

Ch.1 Basic Concepts of Fluid Flow

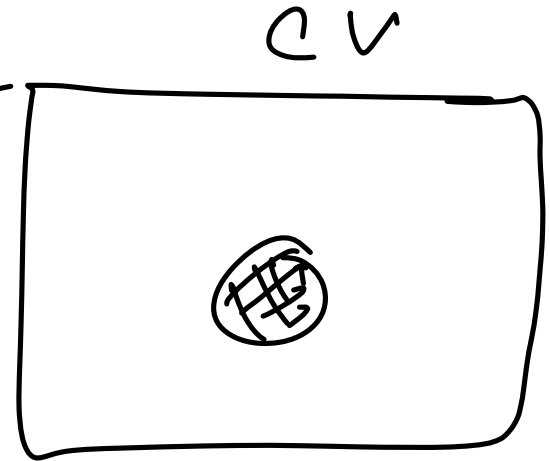
CFD - Computational Fluid Dynamics

~~(Colorful)~~



$$\underline{u}, p \rightarrow \underline{\omega} = \nabla \times \underline{u}, \dots$$

(\underline{x}, t)



① Conservation principles

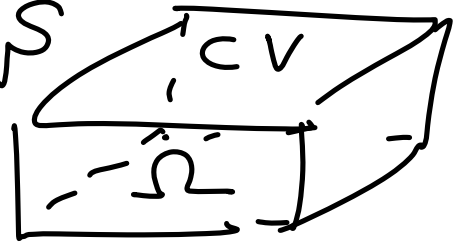
→ Control volume approaches
Reynolds transport theorem

① mass conservation

$$\frac{\partial}{\partial t} \int_{\Omega} \rho d\Omega + \int_S \rho \underline{u} \cdot \underline{n} dS = 0$$

out-normal unit vector \underline{n}
 S

ρ density
 \underline{u} velocity



$$\rightarrow \frac{\partial \rho}{\partial t} + \nabla \cdot (\rho \underline{u}) = 0, \quad \boxed{\frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x_i} (\rho u_i) = 0}$$

$$(\bar{i}=1,2,3) \quad \frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x_1} (\rho u_1) + \frac{\partial}{\partial x_2} (\rho u_2)$$

$$+ \frac{\partial}{\partial x_3} (\rho u_3) = 0$$

② momentum conservation

$$\frac{\partial}{\partial t} \int_{\Omega} \rho \underline{u} d\Omega + \int_S \rho \underline{u} (\underline{u} \cdot \underline{n}) dS = \sum \underline{f} = \left(\frac{d}{dt} (\rho \underline{u}) \right)_{sys} d\Omega$$

$$\underline{\underline{T}} = - \left(p + \frac{2}{3} \mu \nabla \cdot \underline{u} \right) \underline{\underline{1}}$$

pressure

surface forces
body forces

$$\begin{bmatrix} T_{11} & T_{12} & T_{13} \\ T_{21} & T_{22} & T_{23} \\ T_{31} & T_{32} & T_{33} \end{bmatrix} \text{ stress tensor}$$

$$+ 2\mu \underline{\underline{D}} \quad \text{viscosity}$$

$$\underline{\underline{D}} = \frac{1}{2} \left(\nabla \underline{u} + (\nabla \underline{u})^T \right) ; \text{ strain-rate tensor (deformation)}$$

$$T_{\bar{i}\bar{j}} = - \left(p + \frac{2}{3} \mu \frac{\partial u_{\bar{k}}}{\partial x_{\bar{k}}} \right) \delta_{\bar{i}\bar{j}} + 2\mu D_{\bar{i}\bar{j}} \quad \left(\begin{array}{l} \bar{i} = 1, 2, 3 \\ \bar{j} = 1, 2, 3 \end{array} \right)$$

kronedker delta

$$\begin{cases} 1 & \text{if } \bar{i} = \bar{j} \\ 0 & \text{otherwise} \end{cases}$$

$$\underline{\underline{1}} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$D_{\bar{i}\bar{j}} = \frac{1}{2} \left(\frac{\partial u_{\bar{i}}}{\partial x_{\bar{j}}} + \frac{\partial u_{\bar{j}}}{\partial x_{\bar{i}}} \right)$$

$$\bar{T}_{ij} = -p \delta_{ij} + \tau_{ij}$$

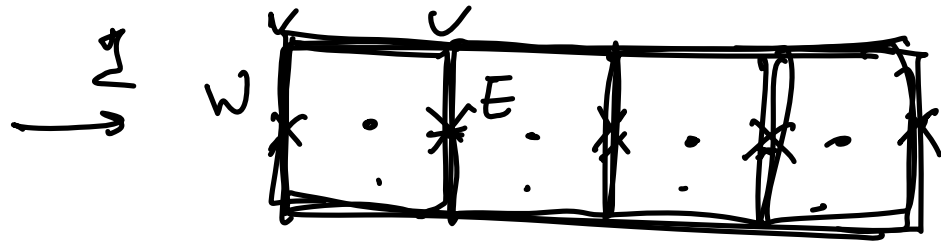
$$\tau_{ij} = 2\mu D_{ij} - \frac{2}{3}\mu \delta_{ij} \frac{\partial u_k}{\partial x_k} \quad \text{: viscous part of stress tensor}$$

$$\rightarrow \frac{\partial}{\partial t} \int_{\Omega} \rho \underline{u} \, d\Omega + \int_S \rho \underline{u} (\underline{u} \cdot \underline{n}) \, dS = \int_S \underline{T} \cdot \underline{n} \, dS + \int_{\Omega} \rho \underline{b} \, d\Omega$$

↑
body force
per unit mass

$$\rightarrow \frac{\partial}{\partial t} (\rho \underline{u}) + \nabla \cdot (\rho \underline{u} \underline{u}) = \nabla \cdot \underline{T} + \rho \underline{b}$$

$$\boxed{\frac{\partial}{\partial t} (\rho u_i) + \frac{\partial}{\partial x_j} (\rho u_j u_i) = \frac{\partial (\tau_{ij})}{\partial x_j} + \rho b_i} \quad \text{: strong conservation form}$$



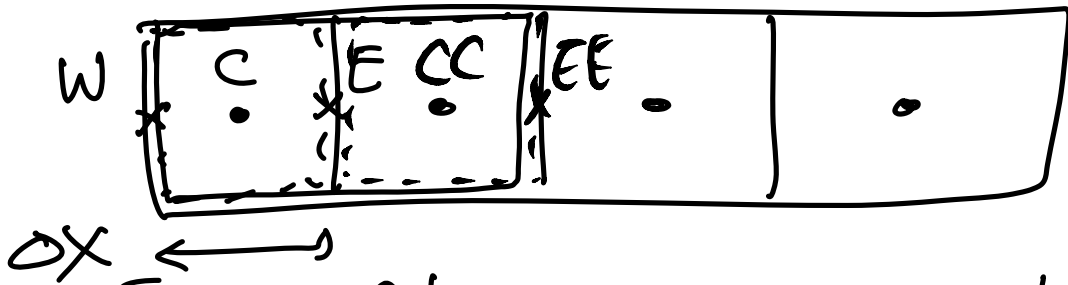
$$\int \frac{\partial}{\partial x_j} (\rho u_j u_i) \rightarrow \int \frac{\partial}{\partial x_1} (\rho u_1 u_i) dx, = \rho u_1 u_i|_E - \rho u_1 u_i|_w$$

$$\frac{\partial}{\partial x_j} (\rho u_j u_i) = \rho u_j \frac{\partial u_i}{\partial x_j} + u_i \frac{\partial}{\partial x_j} (\rho u_j)$$

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

$$\rightarrow \rho \frac{\partial u_i}{\partial t} + \rho u_j \frac{\partial u_i}{\partial x_j} = \frac{\partial T_{ij}}{\partial x_j} + \rho b_i$$

weakly conservative form



$$\int p u_i \frac{\partial u_i}{\partial x_i} dx_i = p u_i \Big|_c \cdot \frac{\partial u_i}{\partial x_i} \Big|_c \Delta x$$

$$= p_c u_{i,c} \cdot \frac{u_{i,E} - u_{i,w}}{\Delta x} \cdot \Delta x$$

$$= p_c u_{i,c} \cdot (u_{i,E} - u_{i,w})$$

$$+ \quad p_{cc} u_{i,cc} \cdot (u_{i,EE} - u_{i,E})$$

$$\rightarrow \frac{\partial}{\partial t} (\rho u_i) + \frac{\partial}{\partial x_j} (\rho u_j u_i) = \frac{\partial \tau_{ij}}{\partial x_j} - \frac{\partial p}{\partial x_i} + \rho b_i$$

$$(\rho b_i = \rho g_i)$$

$$\rho \frac{\partial u_i}{\partial t} + \rho u_j \frac{\partial u_i}{\partial x_j} = \frac{\partial \tau_{ij}}{\partial x_j} + \rho b_i \quad ; \quad \text{non-conservative form}$$

③ conservation of scalar quantities

$$\frac{\partial}{\partial t} \int_{\Omega} \rho \phi d\Omega + \int_{\Sigma} \rho \phi (\underline{u} \cdot \underline{n}) ds = \Sigma f_{\phi}$$

$$f_{\phi}^d = \int_{\Sigma} \Gamma \nabla \phi \cdot \underline{n} ds \quad ; \quad \text{diffusion}$$

↑
diffusivity

heat \rightarrow Fourier law

mass diffusion \rightarrow Fick's law

$$\rightarrow \frac{\partial}{\partial t} \int_{\Omega} \rho \phi \, d\Omega + \int_S \rho \phi (\underline{u} \cdot \underline{n}) \, dS = \int_S \Gamma \nabla \phi \cdot \underline{n} \, dS + \int_{\Omega} \delta \phi \, d\Omega$$

↑
source/sink

$$\rightarrow \frac{\partial}{\partial t} (\rho \phi) + \nabla \cdot (\rho \phi \underline{u}) = \nabla \cdot (\Gamma \nabla \phi) + \delta \phi$$

$$\frac{\partial}{\partial t} (\rho \phi) + \frac{\partial}{\partial x_j} (\rho \phi u_j) = \frac{\partial}{\partial x_j} \left(\Gamma \frac{\partial \phi}{\partial x_j} \right) + \delta \phi$$

energy eq. $\dot{=} \frac{\partial}{\partial t} (\rho c_p T) + \frac{\partial}{\partial x_j} (\rho c_p T u_j) = \frac{\partial}{\partial x_j} (k \frac{\partial T}{\partial x_j}) + \dot{q}_v$

↑
temperature

① Dimensionless form of equations

$$x_i^* = x_i / L_0, \quad u_i^* = u_i / v_0, \quad t^* = t / (L_0 / v_0) \quad \cancel{t^* = t / t_0}$$

$$p^* = p / \rho v_0^2, \quad T^* = \frac{T - T_0}{T_1 - T_0}$$

normalized
gravitational
acceleration vector

$$\rightarrow \left\{ \begin{aligned} \frac{\partial u_i^*}{\partial x_i^*} &= 0 \\ \frac{\partial u_i^*}{\partial t^*} + \frac{\partial}{\partial x_j^*} (u_j^* u_i^*) &= -\frac{\partial p^*}{\partial x_i^*} + \frac{1}{Re} \nabla^{*2} u_i^* + \frac{1}{Fr} \gamma_i \\ \frac{\partial T^*}{\partial t^*} + \frac{\partial}{\partial x_j^*} (u_j^* T^*) &= \frac{1}{Re Pr} \nabla^{*2} T^* \end{aligned} \right.$$

$$Re = \frac{\rho v_0 L_0}{\mu}, \quad Fr = \frac{v_0}{\sqrt{g L_0}}, \quad Pr = \mu c_p / k$$

① Simplified mathematical models

Continuity eq > very difficult to solve.
N-S eq

Sol. unique? exist? \leftarrow Navier-Stokes eqs.
by Roger Temam

① Incompressible flow: $\rho \equiv \text{const.}$ ($Ma < 0.3$)

$$\rightarrow \left\{ \begin{array}{l} \frac{\partial \rho_i}{\partial x_i} = 0 \end{array} \right.$$

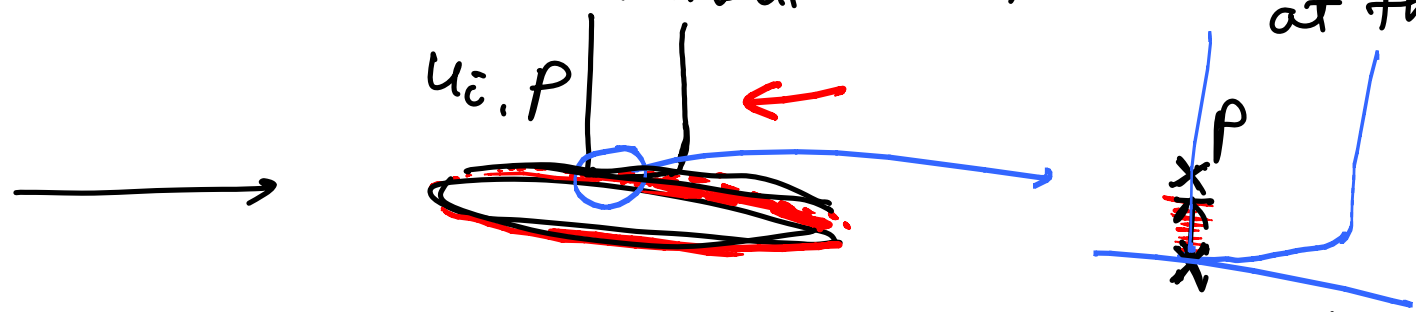
$$\left\{ \begin{array}{l} \frac{\partial \rho_i}{\partial t} + \frac{\partial}{\partial x_j} (\rho_i u_j) = -\frac{1}{\rho} \frac{\partial p}{\partial x_i} + \frac{1}{\rho} \frac{\partial}{\partial x_j} \left(\mu \frac{\partial \rho_i}{\partial x_j} \right) + b_i \end{array} \right.$$

still very difficult to solve.

② Inviscid (Euler) Flow : $\nu \equiv 0$, $\underline{T} = -p \underline{I}$

$$\begin{cases} \frac{\partial p}{\partial t} + \frac{\partial}{\partial x_i} (\rho u_i) = 0 \\ \frac{\partial}{\partial t} (\rho u_i) + \frac{\partial}{\partial x_j} (\rho u_i u_j) = - \frac{\partial p}{\partial x_i} + \rho b_i \end{cases} \quad \text{Euler eq.}$$

At high Re, viscous effect } are important only
 turbulence " } at the near-wall region



Euler eq. → there is no need to put dense grid near the wall

③ Potential flow (inviscid irrotational flow)

$$\underline{\nu} \equiv 0$$

ϕ : velocity potential

$$\underline{u} = \nabla \phi, \quad \nabla \cdot \underline{u} = 0$$

$$\underline{\omega} = \nabla \times \underline{u} \equiv 0$$

$$\underline{u} = \nabla \phi$$

$$\nabla^2 \phi = 0$$

$$\frac{\partial \phi}{\partial x} \quad \frac{\partial \phi}{\partial y}$$

$$\phi(x, y, z)$$

Bernoulli eq. $\frac{p}{\rho} + \frac{v^2}{2} + \gamma z = C$

④ Creeping (Stokes) flow

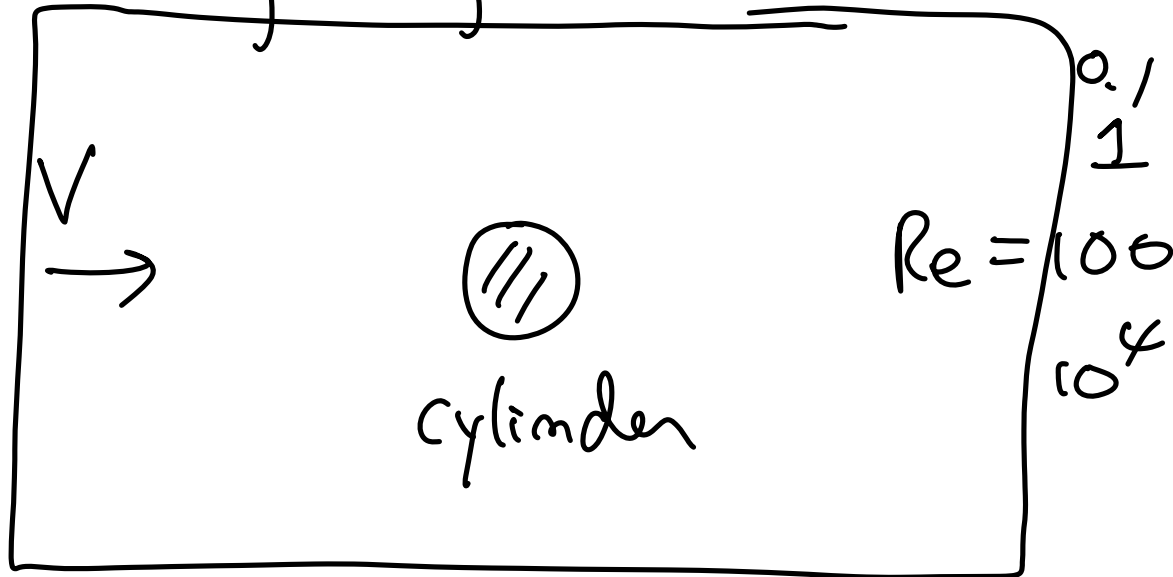
viscous term
pressure "
body-force "

\gg inertia term
(nonlinear)

\rightarrow Stokes eq.
(linear eq.)

$$\frac{\partial h_i}{\partial x_i} = 0$$

$$\frac{\partial}{\partial x_j} \left(\mu \frac{\partial h_i}{\partial x_j} \right) - \frac{\partial p}{\partial x_i} + \rho b_i = 0$$

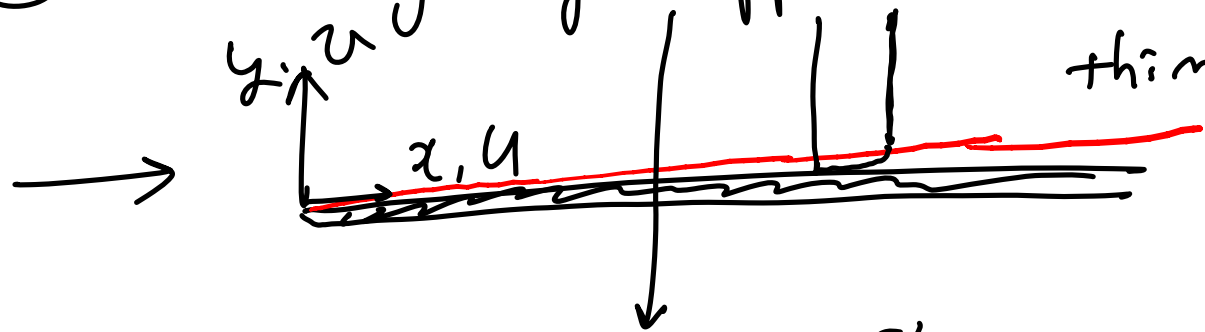


coeff. of
volumetric expansion
↓
-sion

- ⑤ Boussinesq approximation: $(\rho - \rho_0) g_i = -\rho_0 g_i \beta (T - T_0)$
 if density variation is not large, treat ρ as constant
 in unsteady and convection terms and treat ρ

as variable only in the gravitational form

⑥ Boundary layer approximation



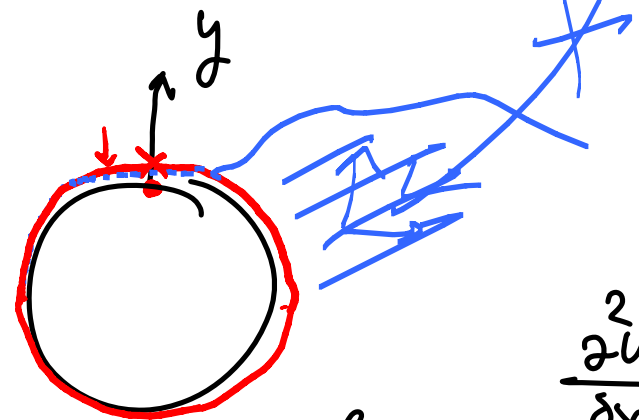
thin shear layer

$$\frac{\partial u}{\partial y} \neq 0$$

$$\frac{\partial}{\partial x} \ll \frac{\partial}{\partial y}$$

$$v \ll u$$

no reverse flow
no recirculation



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

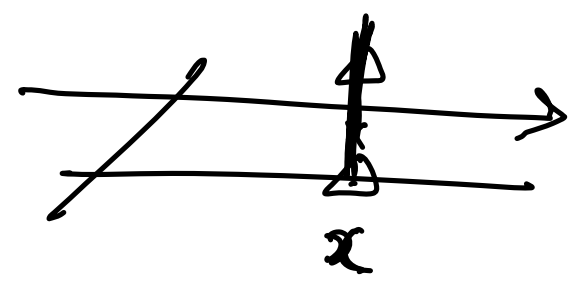
$$\Rightarrow \int \left[\frac{\partial}{\partial t} (\rho u) + \frac{\partial}{\partial x} (\rho u^2) + \frac{\partial}{\partial y} (\rho uv) \right] = -\frac{\partial p}{\partial x} + \mu \frac{\partial^2 u}{\partial y^2}$$

$$\frac{\partial^2 u}{\partial x^2}$$

$$\left(\frac{d}{dt} (p v) \right)$$

$$Q = \frac{p v}{dt}$$

$$= \frac{p v}{dt}$$



① Mathematical Classification of Flows

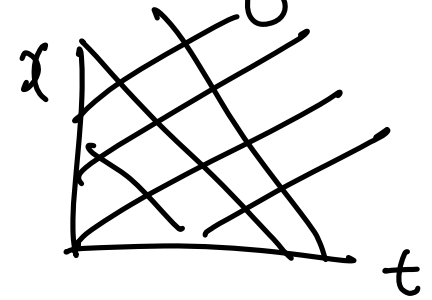
$$au_{xx} - 2bu_{xy} + cu_{yy} = f$$

$b^2 - ac > 0 \rightarrow$ hyperbolic \rightarrow two real eigenvalues
 $= 0 \rightarrow$ parabolic \rightarrow one " eigenvalue
 $< 0 \rightarrow$ elliptic \rightarrow imaginary or complex eigenvalues

Numerical method should respect the properties of the eqs.

① Hyperbolic flows

unsteady inviscid compress. flow
 Steady comp. supersonic flow

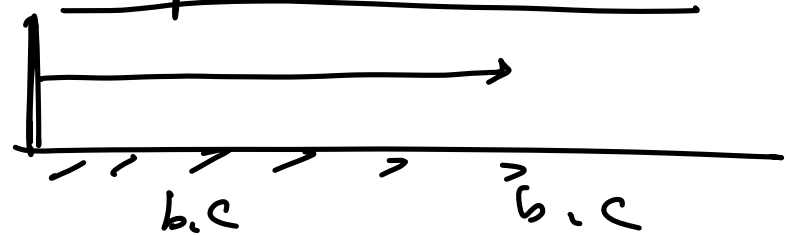


② parabolic flows

boundary layer approx. → parabolic char.

but pressure should be obtained by potential flow approach → elliptic b.c

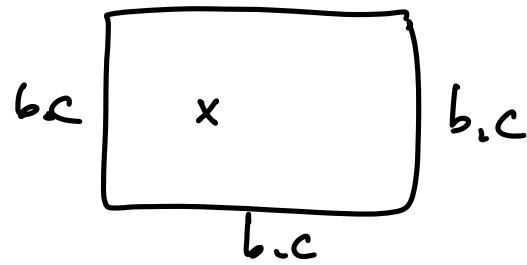
→ mixed type



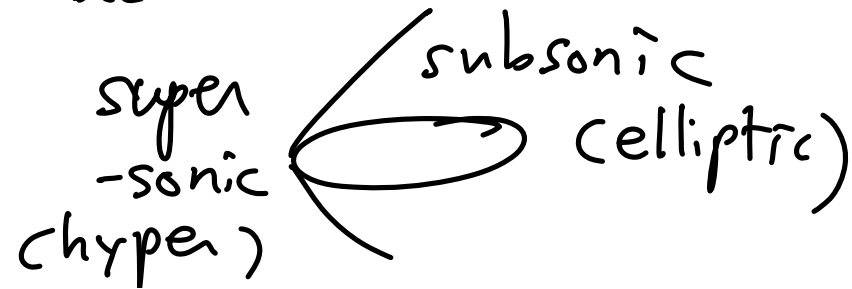
③ elliptic flows

recirculating flow

unsteady incomp. flow (mixed)



④ mixed flow types
steady transonic flow



CFD \rightarrow $\left(\begin{array}{l} \text{Fluid mechanics} \\ \text{Numerical analysis} \end{array} \right) \Rightarrow \text{J. Comput. Phys.}$

Ch. 2 Introduction to Numerical Methods

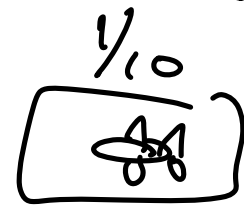
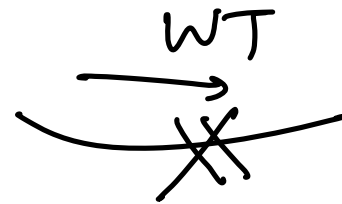
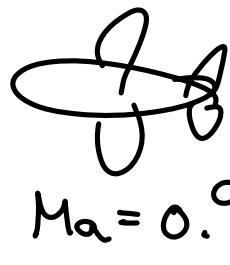
1. Approaches to fluid dynamical problems

- Dimensional analysis & experiments

$$F_D = C_D \cdot A \cdot \frac{1}{2} \rho U^2$$

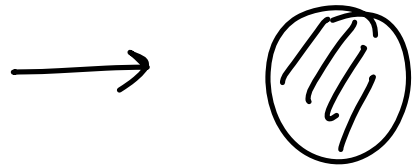
C_D vs. Re

$$Re = \frac{U_{\infty} d}{\nu}$$



$U \uparrow 10 \rightarrow Ma = 9$

Re & Ma , Re & Fr , Re & We

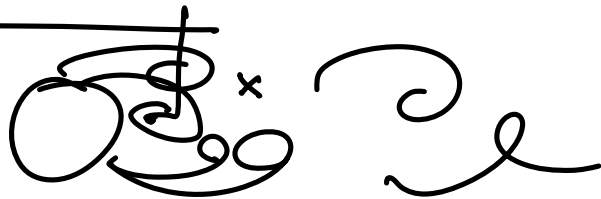


experiment : efficient means to measure global parameters like drag, lift, heat transfer rate, etc.

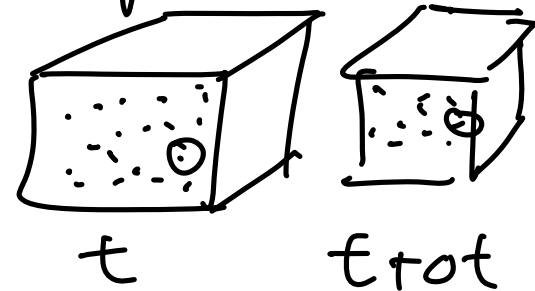
→ most of time, no detailed measurement.

i.e. some important phenomena are missing like flow separation

not true from PIV people.



$$\vec{\omega} = \nabla \times \vec{u}$$



⇒ use CFD

but errors in CFD

CFD requires "good" experiments

turbulence

↓
constants

2. Components of a numerical sol. method

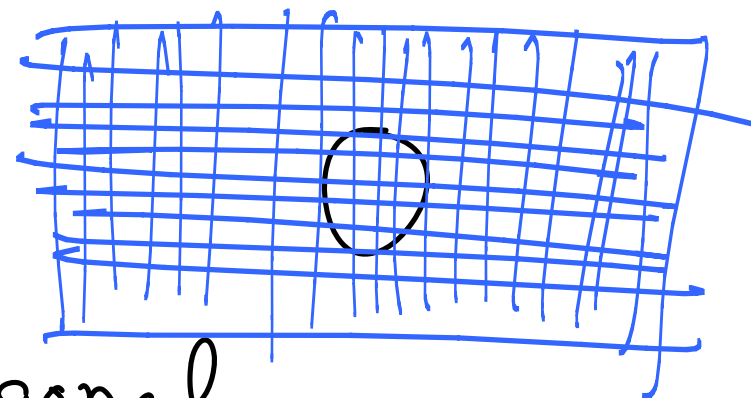
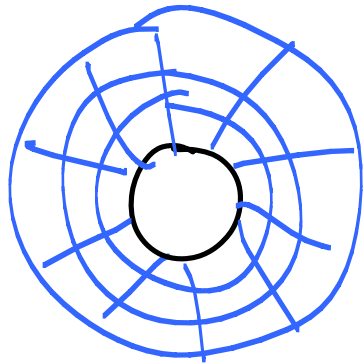
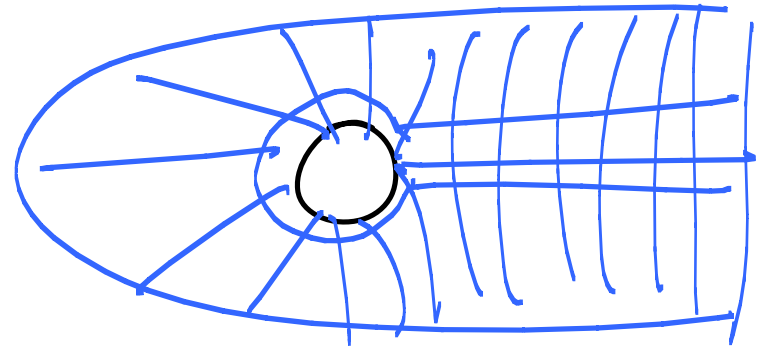
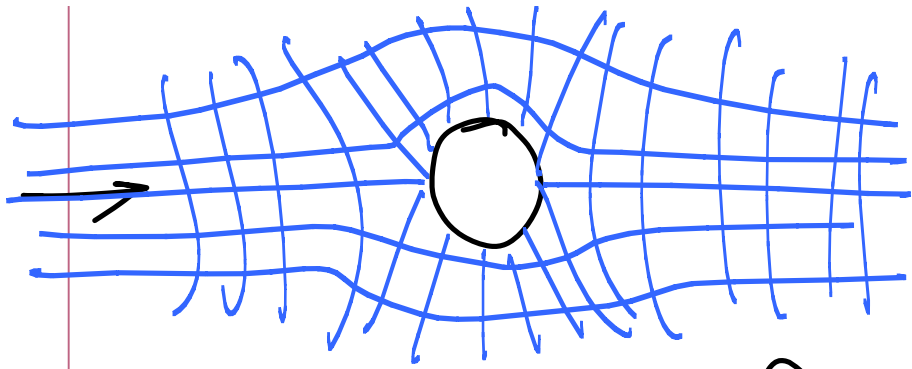
① mathematical model

Governing eqs., b.c.'s, i.c.'s.

② discretization method

FDM, FVM, FEM, etc.

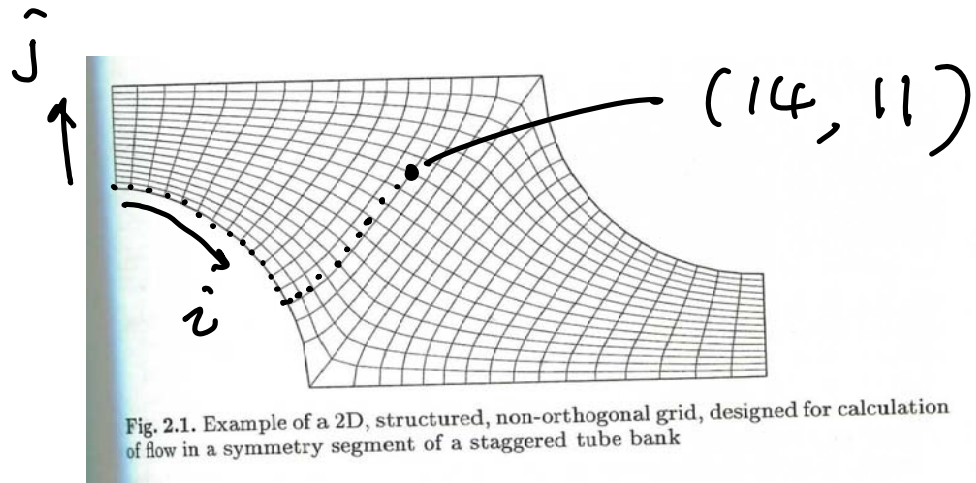
③ coordinates and basis vector systems



{ Cartesian
cylindrical
spherical
curvilinear orthogonal
" non-orthogonal
covariant or contra-variant

④ Grid

- Structured grid



- Block-structured grid with matching interface

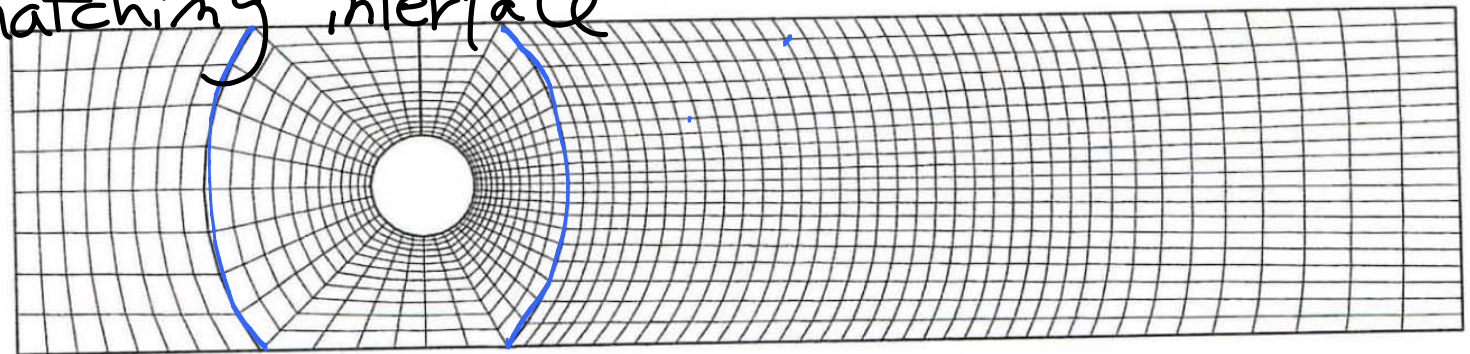


Fig. 2.2. Example of a 2D block-structured grid which matches at interfaces, used to calculate flow around a cylinder in a channel

can be treated in a fully conservative manner

without " "

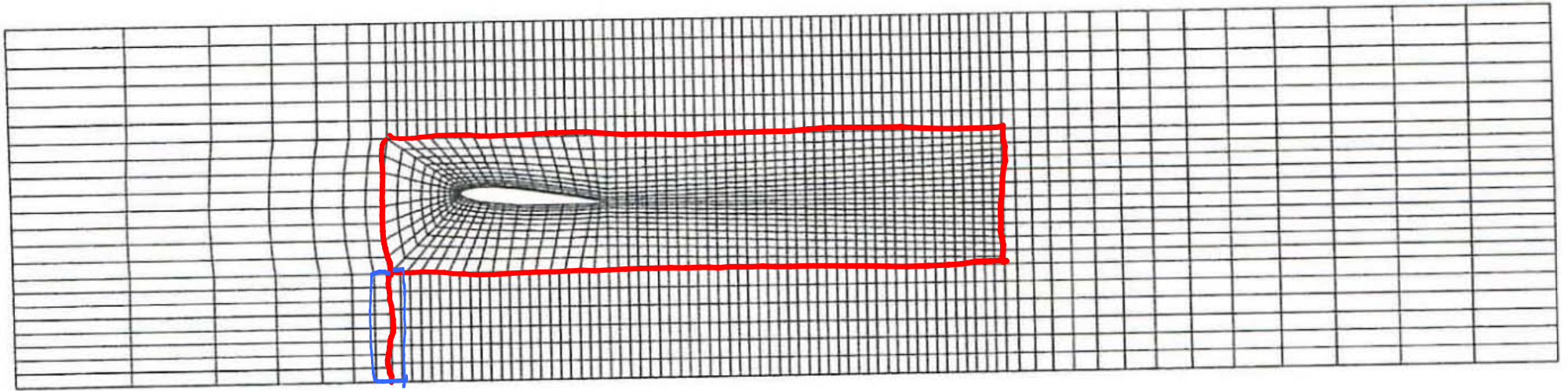


Fig. 2.3. Example of a 2D block-structured grid which does not match at interfaces, designed for calculation of flow around a hydrofoil under a water surface

with overlapping blocks
(composite grid, chimera grid) → difficulty in
conserving properties
good for complex domain.

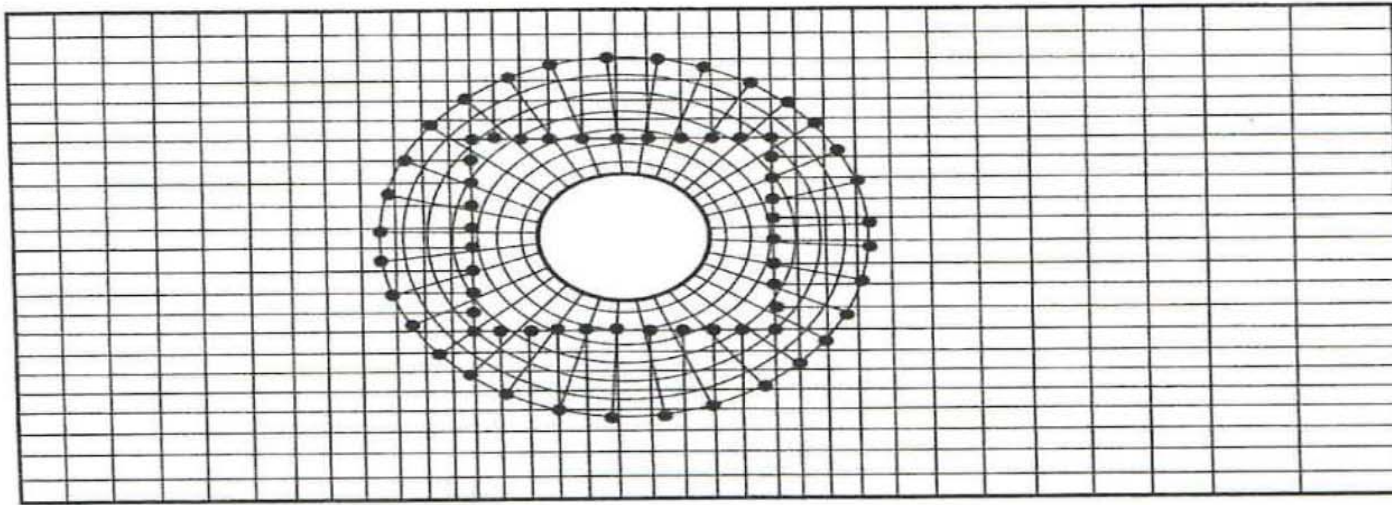
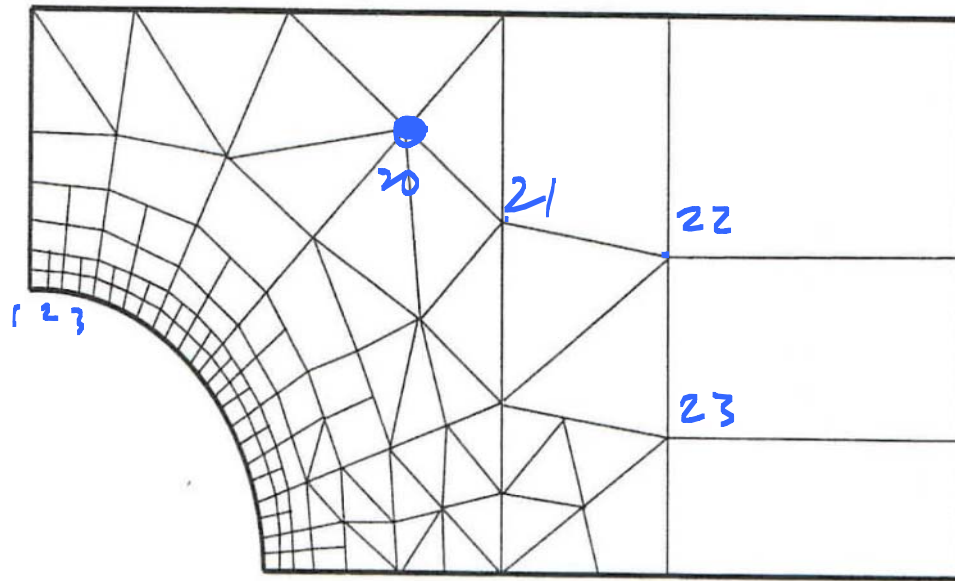


Fig. 2.4. A composite 2D grid, used to calculate flow around a cylinder in a channel

- Unstructured grid
 - very complex geometries
 - good for FVM, FEM
 - irregularity of the data structure
 - sparse matrix

$$\underline{A} \underline{u} = \underline{\phi}$$

$$\begin{matrix} \downarrow \\ u_1 \\ u_2 \\ \vdots \\ u_{100} \end{matrix}$$



grid generation
is difficult.

