the values of u_{i+1} and u_{i-1} are fixed an increase in u_i^n should increase u_i^{n+1}
- If f is greater than 0.5, an increase in u_i^n will cause a decrease in u_i^{n+1}
- We can avoid this incorrect result by keeping $f = \alpha\Delta t/(\Delta x)^2 \leq 0.5$
- This imposes a time step limit that may be less than the limit required for accuracy in the solution

FTCS Truncation Error

- Derivation in appendix for notes on solving PDEs gives this equation

$$TE_i^n = \alpha \sum_{k=2}^{\infty} (\Delta x)^{2k-2} \left[\frac{2}{(2k)!} - \frac{f^{k-1}}{k!} \right] \frac{\partial^{2k} T}{\partial x^{2k}} \Big|_i^n$$

$$TE_i^n = \frac{\alpha(\Delta x)^2}{2} \left(\frac{1}{6} - f \right) \frac{\partial^4 T}{\partial x^4} \Big|_i^n + \frac{\alpha(\Delta x)^4}{6} \left(\frac{1}{60} - f^2 \right) \frac{\partial^6 T}{\partial x^6} \Big|_i^n + \dots$$

- Setting $f = \alpha\Delta t/(\Delta x)^2 = 1/6$ eliminates first term in the truncation error

Alternative Methods

- Next class will show how to avoid stability limit by using explicit methods
 - Solution for u_i^{n+1} depends on other u values at new time step
 - Requires solution of simultaneous equations for all u_i^{n+1}
 - Crank-Nicholson uses values for old and new time steps in space derivatives
 - Fully implicit uses new time values only