

Computational Fluid Dynamics: Lecture 3 (ME EN 6720)

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Conservation Equations

summing up our equations:

- conservation of mass

$$\frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x_i}(\rho u_i) = 0$$

- conservation of momentum

$$\rho \frac{\partial (u_i)}{\partial t} + \rho u_j \frac{\partial (u_i)}{\partial x_j} = \frac{\partial (\tau_{ij})}{\partial x_j} - \frac{\partial (P)}{\partial x_i} + \rho g_i \quad \left\{ \begin{array}{l} \tau_{ij} = 2\mu D_{ij} - \frac{2}{3}\mu \delta_{ij} \frac{\partial u_k}{\partial x_k} \\ D_{ij} = \frac{1}{2} \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right) \end{array} \right.$$

- conservation of energy

$$-\frac{\partial}{\partial x_j} \left[\rho u_j \left(e + \frac{1}{2} u_i u_i \right) \right] - \frac{\partial q_i}{\partial x_i} - \frac{\partial (P u_i)}{\partial x_j} + \frac{\partial (\tau_{ij} u_i)}{\partial x_j} = \frac{\partial}{\partial t} \left[\rho \left(e + \frac{1}{2} u_i u_i \right) \right]$$

- conservation of scalar concentration

$$\frac{\partial \phi}{\partial t} + \frac{\partial}{\partial x_i}(\phi u_i) = r_g - r_d$$

Conservation of Momentum

- An equivalent index-notation form of the momentum equation is:

$$\frac{\partial(\rho u_i)}{\partial t} + \frac{\partial(\rho u_i u_j)}{\partial x_j} = \frac{\partial(\tau_{ij})}{\partial x_j} - \frac{\partial(P)}{\partial x_i} + \rho g_i$$

- Forms of the Momentum Equation

☞ **Differential** vs. **Integral**

☞ **Eulerian** vs. **Lagrangian**

☞ **Conservative** vs. **Non conservative**

- For the integral equations, **conservative** means all terms are written in the form of a divergence (e.g., $\nabla \cdot \vec{V}$)
- For Finite Volume formulations, using a conservative form instead of the non conservative form guarantees the velocity field will be divergent free (a requirement for realizeability)

Systems of Equations

• Last Notes on Equations:

- In Lecture 2, we mentioned a plus of the conservative form (flux form) of the equations.
- All equations (mass, momentum, energy, scalar) have a divergence term for the integral form (e.g., $\nabla \cdot \vec{V}$) or flux term for the differential form of the equations
- in 1-D

- ρu flux of mass
- ρuu flux of momentum
- $\rho \left(e + \frac{u^2}{2} \right) u$ flux of total energy (E)

- We can summarize the equations for these fluxes into one compact form:

$$\frac{\partial \mathbf{U}}{\partial t} + \frac{\partial \mathbf{F}}{\partial x} = \mathbf{G}$$

$$\mathbf{U} = \begin{bmatrix} \rho \\ \rho u \\ \rho E \end{bmatrix}$$

Solution Vector

$$\mathbf{F} = \begin{bmatrix} \rho u \\ (\rho u^2 + P - \tau_{xx}) \\ \left(Eu + Pu - k \frac{\partial T}{\partial x} - u \tau_{xx} \right) \end{bmatrix}$$

Flux Vector

$$\mathbf{G} = \begin{bmatrix} 0 \\ \rho f_x \\ (\rho(u f_x) + \rho \dot{q}) \end{bmatrix}$$

Source Term Vector

- Note, the form described above could be termed the “strong” conservative form since all of the terms appear inside the spatial derivatives.
- Practically, this indicates that all of our equations are a form of the same PDE.