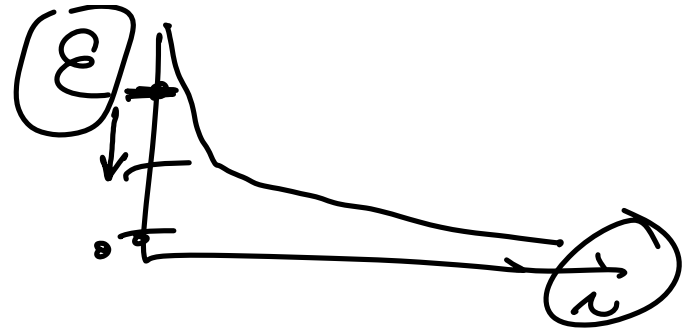
Fig. 2.5. Example of a 2D unstructured grid

Book : Numerical grid generation by Thompson et al.
(1985)

- ⑤ Finite approximations ... accuracy, memory
- ⑥ Solution method for nonlinear algebraic eqs.

⑦ convergence criteria

⑧ "good" physics



3. Properties of numerical sol. methods

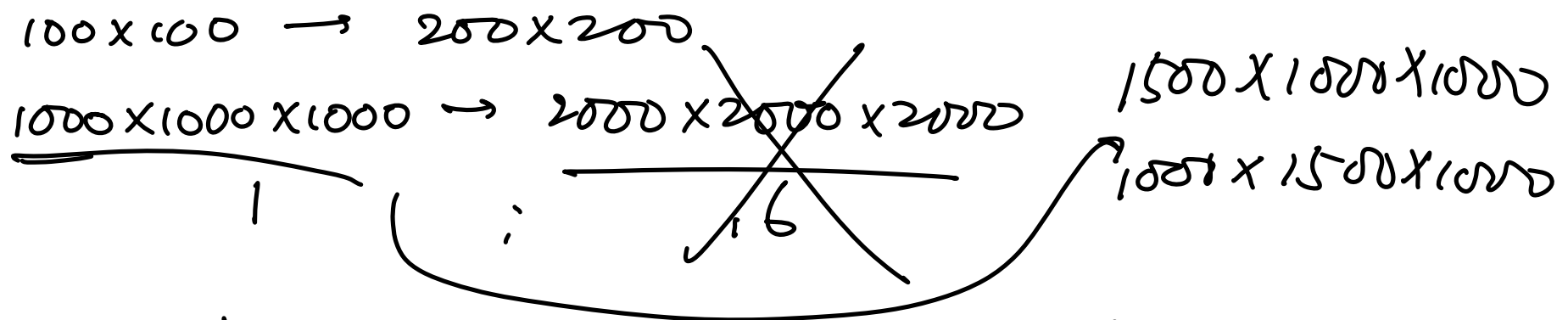
① consistency : modified PDE \rightarrow truncation error
 \rightarrow should go to zero when $\Delta x \rightarrow 0$
& $\Delta t \rightarrow 0$

② stability : von Neumann stability analysis
modified wavenumber analysis

\rightarrow absolutely stable
conditionally stable
unstable

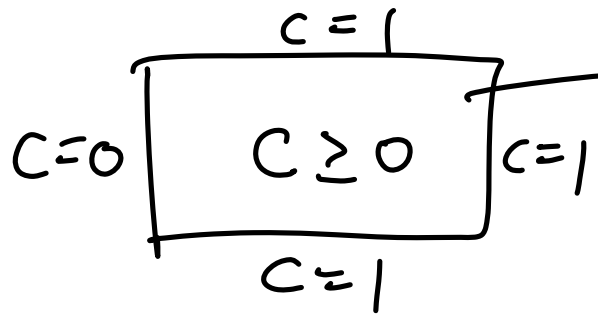
efc.snu.ac.kr

③ convergence : resolution test
(only by numerical experiments)



④ conservation : strong conservative form of G.E.,
+ FVM → conservative
non-conservative scheme → artificial source/sink
→ goes to zero as $\Delta x \rightarrow 0$

⑤ boundedness : difficult to guarantee



$C < 0$ or $C > 1$
some 1st-order schemes guarantees
this property.

but higher-order schemes can produce
unbounded sols,

⑥ realizability : models of phenomena should be
correct.

turbulence, combustion, multi-phase
flow

⑦ accuracy

i) modeling error : actual flow vs. exact sol. of math model

ii) discretization error: $N-S$ eq. \rightarrow $(N-S \text{ eq} + TM)$
 exact sol. of G.E vs. exact sol. of algebraic eqs.

iii) iteration error: iterative vs. exact sol. of algebraic eqs.

4. Discretization approaches

① FDM: oldest method

PDE \longrightarrow algebraic eq.

simple & effective

easy to obtain higher-order scheme on regular grids

conservation is not enforced
restricted to simple geometries

② FVM : integral eq. \rightarrow algebraic eq.
suitable for complex geometries
higher order schemes are difficult
most popular in engineering

③ FEM : eqs. are multiplied by a weighting f_t
in a way that guarantees continuity of the
sol. across element boundaries
arbitrary geometries

sparse matrix.

ch. 3 Finite Difference Methods (FDM)

$$\frac{\partial(\rho\phi)}{\partial t} + \underbrace{\frac{\partial}{\partial x_j} (\rho u_j \phi)}_{\text{convection}} = \underbrace{\frac{\partial}{\partial x_j} \left(\Gamma \frac{\partial \phi}{\partial x_j} \right)}_{\text{diffusion}} + \mathcal{S}\phi$$

Steady

$$\longrightarrow \frac{\partial}{\partial x_j} (\rho u_j \phi) = \frac{\partial}{\partial x_j} \left(\Gamma \frac{\partial \phi}{\partial x_j} \right) + \mathcal{S}\phi$$

1. Approximation of the 1st derivative

c1) Taylor series expansion

$$\phi(x) = \phi(x_i) + (x - x_i) \left(\frac{\partial \phi}{\partial x} \right)_i + \frac{(x - x_i)^2}{2!} \left(\frac{\partial^2 \phi}{\partial x^2} \right)_i$$

$$+ \frac{(x-x_i)^3}{3!} \left(\frac{\partial^3 \phi}{\partial x^3} \right)_i + \dots + \frac{(x-x_i)^n}{n!} \left(\frac{\partial^n \phi}{\partial x^n} \right)_i + \text{HOT}$$

② $x = x_{i+1}$,

$$\left(\frac{\partial \phi}{\partial x} \right)_i = \frac{\phi_{i+1} - \phi_i}{x_{i+1} - x_i} - \frac{x_{i+1} - x_i}{2} \left(\frac{\partial^2 \phi}{\partial x^2} \right)_i - \frac{(x_{i+1} - x_i)^2}{6} \left(\frac{\partial^3 \phi}{\partial x^3} \right)_i + \text{HOT}$$

leading 6 truncation error \times

forward difference (FD)

③ $x = x_{i-1}$

$$\left(\frac{\partial \phi}{\partial x} \right)_i = \frac{\phi_i - \phi_{i-1}}{x_i - x_{i-1}} + \frac{x_i - x_{i-1}}{2} \left(\frac{\partial^2 \phi}{\partial x^2} \right)_i - \frac{(x_i - x_{i-1})^2}{6} \left(\frac{\partial^3 \phi}{\partial x^3} \right)_i + \text{HOT}$$

backward difference (BD)

④ both

$$\left(\frac{\partial \phi}{\partial x} \right)_i = \frac{\phi_{i+1} - \phi_{i-1}}{x_{i+1} - x_{i-1}} - \frac{(x_{i+1} - x_i)^2 - (x_i - x_{i-1})^2}{2(x_{i+1} - x_{i-1})} \left(\frac{\partial^2 \phi}{\partial x^2} \right)_i$$

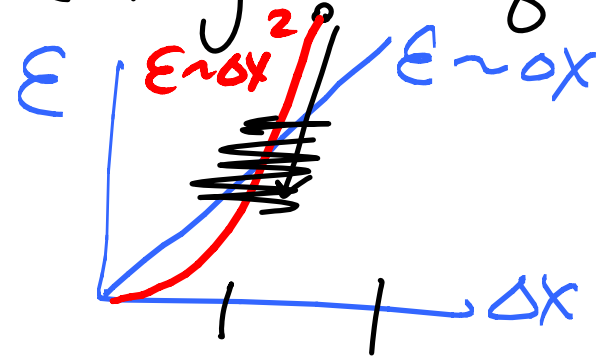
central difference (CD)

$$- \frac{(\bar{x}_{i+1} - x_e)^3 + (x_e - \bar{x}_{i-1})^3}{6(\bar{x}_{i+1} - \bar{x}_{i-1})} \left(\frac{\partial^3 \phi}{\partial x^3} \right)_{\bar{x}} + \text{HOT}$$

truncation error $\epsilon_T = (\Delta x)^n \alpha_{n+1} + (\Delta x)^{n+1} \alpha_{n+2} + \dots$

- The order of an approximation indicates how fast the error is reduced when the grid is refined.

It does not indicate the absolute magnitude of the error.

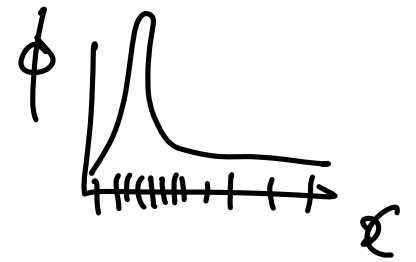


- When $\Delta x_{i+1} (= x_{i+1} - x_i) = \Delta x_i (= x_i - x_{i-1})$,
 CD becomes 2nd-order accurate.

- ϵ_T for CD:

$$\epsilon_T = - \frac{(\Delta x_{i+1})^2 - (\Delta x_i)^2}{2(\Delta x_{i+1} + \Delta x_i)} \left(\frac{\partial^2 \phi}{\partial x^2} \right)_i - \frac{(\Delta x_{i+1})^3 + (\Delta x_i)^3}{6(\Delta x_{i+1} + \Delta x_i)^2} \left(\frac{\partial^3 \phi}{\partial x^3} \right)_i + \text{HOT}$$

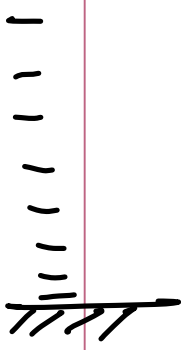
$\underbrace{\hspace{10em}}_{\uparrow \mathcal{O}(\Delta x)}$



why use non-uniform grid?

→ to resolve large gradients of ϕ

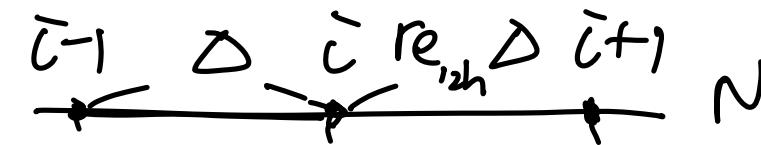
If $\boxed{\Delta x_{i+1} = r_e \Delta x_i}$, then



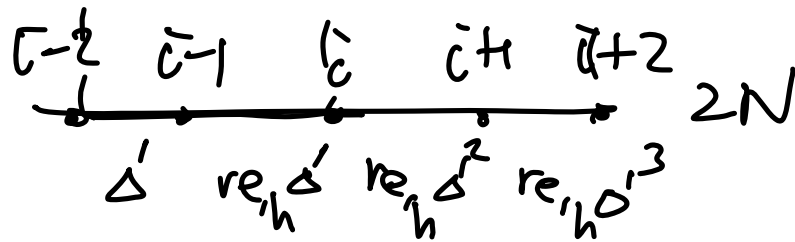
$$\epsilon_T = \frac{1-r_e}{2} \delta x_i \left(\frac{\partial^2 \Phi}{\partial x^2} \right)_i - \frac{1-r_e+r_e^2}{6} \delta x_i^2 \left(\frac{\partial^3 \Phi}{\partial x^3} \right)_i + \text{HOT}$$

As $r_e \rightarrow 1$, $\frac{1-r_e}{2} \rightarrow 0$

What happens when grid is refined in CD?



$$\Delta = \Delta' (1+r_{e,h})$$



$$\Delta (1+r_{e,h}) = \Delta' (1+r_{e,h}+r_{e,h}^2+r_{e,h}^3)$$

$$\checkmark \quad r_{e,h} = \sqrt{r_{e,2h}}$$

$$\checkmark \quad (\delta x_i)_{2h} = (\delta x_i)_h + (\delta x_{i-1})_h = (1+r_{e,h}) (\delta x_{i-1})_h$$

the ratio of the leading truncation errors

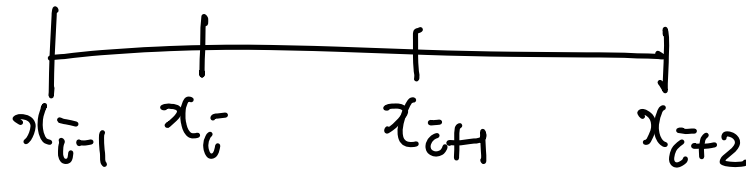
$$= \frac{(1-r_{e,2h}) \alpha x_{i,2h}}{(1-r_{e,h}) \alpha x_{i,h}} = \frac{(1+r_{e,h})^2}{r_{e,h}} \geq 4$$

when $r_e = 1 \rightarrow 4$ (second order)

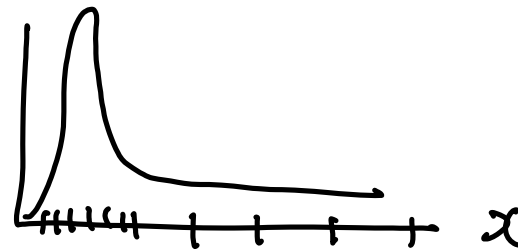
$r_e \neq 1$
 (i.e. $r_e > 1$
 $r_e < 1$) $\rightarrow > 4$ (faster than 2nd-order)
 properly good!

Even when the grid is non-uniform, the truncation error is reduced as in a second-order scheme when the grid is refined!

non-uniform grids



CD \rightarrow $O(\Delta x)$

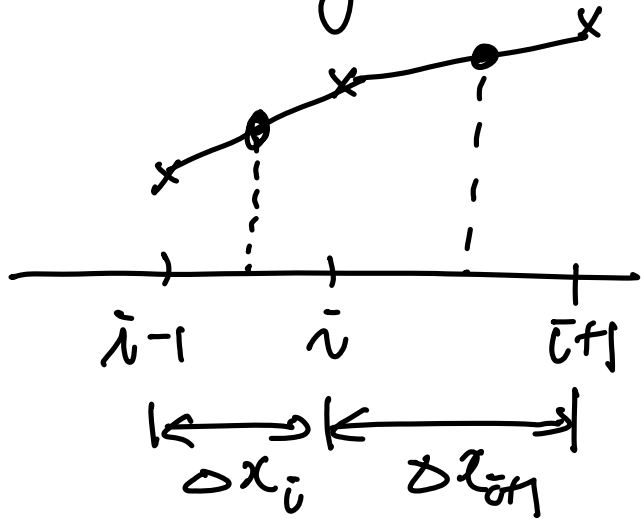


$\Delta x_{i+1} = r_e \Delta x_i$

the ratio of the leading truncation errors

$$r_T = \frac{\epsilon_{T_h}}{\epsilon_{T_{2h}}} = \frac{(1+r_e h)^2}{r_e h} \geq 4 \rightarrow \text{at least second order!}$$

• Absolutely second order



$$\begin{aligned} \frac{\partial \phi}{\partial x} \Big|_i &= \frac{\Delta x_{i+1} \frac{\partial \phi}{\partial x} \Big|_{i-\frac{1}{2}} + \Delta x_i \frac{\partial \phi}{\partial x} \Big|_{i+\frac{1}{2}}}{\Delta x_i + \Delta x_{i+1}} \\ &= \frac{\phi_{i+1} (\Delta x_i)^2 - \phi_{i-1} (\Delta x_{i+1})^2 + \phi_i [(\Delta x_{i+1})^2 - (\Delta x_i)^2]}{\Delta x_{i+1} \Delta x_i (\Delta x_i + \Delta x_{i+1})} \\ &\quad - \frac{\Delta x_{i+1} \Delta x_i}{6} \frac{\partial^3 \phi}{\partial x^3} \Big|_i + \text{HOT} \\ &\quad \underbrace{\hspace{10em}}_{\hookrightarrow \mathcal{O}(\Delta x^2)} \end{aligned}$$

(2) Polynomial fitting

to fit the function to an interpolation curve

and differentiate the resulting curve.

ex) parabola for x_{i-1}, x_i, x_{i+1}

$$\rightarrow \left. \frac{\partial \phi}{\partial x} \right|_i = \frac{\phi_{i+1} (\Delta x_i)^2 - \phi_{i-1} (\Delta x_{i+1})^2 + \phi_i [(\Delta x_{i+1})^2 - (\Delta x_i)^2]}{\Delta x_{i+1} \Delta x_i (\Delta x_i + \Delta x_{i+1})}$$

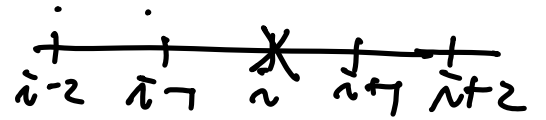
second order

other polynomials, splines etc.

In general, approximation of the 1st derivative possesses a truncation error of the same order as the degree of the polynomial used to approximate the function.

3rd-order polynomials

$$\rightarrow \frac{\partial \phi}{\partial x} \Big|_{\tilde{i}} = \frac{2\phi_{\tilde{i}+1} + 3\phi_{\tilde{i}} - 6\phi_{\tilde{i}-1} + \phi_{\tilde{i}-2}}{6\Delta x} + \mathcal{O}(\Delta x^3) \quad \text{BD(S)}$$



$$\frac{\partial \phi}{\partial x} \Big|_{\tilde{i}} = \frac{-\phi_{\tilde{i}+2} + 6\phi_{\tilde{i}+1} - 3\phi_{\tilde{i}} - 2\phi_{\tilde{i}-1}}{6\Delta x} + \mathcal{O}(\Delta x^3) \quad \text{FD}$$

4th-order polynomials

$$\frac{\partial \phi}{\partial x} \Big|_{\tilde{i}} = \frac{-\phi_{\tilde{i}+2} + 8\phi_{\tilde{i}+1} - 8\phi_{\tilde{i}-1} + \phi_{\tilde{i}-2}}{12\Delta x} + \mathcal{O}(\Delta x^4) \quad \text{CD}$$

For convection terms,

$$u \frac{\partial \phi}{\partial x}$$



$u > 0 \rightarrow$ BD

$u < 0 \rightarrow$ FD

} upwind schemes

$$u \frac{\partial \phi}{\partial x} \Big|_i = \begin{cases} u_i \cdot \frac{\phi_i - \phi_{i-1}}{\Delta x} & \text{BD if } u_i > 0 \\ u_i \cdot \frac{\phi_{i+1} - \phi_i}{\Delta x} & \text{FD if } u_i < 0 \end{cases}$$

$$\frac{\partial \phi}{\partial x} \Big|_i = \frac{\phi_i - \phi_{i-1}}{\Delta x} - \underbrace{\frac{\Delta x}{2} \frac{\partial^2 \phi}{\partial x^2} \Big|_i}_{O(\Delta x)} + \dots$$

First-order upwind scheme

very inaccurate
truncation error sometimes bigger than
actual diffusivity.

$$\odot \quad u \frac{\partial \phi}{\partial x} + v \frac{\partial \phi}{\partial y} = \underbrace{\Gamma}_{\text{false diffusion}} \left(\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} \right) + \underbrace{\frac{\Delta x}{2} \frac{\partial^2 \phi}{\partial x^2}}_{\text{false diffusion}}$$

Higher-order upwind scheme is costly.

→ use CD.

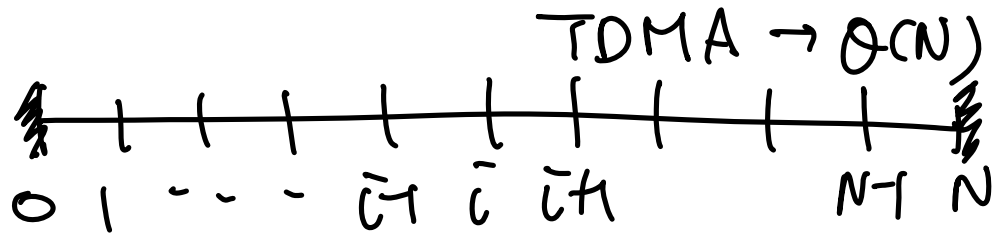
(3) compact schemes $\frac{}{\bar{i}-1} \quad \frac{}{\bar{i}} \quad \frac{}{\bar{i}+1} \rightarrow$ 2nd order accuracy

→ increase the accuracy w/ the same grids but by including derivatives at $\bar{i}-1$ and $\bar{i}+1$.

(Padé scheme)

$$a \frac{\partial \phi}{\partial x} \Big|_{\bar{i}-1} + b \frac{\partial \phi}{\partial x} \Big|_{\bar{i}} + c \frac{\partial \phi}{\partial x} \Big|_{\bar{i}+1} + d \phi_{\bar{i}-1} + e \phi_{\bar{i}} + f \phi_{\bar{i}+1} = 0$$

$$\rightarrow \frac{\partial \phi}{\partial x} \Big|_{\bar{i}-1} + 4 \frac{\partial \phi}{\partial x} \Big|_{\bar{i}} + \frac{\partial \phi}{\partial x} \Big|_{\bar{i}+1} = 3 \frac{\phi_{\bar{i}+1} - \phi_{\bar{i}-1}}{\Delta x} + \underline{\underline{O(\Delta x^4)}}$$

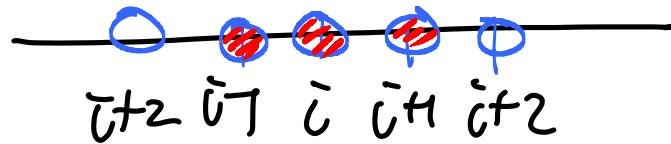


$\bar{i} = 1, \dots, N-1$ compact

* A family of compact centered approximations of up to sixth order can be written as

$$\Rightarrow \alpha \frac{\partial \phi}{\partial x} \Big|_{i+1} + \frac{\partial \phi}{\partial x} \Big|_i + \alpha \frac{\partial \phi}{\partial x} \Big|_{i-1} = \beta \frac{\phi_{i+1} - \phi_{i-1}}{2\Delta x} + \gamma \frac{\phi_{i+2} - \phi_{i-2}}{4\Delta x}$$

$A\phi = b$
 \uparrow 3 pts \rightarrow TDMA
 \searrow 5 pts



Scheme	Truncation error	α	β	γ
--------	------------------	----------	---------	----------

CDS-2	$\frac{(0x)^2}{3!} \frac{\partial^3 \phi}{\partial x^3}$	0	1	0
CDS-4	$\frac{13(0x)^4}{3 \cdot 3!} \frac{\partial^5 \phi}{\partial x^5}$	0	$\frac{4}{3}$	$-\frac{1}{3}$
Padé-4	$\frac{(0x)^4}{5!} \frac{\partial^5 \phi}{\partial x^5}$	$\frac{1}{4}$	$\frac{3}{2}$	0
Padé-6	$\frac{4(0x)^6}{7!} \frac{\partial^7 \phi}{\partial x^7}$	$\frac{1}{3}$	$\frac{14}{9}$	$\frac{1}{9}$

2. Approximation of the 2nd derivative

(1) Taylor series expansion

(2) Use a formula for 1st derivative

$$\text{ex) } \frac{\partial^2 \phi}{\partial x^2} = \frac{\partial}{\partial x} \left(\frac{\partial \phi}{\partial x} \right) = \frac{\frac{\partial \phi}{\partial x} \Big|_{i+\frac{1}{2}} - \frac{\partial \phi}{\partial x} \Big|_{i-\frac{1}{2}}}{\frac{1}{2}(x_{i+1} - x_{i-1})} + \dots$$

(3) Polynomial fitting

In general, the truncation error of the approx. to the 2nd derivative is the degree of the interpolating polynomial minus one.

One order is gained when the spacing is uniform and even-order polynomials are used.

* One can use approx. of the 2nd derivative to increase the accuracy of approx. to the 1st derivative.

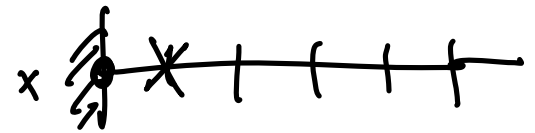
$$\frac{\partial \phi}{\partial x} \Big|_i = \frac{\phi_{i+1} - \phi_i}{x_{i+1} - x_i} - \frac{1}{2} \underbrace{(x_{i+1} - x_i)}_{O(\Delta x)} \underbrace{\frac{\partial^2 \phi}{\partial x^2} \Big|_i}_{O(\Delta x)} + \dots$$

FD CD + O(\Delta x)

→ 2nd-order accurate O(\Delta x^2)

* higher order schemes need more grid points

→ more complex eqs to solve
more complex to treat b.c.



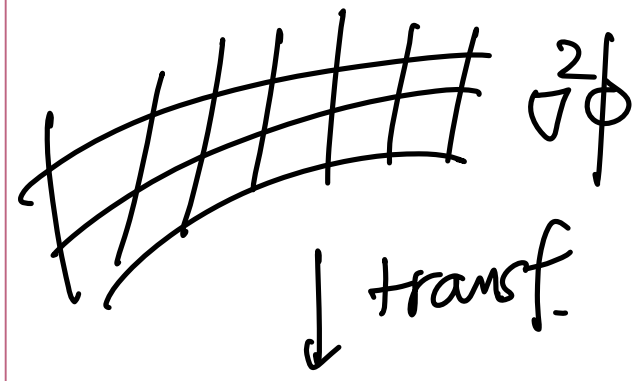
* second-order accuracy is good enough for engineering applications.

$$\frac{\partial}{\partial x} \left(\Gamma \frac{\partial \phi}{\partial x} \right) = \frac{1}{\frac{1}{2}(x_{i+1} - x_{i-1})} \left[\left(\Gamma \frac{\partial \phi}{\partial x} \right)_{i+\frac{1}{2}} - \left(\Gamma \frac{\partial \phi}{\partial x} \right)_{i-\frac{1}{2}} \right]$$

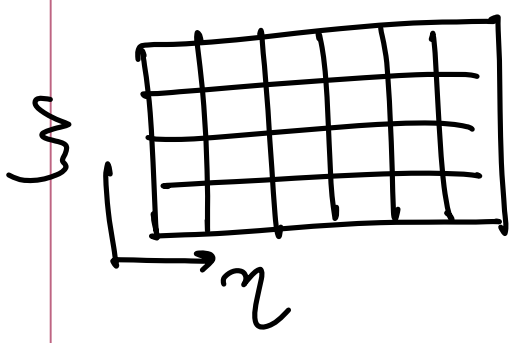
3. Mixed derivatives

$\frac{\partial^2 \phi}{\partial x \partial y}$ occurs

when non-orthogonal grids are used.



↓ transf.



$$\frac{\partial^2 \phi}{\partial \eta^2}$$

$$\frac{\partial^2 \phi}{\partial \eta \partial \xi}$$

$$\frac{\partial^2 \phi}{\partial \xi^2}$$

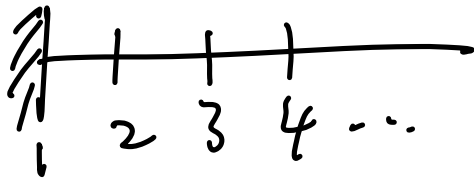
$$\frac{\partial^2 \phi}{\partial \eta \partial \xi} = \frac{\partial}{\partial \eta} \left(\frac{\partial \phi}{\partial \xi} \right)$$

↑ ↑

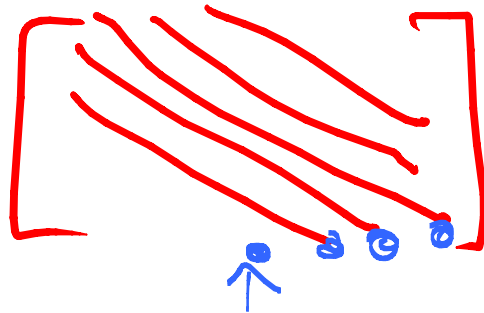
4. Implementation of boundary conditions

(Dirichlet b.c. : $\phi_1 = c$ \longrightarrow When high order scheme is used at interior pt, one may use a different scheme at $\tilde{j} = 2$: eg,

Neumann b.c. : $\frac{\partial \phi}{\partial x}|_1 = 0$



$$\frac{\partial \phi}{\partial x}|_2 = \frac{-\phi_5 + 6\phi_4 + 18\phi_3 + 10\phi_2 - 33\phi_1}{60 \Delta x} + O(\Delta x^4)$$



Neumann b.c. : $\frac{\partial \phi}{\partial x} \Big|_1 = 0 = \frac{\phi_2 - \phi_1}{x_2 - x_1}$

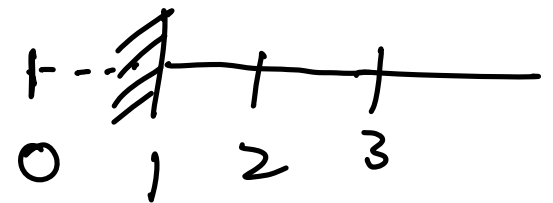
Usually the approx. of the b.d. value or near the b.d. is of lower order than the approx. used deeper in the interior and may be one-sided difference.

* Issues of global accuracy (Fletcher's book)

Dirichlet b.c. → no prob.

Neumann b.c. → $\frac{\partial \phi}{\partial x} \Big|_1 = c(\epsilon)$

$\frac{\partial \phi}{\partial t} - \alpha \frac{\partial^2 \phi}{\partial x^2} = 0$ ——— ①



CD2 + "Implicit" Euler (IE) $\frac{\partial y}{\partial t} = f(y) \rightarrow \frac{y^{n+1} - y^n}{\Delta t} = f(y^{n+1}) + O(\Delta t)$

$$\frac{\phi_j^{n+1} - \phi_j^n}{\Delta t} - \alpha \frac{\phi_{j+1}^{n+1} - 2\phi_j^{n+1} + \phi_{j-1}^{n+1}}{\Delta x^2} = 0 \quad \text{--- (2)}$$

$O(\Delta t, \Delta x^2)$

CD2 + "Explicit" Euler (EE) $\frac{y^{n+1} - y^n}{\Delta t} = f(y^n) + O(\Delta t)$

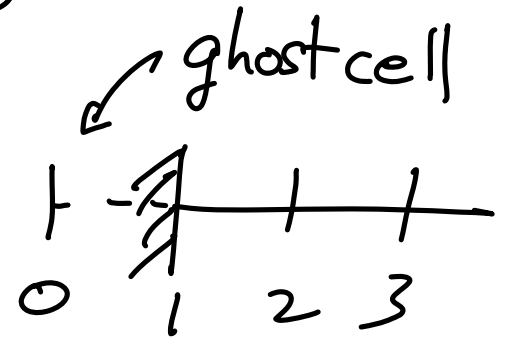
$$\frac{\phi_j^{n+1} - \phi_j^n}{\Delta t} - \alpha \frac{\phi_{j+1}^n - 2\phi_j^n + \phi_{j-1}^n}{\Delta x^2} = 0 \quad \text{--- (2)'}$$

$O(\Delta t, \Delta x^2)$

1st order b.c. $\frac{\phi_2^{n+1} - \phi_1^{n+1}}{\Delta x} = c^{n+1}$

$\rightarrow \phi_1^{n+1} = \phi_2^{n+1} - c^{n+1} \Delta x \quad \text{--- (3)}$

or $\phi_1^n = \phi_2^n - c^n \Delta x \quad \text{--- (3)'}$



2nd order b.c. $\frac{\phi_2^{n\pi} - \phi_0^{n\pi}}{2\Delta x} = c^{n\pi} \rightarrow \phi_0^{n\pi} = \phi_2^{n\pi} - 2c^{n\pi}\Delta x \quad \text{--- (4)}$

or $\phi_0^n = \phi_2^n - 2c^n\Delta x \quad \text{--- (4')}$

Apply (2) (or (2')) at $j=1$ and use $\phi_0^{n\pi}$ from (4) (or (4')).

$\rightarrow (1 + 2\frac{\alpha\Delta t}{\Delta x^2})\phi_1^{n\pi} - 2\frac{\alpha\Delta t}{\Delta x^2}\phi_2^{n\pi} = \phi_1^n - 2\frac{\alpha\Delta t}{\Delta x^2} \cdot \Delta x \cdot c^{n\pi} \quad \text{--- (5)}$

or (2) & (4') $\phi_1^{n\pi} = -2\frac{\alpha\Delta t}{\Delta x^2} \cdot \Delta x \cdot c^n + (1 - 2\frac{\alpha\Delta t}{\Delta x^2})\phi_1^n + 2\frac{\alpha\Delta t}{\Delta x^2}\phi_2^n \quad \text{--- (5')}$

Ex) $\frac{\partial \phi}{\partial t} - \alpha \frac{\partial^2 \phi}{\partial x^2} = 0 \quad 0.1 \leq x \leq 1 \quad \leftarrow \text{HW! (1 week)}$

b.c. $\frac{\partial \phi}{\partial x} = 2 - 2\pi \sin 0.5\pi \cdot e^{-\alpha(\frac{\pi}{2})^2 t} \quad \text{--- (a) } x = 0.1$

$\phi = 2 \quad \text{--- (b) } x = 1$

i.c. $\phi = 2x + 4 \cos 0.5\pi x \cdot e^{-\alpha \left(\frac{\pi}{2}\right)^2 \cdot 0.8}$ @ $t = 0.8$

$\rightarrow \phi_{\text{exact}} = 2x + 4 \cos 0.5\pi x e^{-\alpha \left(\frac{\pi}{2}\right)^2 t}$

$\left(S \equiv \frac{\alpha \Delta t}{\Delta x^2} = 0.3 \right)$

@ $t = 9$

(2)' + (3)'

(2)' + (5)'

Δx	0.225	0.1125	0.05625	convergence rate, r
rms error	0.1958	0.07978	0.03538	1.2
//	0.1753×10^{-2}	0.4235×10^{-3}	0.1064×10^{-3}	2.0

(error distribution) $\Delta x = 0.225$

x	0.1	0.325	0.550	0.775	1
$\textcircled{2}' + \textcircled{3}'$	-0.3799	-0.1989	-0.08366	-0.02610	0
$\textcircled{2}' + \textcircled{1}'$	0.332×10^{-3}	-0.221×10^{-2}	-0.272×10^{-2}	-0.173×10^{-2}	0

biggest error due to 1st-order b.c. approx.

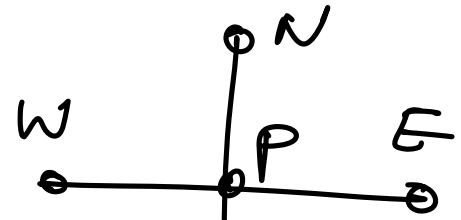
HW 1 : do it for max. error ! (due: Mar. 28)

In conclusion, 1st order b.c. approx. deteriorates numerical sol.

5. Algebraic eqs.

$$\text{FDM} \rightarrow A_p \phi_p + \sum_{\ell} A_{\ell} \phi_{\ell} = Q_p$$

$$\text{2nd order} \rightarrow A_w \phi_w + A_s \phi_s + A_p \phi_p + A_N \phi_N + A_E \phi_E = Q_p$$



$$\begin{bmatrix}
 0 & 0 & 0 & 0 & 0 \\
 A_w & A_s & A_p & A_N & A_E \\
 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0
 \end{bmatrix}
 \begin{bmatrix}
 \vdots \\
 \phi_w \\
 \vdots \\
 \phi_N \\
 \vdots \\
 \phi_p \\
 \vdots \\
 \phi_s \\
 \vdots \\
 \phi_E \\
 \vdots
 \end{bmatrix}
 =
 \begin{bmatrix}
 \vdots \\
 Q_p \\
 \vdots \\
 \vdots \\
 \vdots
 \end{bmatrix}$$

6. Discretization errors

* truncation error (Taylor series truncation)

$$\underset{\substack{\uparrow \\ \text{diff'l operator}}}{L(\underline{\Phi})} = \underset{\substack{\uparrow \\ \text{difference} \\ \text{operator}}}{L_h(\underline{\Phi})} + \underset{\substack{\uparrow \\ \text{truncation error}}}{\tau_h} = 0 \quad \underline{\Phi}: \text{exact sol.}$$

h : grid size

$$L_h(\phi_h) = (A\phi - Q)_h = 0 \quad \phi_h: \text{exact sol. of } L_h(\underline{\Phi})$$

Then, discretization error ϵ_h^d

$$\epsilon_h^d = \underline{\Phi} - \phi_h$$

$$L_h(\Phi_h) = L_h(\Phi - \epsilon_h^d) = -\tau_h - L_h(\epsilon_h^d) = 0$$

$\rightarrow L_h(\epsilon_h^d) = -\tau_h$ \Leftrightarrow truncation error is a source of the discretization error.

Since we don't know the mag. of τ_h , we have to do grid refinement test.

For sufficiently fine grids,

$$\epsilon_h^d \simeq \alpha h^p + \text{HOT} \quad p: \text{order of scheme.}$$

$$\Phi = \Phi_h + \epsilon_h^d = \Phi_{2h} + \epsilon_{2h}^d$$

$$\rightarrow \Phi_h + \alpha h^p + \text{HOT} = \Phi_{2h} + \alpha (2h)^p + \text{HOT}$$

$$\rightarrow \begin{cases} \phi_h - \phi_{2h} = \alpha h^p (2^p - 1) \\ \phi_{2h} - \phi_{4h} = \alpha h^p 2^p (2^p - 1) \end{cases}$$

$$\rightarrow p = \log \left(\frac{\phi_{2h} - \phi_{4h}}{\phi_h - \phi_{2h}} \right) / \log 2$$

↑ useful tool to check the order of accuracy in practice when h is fine enough and convergence is monotonic.

Also, $\varepsilon_h^d = \alpha h^p = \Phi - \phi_h$

$\varepsilon_{2h}^d = 2^p \alpha h^p = \Phi - \phi_{2h}$

$$\alpha h^p (2^p - 1) = \phi_h - \phi_{2h}$$

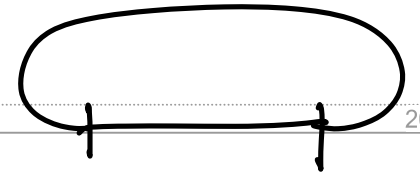
$$\rightarrow \underbrace{\epsilon_h^d}_{\text{error}} = \frac{\phi_h - \phi_{2h}}{2^p - 1} \quad \uparrow \text{to get } p$$

If we have sds. of ϕ_h , ϕ_{2h} (and ϕ_{4h}), we could get better sol. of Φ than ϕ_h by using ϵ_h^d

$$\rightarrow \Phi \doteq \phi_h + \epsilon_h^d$$

↳ Richardson extrapolation.

6. Introduction to spectral method



(1) Fourier transform

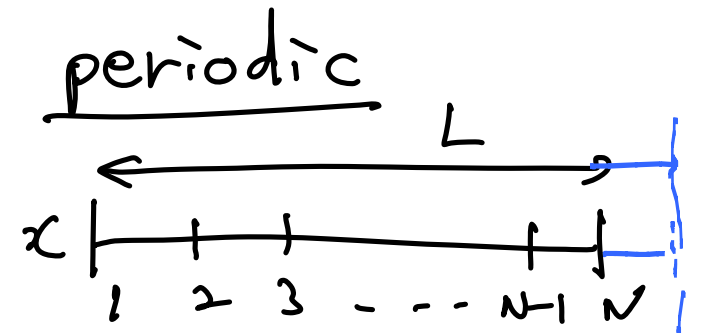
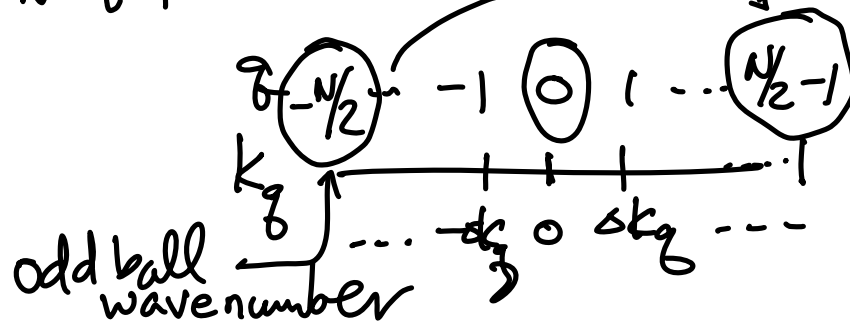
a uniformly spaced set of pts., periodic

discrete Fourier series

$$f(x_i) = \sum_{q=-N/2}^{N/2-1} \hat{f}(k_q) e^{ik_q x_i}$$

$$\hat{f}(k_q) = \frac{1}{N} \sum_{i=1}^N f(x_i) e^{-ik_q x_i}$$

↑
Fourier coeff.
of f .



$$\Delta k_q \cdot L = 2\pi \quad N: \text{even number}$$

$$\Delta k_q = \frac{2\pi}{L} = \frac{2\pi}{N \Delta x}$$

$$k_q = \Delta k_q \cdot q = \frac{2\pi q}{N \Delta x}$$

$$q = -N/2, \dots, 0, 1, \dots, N/2-1$$

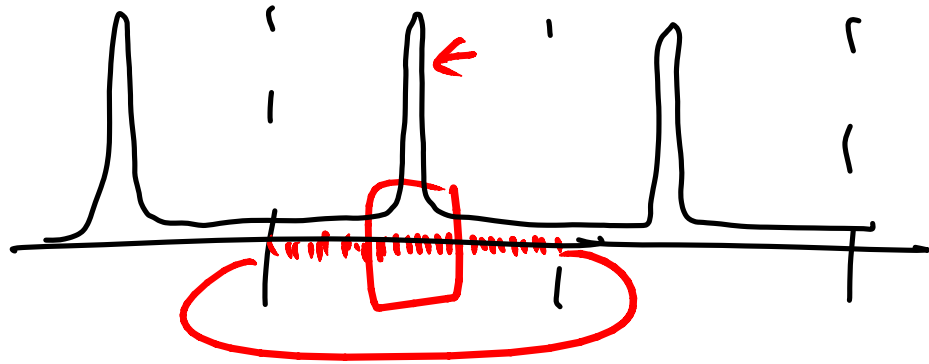
$$\frac{df}{dx} = \sum_{k=-N/2}^{N/2-1} \boxed{\bar{c}_k \hat{f}(k)} e^{ikx} \quad \text{Fourier coeff. of } \frac{df}{dx}$$

To get df/dx . (1) FT of $f(x)$ to get $\hat{f}(k)$

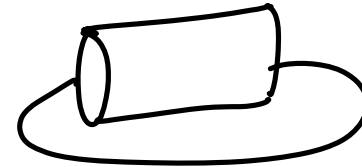
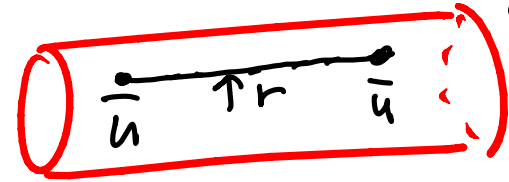
(2) obtain $ik \hat{f}(k)$

(3) IFT to get df/dx

- higher derivatives are easy to obtain $\frac{d^2 f}{dx^2} \rightarrow -k^2 \hat{f}$
 - error decreases exponentially with N when N is large enough.
- much more accurate than FD, FE, FV.



pipe flow fully developed flow



but it may be worse

than FD, FV, FE when N is small.

cost of FT $\rightarrow O(N^2)$

" FFT $\rightarrow O(N \log_2 N)$

(2) Modified wavenumber

$$\frac{d\phi}{dx} \rightarrow \sum_w \hat{\phi} e^{ikx}$$

(spectral or exact)

- $\phi \equiv e^{ikx}$

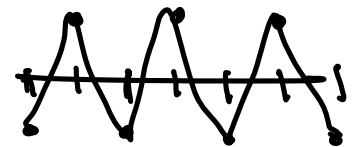
$$\text{CD 2 : } \frac{d\phi}{dx} = \frac{\phi_{i+1} - \phi_{i-1}}{2\Delta x} = \frac{e^{ik(x+\Delta x)} - e^{ik(x-\Delta x)}}{2\Delta x}$$

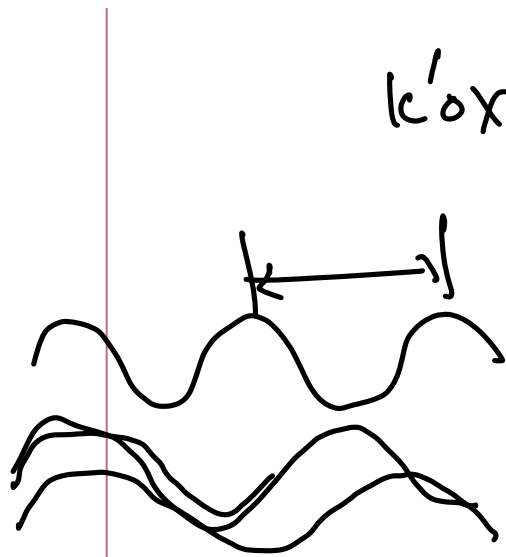
$$= i \underbrace{\frac{\sin k\Delta x}{\Delta x}}_{\text{|||}} e^{ikx}$$

k' : modified wave number

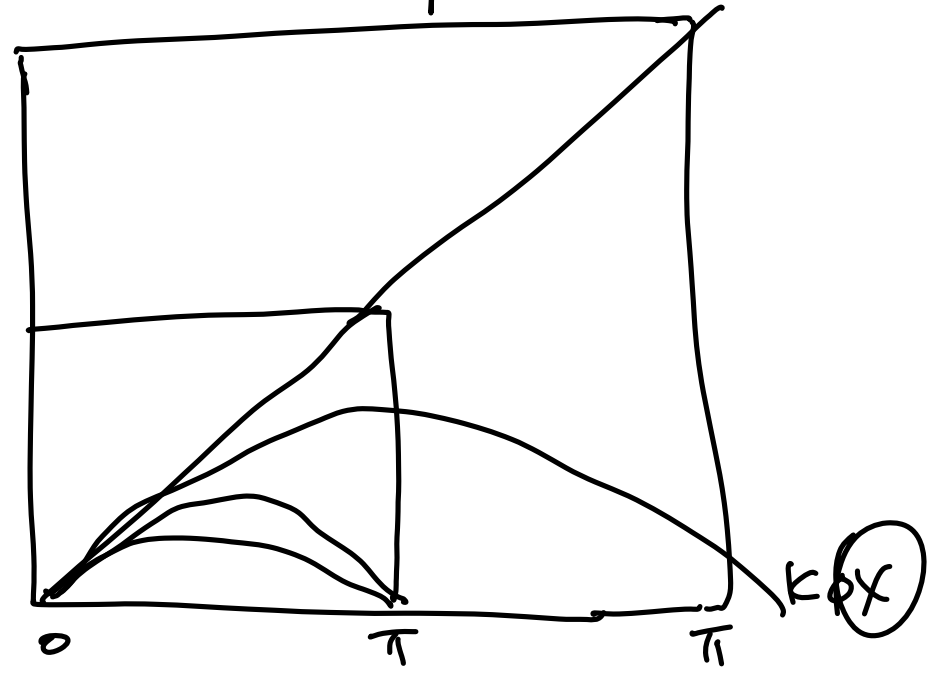
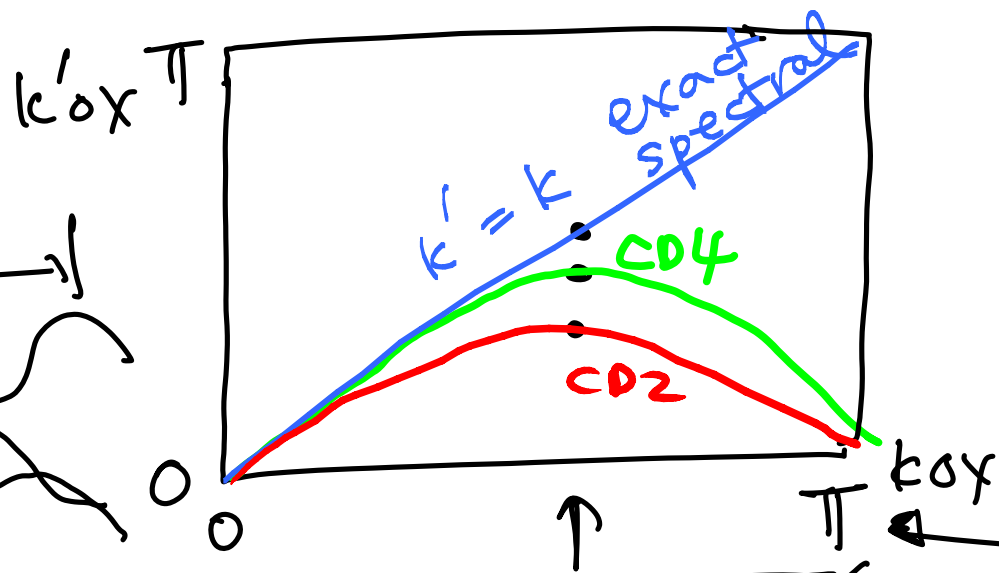
$$k'\Delta x = \sin k\Delta x \quad (\text{CD 2})$$

$$k'\Delta x = \sin k\Delta x \cdot \frac{4 - \cos k\Delta x}{3} \quad (\text{CD 4})$$





↓



$$\frac{L}{N}$$

$$\Delta x = \frac{L}{N}$$

$$K \cdot (2\Delta x) = 2\pi$$

$$K_{ox} = \pi$$

↖

upwind scheme

$$\frac{d\phi}{dx} = \frac{\phi_i - \phi_{i-1}}{\Delta x} = \frac{e^{ikx} - e^{ik(x-\Delta x)}}{\Delta x} = \frac{1}{\Delta x} e^{ikx} (1 - e^{-ik\Delta x})$$

$$= i \underbrace{\frac{-i(1 - e^{-ik\Delta x})}{\Delta x}}_{k'} e^{ikx}$$

k' : complex number

$$k' = -\frac{i}{\Delta x} (1 - \cos k\Delta x + i \sin k\Delta x)$$

$$= \frac{1}{\Delta x} (\sin k\Delta x - i(1 - \cos k\Delta x)) = \frac{\sin k\Delta x}{\Delta x} - \frac{i(1 - \cos k\Delta x)}{\Delta x}$$

$$\bar{c} k' e^{ikx} = i \frac{\sin k\Delta x}{\Delta x} e^{ikx} + \frac{1 - \cos k\Delta x}{\Delta x} e^{ikx}$$

$$\frac{\partial \phi}{\partial x} \Big|_{UD} \quad \frac{\partial \phi}{\partial x} \Big|_{CD}$$

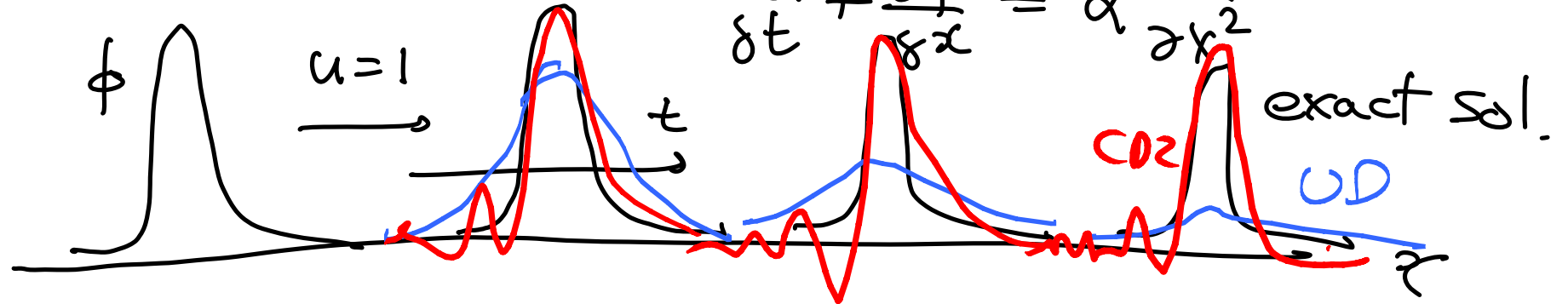
$$-\frac{d^2 \phi}{dx^2} \Big|_{CD}$$

dissipation

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

UD

$$\frac{\partial \phi}{\partial t} + \frac{\partial \phi}{\partial x} = \alpha \frac{\partial^2 \phi}{\partial x^2}$$



7. Examples 1-D convection/diffusion eq.

$$\frac{\partial}{\partial x}(\rho u \phi) = \frac{\partial}{\partial x} \left(\Gamma \frac{\partial \phi}{\partial x} \right) \quad \left(\begin{array}{l} \phi = \phi_0 \text{ @ } x=0 \\ \phi = \phi_L \text{ @ } x=L \end{array} \right)$$

ρ, u, Γ const.

Exact sol.
$$\phi = \phi_0 + \frac{e^{xPe/L} - 1}{e^{Pe} - 1} (\phi_L - \phi_0)$$

$$Pe = \rho u L / \Gamma \quad : \quad \text{Peclet number}$$

Physics : convection is balanced by diffusion

Almost no flows are in this kind of balance.
Convection is balanced by pressure gradient

$$\frac{\partial}{\partial x}(uu) + \frac{\partial}{\partial y}(uv) = \mu \frac{\partial^2 u}{\partial y^2}$$

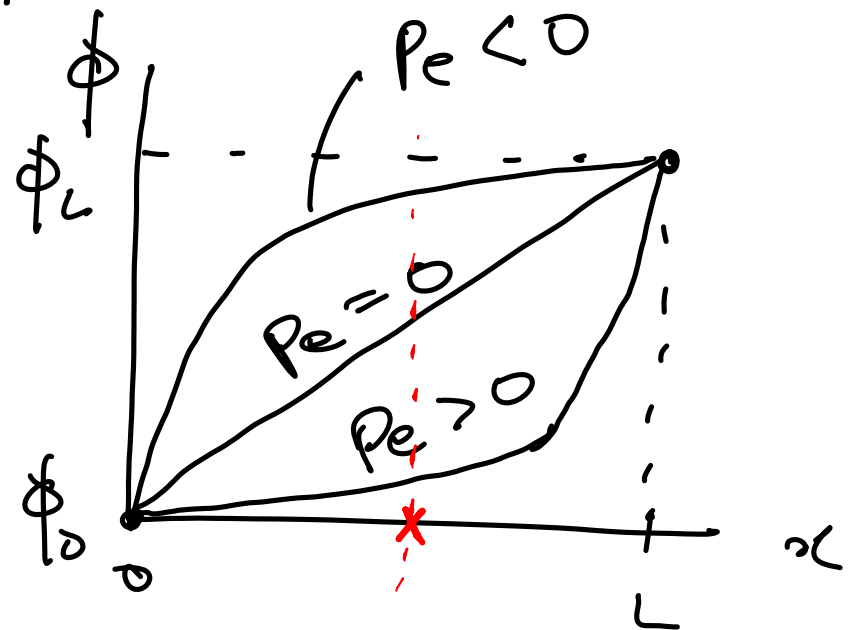
or diffusion in the direction normal to the flow

★ Thus, it is very misleading to conclude anything from this example.

Anyway, let's do it.

Let $u \geq 0$, $\phi_0 < \phi_L$

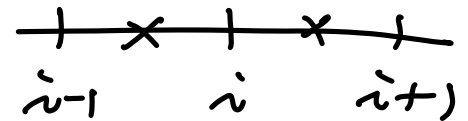
$$(Pe = \rho u L / \mu)$$



$$\frac{\partial}{\partial x} (\rho u \phi) = \frac{\partial}{\partial x} \left(\Gamma \frac{\partial \phi}{\partial x} \right) \quad \left(\begin{array}{l} \phi = \phi_0 \text{ at } x=0 \\ \phi = \phi_L \text{ at } x=L \end{array} \right)$$

Let's do UDS and CDS

Diffusion term - CDS

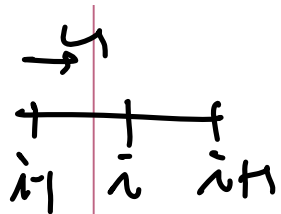


$$\frac{\partial}{\partial x} \left(\Gamma \frac{\partial \phi}{\partial x} \right)_i = \frac{1}{\frac{1}{2}(x_{i+1} - x_{i-1})} \left[\Gamma \frac{\partial \phi}{\partial x} \Big|_{i+\frac{1}{2}} - \Gamma \frac{\partial \phi}{\partial x} \Big|_{i-\frac{1}{2}} \right]$$

$$\Downarrow \qquad \qquad \qquad \Downarrow$$

$$\Gamma \frac{\phi_{i+1} - \phi_i}{x_{i+1} - x_i} \qquad \qquad \qquad \Gamma \frac{\phi_i - \phi_{i-1}}{x_i - x_{i-1}}$$

Convection term $\frac{\partial}{\partial x} (\rho u \phi)$



UDS

$$\frac{\partial}{\partial x} (\rho u \phi) = \begin{cases} \rho u \frac{\phi_i - \phi_{i-1}}{x_i - x_{i-1}} & \text{if } u > 0 \\ \rho u \frac{\phi_{i+1} - \phi_i}{x_{i+1} - x_i} & \text{if } u < 0 \end{cases}$$

CDS

$$\frac{\partial}{\partial x} (\rho u \phi) = \rho u \frac{\phi_{i+1} - \phi_{i-1}}{x_{i+1} - x_{i-1}}$$

$$\Rightarrow \text{construct } A_w \phi_w + A_p \phi_p + A_E \phi_E = Q_p$$

\Rightarrow TDMA \rightarrow obtain ϕ

① $L=1, \rho=1, u=1, \Gamma=0.02, \phi_0=0, \phi_L=1$

$$Pe = \rho u L / \Gamma = 50$$

① $N = 11$ including bdry pts. $\begin{array}{c} 1 \quad 2 \quad \dots \quad 10 \quad 11 \\ | \quad | \quad \dots \quad | \quad | \\ 0 \quad \quad \quad \quad \quad L \end{array}$

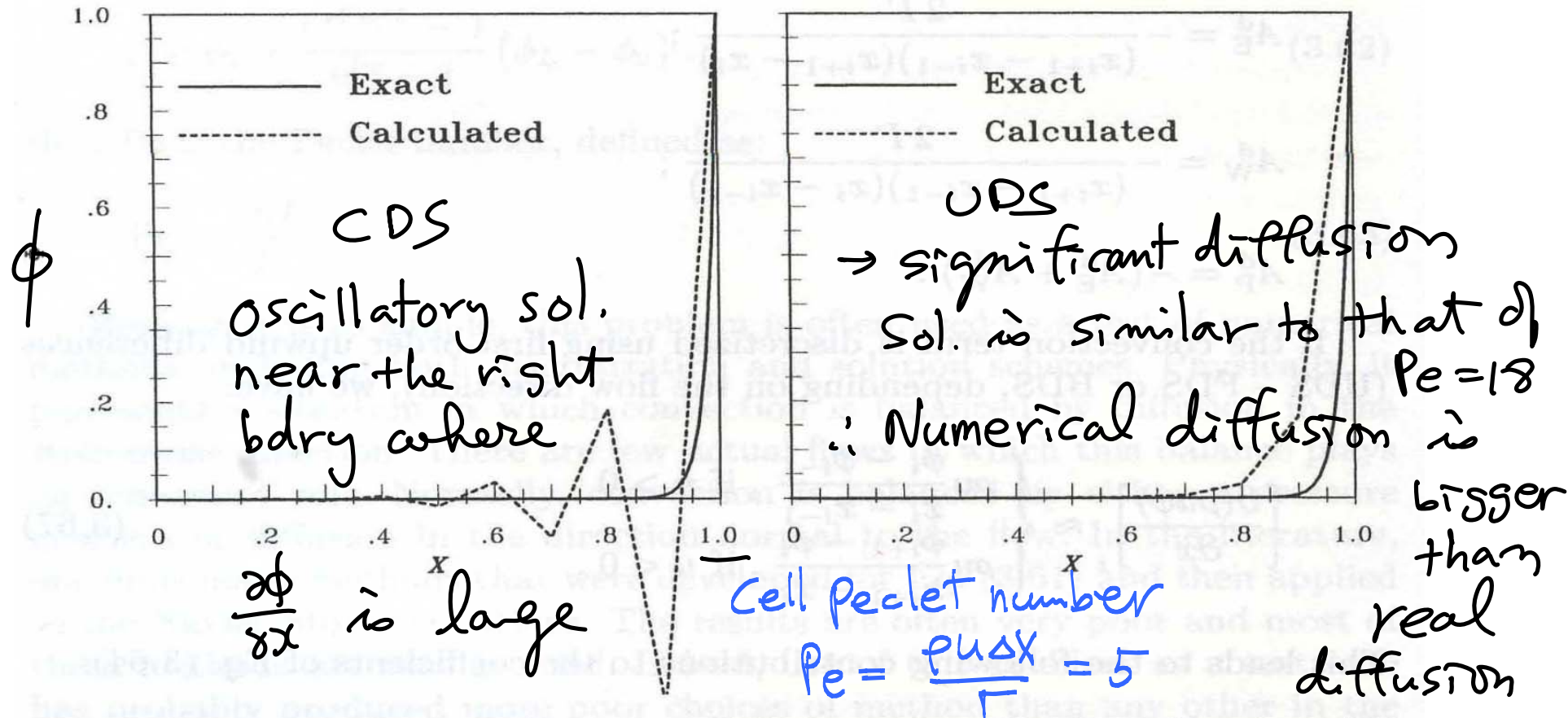


Fig. 3.8. Solution of the 1D convection/diffusion equation at $Pe = 50$ using CDS (left) and UDS (right) for convection terms and a uniform grid with 11 nodes

② $N = 41$

$$\text{cell Pe} = \frac{5}{4} = 1.25$$

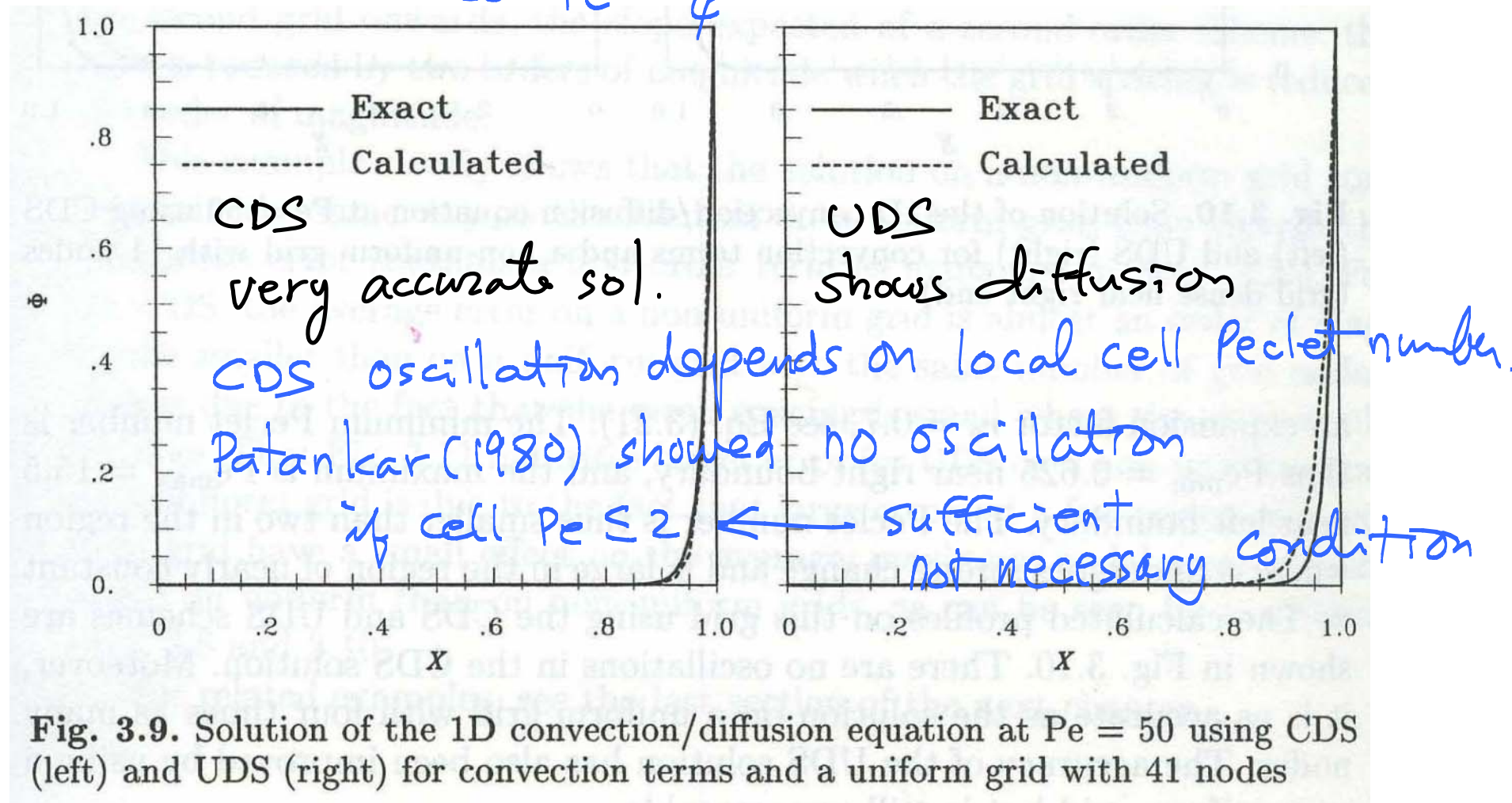


Fig. 3.9. Solution of the 1D convection/diffusion equation at $Pe = 50$ using CDS (left) and UDS (right) for convection terms and a uniform grid with 41 nodes

• Hybrid scheme (Spalding 1972)

Switch from CDS to UDS when $Pe > 2$.

(let diff. coeff = 0)

→ too restrictive
reduces accuracy

Oscillation appears only when sol. changes rapidly
in a region of high cell Pe .

③ Non-uniform grids w/ $N=11$

$$\Delta x_{\max} = 0.31, \quad \Delta x_{\min} = 0.0125; \quad r_e = 0.75$$

$$\frac{\rho u \Delta x}{\Gamma} = Pe_{\max} = 15.5$$

$$Pe_{\min} = 0.625$$

expansion factor

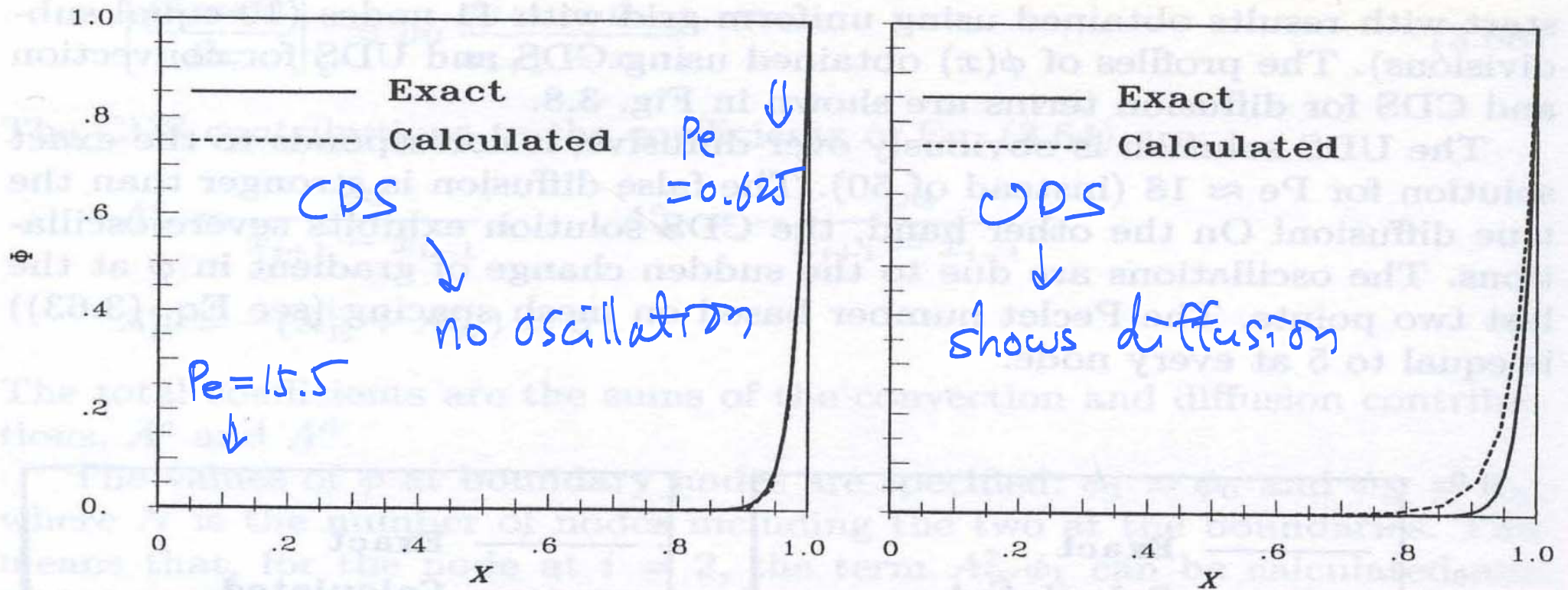
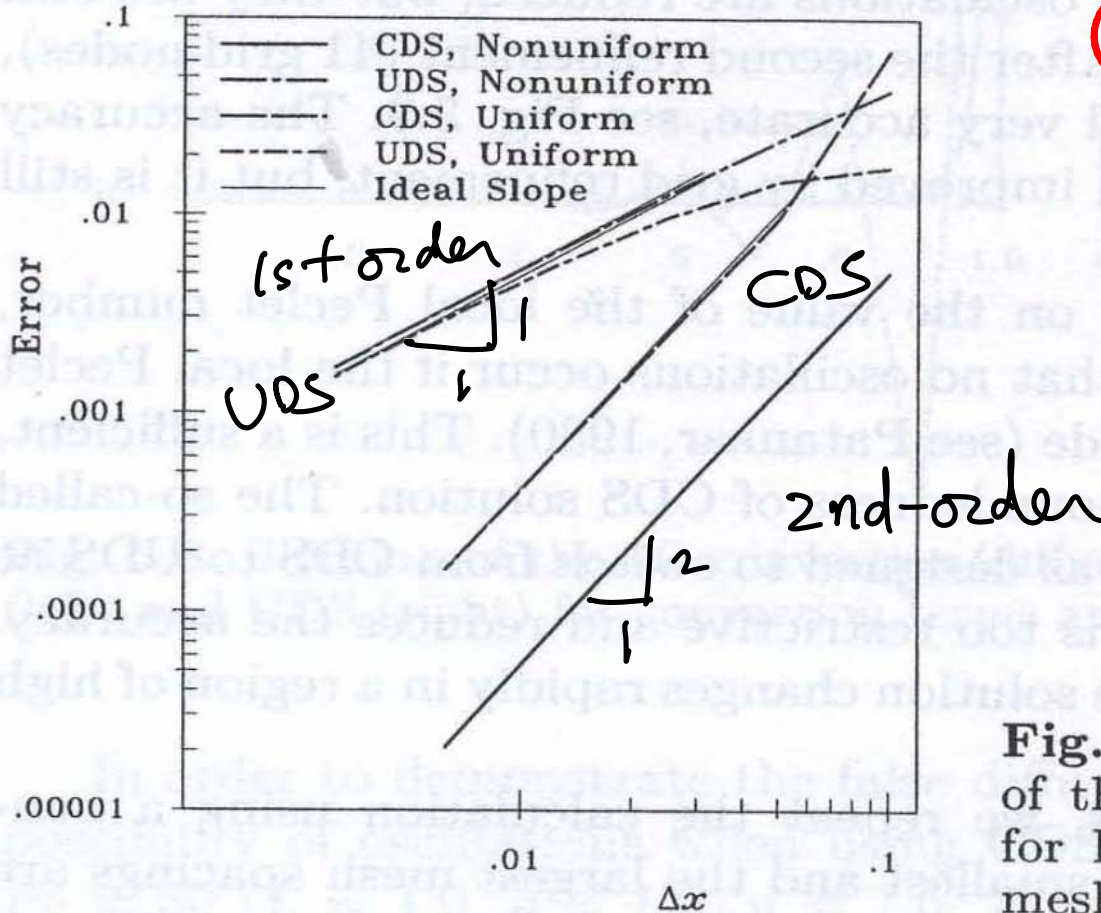


Fig. 3.10. Solution of the 1D convection/diffusion equation at $Pe = 50$ using CDS (left) and UDS (right) for convection terms and a non-uniform grid with 11 nodes (grid dense near right end)

$$L_1 \text{ norm} \quad \epsilon = \frac{1}{N} \sum_{i=1}^M |\phi_i^{\text{exact}} - \phi_i| \quad L_2 \text{ norm} \quad \frac{1}{N} \sum (\)^2$$



HW2

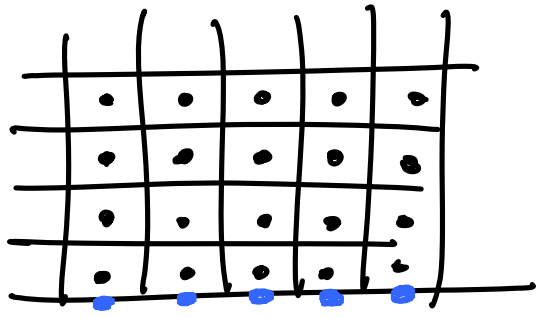
Do it again for
 $\epsilon = \max |\phi_i^{\text{exact}} - \phi_i|$

Fig. 3.11. Average error in the solution of the 1D convection/diffusion equation for $Pe=50$ as a function of the average mesh spacing

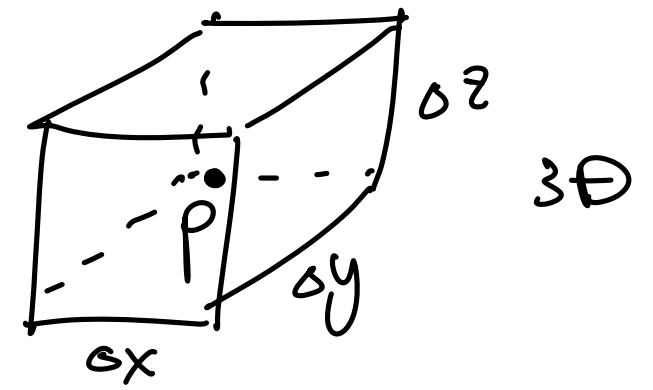
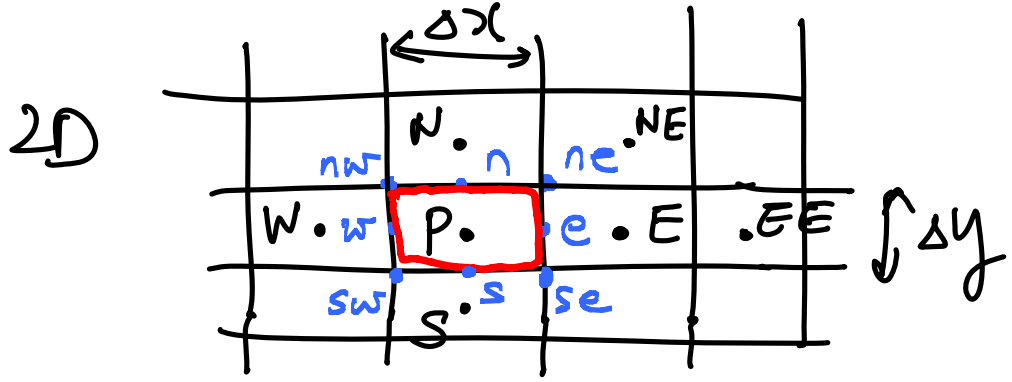
Ch. 4 Finite volume methods

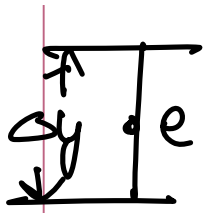
FVM \rightarrow integral form of governing eq.

$$\int_S \rho \phi (\underline{u} \cdot \underline{n}) dS = \int_S \Gamma (\nabla \phi \cdot \underline{n}) dS + \int_{\Omega} q \phi d\Omega$$



- : location for variables
- : control volume

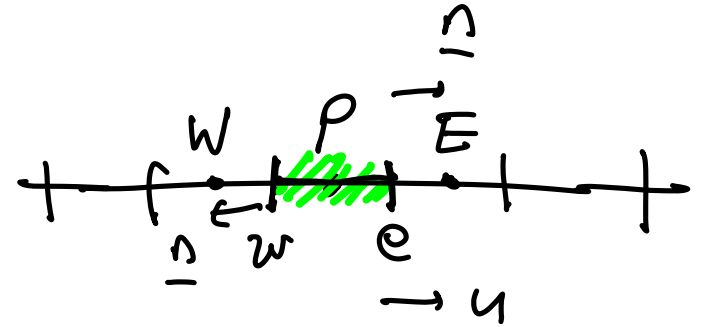




$$\int_{S_e} f dS = f_e S_e = f_e \Delta y \quad \text{how to determine?}$$

$O(\Delta y^2)$ - mid-point rule

$$\int_{\Omega} \phi d\Omega = \phi_p \Delta\Omega$$



• upwind interpolation (UDS)

$$\phi_e = \begin{cases} \phi_p & \text{if } (u \cdot \underline{n})_e > 0 \\ \phi_E & \text{if } \quad \quad \quad < 0 \end{cases}$$

$$\phi_e = \phi_p + (x_e - x_p) \frac{\partial \phi}{\partial x} \Big|_p + \frac{(x_e - x_p)^2}{2} \frac{\partial^2 \phi}{\partial x^2} \Big|_p + \text{HOT}$$

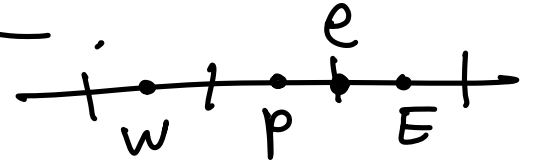
UDS

leading error $\sim \Gamma_e \frac{\partial \phi}{\partial x}$

$$\Rightarrow \Gamma_e^{\text{num}} = (\rho u)_e \frac{\Delta x}{2}$$

numerical diffusion

∴ this scheme never yields oscillatory sols.,
but is numerically diffusive.