Process of CFD

As discussed in previous lectures our general process in CFD is:

Determine physics → determine appropriate PDE → **Discretize**

Discretization techniques:

finite differences

- use differential form of the equations
- construct approximate difference operators

e.g.
$$\frac{\partial \phi}{\partial x} \approx \frac{\phi_{i+1} - \phi_{i-1}}{2\Delta x}$$

finite volumes

- use integral form of the equations
- approximate fluxes at “volume” interfaces and gradients discretely

variational method

- leave differential operators unchanged
- approximate the solution space:

$$\phi(x) = \sum_{i=1}^N a_i \Psi_i(x)$$

expansion
coefficients

trial
functions

- 2 types of trial functions
 - local → finite elements
 - Global → spectral methods

Finite difference approximations

Finite Difference Approximations (Ferziger chapter 3)

- goal: build a discrete analog to the continuous problem

Discretization

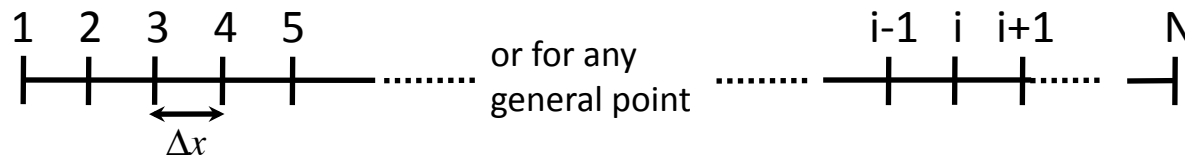


Finite Difference Equation (FDE)

Represent variables at points

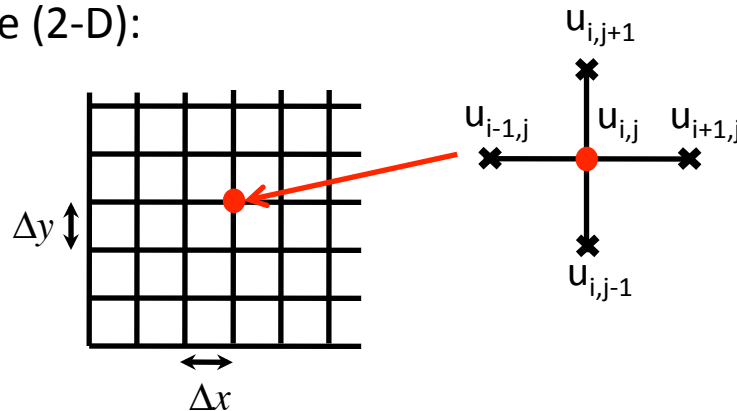
- Algebraic representation of PDE
- Solve using Linear algebra

- space (1-D):



where $\Delta x = \frac{1}{N-1}$
and $x_i = (i-1)\Delta x$
for $i=1,2,\dots,N$

- space (2-D):



definitions:

$u_{i,j} = u(x_o, y_o)$
 $u_{i+1,j} = u(x_o + \Delta x, y_o)$
 $u_{i,j-1} = u(x_o, y_o - \Delta y)$
 and so forth...

Finite difference approximations

- Recall the definition of a derivative:

for a function $u(x, y)$ at a point $(x_o, y_o) \Rightarrow \frac{\partial u}{\partial x} = \lim_{\Delta x \rightarrow 0} \frac{u(x_o + \Delta x, y_o) - u(x_o, y_o)}{\Delta x}$

-for a continuous u (continuum assumption for fluids), the above will be “reasonable” for a “sufficiently” small Δx

- we can develop a more formal definition using a Taylor series expansion of u

$$u_{i+1,j} = u_{i,j} + \left(\frac{\partial u}{\partial x} \right)_{i,j} \Delta x + \left(\frac{\partial^2 u}{\partial x^2} \right)_{i,j} \frac{(\Delta x)^2}{2!} + \left(\frac{\partial^3 u}{\partial x^3} \right)_{i,j} \frac{(\Delta x)^3}{3!} + \dots + \underbrace{\text{H.O.T}}_{\text{Higher-Order Terms}}$$

This is mathematically exact if [one or both]:

- 1) $\Delta x \rightarrow 0$
- 2) The number of terms is infinite and the series converges