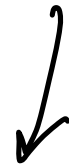


• Linear interpolation (CDS)

" " " between two nearest nodes

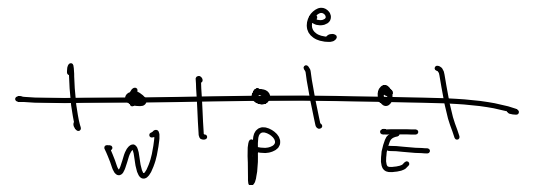
$$\phi_e = \phi_E \lambda_e + \phi_p (1 - \lambda_e) \quad : \text{ 2nd-order}$$

$$\lambda_e = \frac{x_e - x_p}{x_E - x_p}$$



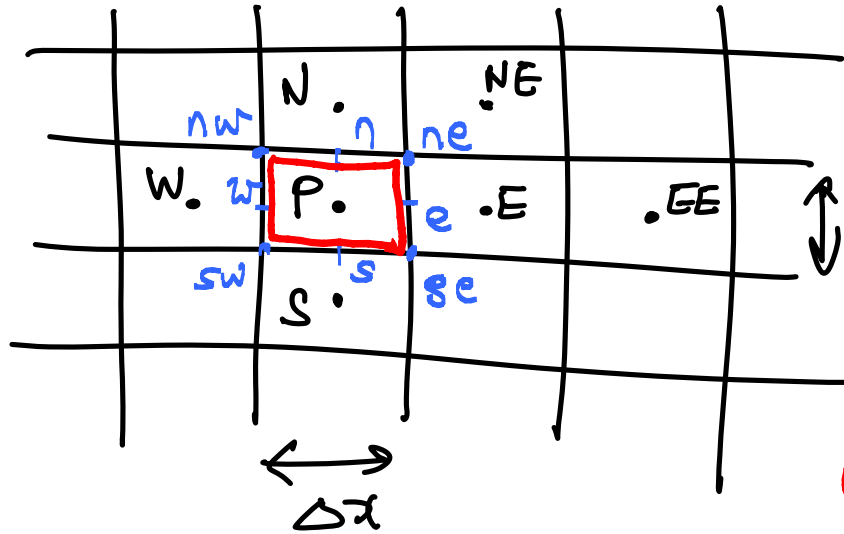
$$\phi_e = \phi_E \lambda_e + \phi_p (1 - \lambda_e) - \frac{(x_e - x_p)(x_E - x_e)}{2} \left. \frac{\partial^2 \phi}{\partial x^2} \right|_p + \text{HOT}$$

viscous term $\rightarrow \frac{\partial \phi}{\partial x}|_e$ CDS



$$\left. \frac{\partial \phi}{\partial x} \right|_e = \frac{\phi_E - \phi_p}{x_E - x_p} + \frac{(x_e - x_p)^2 - (x_E - x_e)^2}{2(x_E - x_p)} \left. \frac{\partial^2 \phi}{\partial x^2} \right|_e + \text{HOT}$$

$$\Delta x_{TH} = r_e \delta x_i \rightarrow \text{second-order}$$



$$\int_S \rho \phi (\underline{u} \cdot \underline{n}) dS$$

$$= \int_S \Gamma \nabla \phi \cdot \underline{n} dS + \int_{\Omega} \partial_t \phi d\Omega$$

$$\int_{S_e} f dS \doteq f_e S_e = f_e \Delta y$$

• QUICK (quadratic upwind interpolation)

Leonard (1979)

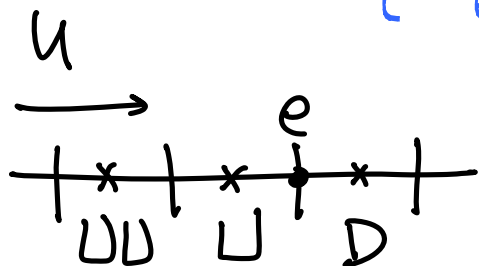
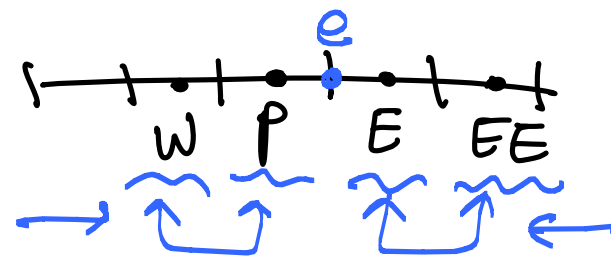
Q(uadratic) U(pwind) I(terpolation) for C(onvective) K(inematics)

→ "parabolic" interpolation between P and E to

evaluate variables at e .

→ need to use data at one more pt.

$\left\{ \begin{array}{l} W \text{ for } u > 0 \\ EE \text{ for } u < 0 \end{array} \right.$

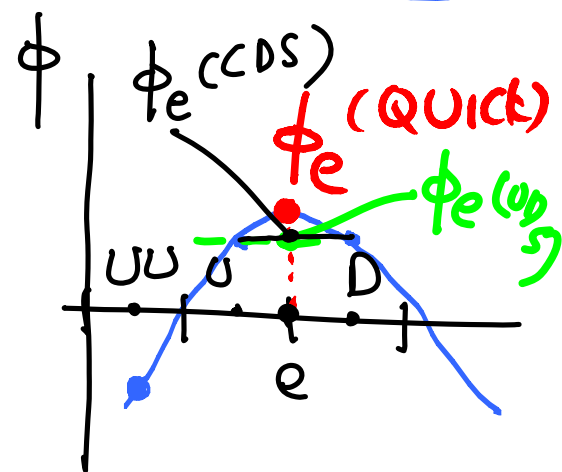


$$\phi_e = \phi_U + g_1 (\phi_D - \phi_U) + g_2 (\phi_U - \phi_{UU})$$

$$g_1 = \frac{(\lambda_e - \lambda_U)(\lambda_e - \lambda_{UU})}{(\lambda_D - \lambda_U)(\lambda_D - \lambda_{UU})}$$

$$g_2 = \frac{(\lambda_e - \lambda_U)(\lambda_D - \lambda_e)}{(\lambda_U - \lambda_{UU})(\lambda_D - \lambda_{UU})}$$

does not guarantee bounded sols.



For uniform grids,

$$\phi_e = \begin{cases} \frac{3}{8} \phi_E + \frac{6}{8} \phi_P - \frac{1}{8} \phi_W & \text{if } u > 0 \\ \frac{3}{8} \phi_P - \frac{1}{8} \phi_{EE} + \frac{6}{8} \phi_E & \text{if } u < 0 \end{cases}$$

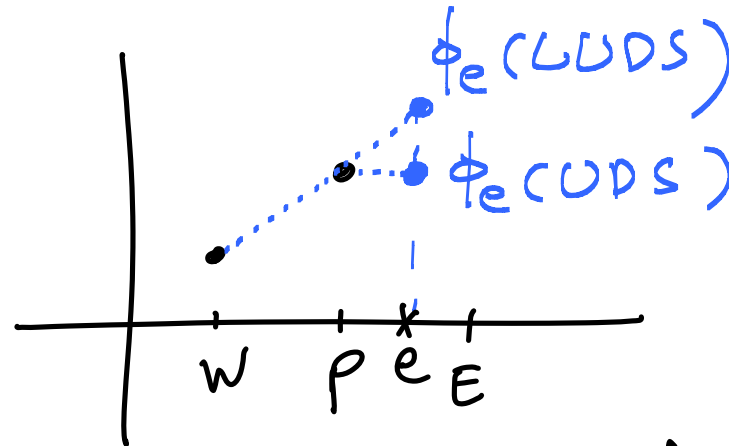
Taylor series exp.

$$\phi_e = \underbrace{\frac{6}{8} \phi_P + \frac{3}{8} \phi_E - \frac{1}{8} \phi_W}_{\text{QUICK}} - \underbrace{\frac{3(\Delta x)^3}{48} \left(\frac{\partial^3 \phi}{\partial x^3} \right)_P}_{\text{3rd-order approx.}} + \text{HOT}$$

$\int \phi dy \approx \underbrace{\phi_e}_{\text{QUICK}} \Delta y \rightarrow$ 2nd-order
 \therefore overall 2nd-order.

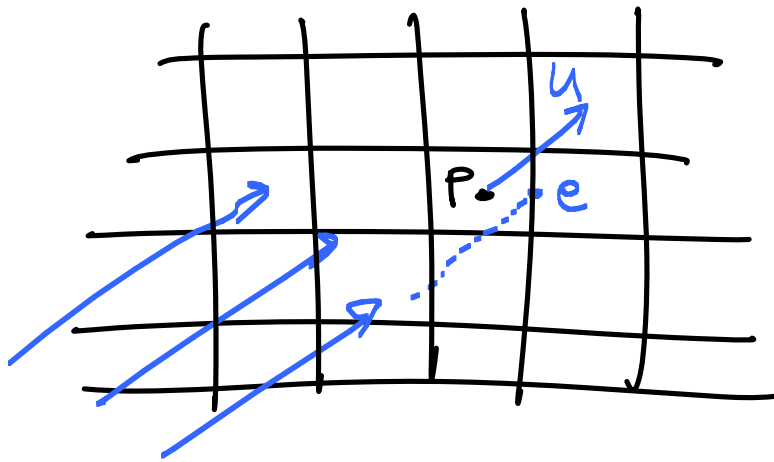
- Linear upwind scheme (LUDS) : 2nd-order


 $\phi_e = \begin{cases} \text{linear interpolation from } \phi_p \text{ and } \phi_w & \text{if } u > 0 \\ \phi_E \text{ and } \phi_{EE} & \text{if } u < 0 \end{cases}$



LUDS can produce unbounded sols.

- Skew upwind scheme (Raithby 1976)
upstream nodes are from streamline rather than from grid line.

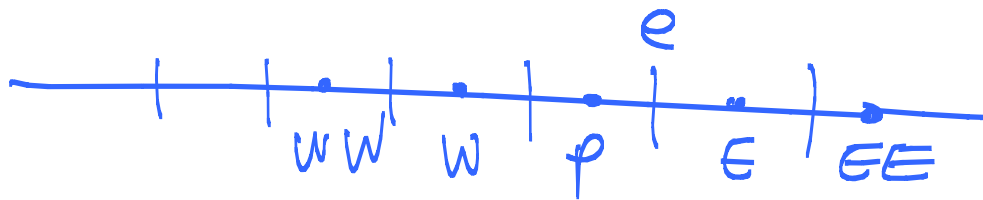


complex
may produce oscillatory sol.

$$\frac{s_e}{6} (f_{ne} + 4f_e + f_{se})$$

• Higher-order schemes

↳ makes sense only if the integral (i.e. $\int f_e dy$) are approximated using higher-order formulae.



uniform mesh

$$\phi_e = \frac{1}{48} (27\phi_p + 27\phi_{EE} - 3\phi_w - 3\phi_{EE}) + O(\Delta x^6)$$

CBS

$$\left. \frac{\partial \phi}{\partial x} \right|_e = \frac{1}{24\Delta x} (27\phi_E - 27\phi_P + \phi_W - \phi_{EE}) + \dots$$

or, cubic splines

or, Padé (compact) scheme

$$\phi_e = \frac{\phi_P + \phi_E}{2} + \frac{\Delta x}{8} \left[\left. \frac{\partial \phi}{\partial x} \right|_P - \left. \frac{\partial \phi}{\partial x} \right|_E \right] + \mathcal{O}(\Delta x^4)$$

$$= \dots + \frac{\phi_E - \phi_W}{2\Delta x} + \frac{\phi_{EE} - \phi_P}{2\Delta x}$$

$$= \dots + \frac{\phi_P + \phi_E - \phi_W - \phi_{EE}}{16} + \mathcal{O}(\Delta x^4)$$

(large stencil size!)

- Deferred correction

higher-order interpolation \rightarrow sparse matrix

$$\phi_e \approx \phi_e^L + (\phi_e^H - \phi_e^L)^{\text{old}}$$

\uparrow
low-order interpolation

\nwarrow
high-order interpolation

usually upwind scheme

L.H.S

R.H.S

$$A\phi = b \quad \longrightarrow \quad (A_1 + (A - A_1))\phi = b$$

\uparrow
 ϕ_e^H sparse

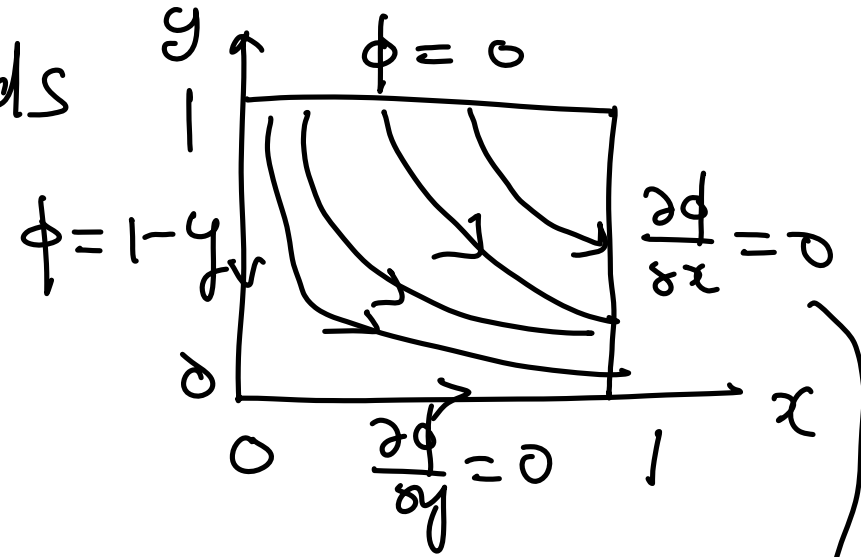
$$A_1\phi^{k+1} = (A_1 - A)\phi^k + b \quad \left[\begin{array}{l} \text{index} \\ k: \text{iteration} \end{array} \right.$$

\uparrow
upwind, CDS

Ex. $\int_S \rho \phi (\underline{u} \cdot \underline{n}) ds = \int_V \Gamma \nabla \phi \cdot \underline{n} ds$

HW3

$\underline{u} = (u_x, u_y)$ $u_x = x$, $u_y = -y$
(stagnation flow)



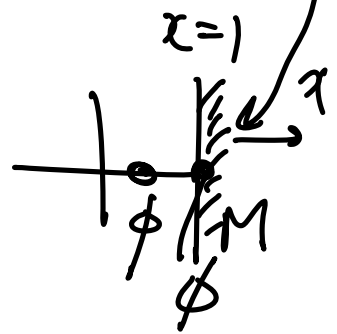
until April 30.

Conv. term: UDS or CDS

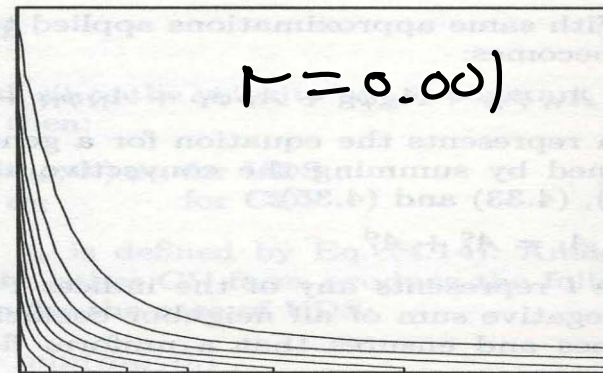
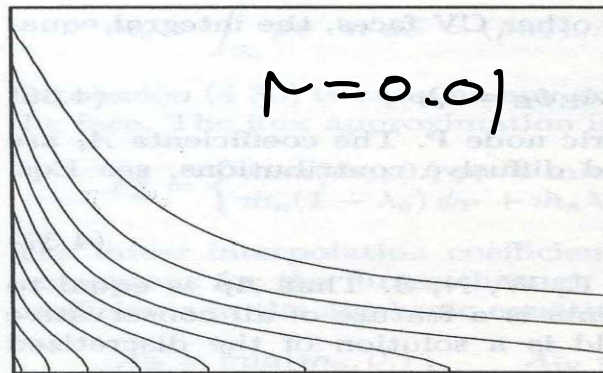
viscous term: CDS

@ $x=1$, $\frac{\partial \phi}{\partial x} = 0$: one-side difference


@ $y=0$, $\frac{\partial \phi}{\partial y} = 0$: // //



40x40 uniform grids, $\rho = 1$, $\Gamma = 0.001$ or 0.01



CDS

 **Fig. 4.5.** Isolines of ϕ , from 0.05 to 0.95 with step 0.1 (top to bottom), for $\Gamma = 0.01$ (left) and $\Gamma = 0.001$ (right)

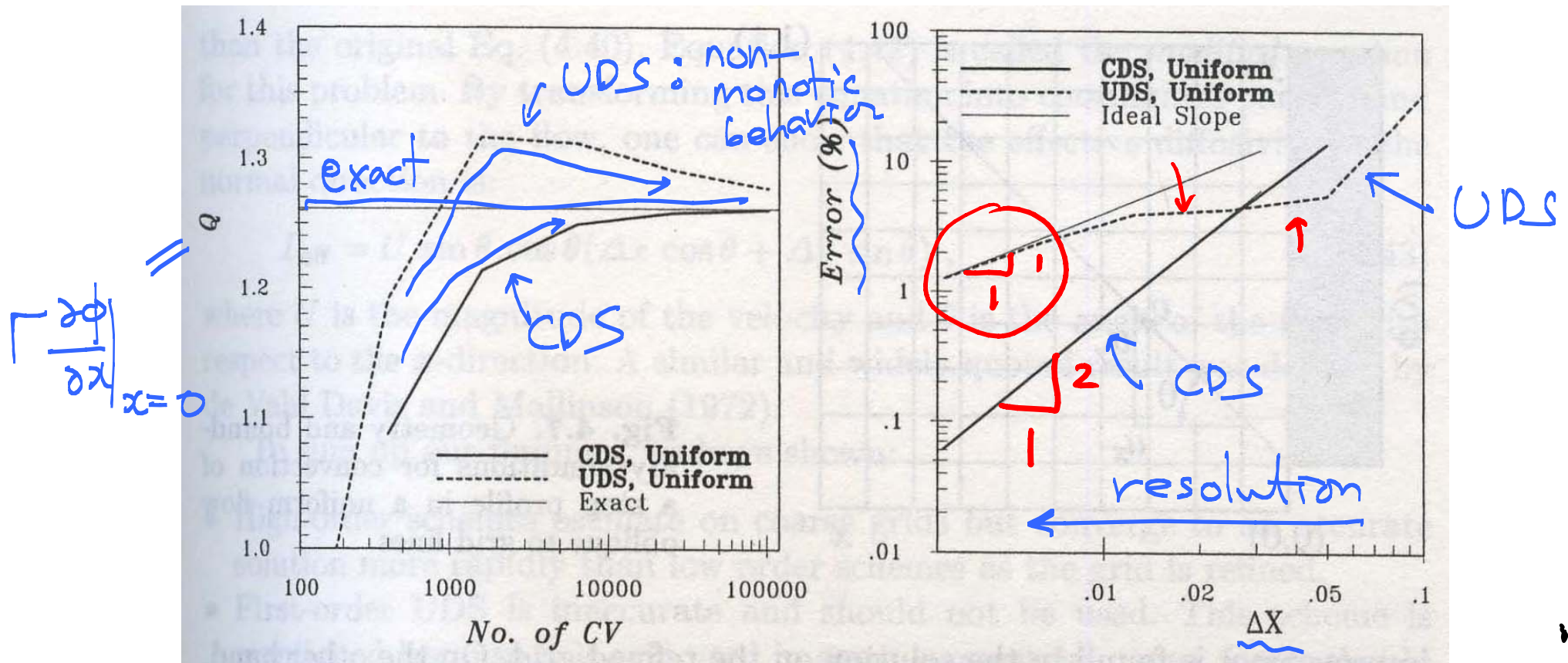
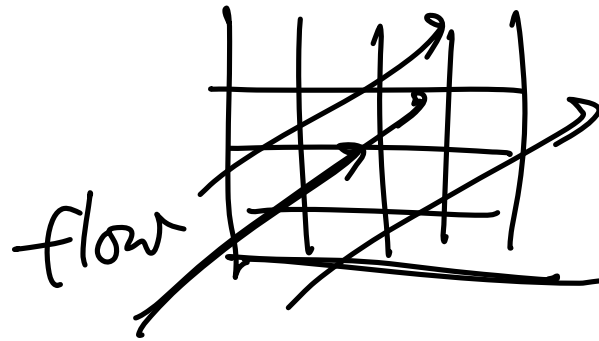


Fig. 4.6. Convergence of total flux of ϕ through the west wall (left) and the error in computed flux as a function of grid spacing, for $\Gamma = 0.001$

Ex. no diffusion

$$u_x \frac{\partial \phi}{\partial x} + u_y \frac{\partial \phi}{\partial y} = 0$$



$$u_x = u_y$$

$$\text{UDS : } u_x \frac{\phi_p - \phi_w}{\Delta x} + u_y \frac{\phi_p - \phi_s}{\Delta y} = 0$$

$$\phi_p = f(\phi_w, \phi_s)$$

↑ known

→ modified PDE

$$u_x \frac{\partial \phi}{\partial x} + u_y \frac{\partial \phi}{\partial y} = \underbrace{u_x \Delta x \frac{\partial^2 \phi}{\partial x^2} + u_y \Delta y \frac{\partial^2 \phi}{\partial y^2}}_{\text{diffusion!}}$$

CDS : oscillatory sol.

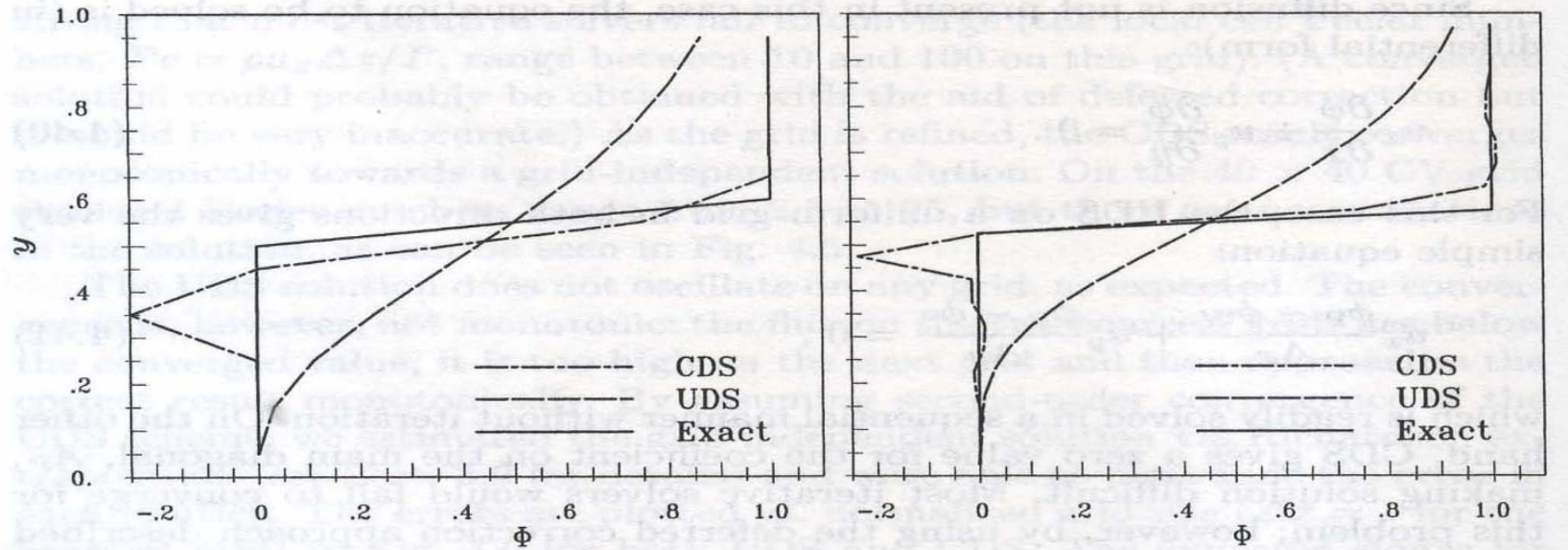


Fig. 4.8. Profile of ϕ at $x = 0.45$, calculated on a 10×10 CV grid (left), and at $x = 0.475$, calculated on a 20×20 CV grid (right)

Summary

1. High-order schemes oscillates on coarse grids but converges to an accurate sol. more rapidly than low order schemes as the grid is refined.
2. First-order UDS is inaccurate and should NOT be used (but still used in commercial codes). High accuracy cannot be obtained on affordable grids w/ this method, especially in 3D.
3. CDS is the simplest scheme of 2nd-order accuracy and offers a good compromise among accuracy, simplicity and efficiency.

< Finite Element Method >

Fletcher I pp. 98-162 참고

Weighted residual method.

Heat eq. $\frac{d^2 T}{dx^2} + T = F$

Residual: $R = \frac{d^2 T}{dx^2} + T - F$

Approx. sol. of $T = \sum_{j=1}^J T_j \phi_j(x)$

approx. ft
 trial ft
 interpolating ft
 shape ft
 basis ft

The coefficients T_j are determined by requiring that the integral of the weighted residual over the computational domain is zero.

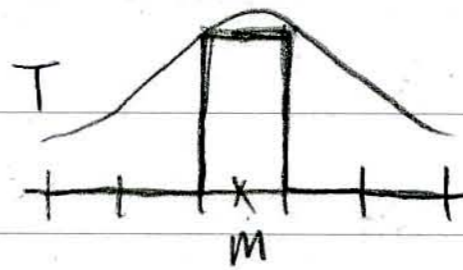
i.e., $\int w_m(x) R dx = 0 \quad (m=1, 2, \dots, J)$

↳ weighting ft

determine w_m : ① Subdomain method.

$w_m = 1$ in subdomain

$= 0$ otherwise



↓

Finite Volume Method

② Collocation method

$w_m(x) = \delta(x - x_m)$

→ $R_m = 0$

↓

Finite Difference Method

③ Least square method

Minimize $\int R^2 dx = Q$

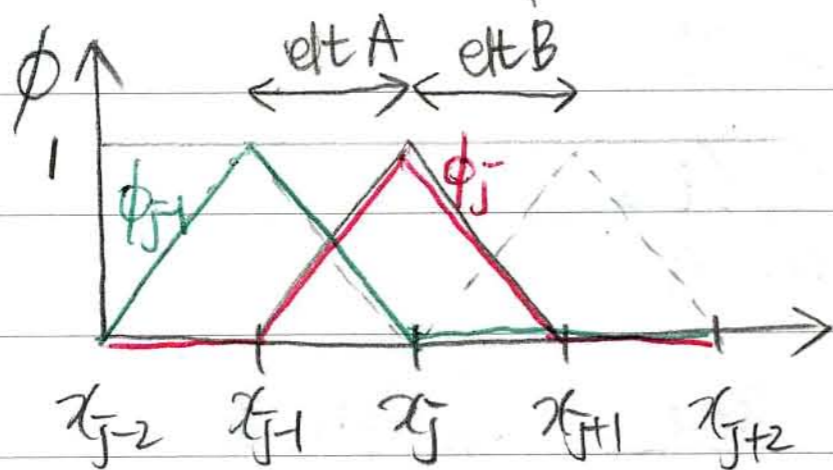
$$\frac{dT_m}{dT_m} = \int \left(\frac{\partial R}{\partial T_m} \right)_{w_m} R dx = 0$$

⊕ Galerkin method

$$w_m = \phi_m \rightarrow \text{Standard FEM}$$

1) Linear interpolation

elt: element



$$\phi_j(x)$$

$$\phi_j = 0 \quad \text{for } x < x_{j-1}$$

$$\phi_j = \frac{x - x_{j-1}}{x_j - x_{j-1}} \quad \text{for } x_{j-1} \leq x \leq x_j$$

$$\phi_j = 1 \quad \text{for } x = x_j$$

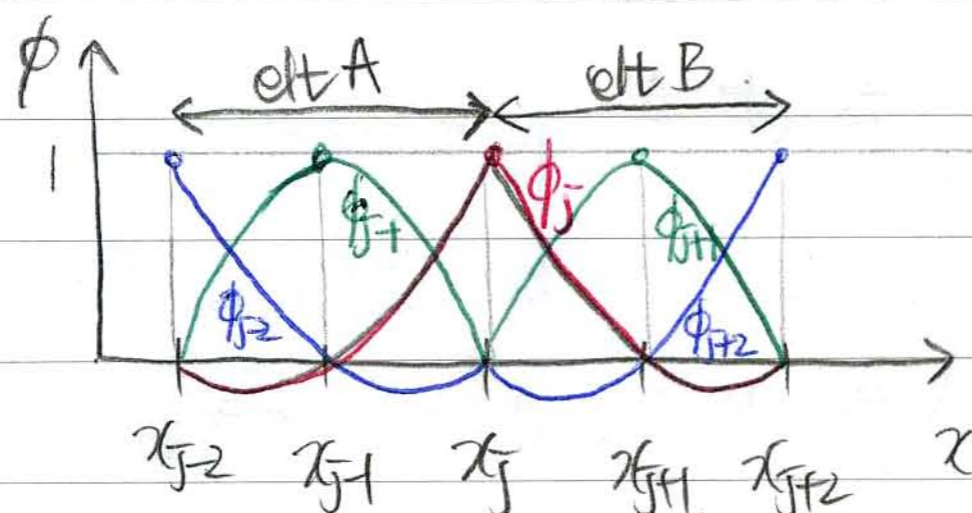
$$\phi_j = \frac{x_{j+1} - x}{x_{j+1} - x_j} \quad \text{for } x_j \leq x \leq x_{j+1}$$

$$\phi_j = 0 \quad \text{for } x > x_{j+1}$$

$$T = T_{j-1} \phi_{j-1} + T_j \phi_j \quad \text{in elt A}$$

$$T = T_j \phi_j + T_{j+1} \phi_{j+1} \quad \text{in elt B}$$

2) Quadratic interpolation

parabola (3pts)
for basis ft

$$\phi_j(x)$$

$$\phi_j = 0 \quad \text{for } x < x_{j-2}$$

$$\phi_j = \frac{x - x_{j-2}}{x_j - x_{j-2}} \cdot \frac{x - x_{j+1}}{x_j - x_{j+1}} \quad \text{for } x_{j-2} \leq x \leq x_j$$

$$\phi_j = \frac{x - x_{j+1}}{x_j - x_{j+1}} \cdot \frac{x - x_{j+2}}{x_j - x_{j+2}} \quad \text{for } x_j \leq x \leq x_{j+2}$$

$$\phi_j = 0 \quad \text{for } x > x_{j+2}$$

FDM이라 하면
 $T = T_j$

$$T = T_{j-2} \phi_{j-2} + T_{j-1} \phi_{j-1} + T_j \phi_j \quad \text{in element A}$$

$$T = T_j \phi_j + T_{j+1} \phi_{j+1} + T_{j+2} \phi_{j+2} \quad \text{in element B}$$

$$\phi_{j-2} = \frac{x - x_{j-1}}{x_{j-2} - x_{j-1}} \cdot \frac{x - x_j}{x_{j-2} - x_j}$$

$$\phi_{j-1} = \frac{x - x_{j-2}}{x_{j-1} - x_{j-2}} \cdot \frac{x - x_j}{x_{j-1} - x_j}$$

$$\phi_j = \frac{x - x_{j-2}}{x_j - x_{j-2}} \cdot \frac{x - x_{j-1}}{x_j - x_{j-1}} \quad \text{in element A.}$$

Roughly,

Linear interpolation \rightarrow 2nd order FDM

quadratic " \rightarrow 3rd order "

* 2D prob. (bilinear interpolation)

$$\phi_1 = 0.25(1-x)(1-y) \quad \phi_3 = 0.25(1+x)(1+y)$$

$$\phi_2 = 0.25(1+x)(1-y) \quad \phi_4 = 0.25(1-x)(1+y)$$

Example

$$\frac{d^2 T}{dx^2} + T = F \quad \text{use Galerkin FEM, } T(0)=0, \frac{dT}{dx}(1)=0$$

cf. FDM라 하면

$$\left[\frac{T_{j+1} - 2T_j + T_{j-1}}{\Delta x^2} + T_j = F_j \right]$$

$$T = \sum_{j=1}^I T_j \phi_j$$

(linear approx.)

in element A: $\phi_j(\xi) = 0.5(1+\xi)$

$$\xi = 2\left(x - \frac{x_{j-1} + x_j}{2}\right) / \Delta x_j$$

in element B: $\phi_j(\xi) = 0.5(1-\xi)$

$$\xi = 2\left(x - \frac{x_j + x_{j+1}}{2}\right) / \Delta x_j$$

$$R = \frac{d^2 T}{dx^2} + T - F$$

$$\int w_m R dx = \int \phi_m R dx = 0$$

$$\rightarrow \int_0^1 \phi_m(x) \left[\frac{d^2 T}{dx^2} + T - F \right] dx = 0$$

$$\int_0^1 \phi_m \frac{d^2 T}{dx^2} dx = \left[\phi_m \frac{dT}{dx} \right]_0^1 - \int_0^1 \frac{d\phi_m}{dx} \frac{dT}{dx} dx$$

0 from b.c.'s

$$T = \sum T_j \phi_j$$

$$\rightarrow \sum_{j=1}^J \left[\int_0^1 \left(- \frac{d\phi_m}{dx} \frac{d\phi_j}{dx} + \phi_m \phi_j \right) dx \right] T_j = \int_0^1 \phi_m F dx$$

(m=1, 2, ..., J) g_m

b_{mj} : analytically obtained.

\rightarrow $\textcircled{BT} = G$

TDMA

$$\left. \begin{aligned} b_{j,j-1} &= \frac{1}{\Delta x_j} + \frac{\Delta x_j}{6} \\ b_{j,j} &= - \left(\frac{1}{\Delta x_j} + \frac{1}{\Delta x_{j+1}} \right) + \frac{\Delta x_j + \Delta x_{j+1}}{3} \\ b_{j,j+1} &= \frac{1}{\Delta x_{j+1}} + \frac{\Delta x_{j+1}}{6} \\ b_{j,j+1} &= \frac{1}{\Delta x_j} + \frac{\Delta x_j}{6} \\ b_{j,j} &= - \frac{1}{\Delta x_j} + \frac{\Delta x_j}{3} \\ b_{j,j+1} &= 0 \end{aligned} \right\} \text{for } j=1, \dots, J-1$$

$$g_m = \sum_{j=1}^J F_j \int_0^1 \phi_m \phi_j dx \quad \text{for } m=1, \dots, J-1$$

analytically obtained

$$= \frac{\Delta x_m}{6} F_{m-1} + \frac{\Delta x_m + \Delta x_{m+1}}{3} F_m + \frac{\Delta x_{m+1}}{6} F_{m+1}$$

$$g_J = \frac{\Delta x_J}{6} F_{J-1} + \frac{\Delta x_J}{3} F_J$$

for uniform grids,

$$\frac{T_{j+1} - 2T_j + T_{j-1}}{\Delta x^2} + \frac{T_{j-1} + 4T_j + T_{j+1}}{6} = \frac{F_{j-1} + 4F_j + F_{j+1}}{6}$$

\uparrow 2nd-order accuracy (CD2) \uparrow Simpson's rule \rightarrow 4th-order accuracy (FDM: T_j, F_j)

Ch. 7 Solution of the Navier-Stokes Eqs.

1. Conservation properties

- mass conservation
- momentum " : convection term, viscous term, pressure gradient term
- energy "

(conservative approx. — finite volume method
non- " " " kinetic

Biggest problem is how to conserve ^{kinetic} energy.

incompressible flow \rightarrow kinetic energy

u, v, w, p (4) cont. & N-S eqs (4)
 compressible flow \rightarrow (kinetic energy
 thermal "

u, v, w, p, T, ρ (6) \rightarrow (4) + energy eq (1)
 + state eq (1)

$$\bullet \frac{\partial}{\partial t}(\rho u_i) + \frac{\partial}{\partial x_j}(\rho u_i u_j) = -\frac{\partial p}{\partial x_i} + \frac{\partial}{\partial x_j}(\mu \frac{\partial u_i}{\partial x_j}) - \frac{2}{3} \frac{\partial}{\partial x_j}(\mu \frac{\partial u_k}{\partial x_k}) \delta_{ij} + \rho b_i$$

u_i (*) (")

$$\rightarrow \frac{\partial}{\partial t} \left(\frac{1}{2} \rho u_i u_i \right) + u_i \frac{\partial}{\partial x_j} (\rho u_i u_j) = -u_i \frac{\partial p}{\partial x_i} + u_i \frac{\partial}{\partial x_j} \left(\mu \frac{\partial u_i}{\partial x_j} \right) - \frac{2}{3} u_i \frac{\partial}{\partial x_j} \left(\mu \frac{\partial u_k}{\partial x_k} \right) \delta_{ij} + \rho u_i b_i$$

$$(u^2 = u_i u_i)$$

$$\rightarrow \frac{\partial}{\partial t} \left(\frac{1}{2} \rho u^2 \right) + \frac{\partial}{\partial x_j} \left(\frac{1}{2} \rho u^2 u_j \right) = - \frac{\partial}{\partial x_i} (\rho u_i) + \rho \frac{\partial u_i}{\partial x_i}$$

$$+ \frac{\partial}{\partial x_j} \left(u_i \mu \frac{\partial u_i}{\partial x_j} \right) - \frac{\partial u_i}{\partial x_j} \cdot \mu \frac{\partial u_i}{\partial x_j}$$

$$- \frac{2}{3} \frac{\partial}{\partial x_j} \left(u_i \mu \frac{\partial u_k}{\partial x_k} \right) \delta_{ij} + \frac{2}{3} \frac{\partial u_i}{\partial x_j} \mu \frac{\partial u_k}{\partial x_k} \delta_{ij} + \rho u_i b_i$$

$$\int_{\Omega} \odot dV$$

$$\frac{\partial}{\partial t} \int_{\Omega} \frac{1}{2} \rho u^2 dV = - \int_S \frac{1}{2} \rho u^2 u_j n_j dA - \int_S \rho u_i n_i dA$$

$$+ \int_S u_i \left(\mu \frac{\partial u_i}{\partial x_j} - \frac{2}{3} \mu \frac{\partial u_k}{\partial x_k} \delta_{ij} \right) n_j dA$$



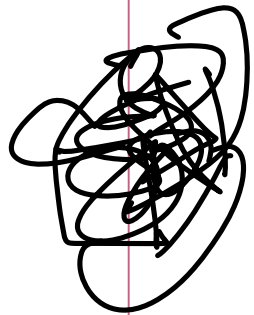
$$+ \int_{\Omega} \left(p \frac{\partial u_i}{\partial x_i} - \frac{\partial u_i}{\partial x_j} \left(\mu \frac{\partial u_i}{\partial x_j} - \frac{2}{3} \mu \frac{\partial u_k}{\partial x_k} \delta_{ij} \right) + \rho u_i b_i \right) dV$$

Discussion

- ① First 3 terms on RHS: Kinetic energy in Ω is NOT changed by the action of convection and pressure within the control volume.

If no viscosity
 no internal energy → kinetic energy is then
 no mfm forcing → globally conserved

↓
 the property that we like to preserve
 from a numerical method.



② We solve mtn eq. (NOT energy eq.)
Conservative scheme for mtn eq. does not
guarantees conservation of energy.

→ difficult to conserve energy numerically.

③ If a numerical method is energy conservative,
total kinetic energy does not grow in time.

→ vel at every grid pt. in the domain must remain
bounded.

→ guarantees numerical stability (NOT accuracy)

→ kinetic energy conservation is important

in computing unsteady flow.

④ pressure gradient term

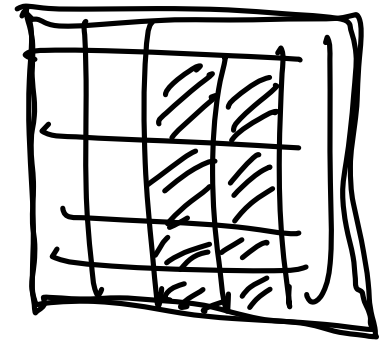
$$u_i \frac{\partial p}{\partial x_i} = \frac{\partial}{\partial x_i} (p u_i) - p \frac{\partial u_i}{\partial x_i}$$

If incomp. flow, $\partial u_i / \partial x_i = 0$

$$\int_{\Omega} u_i \frac{\partial p}{\partial x_i} = \int_{\Omega} \frac{\partial}{\partial x_i} (p u_i) = \int_S p u_i n_i dA$$

⇒ pressure influences the overall kinetic energy budget only by its action at the surface.

→ We have to retain this property.



$G_i p$: numerical approx. of the press. grad.

$\sum u_i \cdot u_i - \text{mtm} \rightarrow \sum_{i=1}^N u_i \overbrace{G_i p}^{\text{used in mtm eq. num. approx. of } \partial u_i / \partial x_i} \Delta \Omega$

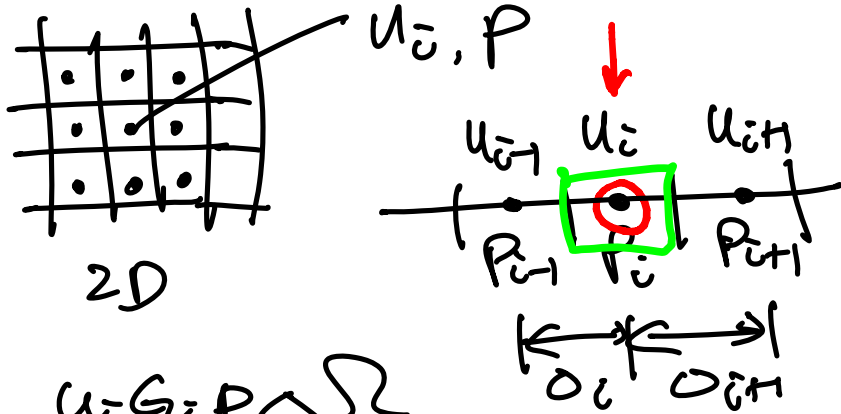
$\left(u_i \frac{\partial p}{\partial x_i} = \frac{\partial}{\partial x_i} (u_i p) - p \frac{\partial u_i}{\partial x_i} \right) \overset{\text{=}}{\parallel} \sum_S p u_n \Delta S - \sum_N p \underbrace{D_i u_i}_{\text{used in the cont. eq.}} \Delta \Omega$

Equality is ensured only if G_i and D_i are compatible.

i.e. $\sum_{i=1}^N (u_i G_i p + p D_i u_i) \Delta \Omega \overset{\text{=}}{\parallel} \text{surface terms.}$

- ex) G_i : backward
- D_i : forward

collocated mesh



$$u_i G_i P \Delta \Omega$$

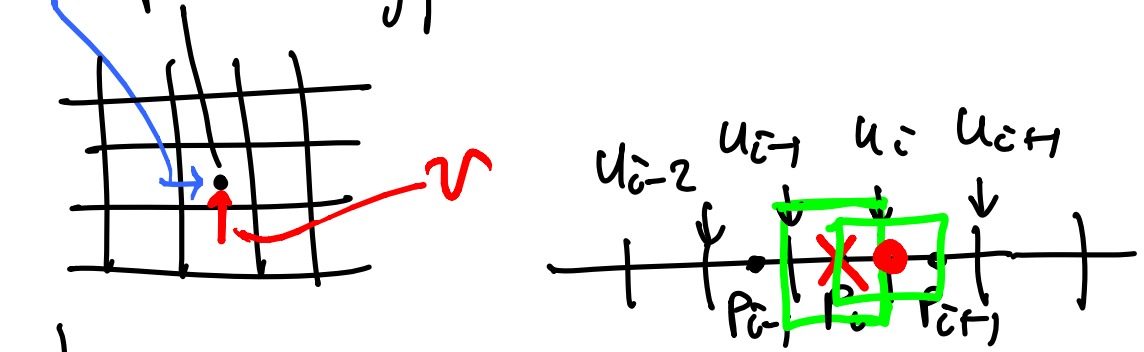
$$= u_i \frac{P_i - P_{i-1}}{\Delta_i} \Delta_i$$

$$P D_i u_i \Delta \Omega$$

$$= P_i \frac{u_{i+1} - u_i}{\Delta_{i+1}} \cdot \Delta_{i+1}$$

$$= -u_i P_{i-1} + u_{i+1} P_i$$

P staggered mesh



Do the same thing

$$u_i G_i P \Delta \Omega = \text{circle}$$

$$P D_i u_i \Delta \Omega = \text{circle}$$

$$\sum_{i=1}^N (-u_i P_{i-1} + u_{i+1} P_i) = -u_1 P_0 + u_{N+1} P_N$$

OK.

⊙ $G_{ip} : CD2$, $D_i u_i : CD2$ ok only for staggered mesh
 not ok for collocated mesh

The requirement that only bdry terms remain when the sum over all c.v.'s is taken is not easily satisfied for other two terms.
 conv. & diffusion

⑤ Pressure \rightarrow Poisson eq (incomp. flow)

$$\underbrace{\frac{\nabla \cdot \nabla p}{\text{Di} \nabla^2 p}} = \frac{q}{\sigma} \quad ; \quad \text{numerical operators should be consistent if mass conserv. is to obtain.}$$

$$\left\{ \begin{array}{l} \frac{\partial u_i}{\partial t} = \dots - \frac{\partial p}{\partial x_i} + \dots \\ \nabla^2 p = q \end{array} \right.$$

⑥ Incomp. flow w/o body force

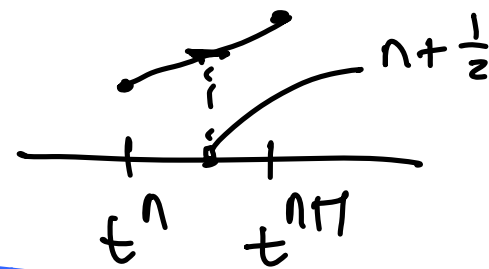
$$-\int_{\Omega} \mu \frac{\partial u_i}{\partial x_j} \frac{\partial u_i}{\partial x_j} dt < 0 \quad \text{dissipation} \rightarrow \text{goes to thermal energy}$$

⑦ $\frac{\partial}{\partial t}(\rho u_i) \rightarrow \rho \frac{u_i^{n+1} - u_i^n}{\Delta t} \Delta \Omega$ n : time step index

If Crank-Nicolson method is used, $u_i^{n+\frac{1}{2}} = \frac{1}{2}(u_i^{n+1} + u_i^n)$

$$\frac{\partial u}{\partial t} = f(u)$$

CN \Rightarrow Trapezoidal method $\frac{u^{n+1} - u^n}{\Delta t} = \frac{1}{2} [f(u^{n+1}) + f(u^n)]$



u_i $\frac{\partial}{\partial t}(\rho u_i) \Rightarrow u_i^{n+\frac{1}{2}} \cdot \rho \frac{u_i^{n+1} - u_i^n}{\Delta t} \cdot \Delta \Omega$

$u_i^{n+\frac{1}{2}}$

$$= \frac{\rho \Delta \Omega}{\Delta t} \left[\frac{1}{2} u_i^{n+1}{}^2 - \frac{1}{2} u_i^n{}^2 \right]$$

\therefore CN is energy conservative

forward Euler

$$u_i^n \cdot \rho \frac{u_i^{n+1} - u_i^n}{\Delta t} \cdot \Delta \Omega$$

\downarrow
not energy conservative

⇒ time difference scheme can destroy the energy conservation property.

FE $\frac{\partial u}{\partial t} = f$

$\frac{u^{n+1} - u^n}{\Delta t} = f^n \cdot \Delta t$

BE $(u^n, f^{n+1}) \cdot \Delta t$

• Alternative method:

to use a different form of mtr eq.

$$H_i = \frac{\partial}{\partial x_j} (u_i u_j) = u_j \left(\frac{\partial u_i}{\partial x_j} - \frac{\partial u_j}{\partial x_i} \right) + \frac{\partial}{\partial x_j} \left(\frac{1}{2} u_j^2 \right) = u_j \frac{\partial u_i}{\partial x_j}$$

conv-term divergence form rotational form convective form

$$= \frac{1}{2} \left(\frac{\partial u_i u_j}{\partial x_j} + u_j \frac{\partial u_i}{\partial x_j} \right)$$

skew-symmetric form

Kravchenko, ...

JCP (19**
20**)

rotational form

$$= \epsilon_{ijk} u_j \omega_k + \frac{\partial}{\partial x_i} \left(\frac{1}{2} u_j u_j \right)$$

↑ vorticity ($\underline{\omega} = \nabla \times \underline{u}$)

$$\left(\begin{array}{l} \epsilon_{111} = 0 \\ \epsilon_{123} = 1 \\ \epsilon_{132} = -1 \end{array} \right)$$

$$\Rightarrow \frac{\partial u_i}{\partial t} + \boxed{\epsilon_{ijk} u_j \omega_k} = -\frac{\partial}{\partial x_i} \left(\frac{p}{\rho} + \frac{1}{2} u_j u_j \right) + \nu \frac{\partial}{\partial x_j} \frac{\partial u_i}{\partial x_j}$$

$u_i \cdot$ (



//

sym. tensor

$$u_i \epsilon_{ijk} u_j \omega_k = \epsilon_{ijk} u_i u_j \omega_k = 0$$

anti-sym. tensor

thus, nonlinear term has no effect.
However, this eq. is nonconservative form
for momentum.

- Kinetic energy conservation
 - turbulence
 - weather prediction

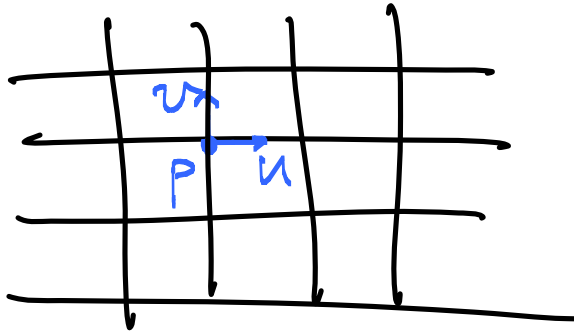
angular m_{tm} conservation - turbomachinery

- CDS on staggered mesh → m_{tm} energy conservation
- CDS is much better than UDS.

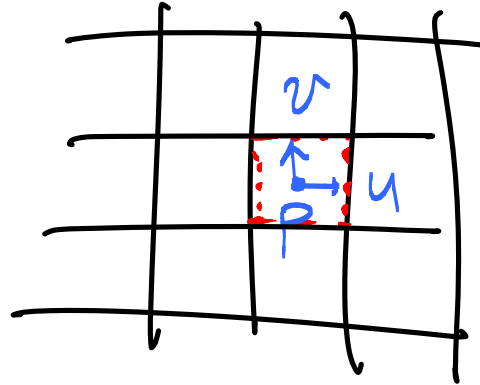
2. Choice of grid system

- ① Collocated (non-staggered) mesh

FDM

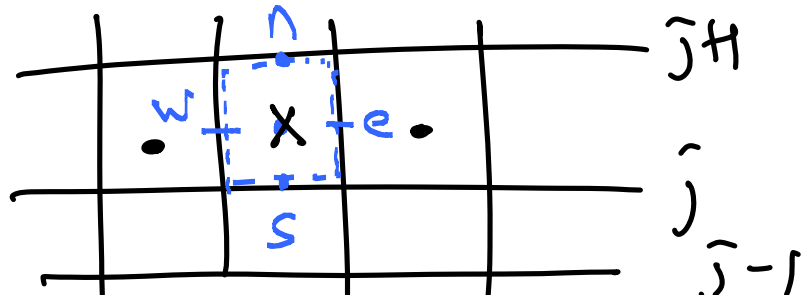


FVM



out of favor for incomp. flow
 due to difficulties with pressure-velocity decoupling
 and occurrence of oscillations in p.

FVM - collocated mesh



Continuity $\frac{\partial u}{\partial x} + \frac{\partial v}{\partial y} = 0$

$$\rightarrow \frac{u_e - u_w}{\Delta x} + \frac{v_n - v_s}{\Delta y} = 0$$

$i-1 \quad i \quad i+1$

→ requires interpolation

$$\left. \begin{aligned} u_e &= \frac{1}{2} (u_{i+1/2} + u_{i/2}) \\ u_w &= \frac{1}{2} (u_{i/2} + u_{i-1/2}) \end{aligned} \right\} \rightarrow \begin{aligned} u_e - u_w &= \frac{1}{2} (u_{i+1/2} - u_{i-1/2}) \end{aligned}$$

likewise $v_n - v_s = \frac{1}{2} (v_{i,j+1} - v_{i,j-1})$

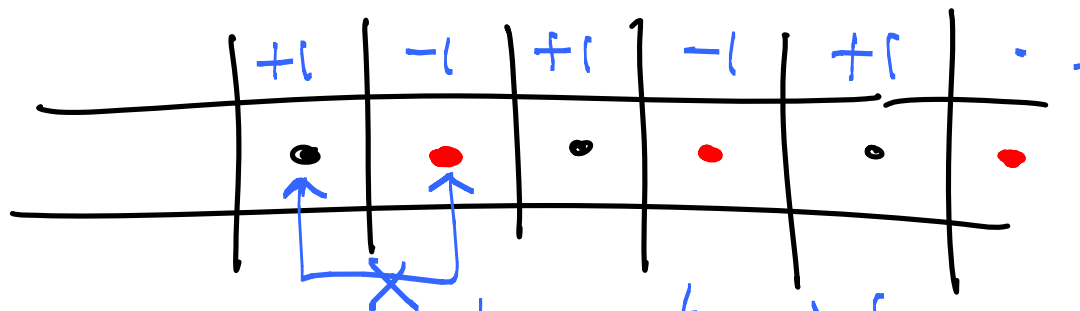
no u_{ij}
& v_{ij}

x-mtm eq:

$$\frac{\partial p}{\partial x} = \frac{P_{i+1/2} - P_{i-1/2}}{2\Delta x} \rightarrow \text{no } P_{ij}$$

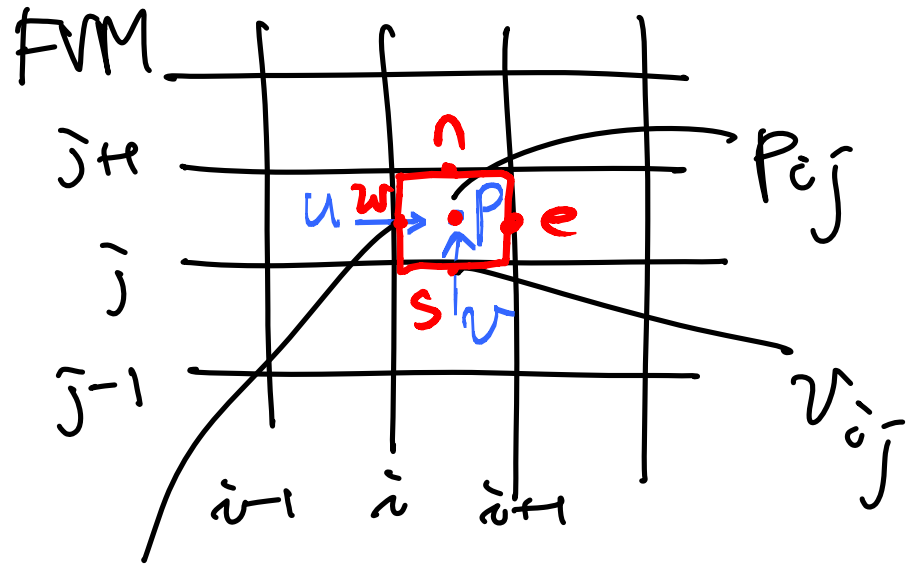
↓

$$\left(\frac{\partial u}{\partial t} \right) = \frac{u_{i+1/2} - u_{i-1/2}}{\Delta t}$$



decoupling bet. p and u .

② Staggered mesh ← invented by Harlow & Welch (1965)
 Phys. Fluids (2176 citations)



Continuity $\frac{\partial u}{\partial x} + \frac{\partial v}{\partial y} = 0$

□ c.v for P.

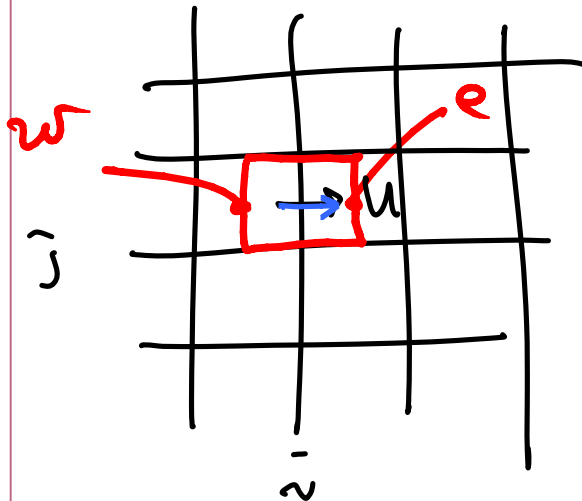
$$\frac{u_e - u_w}{\Delta x} + \frac{v_n - v_s}{\Delta y} = 0$$

Diagram showing the derivation of the continuity equation for a control volume around node P. The x-velocity difference is $\frac{u_e - u_w}{\Delta x}$ and the y-velocity difference is $\frac{v_n - v_s}{\Delta y}$. Arrows point from the velocity terms to their respective grid locations: u_e and u_w are at $(i+1, j)$ and $(i-1, j)$ respectively; v_n and v_s are at $(i, j+1)$ and $(i, j-1)$ respectively.

u_{ij}

no requirement
 of interpolation!
 compact!

x-mtm eq : $\frac{\partial p}{\partial x}$

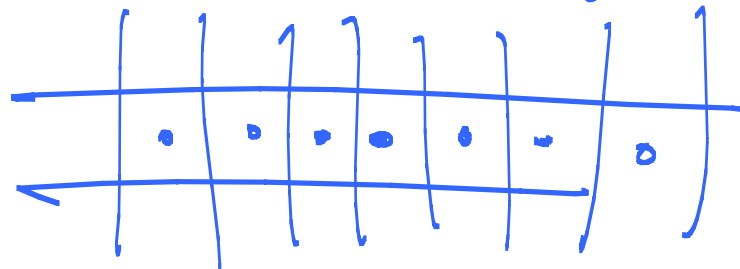


\square : C.V for x-mtm eq.

$$\frac{\partial p}{\partial x} = \frac{P_e - P_w}{\Delta x}$$

$P_{i,j}$ $P_{i+1,j}$

compact
no interpolation
required!



strong coupling
bet. u and p .

- no oscillations in p
- staggered mesh is dominantly used.

However, for complex geometry, one has to transform N-S in "generalized" coordinates with dependent variables of contravariant vel. to satisfy $\nabla \cdot \underline{u} = 0$

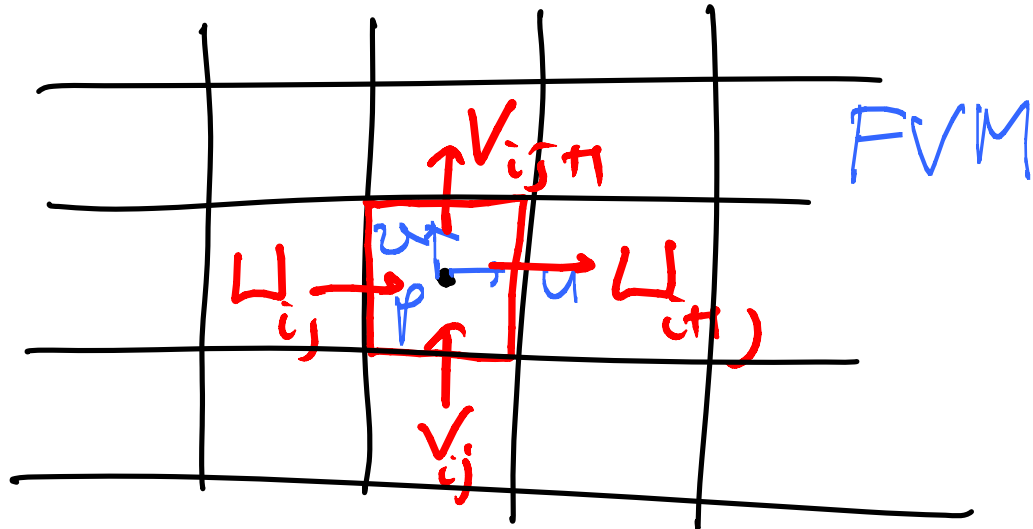
→ very complicated!

→ back to collocated mesh with improved P-U coupling algorithm since 1980's

* momentum interpolation method → getting more popular
invented by Rhie & Chow (AIAA J.)
(1354 citations) Vol. 21, 1525 (1983)

unstructured mesh ← Kim & Choi

(JCP, 162, 411) (2000)



$$\frac{\partial}{\partial x_j} (u u_j)$$

U or V

$$\left(\frac{\partial p}{\partial x} \right) \sim \left(U \right)$$

eliminate the u-p decoupling!

3. Calculation of pressure

Incomp. flow → $\rho \equiv \text{const}$

$P ? \Rightarrow$ a mathematical quantity to enforce the continuity eq.

① pressure eq.

$$N-S: \rho \frac{\partial u_i}{\partial t} + \rho \frac{\partial u_i u_j}{\partial x_j} = - \frac{\partial p}{\partial x_i} + \frac{\partial}{\partial x_j} \left(\mu \frac{\partial u_i}{\partial x_j} \right) \quad \frac{\partial u_i}{\partial x_i} = 0$$

$$\rightarrow \frac{\partial}{\partial x_i} \frac{\partial p}{\partial x_i} = - \frac{\partial}{\partial x_i} \frac{\partial}{\partial x_j} (\rho u_i u_j) \quad \text{Poisson eq.}$$

continuity eq. $\frac{\partial p}{\partial x_i}$
 momentum eq. $\frac{\partial}{\partial x_j} (\rho u_i u_j)$

use consistent discretization method. otherwise, continuity is not satisfied.

② Simple explicit time advance scheme

$$\begin{aligned} \frac{\partial(\rho u_i)}{\partial t} &= -\frac{\partial}{\partial x_j} (\rho u_i u_j) - \frac{\partial p}{\partial x_i} + \frac{\partial \tau_{ij}}{\partial x_j} \\ &= -\frac{\delta}{\delta x_j} (\rho u_i u_j) - \frac{\delta p}{\delta x_i} + \frac{\delta \tau_{ij}}{\delta x_j} \equiv H_i - \frac{\delta p}{\delta x_i} \end{aligned}$$

$$EE: \frac{(\rho u_i)^{n+1} - (\rho u_i)^n}{\Delta t} = H_i^n - \frac{\delta p^n}{\delta x_i} \quad (*)$$

If we have $\frac{\delta u_i^n}{\delta x_i} = 0$ and some p^n ,

then $\frac{\delta(\rho u_i)^{n+1}}{\delta x_i}$ is usually non zero with u_i^{n+1} from $(*)$.

Now, take divergence on $(*)$

$$\frac{\delta}{\delta x_i} (\rho u_i)^{n+1} - \frac{\delta}{\delta x_i} (\rho u_i)^n = \Delta t \left[\frac{\delta}{\delta x_i} (H_i^{\wedge} - \frac{\delta p^{\wedge}}{\delta x_i}) \right]$$

If u_i^{\wedge} is divergence free, $\frac{\delta}{\delta x_i} (\rho u_i)^{\wedge} = 0$

We require $\frac{\delta}{\delta x_i} (\rho u_i)^{n+1} = 0$

$$\frac{\delta}{\delta x_i} \left(\frac{\delta p^{\wedge}}{\delta x_i} \right) = \frac{\delta H_i^{\wedge}}{\delta x_i} \rightarrow \text{obtain } p^{\wedge}$$

higher-order method

$\Rightarrow \frac{\delta}{\delta x_i} (\rho u_i)^{n+1} = 0$ with u_i^{\wedge} from $\textcircled{*}$.

EE : 1st-order accurate \rightarrow inaccurate.

small $\Delta t \rightsquigarrow$ implicit method

② Simple implicit time advance method

Implicit Euler method (backward Euler method)

$$\frac{\partial(\rho u_i)}{\partial t} = -\frac{\partial}{\partial x_j} (\rho u_i u_j) - \frac{\partial p}{\partial x_i} + \frac{\partial \tau_{ij}}{\partial x_j} + O(\Delta t^2)$$

$$IE: \frac{(\rho u_i)^{n+1} - (\rho u_i)^n}{\Delta t} = -\frac{\delta(\rho u_i u_j)^{n+1}}{\delta x_j} - \frac{\delta p^{n+1}}{\delta x_i} + \frac{\partial \tau_{ij}^{n+1}}{\partial x_j} \quad \text{--- } \textcircled{*}$$

We require $\frac{\delta(\rho u_i)^{n+1}}{\delta x_i} = 0$ $\tau_{ij} = \mu \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right)$

Take divergence on $\textcircled{*}$ (ρu_i^n is divergence free)

$$\frac{\delta}{\delta x_i} \left(\frac{\delta p^{NH}}{\delta x_i} \right) = \frac{\delta}{\delta x_i} \left[- \frac{\delta (p u_i u_j)^{NH}}{\delta x_j} \right] - \underbrace{\text{(*)}}_{\text{unknown}}$$

Thus, one has to solve ~~(*)~~ & ~~(*)~~ simultaneously.
 → rely on iterative procedure.

Even if we know p^{NH} , ~~(*)~~ contains nonlinear term
 → requires iterative procedure. $\sim O(\delta t)$

Or, introduce linearization $u_i^{NH} = u_i^{\wedge} + \Delta u_i$

$$\rightarrow \underline{u_i^{NH} u_j^{NH}} = (u_i^{\wedge} + \Delta u_i)(u_j^{\wedge} + \Delta u_j)$$

$$= u_i^n u_j^n + u_i^n \Delta u_j + \Delta u_i u_j^n + \underbrace{\Delta u_i \Delta u_j}_{O(\Delta t^2)}$$

⇓

linear term

∴ neglect

④ Implicit pressure-correction methods

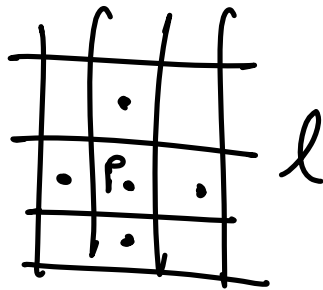
for steady state flow

→ use implicit method and take large Δt .

Why implicit method?
large Δt !

mtm eq.

$$\Rightarrow A_p^{u_i} u_{i,p}^{n+1} + \sum_l A_l^{u_i} u_{i,l}^{n+1} = Q_{u_i}^{n+1} - \left(\frac{\delta p}{\delta x_i} \right)_p$$



contains all the terms treated explicitly as well as the body force term.

→ iterative solver

$$A_p^{u_i} u_{i,p}^{m*} + \sum_l A_l^{u_i} u_{i,l}^{m*} = Q u_i^{m-1} - \left(\frac{\delta p^{m-1}}{\delta x_i} \right)_p$$

m : iteration index

get $u_{i,p}^{m*}$

- $u_{i,p}^{m*}$ does not satisfy continuity.
- We need continuity enforcement.
 - vel. has to be corrected.
 - modification of the press.

$$\hookrightarrow u_{i,p}^{m*} = \frac{1}{A_p^{u_i}} \left(Q_{u_i}^{m*} - \sum_l A_l^{u_i} u_{i,l}^{m*} \right) - \frac{1}{A_p^{u_i}} \left(\frac{\delta p^{m-1}}{\delta x_i} \right)_p$$

$\equiv \tilde{u}_{i,p}^{m*}$: one from which the contribution of the press. grad. has been removed.

continuity $\frac{\delta}{\delta x_i} (\rho u_i^m) = 0$

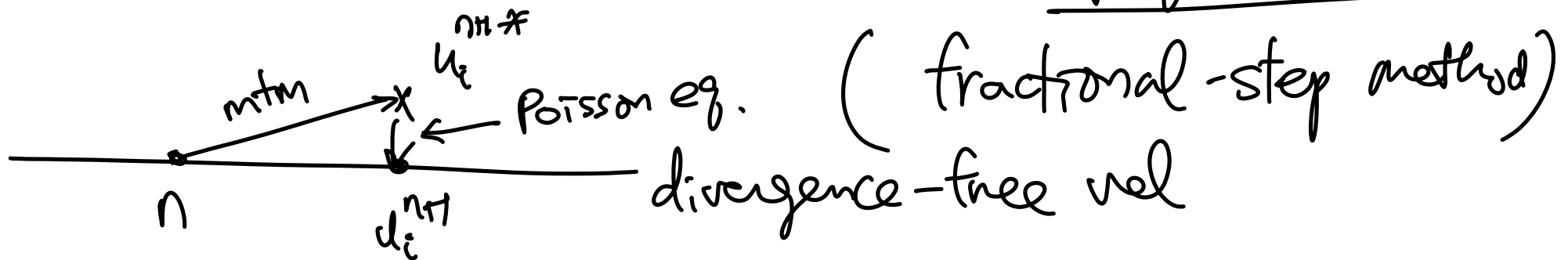
Let $u_{i,p}^m = \tilde{u}_{i,p}^{m*} - \frac{1}{A_p^{u_i}} \left(\frac{\delta p^m}{\delta x_i} \right)_p$
 vel. correction

$u_{i,p}^m = \tilde{u}_{i,p}^{m*} - \frac{1}{A_p^{u_i}} \left(\frac{\delta p^m}{\delta x_i} \right)_p$
 ~~$\left(\frac{\delta p^m}{\delta x_i} \right)_p$~~
 $p^m - p^{m-1}$

take divergence : $\frac{\delta}{\delta x_i} \left[\frac{\rho}{A_p^{u_i}} \left(\frac{\delta p^m}{\delta x_i} \right) \right]_p = \left(\frac{\delta \rho \tilde{u}_i^{m*}}{\delta x_i} \right)_p$

- obtain p^m
- update $u_{i,p}^m$
- keep iteration until converges.

This kind of method is called projection method.



$$u_{i,p}^m = \tilde{u}_{i,p}^{n+1} - \left(\frac{1}{A_p^{u_i}} \frac{\partial p^m}{\partial x_i} \right)_p \quad ?$$

$$A_p u_{i,p}^{n+1} + \sum_l A_l u_{i,l}^{n+1} = - \left(\frac{\partial p}{\partial x_i} \right)_p$$

iteration :
$$A_p u_{i,p}^m + \sum_l A_l u_{i,l}^m = - \left. \frac{\delta \rho^m}{\delta x_i} \right|_p$$

$$A_p u_{i,p}^{m*} + \sum_l A_l u_{i,l}^{m*} = - \left. \frac{\delta \rho^{m-1}}{\delta x_i} \right|_p$$

$$\rightarrow u_{i,p}^{m*} = \frac{1}{A_p} \left(- \sum_l A_l u_{i,l}^{m*} \right) - \left. \frac{\delta \rho^{m-1}}{\delta x_i} \right|_p$$

$$\equiv \underbrace{u_{i,p}^{m*}}_{\tilde{u}_{i,p}^{m*}}$$

$$\frac{\delta}{\delta x_i} (\rho u_i^m) = 0$$

$$u_{i,p}^m = \underbrace{\tilde{u}_{i,p}^{m*}}_{u_{i,p}^{m*}} - \left. \frac{\delta \phi}{\delta x_i} \right|_p$$

$$u_{i,p}^m = u_{i,p}^{m*} + \left. \frac{\delta}{\delta x_i} (\rho^{m-1} - \phi) \right|_p$$

$$\downarrow A_p u_{i,p}^{m*} = A_p \underbrace{u_{i,p}^{m*}}_{=} - \frac{\delta p^{m-1}}{\delta x_i} \Big|_p$$

$$u_{i,p}^{m*} = u_{i,p}^m + \frac{1}{A_p} \frac{\delta}{\delta x_i} (\phi - p^{m-1}) \Big|_p$$

$$- \sum_l A_l u_{i,l}^{m*}$$

$$\rightarrow A_p u_{i,p}^m + \sum_l A_l u_{i,l}^m + \sum_l \frac{A_l}{A_p} \left(\frac{\partial \phi}{\partial x_i} - \frac{\delta p^{m-1}}{\delta x_i} \right) = - \frac{\delta \phi}{\delta x_i} \Big|_p$$

$$\left(\quad \quad \quad \right) = - \frac{\delta p^m}{\delta x_i}$$

\Rightarrow then, $\phi = p^m$

goes to zero when converged.

① SIMPLE (Semi-Implicit Method for Pressure-Linked Equations)

$$\begin{cases} u_i^m = u_i^{m*} + u_i' \\ p^m = p^{m-1} + p' \end{cases}$$

$$\text{mtm} : A_p u_{i,p}^m + \sum_l A_l u_{i,l}^m = \cancel{Q_{i,p}^m} - \frac{\delta p^m}{\delta x_i} \Big|_p$$

obtain u_i^{m*} ← $A_p u_{i,p}^{m*} + \sum_l A_l u_{i,l}^{m*} = \cancel{Q_{i,p}^{m-1}} - \frac{\delta p^{m-1}}{\delta x_i} \Big|_p$

$$A_p u_{i,p}' + \sum_l A_l u_{i,l}' = - \frac{\delta p'}{\delta x_i} \Big|_p$$

$$u_{i,p}' = \frac{1}{A_p} \left(-\sum_l A_{l,p} u_{i,l}' \right) - \frac{1}{A_p} \left(\frac{\delta p'}{\delta x_i} \right)_p$$

$\sim u_{i,p}'$

$\nabla.$

$$\left(\text{cont: } \frac{\delta u_i}{\delta x_i} = \frac{\delta u_i^{m*}}{\delta x_i} + \frac{\delta u_i'}{\delta x_i} = 0 \right)$$

press-corrector eq.

$$\frac{\delta}{\delta x_i} \left[\frac{\rho}{A_p} \left(\frac{\delta p'}{\delta x_i} \right) \right]_p = \left[\frac{\delta}{\delta x_i} (\rho u_i^{m*}) \right]_p + \frac{\left(\frac{\delta u_i'}{\delta x_i} \right)_p}{\text{unknown}}$$

- obtain p'
 - update $u_{i,p}' (= -\frac{1}{A_p} \frac{\delta p'}{\delta x_i})$ causes slow convergence
- no justification $\leftarrow \therefore$ neglect!

→ obtain u_i^m and p^m , → keep iteration

SIMPLE

$$u_i^m = u_i^{m*} + u_i', \quad p^m = p^{m-1} + p' \quad m: \text{iteration index}$$

$$\text{mtm: } A_p u_{i,p}^m + \sum_{\ell} A_{\ell} u_{i,\ell}^m = - \left. \frac{\delta p^m}{\delta x_i} \right|_p + \cancel{Q_{i,p}^m}$$

$$\bullet \textcircled{1} \quad A_p u_{i,p}^{m*} + \sum_{\ell} A_{\ell} u_{i,\ell}^{m*} = - \left. \frac{\delta p^{m*}}{\delta x_i} \right|_p \Rightarrow \text{obtain } u_i^{m*}$$

$$A_p u_{i,p}' + \sum_{\ell} A_{\ell} u_{i,\ell}' = - \left. \frac{\delta p'}{\delta x_i} \right|_p$$

$$\lceil \quad u_{i,p}' = \underbrace{\frac{1}{A_p} \left(- \sum_{\ell} A_{\ell} u_{i,\ell}' \right)} - \frac{1}{A_p} \left. \frac{\delta p'}{\delta x_i} \right|_p$$

Continuity $\frac{\delta u_c^m}{\delta x_i} = \frac{\delta u_c^{m*}}{\delta x_i} + \frac{\delta u_c'}{\delta x_i} = 0$ $= \tilde{u}_{c,p}$

② $\frac{\delta}{\delta x_i} \left(\frac{\rho}{A_p} \frac{\delta p'}{\delta x_i} \right)_p = \frac{\delta}{\delta x_i} (\rho u_c^{m*}) \Big|_p + \frac{\delta \tilde{p} u_c'}{\delta x_i} \Big|_p$

known unknown

→ obtain p' → neglect

→ update $u_c' = -\frac{1}{A_p} \frac{\delta p'}{\delta x_i}$ and $\underline{p^m = p^{m-1} + p'}$

→ obtain $\underline{u_c^m = u_c^{m*} + u_c'}$

⑥ SIMPLEC

Don't neglect $\frac{\delta}{\delta x_i} (\tilde{p} u_c')$.

Do approximate as $\sum_{\ell} A_{\ell} u'_{i,\ell} \simeq u'_{i,p} \cdot (\sum_{\ell} A_{\ell})$

then, $\tilde{u}'_{i,p} = -\frac{1}{A_p} \sum_{\ell} A_{\ell} u'_{i,\ell} \doteq -u'_{i,p} \sum_{\ell} A_{\ell} / A_p$

$$\left[u'_{i,p} = \tilde{u}'_{i,p} - \frac{1}{A_p} \frac{\delta p'}{\delta x_i} \right]_p \leftarrow$$

$$\hookrightarrow u'_{i,p} = - \frac{1}{A_p + \sum_{\ell} A_{\ell}} \left(\frac{\delta p'}{\delta x_i} \right)_p$$

Poisson eq: $\frac{\delta}{\delta x_i} \left[\frac{1}{A_p + \sum_{\ell} A_{\ell}} \frac{\delta p'}{\delta x_i} \right]_p = \frac{\delta}{\delta x_i} (p u_c^{M*}) \Big|_p$

→ obtain p' → obtain u'_i

→ update u_i^M and p^M

© SIMPLER

neglect $\frac{\partial}{\partial x_i} (\rho \tilde{u}_c') \Big|_p$ as in SIMPLE

and obtain p' as in SIMPLE

→ update u_c' → obtain u_c^m

How to obtain p^m ? ($p^m = p^{m-1} + p'$ in SIMPLE)

$$A_p u_{c,p}^m + \underbrace{\sum_l A_l u_{c,l}^m}_{\equiv -\tilde{u}_c^m} = -\frac{\partial p^m}{\partial x_i} \Big|_p$$

$$\rightarrow \frac{\partial}{\partial x_i} \left[\frac{\rho}{A_p} \frac{\partial p^m}{\partial x_i} \right]_p = \frac{\partial}{\partial x_i} (\rho \tilde{u}_c^m) \Big|_p$$

→ obtain p^m → iterate until convergence.

(d) PISO $u_c^m = u_c^{m*} + \underline{u_c'} + \underline{u_c''}$, $p^m = p^{m-1} + \underline{p'} + \underline{p''}$

These methods are fairly efficient for solving steady state problems.

4. Other methods

(i) Fractional step methods (called projection method)

Kim & Moin (1985)

- 2nd-order accurate

Chorin (19--)

Temam ()

} 1st-order accurate

$$\left\{ \begin{array}{l} \frac{\partial u_i}{\partial x_i} = 0 \\ \frac{\partial u_i}{\partial t} + \frac{\partial}{\partial x_j} (u_i u_j) = -\frac{\partial p}{\partial x_i} + \frac{1}{Re} \nabla^2 u_i \end{array} \right. \text{in time}$$

non-dimensionalized
N-S eq.