- from the Taylor series we can quickly come up with an approximation for $\left(\frac{\partial u}{\partial x} \right)_{i,j}$ by solving for it:

$$\left(\frac{\partial u}{\partial x} \right)_{i,j} = \frac{u_{i+1,j} - u_{i,j}}{\Delta x} - \underbrace{\left(\frac{\partial^2 u}{\partial x^2} \right)_{i,j} \frac{\Delta x}{2!} - \left(\frac{\partial^3 u}{\partial x^3} \right)_{i,j} \frac{(\Delta x)^2}{3!} + \dots + \text{H.O.T}}_{\mathbf{O}(\Delta x)}$$

[note: $\mathbf{O}(\Delta x)$ means “on the order of” Δx which is the order of error of the approximation

Finite difference approximations

Finite Difference Approximations for 1st derivatives

- using the above index notation and a Taylor series approximation

$$(*) \quad u_{i+1,j} = u_{i,j} + \left(\frac{\partial u}{\partial x} \right)_{i,j} \Delta x + \left(\frac{\partial^2 u}{\partial x^2} \right)_{i,j} \frac{(\Delta x)^2}{2!} + \left(\frac{\partial^3 u}{\partial x^3} \right)_{i,j} \frac{(\Delta x)^3}{3!} + \text{H.O.T}$$

we can get:

$$\left. \frac{\partial u}{\partial x} \right|_{i,j} = \frac{u_{i+1,j} - u_{i,j}}{\Delta x} + \mathcal{O}(\Delta x)$$

Forward difference approximation

(because we expanded “forward” in space)

- we can do the same thing backwards:

$$(**) \quad u_{i-1,j} = u_{i,j} - \left(\frac{\partial u}{\partial x} \right)_{i,j} \Delta x + \left(\frac{\partial^2 u}{\partial x^2} \right)_{i,j} \frac{(\Delta x)^2}{2!} - \left(\frac{\partial^3 u}{\partial x^3} \right)_{i,j} \frac{(\Delta x)^3}{3!} + \text{H.O.T}$$

→

$$\left. \frac{\partial u}{\partial x} \right|_{i,j} = \frac{u_{i,j} - u_{i-1,j}}{\Delta x} + \mathcal{O}(\Delta x)$$

Backwards difference approximation

Finite difference approximations

- if we subtract $(**)$ from $(*)$ and rearrange we can get:

$$\left. \frac{\partial u}{\partial x} \right|_{i,j} = \frac{u_{i+1,j} - u_{i-1,j}}{2\Delta x} + \mathcal{O}(\Delta x^2)$$

centered difference approximation

Finite difference approximations for 2nd derivatives

- if we add $(*)$ and $(**)$ and rearrange we can get:

$$\left. \frac{\partial^2 u}{\partial x^2} \right|_{i,j} = \frac{u_{i+1,j} - 2u_{i,j} + u_{i-1,j}}{(\Delta x)^2} - \underbrace{\left(\frac{\partial^4 u}{\partial x^4} \right)_{i,j} \frac{(\Delta x)^2}{12}}_{\mathcal{O}(\Delta x^2)} + \text{H.O.T}$$

- This is a 2nd order-centered difference scheme for the 2nd derivative. It is one of the most popular schemes for 2nd derivatives in FD