(Adams - Bashforth method for nonlinear term
 and Crank - Nicolson method for viscous term)

($n \rightarrow n+1$)

$$\frac{u_i^{n+1} - u_i^n}{\Delta t} + \frac{1}{2} \left[3 \frac{\partial}{\partial x_j} (u_i u_j^n) - \frac{\partial}{\partial x_j} (u_i^n u_j^n) \right] \rightarrow \text{explicit}$$

AB2

$$= -\frac{1}{2} \left(\frac{\partial p^{n+1}}{\partial x_r} + \frac{\partial p^n}{\partial x_r} \right) + \frac{1}{2} \frac{1}{Re} \nabla^2 (u_i^n + u_i^{n+1})$$

implicit

$$\frac{\partial u_i^{n+1}}{\partial x_i} = 0$$

one way to go w/ this formulation

→ direct inverting → very expensive

- fractional step method \hat{u}_c : intermediate vel.

$$\textcircled{1} \quad \frac{\hat{u}_c - u_c^n}{\Delta t} + \frac{1}{2} \left[3 \frac{\partial}{\partial x_j} u_c^n u_j^n - \frac{\partial}{\partial x_j} u_c^{n+1} u_j^{n+1} \right] = \frac{1}{2} \frac{1}{Re} \nabla^2 (\hat{u}_c + u_c^n)$$

→ Second-order approximation of

$$\frac{\partial u_i}{\partial t} + \frac{\partial}{\partial x_j} (u_i u_j) = \frac{1}{Re} \nabla^2 u_i$$

obtain \hat{u}_c . → \hat{u}_c does not satisfy the continuity.

$$\frac{u_c^{n+1} - \hat{u}_c}{\Delta t} = - \frac{\partial \phi^{n+1}}{\partial x_i}$$

Force u_i^{n+1} to satisfy the continuity.

$$\rightarrow \nabla^2 \phi^{n+1} = \frac{1}{\Delta t} \frac{\partial \hat{u}_c}{\partial x_i} \quad \textcircled{2}$$

→ obtain ϕ^{n+1}

→ $u_i^{n+1} = \hat{u}_i - \Delta t \frac{\partial \phi^{n+1}}{\partial x_i}$ (3) → time marching

• Issue on the computational cost for Eq. (1)

$$N_i \equiv \frac{\partial}{\partial x_j} (u_i u_j)$$

+ $O(\Delta t^2)$

$$\frac{\hat{u}_i - u_i^n}{\Delta t} + \frac{1}{2} (3N_i^n - N_i^{n+1}) = \frac{1}{2Re} \left(\frac{\delta^2}{\delta x^2} + \frac{\delta^2}{\delta y^2} + \frac{\delta^2}{\delta z^2} \right) (\hat{u}_i + u_i^n)$$

$$\rightarrow \left[1 - \frac{\Delta t}{2Re} \left(\frac{\delta^2}{\delta x^2} + \frac{\delta^2}{\delta y^2} + \frac{\delta^2}{\delta z^2} \right) \right] (\hat{u}_i - u_i^n) \quad \left[\begin{array}{l} \delta \hat{u}_i = \hat{u}_i - u_i^n \\ \text{delta form} \end{array} \right]$$

$$= \frac{\Delta t}{2} (-3N_i^n + N_i^{n+1}) + \frac{\Delta t}{Re} \left(\frac{\delta^2}{\delta x^2} + \frac{\delta^2}{\delta y^2} + \frac{\delta^2}{\delta z^2} \right) u_i^n \quad + O(\Delta t^3)$$

$$A_x = \frac{\Delta t}{2Re} \frac{\delta^2}{\delta x^2}, \quad A_y = \frac{\Delta t}{2Re} \frac{\delta^2}{\delta y^2}, \quad A_z = \frac{\Delta t}{2Re} \frac{\delta^2}{\delta z^2}$$

$$\rightarrow \underbrace{(1 - A_x - A_y - A_z)}_{\text{Sparse matrix}} (\hat{u}_c - u_c^n) = R_i + \mathcal{O}(\Delta t^3)$$

Sparse matrix \rightarrow expensive to invert

✓ introduce approximate factorization scheme

$$(1 - A_x - A_y - A_z) (\hat{u}_c - u_c^n) = (1 - A_x)(1 - A_y)(1 - A_z) (\hat{u}_c - u_c^n) + \mathcal{O}(\Delta t^3) \quad \text{do not lose any accuracy}$$

$$\rightarrow \underbrace{(1 - A_x)}_{\text{tridiagonal matrix}} \underbrace{(1 - A_y)(1 - A_z)}_{v_i} (\hat{u}_c - u_c^n) = R_i + \underline{\mathcal{O}(\Delta t^3)}$$

tridiagonal matrix $v_i \rightarrow$ get v_i

$$(1 - A_y)(1 - A_z) (\hat{u}_c - u_c^n) = v_i \xrightarrow{w_i} \text{get } w_i$$

$$(1 - A\Delta t)(\hat{u}_i - u_i) = w_i \rightarrow \text{get } \hat{u}_i$$

$\mathcal{O}(N)$ operation.

Saves a lot of CPU and memory!

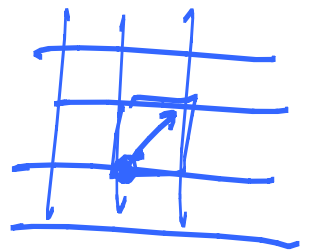
Issue on the numerical stability

AB2 + CN : semi-implicit method
 explicit implicit

↓
 conditionally stable

$$CFL = \left(\frac{|u|}{\Delta x} + \frac{|v|}{\Delta y} + \frac{|w|}{\Delta z} \right) \Delta t \leq 1 \quad \leftarrow$$

$$\Delta t \leq \left(\frac{\Delta x}{|u|} + \frac{\Delta y}{|v|} + \frac{\Delta z}{|w|} \right)^{-1}$$



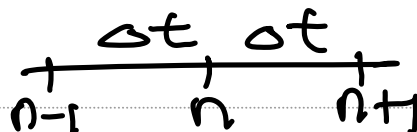
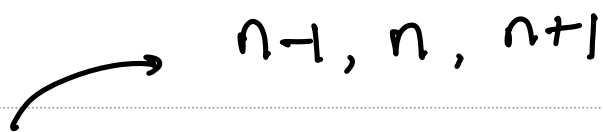
($ABZ + ABZ$ fully explicit $oE \sim \Delta x^2$ too severe)

problem: (not self-starting $\therefore \frac{n \ \& \ n-1}{\downarrow}$
 $ABZ + CN$ (spurious root $\therefore n+1, n, n+1$)

↳ better one is

① RK3 + CN

② CN (fully implicit)



$$CFL \leq 1$$

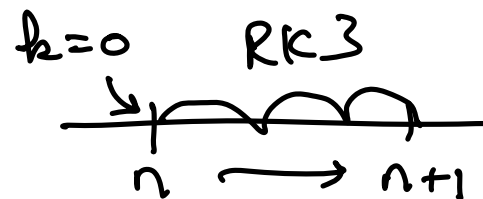
- AB2 + CN
 - RK3 + CN
- } semi-implicit methods

Fractional step method using RK3 + CN

N_i : nonlinear viscous L_i

$$\textcircled{1} \frac{\hat{u}_i^k - u_i^{k-1}}{\Delta t} = (\alpha_k + \beta_k) L_i(u^{k-1}) + \beta_k L_i(\hat{u}^k - u^{k-1}) - \gamma_k N_i(u^{k-1}) - \zeta_k N_i(u^{k-2}) \quad k=1,2,3$$

$$\textcircled{2} + (\gamma_k + \zeta_k) \nabla^2 \phi^k = \frac{1}{\Delta t} \frac{\partial \hat{u}_i^k}{\partial x_i}$$

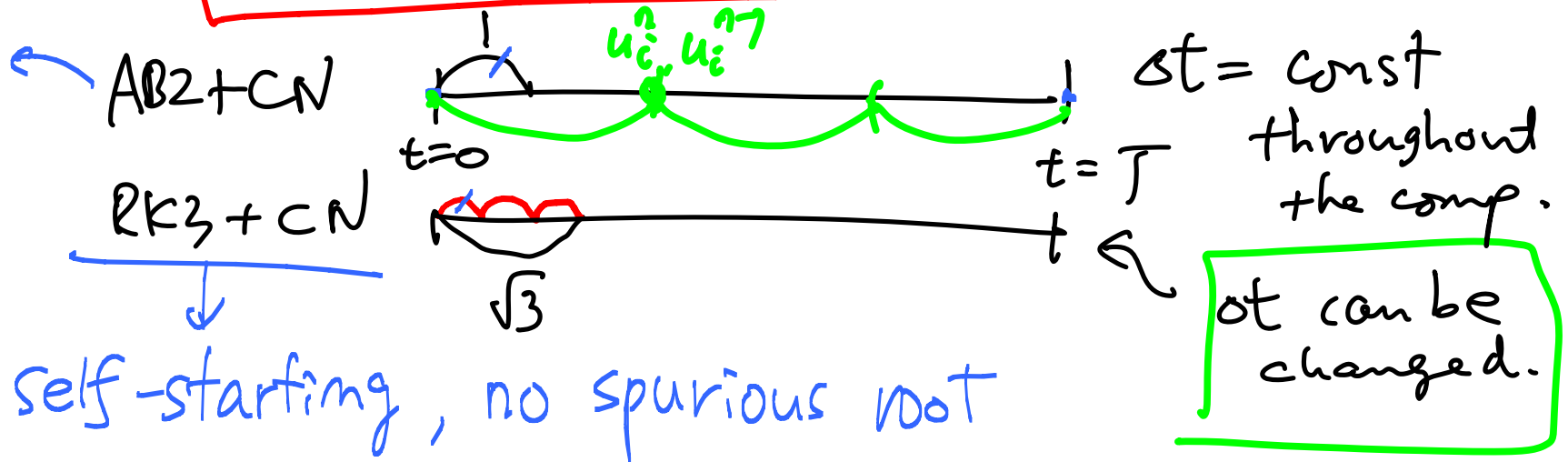


$$\textcircled{3} u_i^k = \hat{u}_i^k - \Delta t (\gamma_k + \zeta_k) \partial \phi^k / \partial x_i$$

$\alpha_1 = \beta_1 = \frac{k}{15}$	$\gamma_1 = \frac{8}{15}$	$\delta_1 = 0$	ϕ : pseudo-pressure
$\alpha_2 = \beta_2 = \frac{1}{15}$	$\gamma_2 = \frac{5}{12}$	$\delta_2 = -\frac{17}{60}$	
$\alpha_3 = \beta_3 = \frac{1}{6}$	$\gamma_3 = \frac{3}{4}$	$\delta_3 = -\frac{5}{12}$	

$$CFL = \left(\frac{|u|}{\Delta x} + \frac{|v|}{\Delta y} + \frac{|w|}{\Delta z} \right) \Delta t \leq \sqrt{3}$$

u_i^n, u_i^{n+1}



- CN only (fully implicit) \rightarrow no limit on Δt .

Choi & Moim (1994, JCP, 113, 1)

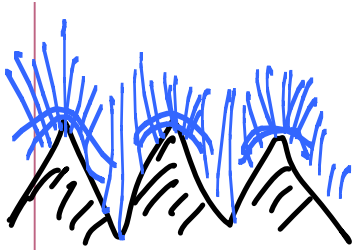
$$\left\{ \begin{array}{l} \frac{u_i^{n+1} - u_i^n}{\Delta t} + \frac{1}{2} \frac{\partial}{\partial x_j} (u_i^{n+1} u_j^{n+1} + u_i^n u_j^n) = -\frac{\partial p^{n+1}}{\partial x_i} + \frac{1}{2Re} \nabla^2 (u_i^{n+1} + u_i^n) \\ \frac{\partial u_i^{n+1}}{\partial x_i} = 0 \end{array} \right. \quad \text{or} \quad -\frac{1}{2} \left(\frac{\partial p^{n+1}}{\partial x_i} + \frac{\partial p^n}{\partial x_i} \right)$$

FSM ① $\frac{\hat{u}_i - u_i^n}{\Delta t} + \frac{1}{2} \frac{\partial}{\partial x_j} (\hat{u}_i \hat{u}_j + u_i^n u_j^n) = -\frac{\partial p^n}{\partial x_i} + \frac{1}{2Re} \nabla^2 (\hat{u}_i + u_i^n)$

② $\frac{u_i^* - \hat{u}_i}{\Delta t} = \frac{\partial p^n}{\partial x_i}$

③ $\nabla^2 p^{n+1} = \frac{1}{\Delta t} \frac{\partial u_i^*}{\partial x_i}$

\hookrightarrow get \hat{u}_i using iterative scheme like Newton iteration or using



$$\textcircled{4} \quad u_i^{n+1} = u_i^* - \Delta t \frac{\partial p^{n+1}}{\partial x_i}$$

linearization

Main difference bet. FSM and SIMPLE:
for unsteady prob.

FSM: solve the Poisson eq 1-3 times / time step

SIMPLE: solve both the mom and pressure-correction eqs several times per time step.

Let's go back to $AB^2 + CN$

$$\frac{u_i^{nH} - \hat{u}_i^n}{\Delta t} + \frac{1}{2} \left[3 \frac{\partial}{\partial x_j} u_i^n u_j^n - \frac{\partial}{\partial x_j} u_i^{nH} u_j^{nH} \right]$$

$$= -\frac{1}{2} \left(\frac{\partial p^{nH}}{\partial x_i} + \frac{\partial p^n}{\partial x_i} \right) + \frac{1}{2 \text{Re}} \nabla^2 (u_i^{nH} + \hat{u}_i^n)$$

$-\frac{\partial p^n}{\partial x_i}$

FSM

$$\frac{\hat{u}_i^n - u_i^n}{\Delta t} + \frac{1}{2} \left[\dots \right] = \frac{1}{2 \text{Re}} \nabla^2 (\hat{u}_i^n + u_i^n)$$

$$\frac{u_i^{nH} - \hat{u}_i^n}{\Delta t} = -\frac{\partial \phi}{\partial x_i} \Rightarrow \hat{u}_i^n = u_i^{nH} + \Delta t \frac{\partial \phi}{\partial x_i}$$

$-\frac{\partial p^{nH}}{\partial x_i}$

$-\frac{\partial p^n}{\partial x_i}$

$$\frac{u_i^{n+1} - u_i^n}{\Delta t} + \frac{\partial \phi}{\partial x_i} + \frac{1}{2} \left[\text{"} \right] = \frac{1}{2Re} \nabla^2 (u_i^{n+1} + u_i^n + \Delta t \frac{\partial \phi}{\partial x_i})$$

$$\rightarrow \frac{u_i^{n+1} - u_i^n}{\Delta t} + \frac{1}{2} \left[\text{"} \right] = -\frac{\partial \phi}{\partial x_i} + \frac{\Delta t}{2Re} \frac{\partial}{\partial x_i} (\nabla^2 \phi) + \frac{1}{2Re} \nabla^2 (u_i^{n+1} + u_i^n)$$

$$-\frac{\partial}{\partial x_i} \left(\frac{1}{2} (p^{n+1} + p^n) \right) = -\frac{\partial}{\partial x_i} \left(\phi - \frac{\Delta t}{2Re} \nabla^2 \phi \right)$$

p^{n+1}

$p^n +$

$$\rightarrow p^{n+1} = -p^n + 2\phi - \frac{\Delta t}{Re} \nabla^2 \phi$$

p^n

$$p^{n+1} = \phi - \frac{\Delta t}{2Re} \nabla^2 \phi$$

Q: following Kim & Moim (1985),
 what is the b.c. for \hat{u}_0 ?

$$\hat{u}_0 = u_0^{n+1} ?$$

$$\rightarrow \hat{u}_0 = u_0^{n+1} + \sigma t \frac{\partial \phi^n}{\partial x_i} \quad (\text{b.c.})$$

w/ explicit
 press.
 grad.

$-\frac{\partial \phi^n}{\partial x_i} \rightarrow$ 1st step
 σt
 FSM

$$\Rightarrow \hat{u}_0 = u_0^{n+1} \quad (\text{b.c.})$$

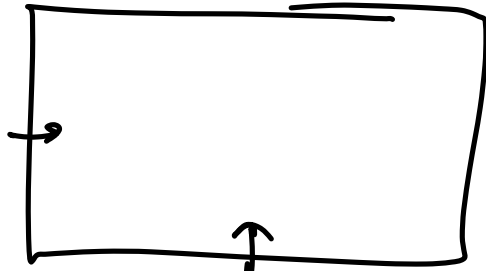
$$\frac{u_0^{n+1} - u_0^n}{\sigma t} = -\frac{\partial \phi}{\partial x_i}$$

b.c for ϕ ?

$$\nabla^2 \phi = \frac{1}{\sigma t} \frac{\partial u_i^n}{\partial x_i}$$

$$\frac{u^{n+1} - \hat{u}}{\Delta t} \Big|_{\text{wall}} = \frac{\partial \phi}{\partial x}$$

0



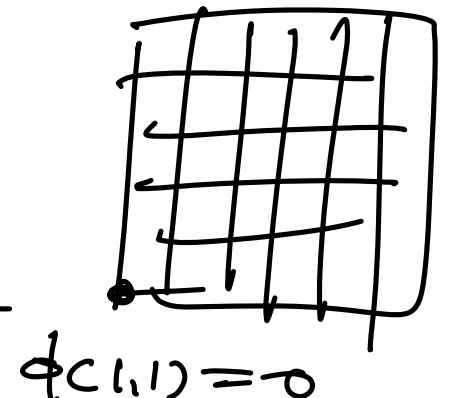
$$\frac{\partial \phi}{\partial y} = (u^{n+1} - \hat{u}) / \Delta t \Big|_{\text{bdry}} = 0 \quad (\omega / - \frac{\partial p^n}{\partial x_i})$$

$$= 0 \quad (\omega / 0 \quad //)$$

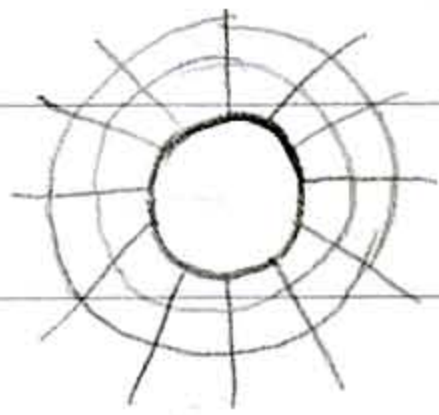
$$\frac{\partial p}{\partial y} \Big|_{\text{bdry}} \neq 0 \quad \frac{1}{Re} \nabla^2 v \Big|_{\text{wall}}$$

$$\nabla^2 \phi = r \quad \omega / \frac{\partial \phi}{\partial n} = 0$$

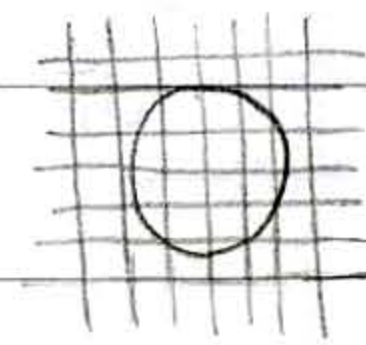
$$\Rightarrow \phi + c$$



$$\phi(x, y) = 0$$

Immersed Boundary method (IB method)

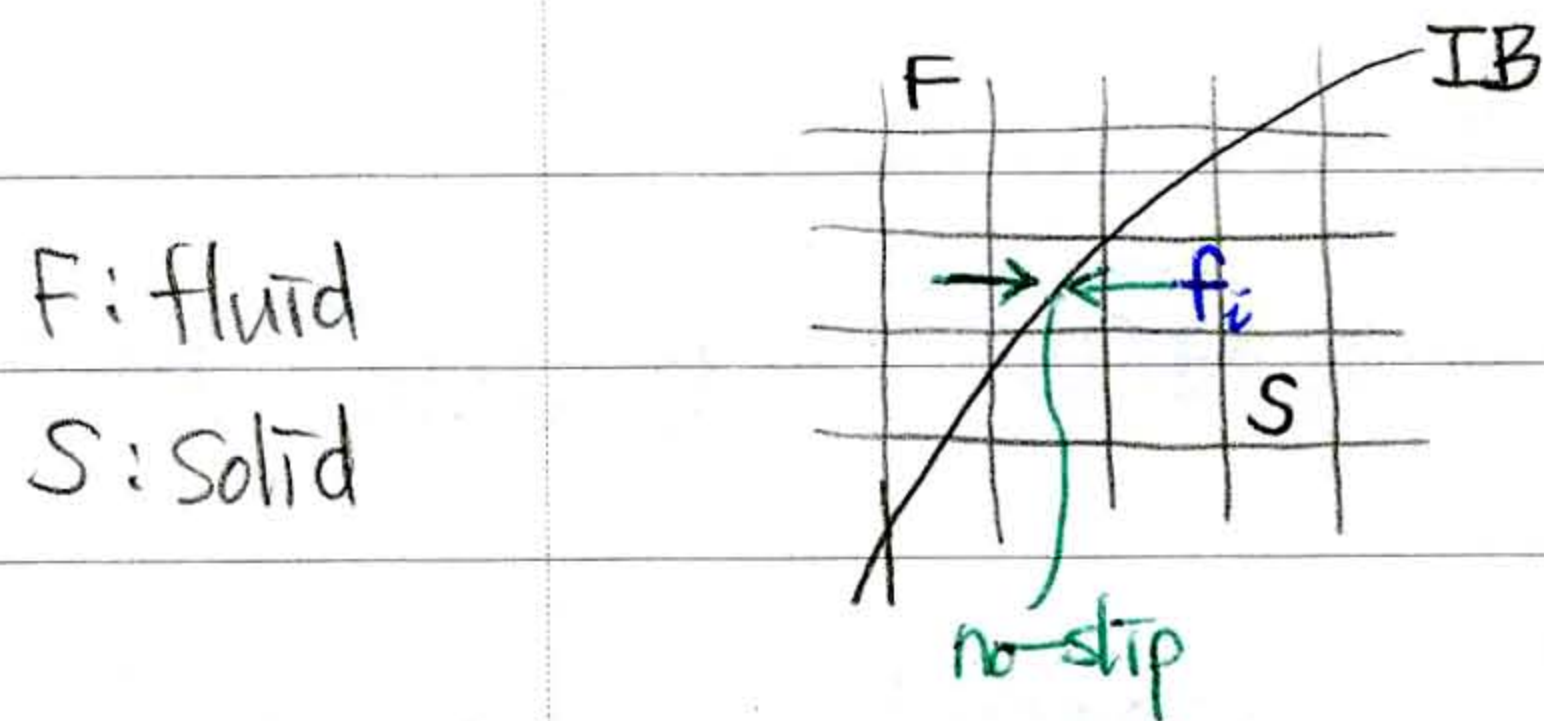
Body-fitted grid



IB method

Kim et al. (2001, JCP)

* Peskin (1982): Original Development

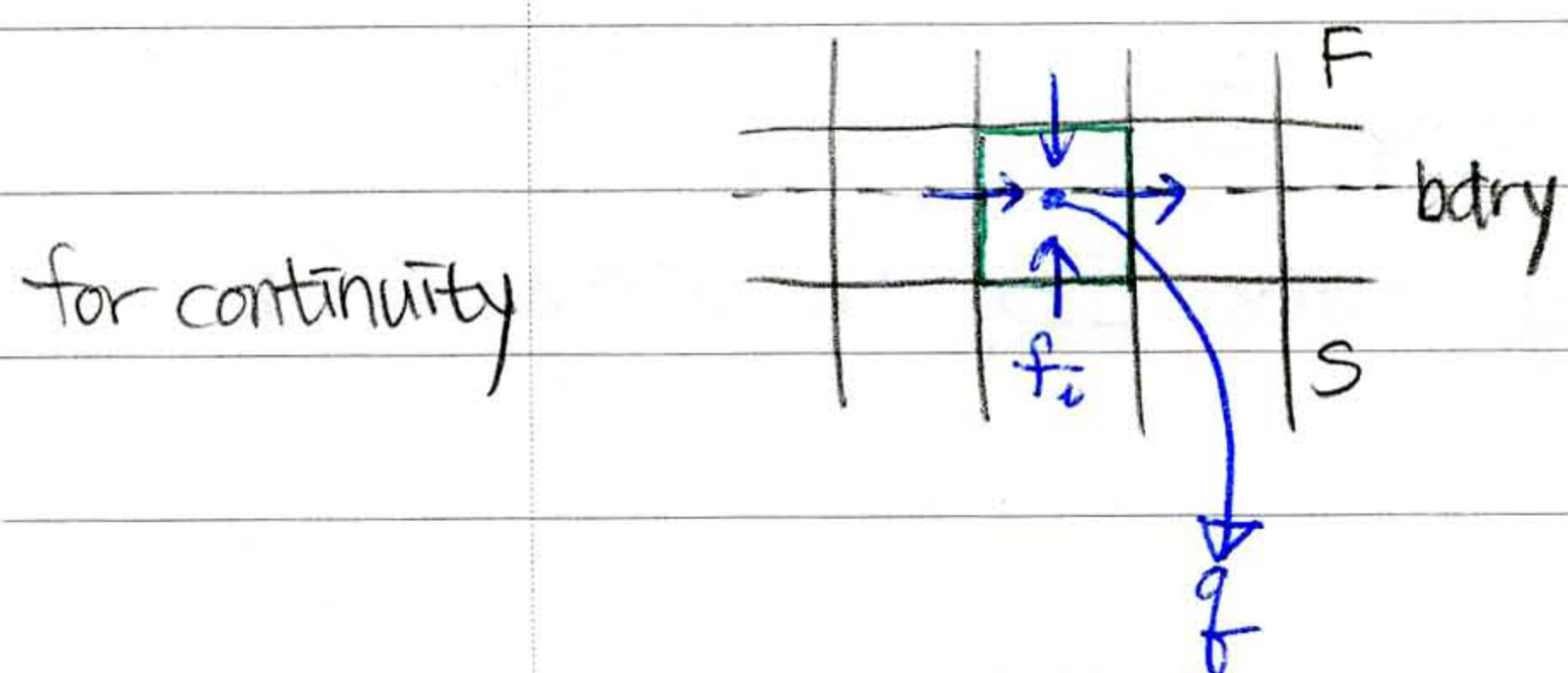


$$\frac{\partial u_i}{\partial t} + \frac{\partial}{\partial x_j} u_i u_j = -\frac{\partial p}{\partial x_i} + \frac{1}{Re} \nabla^2 u_i + \underbrace{f_i}_m$$

m: momentum forcing (Peskin)

$$\frac{\partial u_i}{\partial x_i} - \underbrace{q}_m = 0$$

mass source/sink (Kim et al.)



→ Provides mass source/sink q for cell containing IB

$$\frac{\partial u_i}{\partial x_i} - q = 0$$

Fractional Step method (RK3 + CN2)

$$\textcircled{4} \frac{u_i^k - u_i^{k-1}}{\Delta t} = \alpha_k L(\hat{u}_i^k) + \alpha_k L(u_i^{k-1}) - 2\alpha_k \frac{\partial p^{k-1}}{\partial x_i} - \gamma_k N(u_i^{k-1}) - \beta_k N(u_i^{k-2}) + \underbrace{f_i^k}_m$$

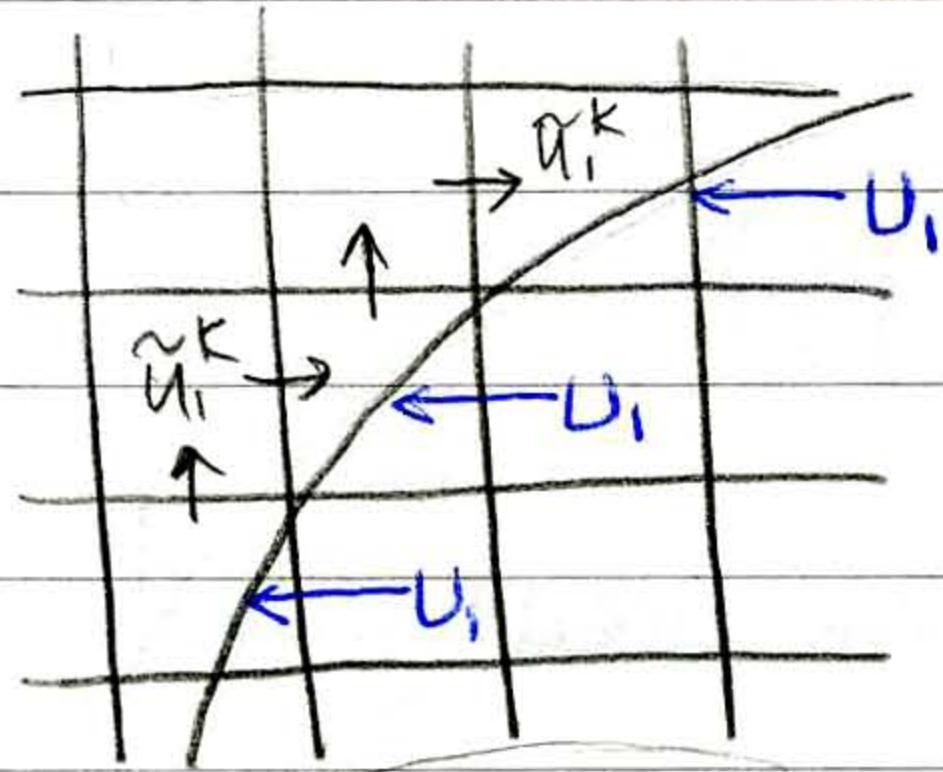
$$\textcircled{6} \nabla^2 \phi^k = \frac{1}{2\alpha_k \Delta t} \left(\frac{\partial u_i^k}{\partial x_i} - \underbrace{q^k}_m \right)$$

$$\textcircled{7} u_i^k = \hat{u}_i^k - 2\alpha_k \Delta t \frac{\partial \phi^k}{\partial x_i}$$

$$\textcircled{8} p^k = p^{k-1} + \phi^k - \frac{\alpha_k \Delta t}{Re} \nabla^2 \phi^k$$

• How to obtain f_i^k ?

$k-1 \rightarrow k$
시간전진
① ~ ⑧



We update N-S eq. with explicit numerical method.

$$\textcircled{1} \frac{\tilde{u}_i^k - u_i^{k-1}}{\Delta t} = \underbrace{2\alpha_k L(u_i^{k-1})}_{EE} - 2\alpha_k \frac{\partial p^{k-1}}{\partial x_i} - \underbrace{\gamma_k N(u_i^{k-1}) - \beta_k N(u_i^{k-2})}_{RK3}$$

By interpolating the provisional vel. near IB \tilde{u}_i^k ,
We obtain u_i^k

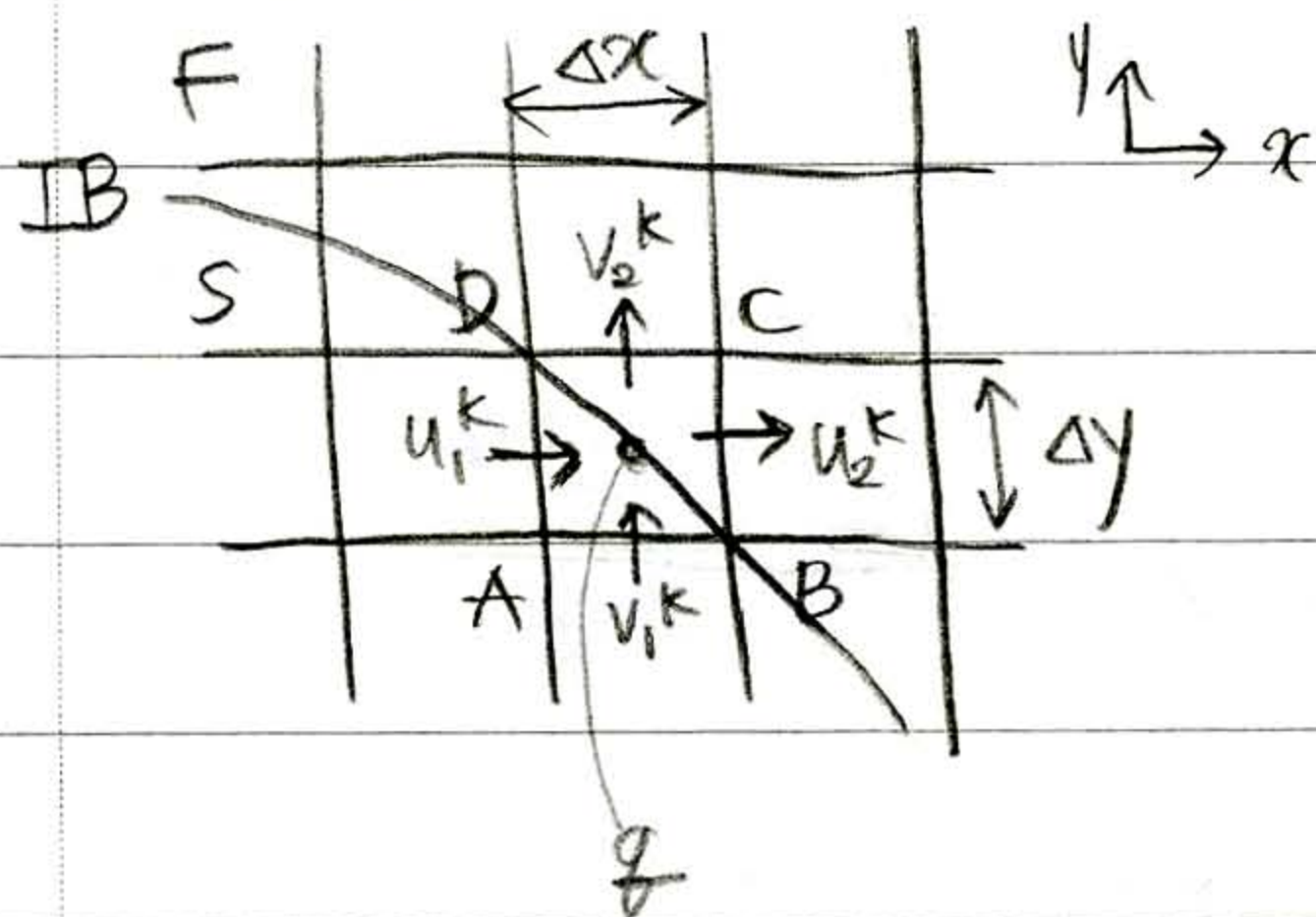
1D, 2D, 3D (Kim et al.)

$$\textcircled{2} \frac{u_i^k - \tilde{u}_i^k}{\Delta t} = 2\alpha_k L(u_i^{k-1}) - 2\alpha_k \frac{\partial p^{k-1}}{\partial x_i} - \gamma_k N(u_i^{k-1}) - \beta_k N(u_i^{k-2}) + f_i$$

$$\textcircled{3} f_i = \frac{u_i^k - \tilde{u}_i^k}{\Delta t} - 2\alpha_k L(u_i^{k-1}) + 2\alpha_k \frac{\partial p^{k-1}}{\partial x_i} + \gamma_k N(u_i^{k-1}) + \beta_k N(u_i^{k-2})$$

\tilde{u}_i^k : Provisional velocity near IB intp.
 u_i^k : Interpolated velocity inside IB to obtain f_i

• How to obtain q^k ?



$$\Delta BCD: u_2^k \Delta y + v_2^k \Delta x = 0$$

$$\square ABCD: u_2^k \Delta y + v_2^k \Delta x = u_1^k \Delta y + v_1^k \Delta x + \Delta x \Delta y q^k$$

$$\textcircled{5} q^k = -\frac{u_1^k}{\Delta x} - \frac{v_1^k}{\Delta y} = -\frac{\hat{u}_1^k}{\Delta x} - \frac{\hat{v}_1^k}{\Delta y}$$

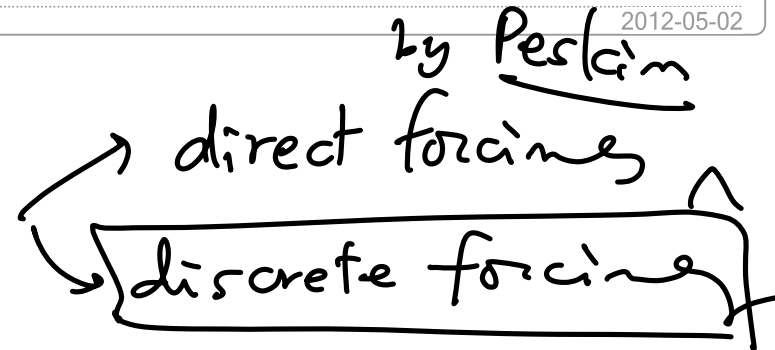
Kim et al. 기준

f_i : always inside IB
 q : inside / outside IB

"Other" methods

① Fractional step methods

⇒ ② Immersed boundary methods



③ Stream function - vorticity methods

→ $u = \frac{\partial \psi}{\partial y}$ $v = -\frac{\partial \psi}{\partial x}$ only (2D) ⇒ weakness

$\omega = \frac{\partial v}{\partial x} - \frac{\partial u}{\partial y} \Rightarrow \Delta^2 \psi = -\omega$ ①

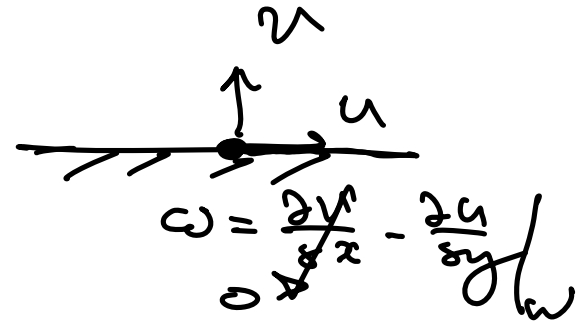
(sgo s-N) x Δ

$$\rightarrow \boxed{\rho \frac{\partial \omega}{\partial t} + \rho u \frac{\partial \omega}{\partial x} + \rho v \frac{\partial \omega}{\partial y} = \mu \nabla^2 \omega} \quad (2)$$

3 eqs (cont + N-S) \rightarrow 2 eqs

* continuity is identically satisfied!
no pressure in governing eqs.

problems: \wedge 2D flows
only for

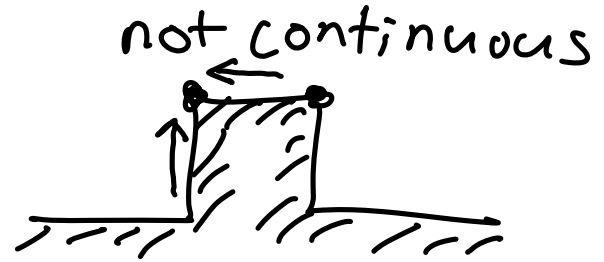


boundary condition of ω @ the wall

↳ usually one-side difference

↳ vorticity is singular at sharp corners.

ψ - ω approaches are less popular nowadays.



④ Artificial compressibility method

↳ by Chorin (1967) JCP, Vol. 2, 12.

↳ for steady flow, less efficient for unsteady flow

incompressible

$$\frac{\partial \rho u_i}{\partial x_i} = 0 \quad \longrightarrow \quad \frac{1}{\beta} \frac{\partial p}{\partial t} + \frac{\partial(\rho u_i)}{\partial x_i} = 0$$

(background) speed of sound $a^2 = \left. \frac{\partial p}{\partial \rho} \right|_s = \gamma R T$ for perfect gas

$$P = \rho R T = \rho R \frac{a^2}{\gamma R} = \rho \cdot \frac{a^2}{\gamma} \equiv \rho \beta$$

$$\frac{\partial p}{\partial t} + \frac{\partial}{\partial x_i} (\rho u_i) = 0 \rightarrow \frac{1}{\rho} \frac{\partial p}{\partial t} + \frac{\partial}{\partial x_i} (\rho u_i) = 0$$

explicit method \rightarrow restriction in Δt
 use implicit method \rightarrow take large Δt .

β : parameter to be given a priori

large $\beta \rightarrow$ more like incompressible

(use IE. : - - -)

\downarrow
 it corresponds to SIMP_E
 w/o (under-relaxation)

SIMPLE

$$\left[\begin{aligned} u^{n+1} &= u^* + u' \\ p^{n+1} &= p^* + p' \end{aligned} \right]$$

u^*

of the press-corrector

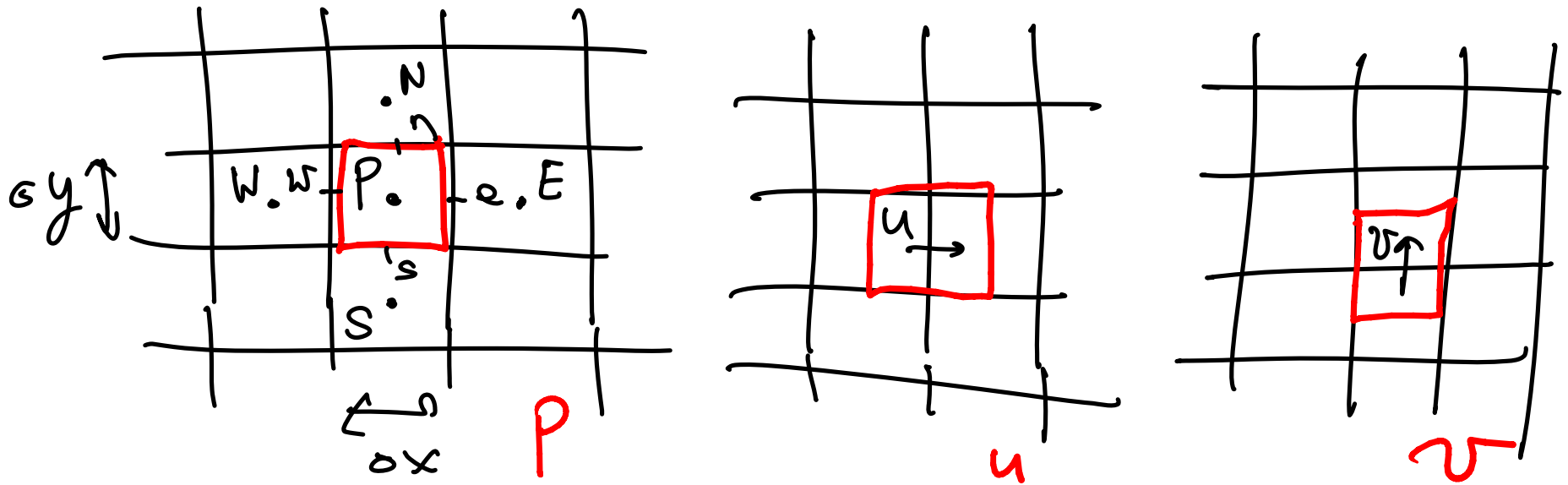
$$u^* = \alpha u_{old}^* + (1-\alpha) u_{new}^*$$

($0 \leq \alpha < 1$)

Kwak et al. (1986) AIAA J. 24, 390.

5. Solution methods for the N-S eqs.

5.1 Implicit scheme using pressure-corrector and a staggered grid (FVM)



$$\int \frac{\partial}{\partial x} (\rho u) dV = O_v + O_e \dots$$

5.2 Treatment of pressure for colocated variables

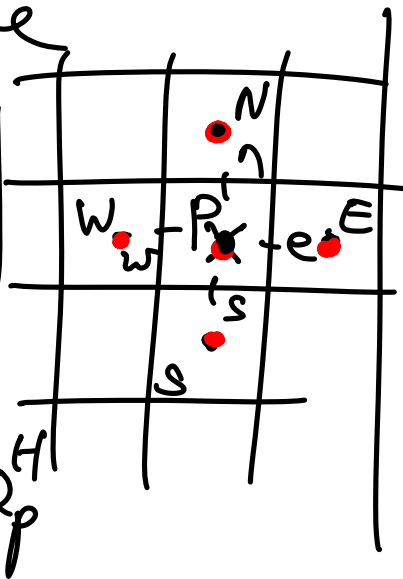
$$\frac{\delta}{\delta x_i} \left(\frac{\delta p^n}{\delta x_i} \right) = \frac{\delta H_i^n}{\delta x_i}, \quad H_i^n = -\frac{\delta}{\delta x_j} (\rho u_i^n u_j^n) + \frac{\delta \tau_{ij}^n}{\delta x_j}$$

\uparrow \uparrow
 grad. operator) should be consistent
 divergence operator) for energy conserving scheme.

e.g. (forward difference for gradient
 backward " for divergence

then, $\frac{1}{\Delta x} \left[\left(\frac{\delta p^n}{\delta x} \right)_p - \left(\frac{\delta p^n}{\delta x} \right)_w \right] + \frac{1}{\Delta y} \left[\left(\frac{\delta p^n}{\delta y} \right)_p - \left(\frac{\delta p^n}{\delta y} \right)_s \right]$

$= \frac{1}{\Delta x} \left[H_{x,p}^n - H_{x,w}^n \right] + \frac{1}{\Delta y} \left[H_{y,p}^n - H_{y,s}^n \right] \equiv Q_p^H$



And then

$$\frac{1}{\Delta x} \left[\frac{1}{\Delta x} \right] \left[(\underline{P_E}^n - \underline{P_P}^n) - (\underline{P_P}^n - \underline{P_W}^n) \right] + \frac{1}{\Delta y} \frac{1}{\Delta y} \left[(\underline{P_N}^n - \underline{P_P}^n) - (\underline{P_P}^n - \underline{P_S}^n) \right] = Q_P^H$$

$$\rightarrow A_P^P P_P^n + \sum_{\ell} A_{\ell}^P P_{\ell}^n = -Q_P^H$$

Same form as the one obtained on a staggered grid w/ central difference.

However, mfm eq : $-\frac{\delta p}{\delta x_i} \rightarrow$ forward difference
 \hookrightarrow 1st order approx.

better to use higher order approx.

Now, CDS for div. & grad. operators. Then

$$\frac{1}{2\Delta x} \left[\left(\frac{\partial p^n}{\partial x} \right)_E - \left(\frac{\partial p^n}{\partial x} \right)_W \right] + \frac{1}{2\Delta y} \left[\left(\frac{\partial p^n}{\partial y} \right)_N - \left(\frac{\partial p^n}{\partial y} \right)_S \right]$$

$$= \frac{1}{2\Delta x} \left[H_{x,E}^n - H_{x,W}^n \right] + \frac{1}{2\Delta y} \left[H_{y,N}^n - H_{y,S}^n \right] \equiv Q_p^H$$

$$\rightarrow \frac{1}{2\Delta x} \frac{1}{2\Delta x} \left[(P_{EE}^n - P_p^n) - (P_p^n - P_{WW}^n) \right]$$

$$+ \frac{1}{2\Delta y} \frac{1}{2\Delta y} \left[(P_{NN}^n - P_p^n) - (P_p^n - P_{SS}^n) \right] = Q_p^H$$

$$\rightarrow A_p^p P_p^n + \sum_l A_l^p P_l^n = -Q_p^H, \quad l = EE, WW, NN, SS$$

This eq. involves nodes which are $2\Delta x$ apart.

→ may create a checkerboard press. distribution.

1	-1	1	-1	1	-1
2	0	2	0	2	0
1	-1	1	-1	1	-1
2	0	2	0	2	0

	N.	P.	E.

Cure?

We could evaluate $\frac{\partial p}{\partial x}|_e$ using CPS $\frac{\partial}{\partial x} \left(\frac{\partial p}{\partial x} \right)$

$$\frac{\partial p^n}{\partial x}|_e \approx \frac{P_E - P_P}{\Delta x}$$

$$\frac{1}{\Delta x} \left(\frac{\partial p^n}{\partial x}|_e - \frac{\partial p^n}{\partial x}|_w \right)$$

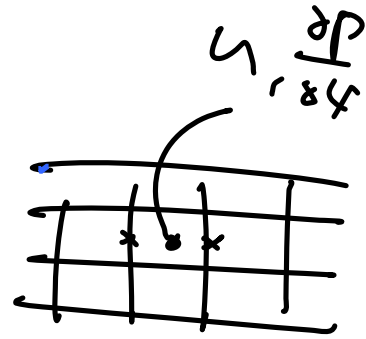
$$\begin{aligned} \Rightarrow \frac{1}{\Delta x^2} (P_E^n - 2P_P^n + P_W^n) + \frac{1}{\Delta y^2} (P_N^n - 2P_P^n + P_S^n) &= Q_P^H \\ &= \frac{1}{\Delta x} (\underline{H_{x,e}^n} - \underline{H_{x,w}^n}) + \frac{1}{\Delta y} (\underline{H_{y,n}^n} - \underline{H_{y,s}^n}) \end{aligned}$$

by interpolation
 this approx. eliminates the oscillation in P

But no energy conservation.

why? \rightarrow mfm: $-\frac{\delta P}{\delta x_i} \rightarrow \frac{\partial P}{\partial x} = \frac{1}{\Delta x} (P_E - P_W)$

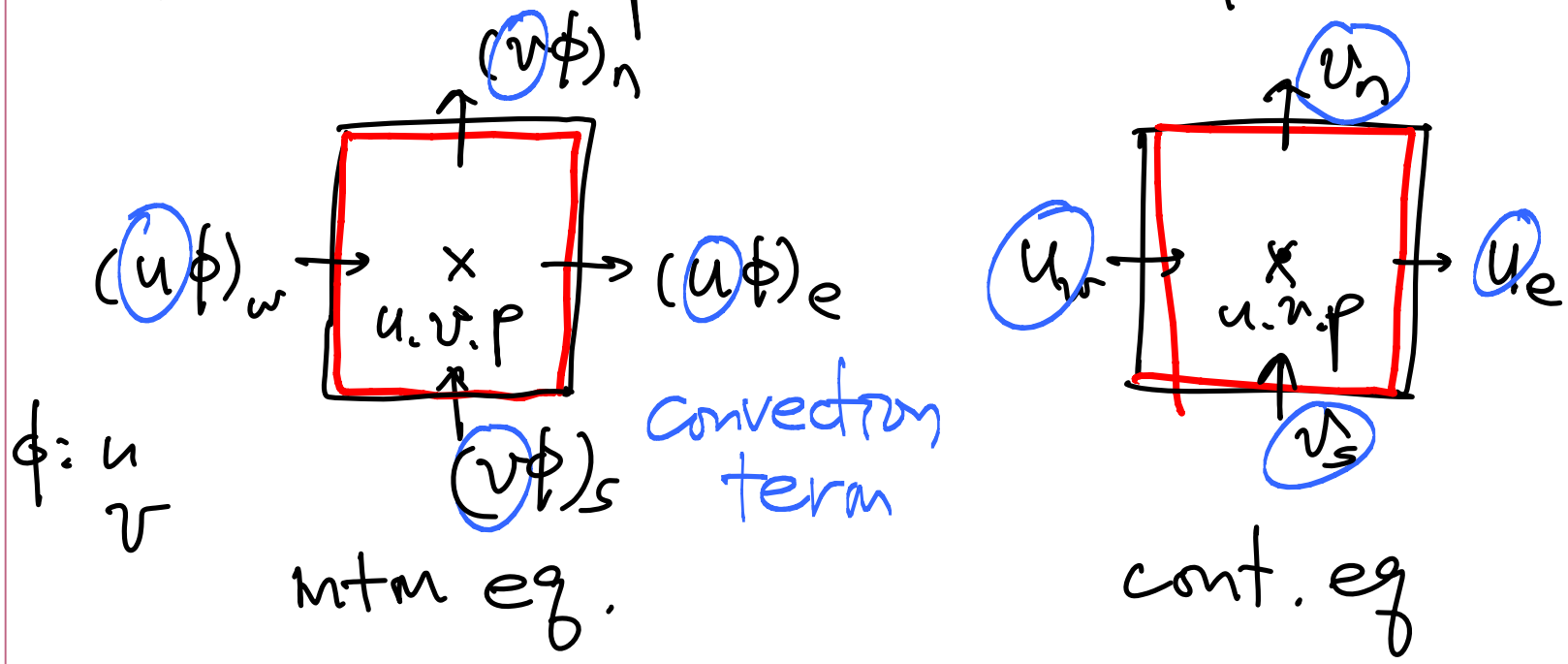
\parallel
 $\frac{1}{2\Delta x} (P_E - P_W)$
 inconsistent.



\downarrow ? \downarrow ?
 undefined

\Rightarrow momentum interpolation method.

- Momentum interpolation method - collocated grid



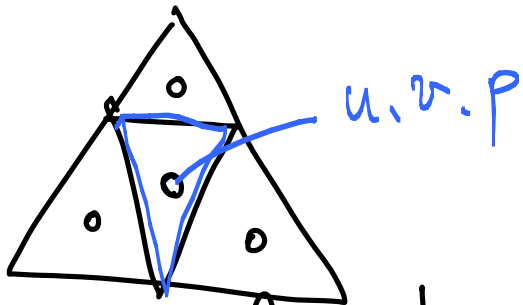
So, in mtm eq., it is important to use u & v at cell faces that satisfy the continuity.

Rhie & Chow (AIAA J. 21, 1525 (1983)) for steady flow

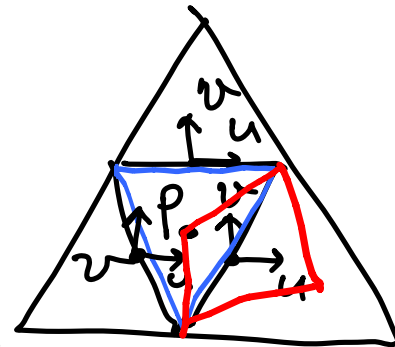
Zang et al. (JCP, 114, 18 (1994)) for unsteady flow
on structured grids

Kim & Choi (JCP, 162, 411 (2000)) for unsteady flow
on unstructured grids

- Kim & Choi's MIM.

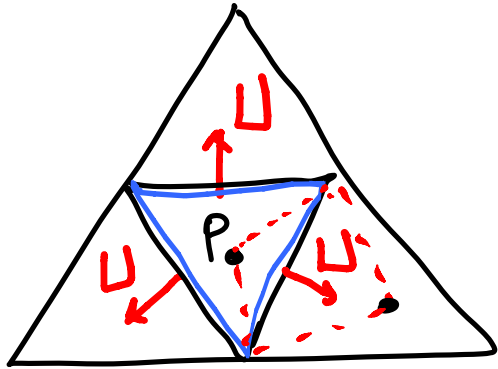


traditional unstructured



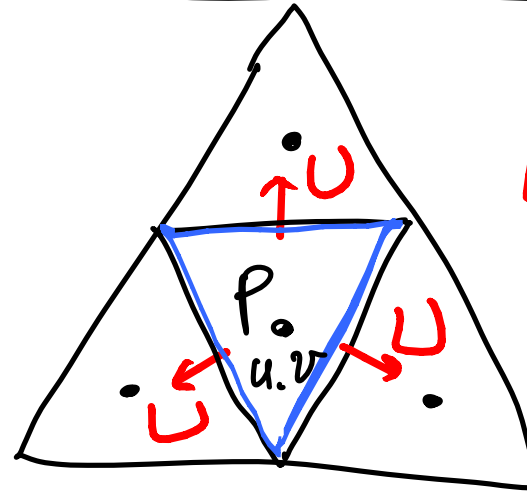
staggered mesh

mesh



staggered mesh
with face-normal
velocity component

with Cartesian
velocity comps.



$$U = (u_i)_{\text{face}} n_i$$

Collocated mesh
w/ face-normal vel. comp

$$U = (u_i)_{\text{face}} n_i$$

$(u_i)_{\text{face}}$: cartesian velocity

n_i : outward-normal unit vector on the cell face

(N-S eq
C-N)

$$\rightarrow \frac{u_i^{\text{NH}} - u_i^{\text{N}}}{\Delta t} + \frac{1}{2} \frac{\partial}{\partial x_j} (u_i^{\text{NH}} u_j^{\text{NH}} + u_i^{\text{N}} u_j^{\text{N}}) = -\frac{\partial p^{\text{NH}}}{\partial x_i} + \frac{1}{2} \frac{1}{\text{Re}} \nabla^2 (u_i^{\text{NH}} + u_i^{\text{N}})$$

nonlinear eq.

linearization

$$u_i^{\text{NH}} u_j^{\text{N}} + u_i^{\text{N}} u_j^{\text{NH}} - u_i^{\text{N}} u_j^{\text{N}} + O(\Delta t^2)$$

Beam & Warming

(AIAA J. 16, 393 (1978))

$$\frac{1}{2} \frac{\partial}{\partial x_j} (u_i^{\text{NH}} u_j^{\text{N}} + u_i^{\text{N}} u_j^{\text{NH}})$$

$$\left[\frac{\partial u_i^{nH}}{\partial x_i} = 0 \right.$$

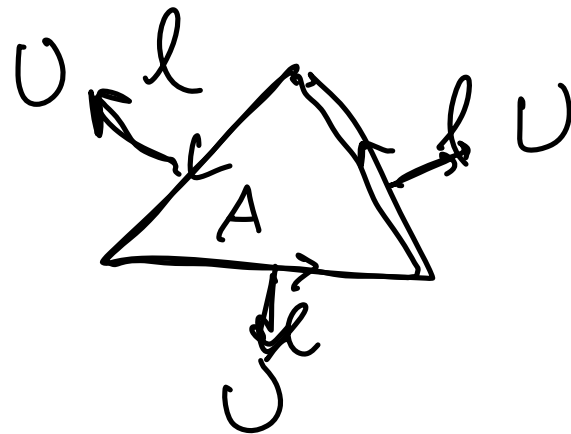
FSM (Choi & Moim) :

$$\frac{\hat{u}_c - u_c^n}{\partial t} + \frac{1}{2} \frac{\partial}{\partial x_j} (\hat{u}_c u_j^n + u_c^n \hat{u}_j) = - \frac{\partial p^n}{\partial x_i} + \frac{1}{2Re} \nabla^2 (\hat{u}_c + u_c^n)$$

$$\frac{u_i^* - \hat{u}_i}{\partial t} = \frac{\partial p^n}{\partial x_i}$$

$$\frac{\partial}{\partial x_i} \frac{\partial p^{nH}}{\partial x_i} = \frac{1}{\partial t} \frac{\partial u_i^*}{\partial x_i}$$

$$\frac{u_i^{nH} - u_i^*}{\partial t} = - \frac{\partial p^{nH}}{\partial x_i}$$



Integrate these eqs over each cell area A
 and apply the divergence theorem: $\delta \hat{u}_i = \hat{u}_c - u_i$

$$\frac{\delta \hat{u}_i}{\delta t} + \frac{1}{A} \oint_{\ell} \frac{1}{2} (\underline{U^n} \delta \hat{u}_i + \underline{u_i^n} n_j \delta \hat{u}_j + \underline{2u_i^n U^n}) dl$$

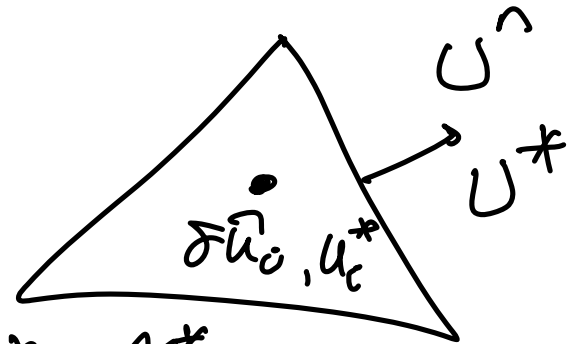
$$= -\frac{1}{A} \int_A \frac{\partial p^n}{\partial x_i} dA + \frac{1}{A} \oint \frac{1}{2} \frac{\partial}{\partial n} (\delta \hat{u}_i + 2u_i^n) dl$$

$$u_i^* - \hat{u}_i = \delta t \frac{\partial p^n}{\partial x_i}$$

$$\frac{1}{A} \oint \frac{\partial p^{n+1}}{\partial n} dl = \frac{1}{A \delta t} \oint U^* dl \rightarrow \text{get } p^n$$

$$u_i^{n+1} - u_i^* = -\delta t \frac{\partial p^{n+1}}{\partial x_i} \rightarrow \text{obtain } u_i^{n+1}$$

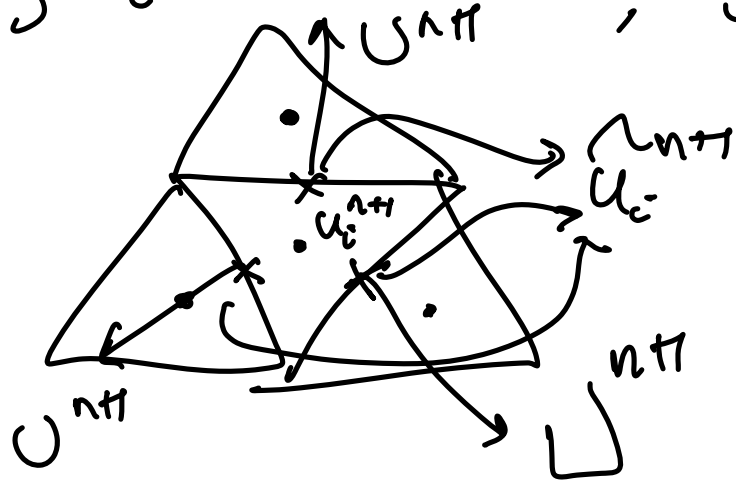
$$U^{n+1} - U^* = -\delta t \frac{\partial p^{n+1}}{\partial n}$$



$$U^* = \underline{\underline{u_i^* n_i}}$$

interpolated from neighboring cell.

$$\rightarrow \oint u_c^{n+1} n_i dl \neq 0, \quad \oint U^{n+1} dl = 0$$



the divergence-free velocity U^{n+1} is used for the calculation of the convective fluxes.

\rightarrow eliminate press. oscillations due to strong coupling bet. U and p .

⊙ Note on pressure and incompressibility

To prove that the press. is a mathematical quantity for the continuity.

$$\exists \underline{v}^* \text{ s.t. } \nabla \cdot \underline{v}^* \neq 0.$$

Want to create \underline{v} s.t. $\nabla \cdot \underline{v} = 0$ & $\underline{v} \approx \underline{v}^*$.

$$\text{Set } \hat{R} = \frac{1}{2} \int_{\Omega} [\underline{v}(\underline{r}) - \underline{v}^*(\underline{r})]^2 d\Omega \quad \left. \vphantom{\hat{R}} \right\} \text{ Find } \underline{v} \text{ minimizing } \hat{R},$$
$$\nabla \cdot \underline{v} = 0$$

→ Calculus of variation

$$R = \frac{1}{2} \int_{\Omega} [\underline{v}(\underline{r}) - \underline{v}^*(\underline{r})]^2 d\Omega - \int_{\Omega} \lambda(\underline{r}) \nabla \cdot \underline{v}(\underline{r}) d\Omega$$

Suppose \underline{u}^{\dagger} s.t. $R_{\min} = \frac{1}{2} \int_{\Omega} [\underline{u}^{\dagger}(\underline{r}) - \underline{u}^*(\underline{r})]^2 d\Omega$ ↙ Lagrange multiplier

$$\& \quad \underline{\nabla} \cdot \underline{u}^{\dagger} = 0$$

Let $\underline{u} = \underline{u}^{\dagger} + \delta \underline{u}$

$$\delta R = R - R_{\min} = \int_{\Omega} \delta \underline{u}(\underline{r}) \cdot [\underline{u}^{\dagger}(\underline{r}) - \underline{u}^*(\underline{r})] d\Omega - \int_{\Omega} \lambda(\underline{r}) \underline{\nabla} \cdot \delta \underline{u}(\underline{r}) d\Omega$$

$$= \int_{\Omega} \delta \underline{u}(\underline{r}) \cdot [\underline{u}^{\dagger}(\underline{r}) - \underline{u}^*(\underline{r}) + \underline{\nabla} \lambda(\underline{r})] d\Omega$$

$$+ \int_{\Sigma} \lambda(\underline{r}) \delta \underline{u}(\underline{r}) \cdot \underline{n} dS$$

↙ make it zero by a natural way or by changing λ .

For arbitrary $\delta \underline{v}$, we need $\delta R = 0$

$$\rightarrow \underline{v}^{\dagger}(\underline{r}) - \underline{v}^*(\underline{r}) + \nabla \lambda(\underline{r}) = 0$$

$$\nabla \cdot (\) \rightarrow \nabla^2 \lambda(\underline{r}) = \nabla \cdot \underline{v}^*(\underline{r})$$

\therefore Lagrange multiplier plays the role of p
and thus the ft. of p is to allow the
continuity to be satisfied.

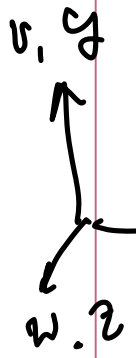
Term project: → Kim² & Choi, JCP (2001), ○, —. this

IB method & code explanation: Friday 1pm - 6pm # 301 - 302

7. Boundary condition for N-S eqs. two b.c's

① wall (x, y, z)

no slip: $u = v = w = 0$

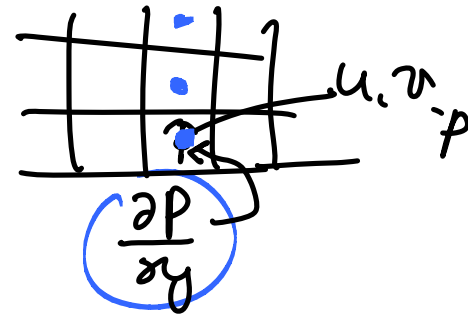


wall → $\frac{\partial u}{\partial x} \Big|_w = 0$ $\frac{\partial w}{\partial z} \Big|_w = 0$ \Rightarrow cont $\frac{\partial v}{\partial y} = 0$

→ One cannot satisfy both b.c.'s simultaneously
 no press. b.c. for staggered mesh

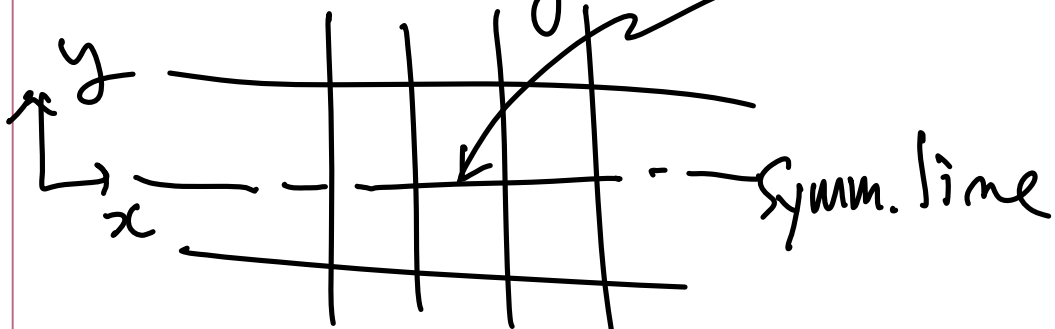
but " " for collocated " "

↳ linear extrapolation

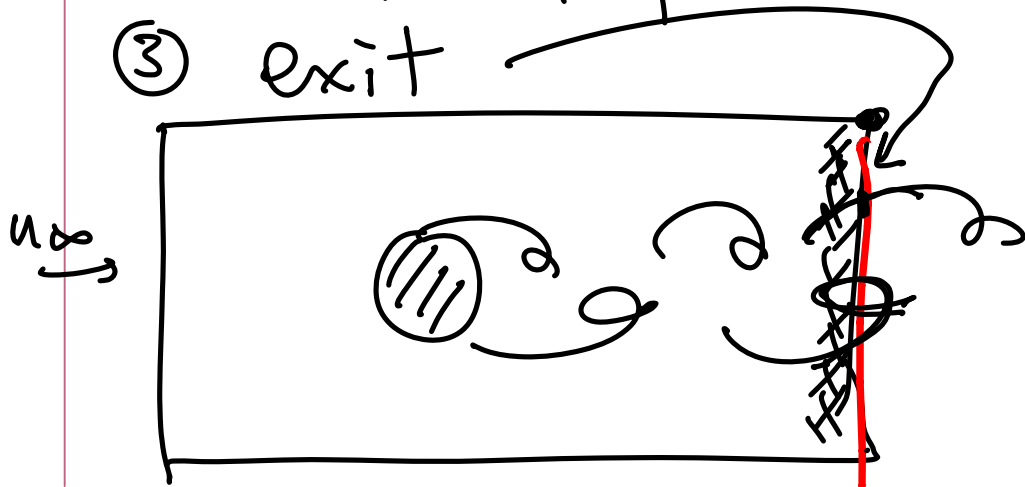


② symmetry

$$v = 0, \quad \frac{\partial u}{\partial y} = 0$$



③ exit



$$\frac{\partial u_i}{\partial x} = 0 \quad \text{Neumann b. c.}$$

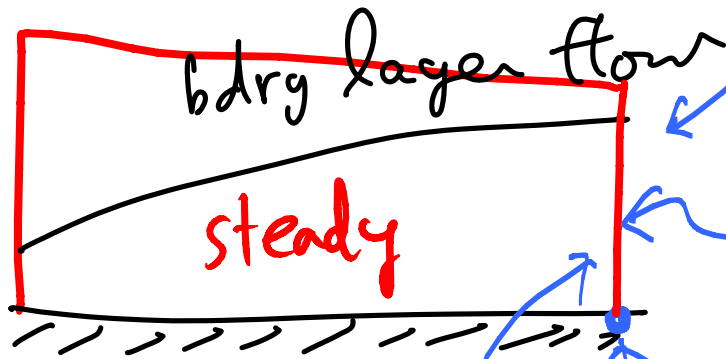
$$\rightarrow \frac{\partial u}{\partial x} = \frac{\partial v}{\partial x} = 0$$

cont. $\hookrightarrow \frac{\partial v}{\partial y} = 0$

two b.c.'s for v

unsteady flow $\rightarrow \frac{\partial u_i}{\partial t} + \underbrace{C}_{\text{local vel } u(y,t)} \frac{\partial u_i}{\partial x} = 0$ Convective outflow b.c.

local vel $u(y,t)$
 mean vel $\bar{u}^y(t) \leftarrow$



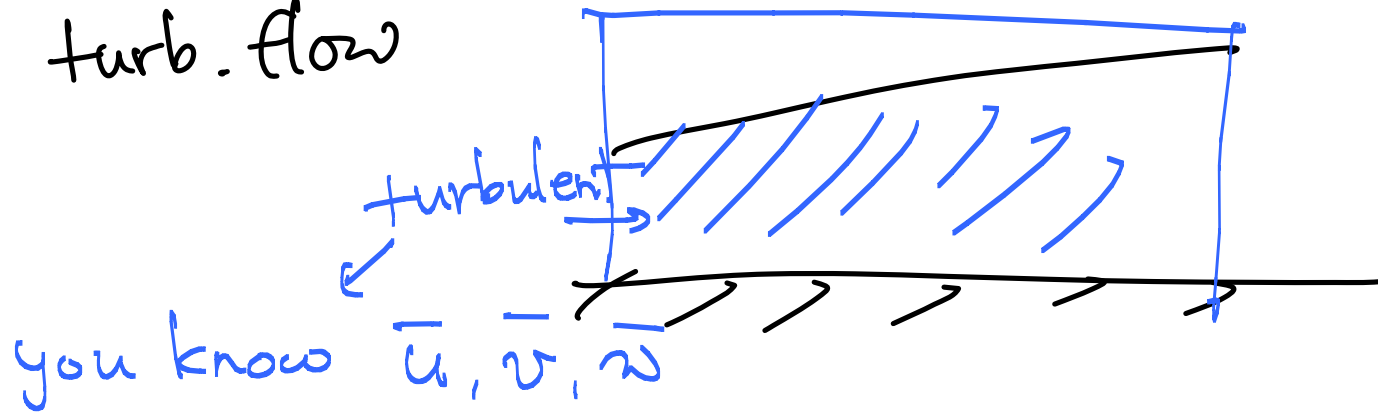
$$\frac{\partial u_z}{\partial x} = 0 \rightarrow \frac{\partial u}{\partial x} = \frac{\partial v}{\partial x} = 0$$

$$\frac{\partial v}{\partial y} = 0$$

$v = 0$ at exit

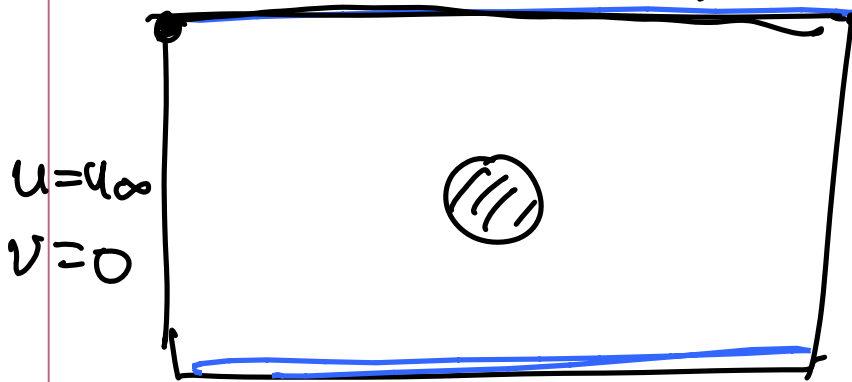
$$\frac{\partial^2 u}{\partial x^2} = \frac{\partial^2 v}{\partial x^2} = 0$$

④ inlet \rightarrow too clear $u = u_\infty, v = 0$
turb. flow



don't " $u', v', w' (y, z, t)$ " \leftarrow Lund et al., JCP, (1998)

⑤ far-field



- $u = u_\infty, v = 0$
 - $\frac{\partial u}{\partial x} = 0$
 - $\frac{\partial v}{\partial y} = 0$
- $u = u_\infty, \frac{\partial u}{\partial y} = 0$
- $\frac{\partial u}{\partial x} = 0, \frac{\partial v}{\partial y} = 0$
 - $\frac{\partial u}{\partial x} = 0$
 - $u = u_\infty$ (inlet)
 - $u = u_\infty$ (far field)

$$\bullet \quad \omega = \frac{\partial u}{\partial x} - \frac{\partial v}{\partial y} = 0, \quad u = u_\infty, \quad \frac{\partial u}{\partial y} = 0$$

• Another formulation of N-S eqs. representation

Kim, Mohan & Moser (1987) JFM

$$\left(\begin{array}{l} \frac{\partial u_i}{\partial t} = -\frac{\partial p}{\partial x_i} + H_i + \frac{1}{Re} \nabla^2 u_i, \quad H_i = -u_j \frac{\partial u_i}{\partial x_j} \\ \frac{\partial u_i}{\partial x_i} = 0 \end{array} \right.$$

vector identity $\nabla^2 \underline{a} = \nabla(\nabla \cdot \underline{a}) - \nabla \times (\nabla \times \underline{a})$

$$(\underline{a} \cdot \nabla) \underline{a} = \frac{1}{2} \nabla(\underline{a} \cdot \underline{a}) - \underline{a} \times (\nabla \times \underline{a})$$

$$\Rightarrow H_i = -(\underline{u} \cdot \nabla) \underline{u} = -\frac{1}{2} \nabla (\underline{u} \cdot \underline{u}) + \underline{u} \times (\nabla \times \underline{u})$$

$$= -\nabla \left(\frac{1}{2} \underline{u} \cdot \underline{u} \right) + \underline{u} \times \underline{\omega}$$

Then, $\frac{\partial \underline{u}}{\partial t} = \underbrace{\underline{u} \times \underline{\omega} - \nabla \left(\frac{1}{2} \underline{u} \cdot \underline{u} \right)}_{\underline{H}} - \nabla p + \frac{1}{\rho_e} \nabla^2 \underline{u}$

$$\frac{\partial}{\partial t} (\nabla^2 \underline{u}) = \nabla^2 \underline{H} - \nabla^2 (\nabla p) + \frac{1}{\rho_e} \nabla^4 \underline{u}$$

$$= \nabla (\nabla \cdot \underline{H}) - \nabla \times (\nabla \times \underline{H}) - \nabla^2 (\nabla p) + \frac{1}{\rho_e} \nabla^4 \underline{u}$$

$$= -\nabla \times (\nabla \times \underline{H}) + \frac{1}{\rho_e} \nabla^4 \underline{u} + \nabla (\nabla \cdot \underline{H} - \nabla^2 p)$$

$$\nabla \cdot (\text{NS}) \rightarrow \frac{\partial}{\partial t} (\nabla \cdot \underline{u}) = -\nabla^2 p + \nabla \cdot \underline{H} + \frac{1}{\rho_e} \nabla^2 (\nabla \cdot \underline{u})$$

$$\rightarrow \frac{\partial}{\partial t} (\nabla^2 \underline{u}) = \underbrace{-\nabla \times (\nabla \times \underline{H})}_{\equiv h_u} + \frac{1}{Re} \nabla^4 \underline{u}$$

eliminate p
but w/ higher derivatives

v component:

$$\frac{\partial}{\partial t} (\nabla^2 v) = h_v + \frac{1}{Re} \nabla^4 v$$

4th-order PDE
→ needs high accurate numerical method

① spectral method

where $h_v = -\frac{\partial}{\partial y} \left(\frac{\partial H_1}{\partial x} + \frac{\partial H_3}{\partial z} \right) + \left(\frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial z^2} \right) H_2$

$\nabla^2 \phi$
 $\rightarrow -k^2 \phi$
 $\nabla^4 \phi \rightarrow -k^4 \phi$

Also, $\frac{\partial}{\partial z} \left(\frac{\partial u}{\partial t} \right) - \frac{\partial}{\partial x} \left(\frac{\partial w}{\partial t} \right) :$

$$\frac{\partial}{\partial t} g = h_g + \frac{1}{Re} \nabla^2 g$$

②

$$g = \frac{\partial u}{\partial z} - \frac{\partial w}{\partial x} = \omega_g$$

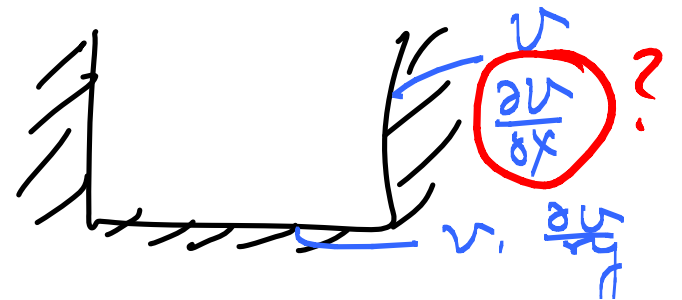
$$h_g = \frac{\partial H_1}{\partial z} - \frac{\partial H_3}{\partial x}$$

③

Also, $f \equiv \frac{\partial v}{\partial x} + \frac{\partial w}{\partial z} \rightarrow \boxed{f + \frac{\partial v}{\partial y} = 0} - \textcircled{4}$

- i) solve Eq. ① to get $v \rightarrow$ b.c.'s : $v|_w = \frac{\partial v}{\partial y}|_w = 0$
- ii) solve Eq. ② to get $g \rightarrow$ " : $g|_w = 0$
- iii) solve Eq. ③ & ④ to get u & w .