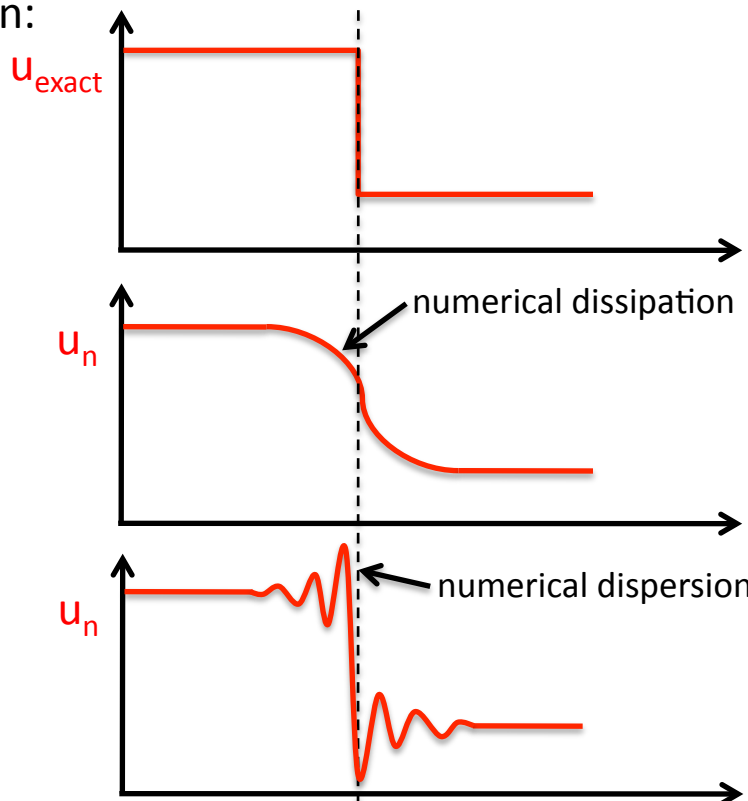
Truncation Error

Effect of Truncation Error (TE) on Solutions

- Two types of errors result from truncation of Taylor series for FD schemes such as those discussed above (and for other types of techniques we haven't talked about).

- **Numerical dissipation** and **Numerical dispersion**

- given a solution:



Why do we get one or the other?

Choice of FD scheme

- CDS \rightarrow dispersion
- FDS \rightarrow dissipation

Numerical dissipation is a direct result of **even** derivatives in the leading term of the TE

Numerical dispersion is a direct result of **odd** derivatives in the leading term of the TE

Accuracy of F.D. schemes

A few notes about the accuracy of numerical schemes

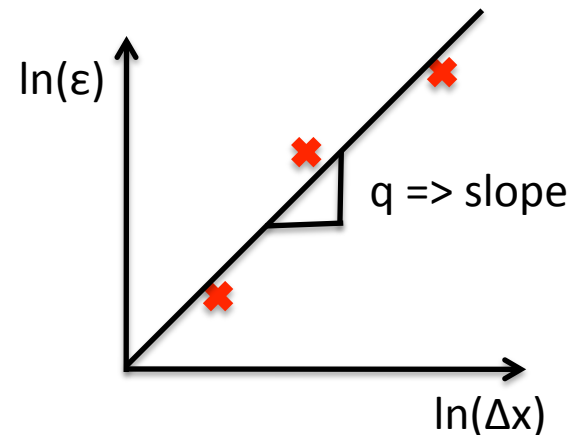
- a definition: If the T.E. of a scheme behaves like $\sim(\Delta x)^q$ then the scheme is called q -order
- **How can we compute the error?** (two measures for now)

a) Error Norms: $\left\{ \begin{array}{l} \text{continuous: } \varepsilon_p = \left[\int_0^1 |u_e - u_n|^p \right]^{\frac{1}{p}} \\ \text{discrete: } \varepsilon_p = \left[\frac{1}{N} \sum_{i=1}^N |u_{e,i} - u_{n,i}|^p \right]^{\frac{1}{p}} \end{array} \right.$

u_e is the exact solution
 u_n is the numerical solution

b) Relative error: $\varepsilon_r = \left| \frac{u_e - u_n}{u_e} \right|$

How can we use the error to verify the order of a scheme?



Discretized equations

Example of a discretized equation

- 1-D heat equation: $\frac{\partial u}{\partial t} = \alpha \frac{\partial^2 u}{\partial x^2}$

- two different FDE representations of this are:

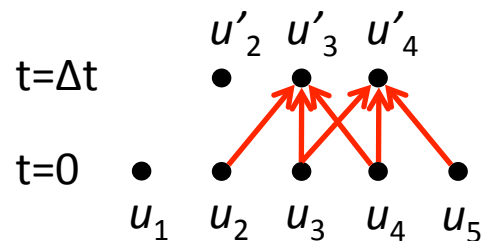
$$\frac{u_i^{n+1} - u_i^n}{\Delta t} = \alpha \frac{1}{(\Delta x)^2} [u_{i+1}^n - 2u_i^n + u_{i-1}^n] \quad (\text{explicit})$$

and

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

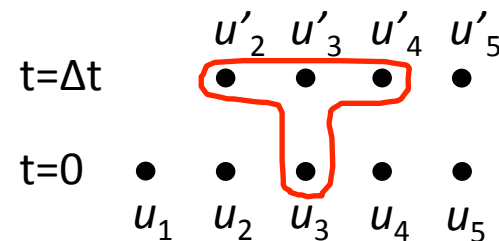
- graphically we can represent the points used by these two schemes as:

explicit



For example to get u'_3 and u'_4

implicit



For example to get u'_3

Truncation error in FDEs

- We can use our FDE equation above (explicit one) to look at the total TE of our solution

- 1st we expand each u as a Taylor series about u_i^n :

$$u_i^{n+1} = u_i^n + \left(\frac{\partial u}{\partial t}\right)_i^n \Delta t + \left(\frac{\partial^2 u}{\partial t^2}\right)_i^n \frac{(\Delta t)^2}{2!} + \mathcal{O}(\Delta t^3)$$

$$u_{i+1}^n = u_i^n + \left(\frac{\partial u}{\partial x}\right)_i^n \Delta x + \left(\frac{\partial^2 u}{\partial x^2}\right)_i^n \frac{(\Delta x)^2}{2!} + \left(\frac{\partial^3 u}{\partial x^3}\right)_i^n \frac{(\Delta x)^3}{3!} + \mathcal{O}(\Delta x^4)$$

$$u_{i-1}^n = u_i^n - \left(\frac{\partial u}{\partial x}\right)_i^n \Delta x + \left(\frac{\partial^2 u}{\partial x^2}\right)_i^n \frac{(\Delta x)^2}{2!} - \left(\frac{\partial^3 u}{\partial x^3}\right)_i^n \frac{(\Delta x)^3}{3!} + \mathcal{O}(\Delta x^4)$$

- substituting these into the FDE:

$$\frac{1}{\Delta t} \left[\cancel{u_i^n} + \left(\frac{\partial u}{\partial t}\right)_i^n \Delta t + \left(\frac{\partial^2 u}{\partial t^2}\right)_i^n \frac{(\Delta t)^2}{2!} + \mathcal{O}(\Delta t^3) - \cancel{u_i^n} \right] =$$

$$\frac{\alpha}{\Delta x^2} \left[\cancel{u_i^n} + \left(\frac{\partial u}{\partial x}\right)_i^n \Delta x + \left(\frac{\partial^2 u}{\partial x^2}\right)_i^n \frac{(\Delta x)^2}{2!} + \left(\frac{\partial^3 u}{\partial x^3}\right)_i^n \frac{(\Delta x)^3}{3!} + \mathcal{O}(\Delta x^4) - 2\cancel{u_i^n} + \cancel{u_i^n} - \left(\frac{\partial u}{\partial x}\right)_i^n \Delta x + \left(\frac{\partial^2 u}{\partial x^2}\right)_i^n \frac{(\Delta x)^2}{2!} - \left(\frac{\partial^3 u}{\partial x^3}\right)_i^n \frac{(\Delta x)^3}{3!} + \mathcal{O}(\Delta x^4) \right]$$

- cancelling terms ($\cancel{}$): $\left(\frac{\partial u}{\partial t}\right)_i^n + \left(\frac{\partial^2 u}{\partial t^2}\right)_i^n \frac{\Delta t}{2} + \mathcal{O}(\Delta t^2) = \alpha \left[\left(\frac{\partial^2 u}{\partial x^2}\right)_i^n + \mathcal{O}(\Delta x^2) \right]$

- recovering the PDE: $\underbrace{\frac{\partial u}{\partial t}}_{\text{PDE}} = \alpha \underbrace{\frac{\partial^2 u}{\partial x^2}}_{\text{PDE}} + \underbrace{\mathcal{O}[\Delta t, \Delta x^2]}_{\text{TE}} \rightarrow \text{PDE} + \text{TE} = \text{FDE}$

- this can be done for any FDE. Mixing schemes doesn't change the TE for individual terms