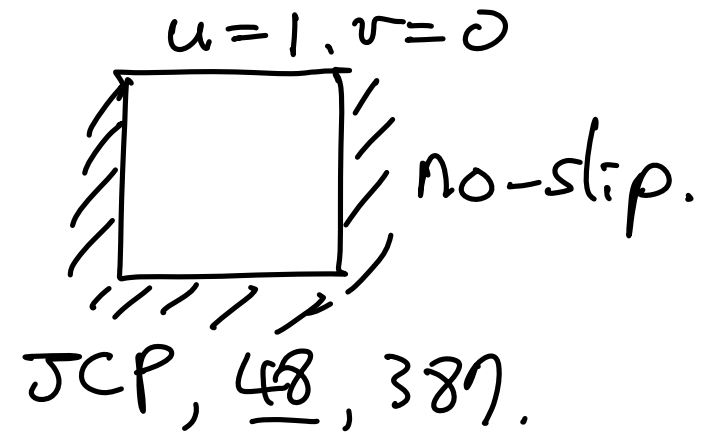
\rightarrow this formulation has limit in that b.c.'s are not well defined in most geometries except channel flow



8. Examples

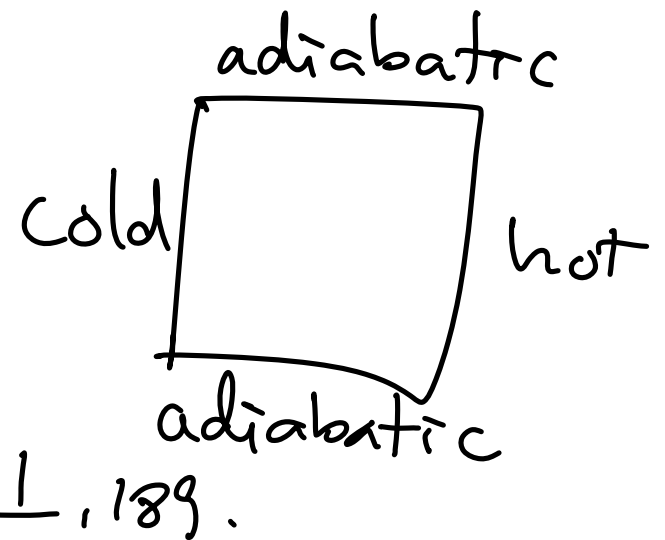
① Lid-driven cavity flow

Ghia, Ghia & Shin (1982)



② Buoyancy-driven cavity flow

Hortmann, Peric & Scheuerer
(1990)



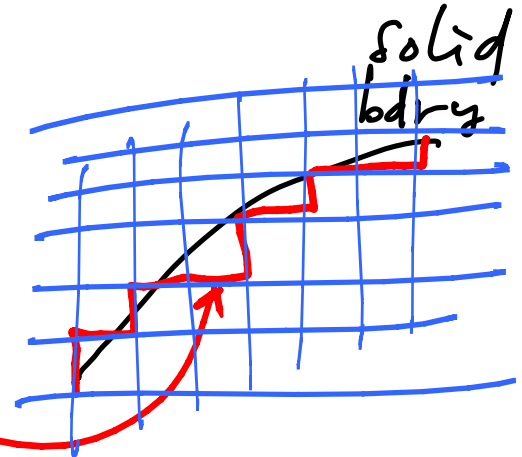
Int. J. Num. Methods Fluids 11, 189.

Ch. 8 Complex geometries

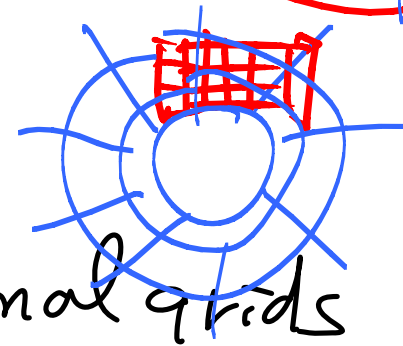
1. Choice of grid

① stepwise approx. using regular grids

provide b.c.'s here

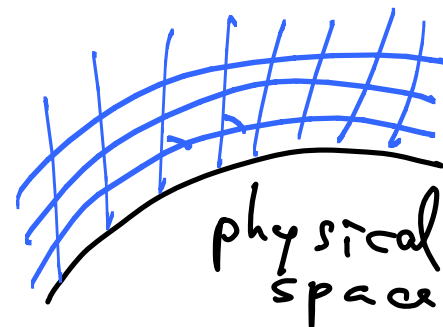


② overlapping grids
ex) chimera grids



③ boundary-fitted non-orthogonal grids

advantage: can be adapted to any geometry



good for b.c.

grid lines follow the streamlines
can

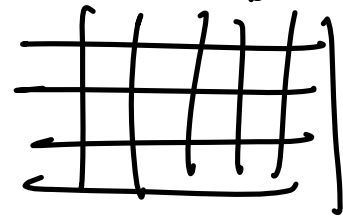
transform

disadvantage:

transformed eqs contain more terms

difficulty of programming

increases cost of solving eqs.



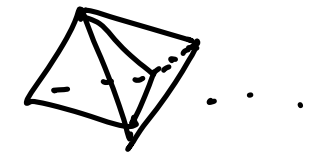
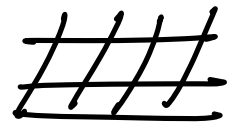
2. Grid generation - very important issue

Thompson et al. (1985)

① make grids as nearly orthogonal as possible

② cell topology

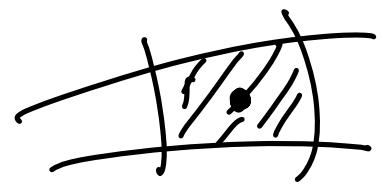
in general, quadrilaterals in 2D are better
hexahedra in 3D
than triangles in 2D
tetrahedra in 3D



if

- i) midpoint rule integral approx.
- linear interpolation
- central difference

are used



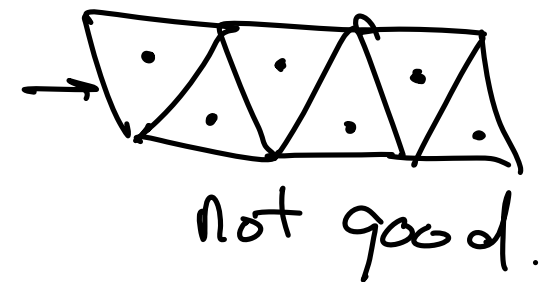
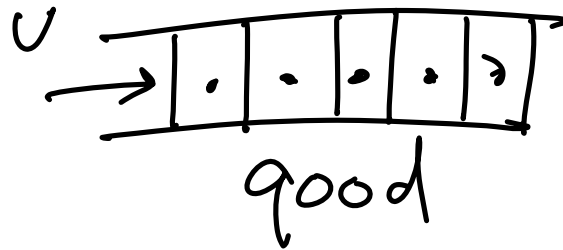
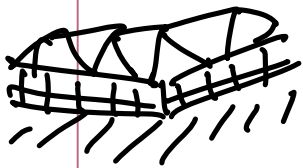
ii) very near the wall



each cell has a same role.

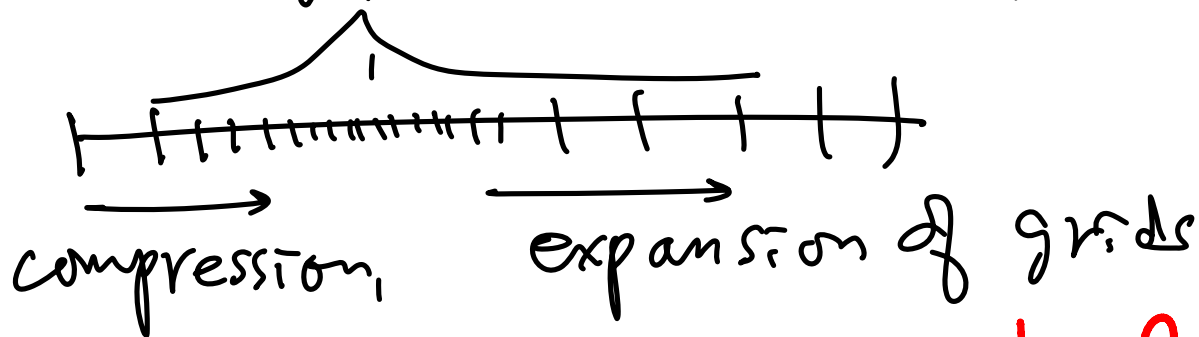
iii) for convection terms

(accuracy is improved if grid lines follow the streamlines)



③ non-uniform grids

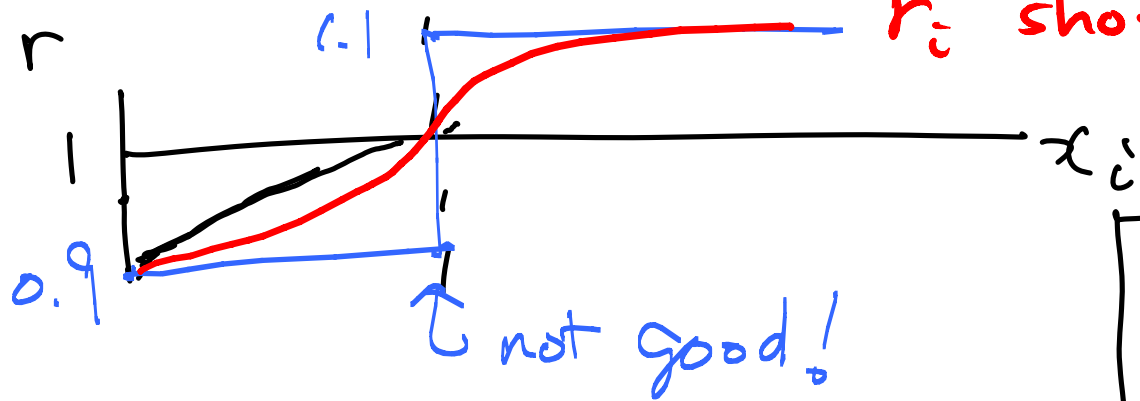
more grids near the region of rapid variation



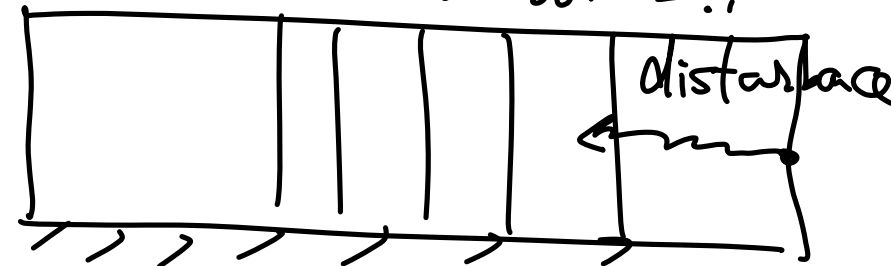
$$r_i = \frac{\Delta x_{i+1}}{\Delta x_i}$$

ratio

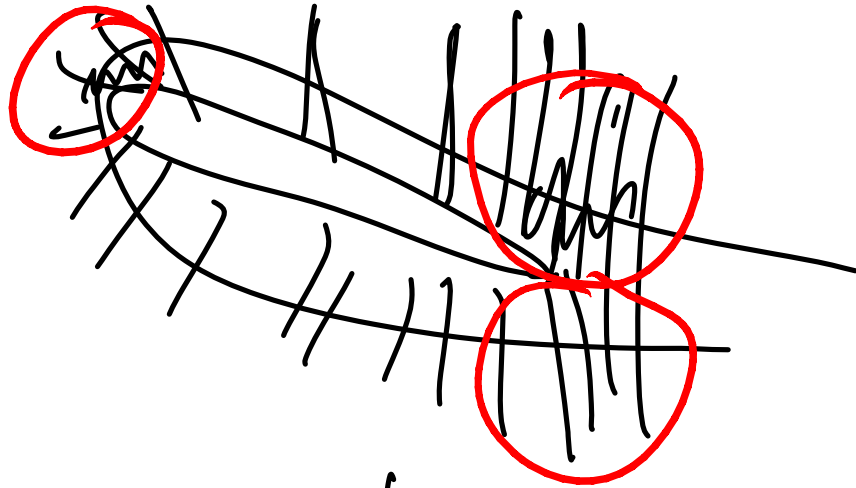
r_i should be kept under control



wiggles!!

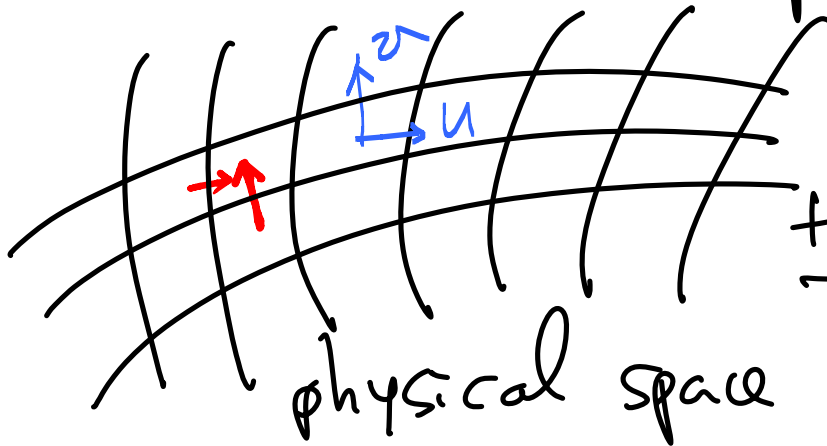


Hahn & Choi (JCP, 1997?) $r_i \uparrow$ expansion \rightarrow

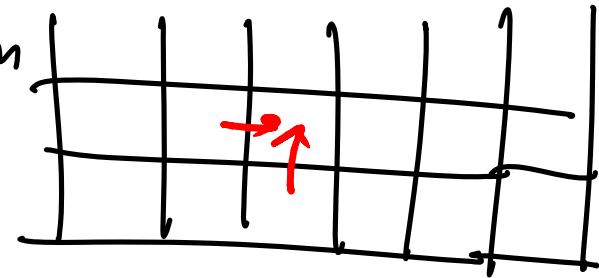


3. Choice of velocity components

① Grid-oriented vel. components satisfy the continuity very well.



transform
→

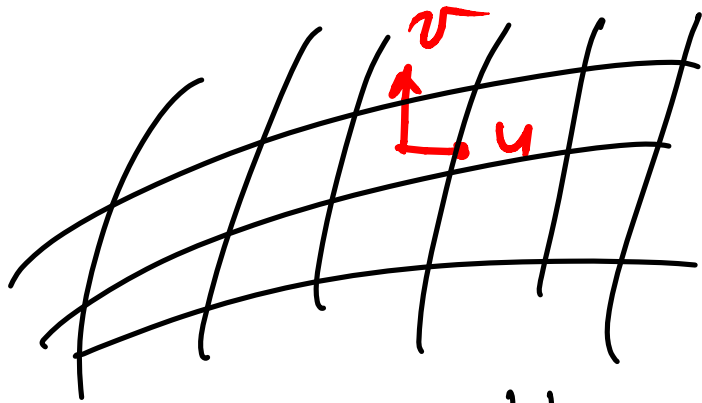


contravariant vel.

very complicated.

Choi, Moin & Kim (1993,
JFM)
(1D Cartesian
2D generalized coord.)

② Cartesian vel comps.



(unstructured
generalized body-fitted.)

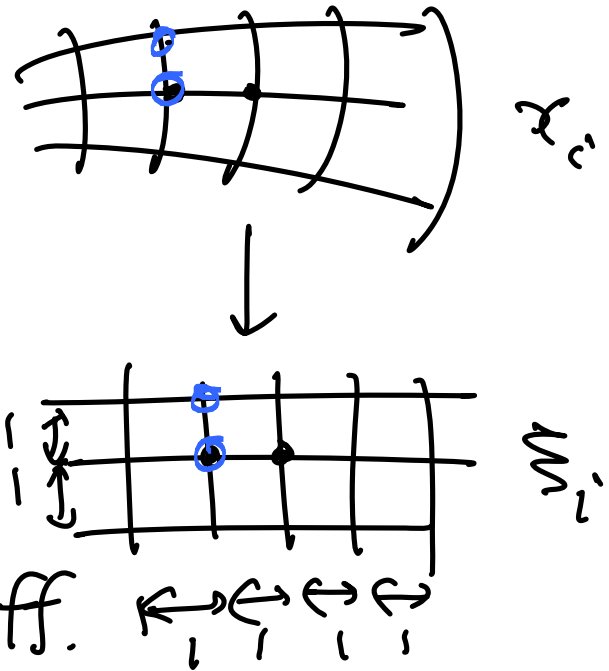
4. choice of variable arrangement
① staggered ② Collocated

5. FDM \rightarrow coordinate transformation

$$x_i \rightarrow \xi_i$$

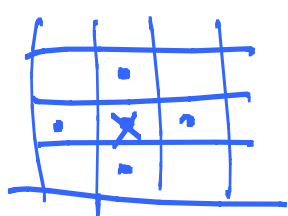
$$\frac{\partial \phi}{\partial x_i} = \frac{\partial \phi}{\partial \xi_j} \frac{\partial \xi_j}{\partial x_i}$$

\uparrow called metric coeff. $\leftarrow \rightarrow \rightarrow \rightarrow$



$$J = \det \left(\frac{\partial x_i}{\partial \xi_j} \right) \text{ Jacobian}$$

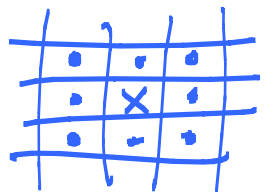
• Transformation usually provides cross derivative terms



$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2}$$

$$\rightarrow \frac{\partial^2 \phi}{\partial \xi^2}, \frac{\partial^2 \phi}{\partial \eta^2},$$

$$\frac{\partial^2 \phi}{\partial \xi \partial \eta}$$

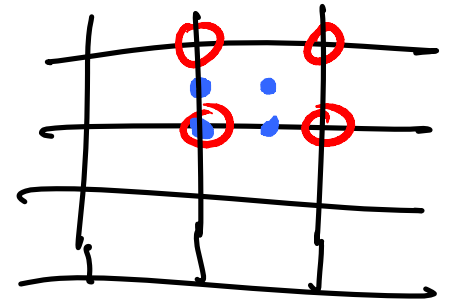


ADI? \leftarrow "0" for orthogonal grids
 No! $\neq 0$ for non-orthogonal grids

- need to store metric coeffs.
- easy to discretize in the transformed space.
- no interpolation of metric coeffs.

\rightarrow interpolation breaks conservation.

(See Thompson et al. 1985)



- skewness (non-orthogonal grid) and large aspect ratios give poor convergence or oscillations in the solution


skewed grid


large aspect ratio

6. FVM

$$\frac{\partial}{\partial t} \int_{\Omega} \rho \phi d\Omega + \int_{\Omega} \rho \phi \underline{v} \cdot \underline{n} dS = \int_{\Omega} \Gamma \nabla \phi \cdot \underline{n} dS + \int_{\Omega} q_{\phi} d\Omega$$

• no need to transform

• need to get ϕ or $\nabla \phi$ at cell surface
by interpolation or some other methods.

7. FEM — natural method for complex geometries

8. Pressure-correction eq. — covered already

9. Axis-symmetric problems — (r, θ , z)

centrifugal force $-\underline{\underline{\Omega}} \times \underline{\underline{\Omega}} \times \underline{\underline{x}}$

Coriolis force $-2 \underline{\underline{\Omega}} \times \underline{\underline{u}}_r$

appear in the mfm eq.

N-S eqs

$-\underline{\underline{\Omega}} \times \underline{\underline{\Omega}} \times \underline{\underline{x}} - 2 \underline{\underline{\Omega}} \times \underline{\underline{u}}_r$

non-conservative form

transform to fully conservative form

Kim & Choi (JCP, —) ←

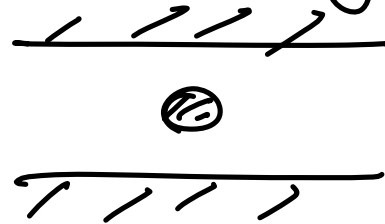
D. Kim

see the reference therein.

10. Implementation of boundary conditions

11. Examples

flow around a circular cylinder in a channel.



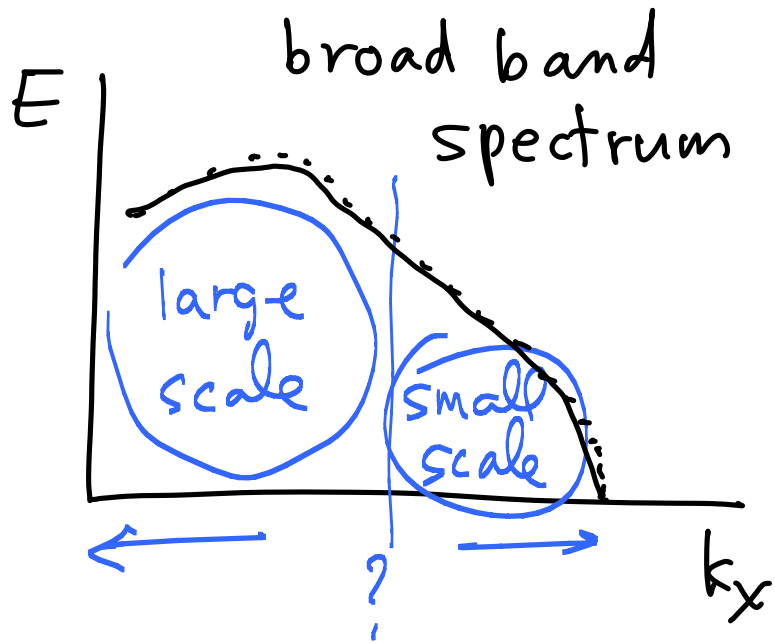
Ch. 9. Turbulent Flows

↳ are unsteady 3-D.

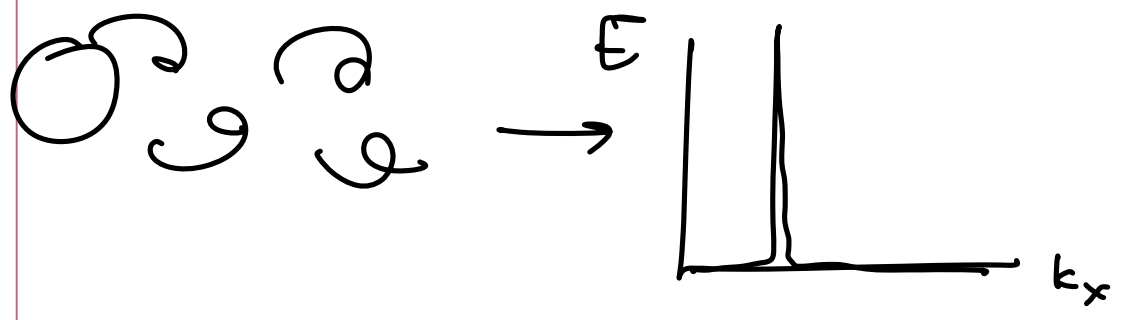
contain a variety of spatial and temporal scales from large to small motions.



⇒ makes the prediction of turbulent flow very difficult.



$$\begin{array}{c}
 u(x) \xrightarrow{\text{FT}} \hat{u}(k_x) \\
 \downarrow \text{Taylor hypothesis} \\
 \hat{u}(\omega) \xrightarrow{\frac{\partial}{\partial x} \rightarrow c \frac{\partial}{\partial x}} \underline{E(k_x)} = \hat{u} \hat{u}^* \\
 \downarrow \\
 \underline{E(\omega)} = \hat{u} \hat{u}^*
 \end{array}$$



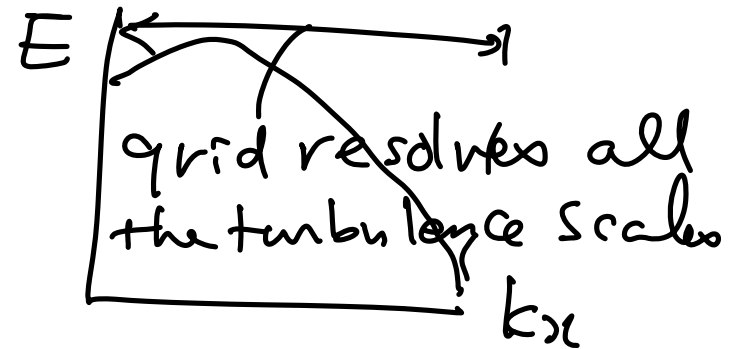
• Prediction methods for turbulent flow

- time ↓
- ① RANS (Reynolds averaged Navier-Stokes eq.) technique
 - ② LES (Large Eddy Simulation)
 - ③ DNS (Direct Numerical Simulation)
 - ④ Wall-modeled LES or Hybrid LES/RANS

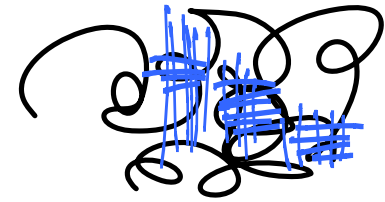
Ⓐ Direct numerical simulation - no model for turbulence

$$\left\{ \begin{array}{l} \frac{\partial u_i}{\partial x_i} = 0 \\ \frac{\partial u_i}{\partial t} + \frac{\partial u_i u_j}{\partial x_j} = -\frac{1}{\rho} \frac{\partial p}{\partial x_i} + \nu \frac{\partial^2 u_i}{\partial x_j \partial x_j} \end{array} \right.$$

unsteady and 3D.



- grid resolves all the turbulent scales \rightarrow no model
- highly accurate
- Since 1980's



- Number of grid points requirement
 smallest scale in turbulence — Kolmogorov $\sqrt{\text{length scale}}$
 DNS should resolve the Kolmogorov scale.

grid resolution \uparrow

$$\Delta \sim \eta = \left(\frac{\nu^3}{\epsilon} \right)^{\frac{1}{4}}$$

ν : kinematic viscosity
 ϵ : dissipation rate

$\Delta \sim O(1) \eta \sim (\sim 10) \eta$
 in real simulation



N
 (number of grid pts
 in 1-direction)

l : largest turbulent length scale
 u : " " " velocity scale

$$N = \frac{l}{\Delta} \doteq \frac{l}{\eta} = l \left(\frac{\epsilon}{\nu^3} \right)^{\frac{1}{4}} \sim l \left(\frac{P}{\nu^3} \right)^{\frac{1}{4}}$$

$$\left(P \sim -\overline{u_i u_j} \frac{\partial \overline{u_i}}{\partial x_j} \sim \frac{u^3}{l} \right)$$

$$\left(Re_l = \frac{ul}{\nu} \right)$$

$$Re_l^{\frac{3}{4}} = \left(\frac{u^3 l^3}{\nu^3} \right)^{\frac{1}{4}} = l \left(\frac{u^3}{\nu^3} \right)^{\frac{1}{4}}$$

total # of grid pts
 in 3D

$$N^3 = Re_l^{\frac{9}{4}}$$

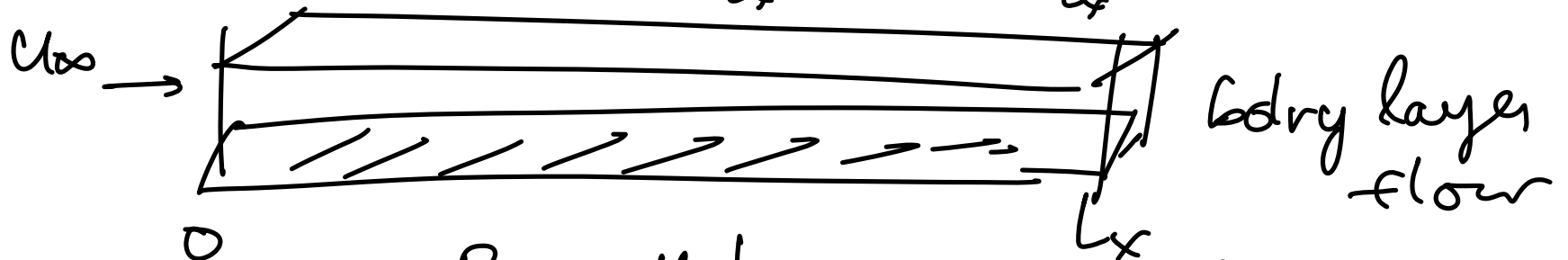
\Rightarrow computationally
 expensive
 |

but very useful for
academic purpose.

rarely used for eng.
applications.

Choi & Moim (2012, PoF)

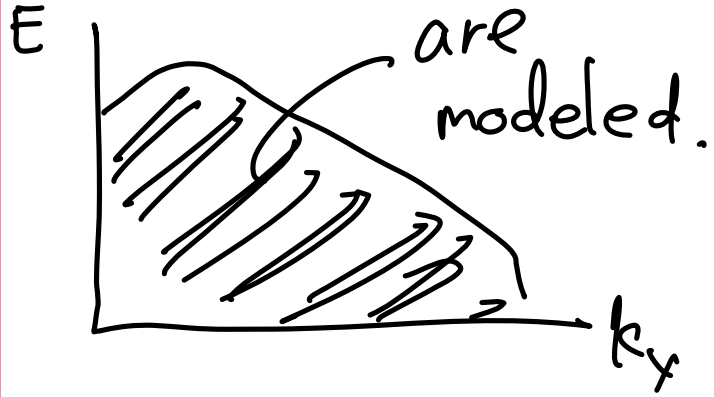
$$N^3 \sim Re_{L_x}^{\frac{37}{14}} \sim Re_{\delta_{L_x}}^{\frac{37}{12}} \sim Re_{\tau_{L_x}}^{\frac{37}{11}}$$



$$Re_{L_x} = \frac{U_{\infty} L_x}{\nu}, \quad Re_{\delta_{L_x}} = \frac{U_{\infty} \delta_{L_x}}{\nu}$$

$$Re_{\tau_{L_x}} = \frac{U_{\tau_{L_x}} L_x}{\nu}$$

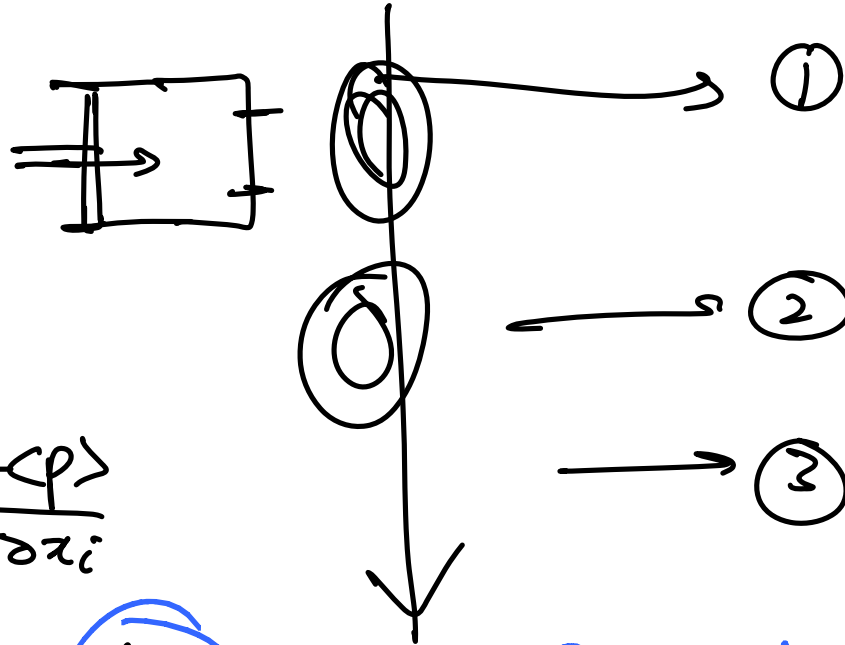
⑧ RANS



ensemble averaging
 Reynolds decomposition

$$u_i = \langle u_i \rangle + u_i'$$

$$= \overline{u_i} + u_i'$$



$$\frac{\partial \langle u_i \rangle}{\partial x_i} = 0$$

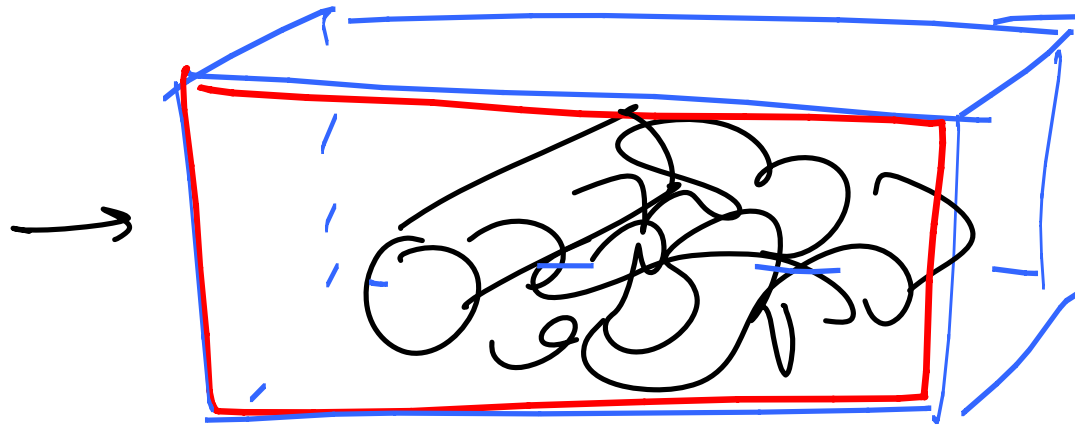
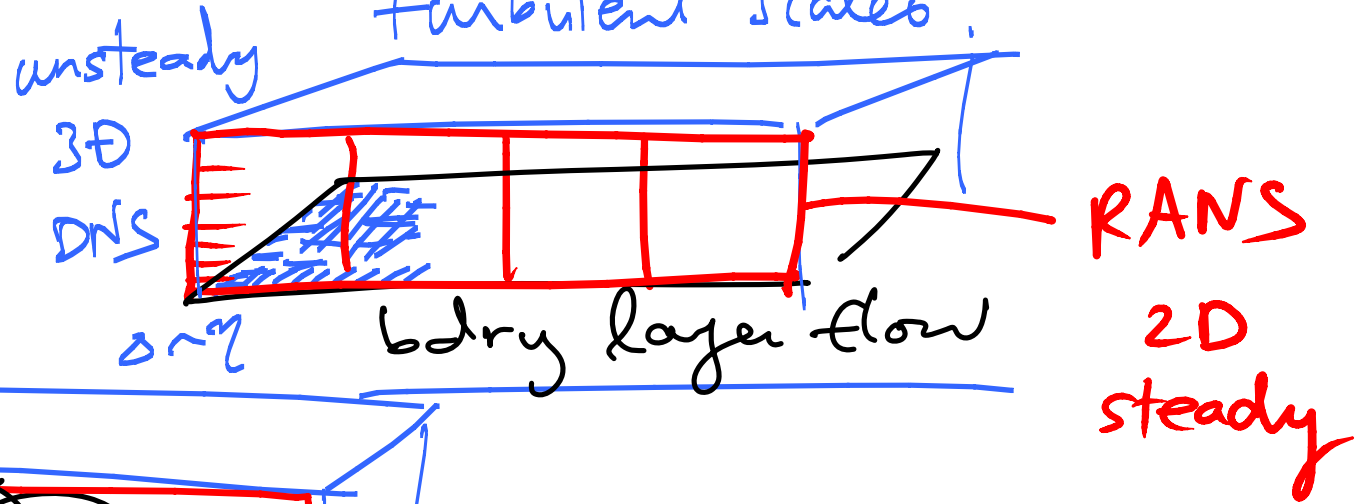
$$\frac{\partial \langle u_i \rangle}{\partial t} + \frac{\partial \langle u_i u_j \rangle}{\partial x_j} = -\frac{1}{\rho} \frac{\partial \langle p \rangle}{\partial x_i}$$

$$+ \nu \frac{\partial^2 \langle u_i \rangle}{\partial x_j \partial x_j} - \frac{\partial}{\partial x_j} \langle u_i' u_j' \rangle$$

Reynolds stress

- low cost!

is obtained by modeling whole turbulent scales.



- inaccurate for the prediction of massively separated flow or flow w/ high curvature.

- How to model the Reynolds stress term $\langle u_i' u_j' \rangle$?

Boussinesq eddy viscosity hypothesis

in laminar flow,

$$\tau_{ij} - \frac{1}{3} \delta_{ij} \tau_{kk} = \underbrace{\rho}_{\substack{\text{ft. of} \\ \text{fluid}}} 2\nu S_{ij}$$

$$S_{ij} = \frac{1}{2} \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right)$$

$$\rightarrow -\langle u_i' u_j' \rangle + \frac{2}{3} k \delta_{ij} = 2\nu_T \langle S_{ij} \rangle$$

$$k = \frac{1}{2} \langle u_i u_i \rangle$$

eddy viscosity
ft. of flow

$$\frac{\partial}{\partial x_j} (\tau_{ij} - \rho \langle u_i' u_j' \rangle)$$

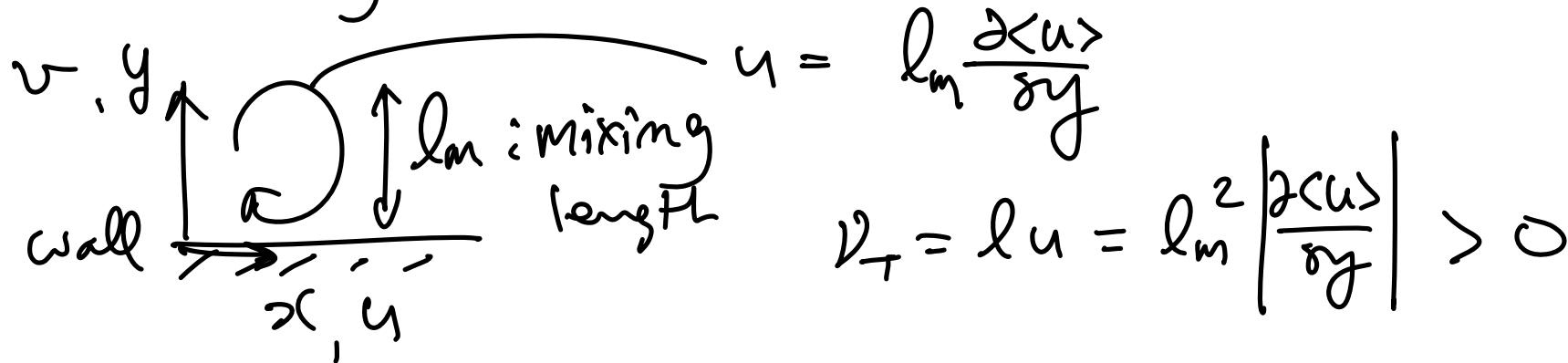
• Eddy viscosity ν_T (m^2/s) [$L^2 T^{-1}$]

$\nu_T \sim l$ (length scale) $\times u$ (velocity scale)

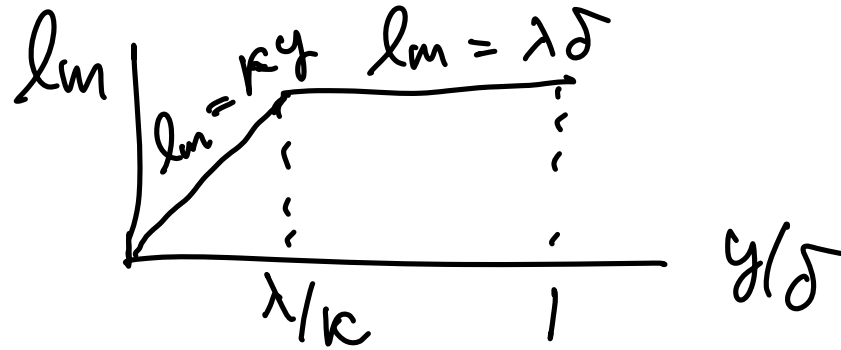
What are the most relevant length and vel. scales?

① zero-equation model: no transport eq.

mixing length model (Prandtl, 1925)



bdry layer flow



$$\kappa = 0.425$$

$$\lambda = 0.09$$

② One-equation model : one transport eq.

ex) k-eg : modeling of the velocity scale only

$$u \sim \sqrt{k}, \quad \nu_T = \zeta \mu' \sqrt{k} l$$

$$\frac{\partial k}{\partial t} + \langle u_j \rangle \frac{\partial k}{\partial x_j} = \frac{\partial}{\partial x_j} \left(\frac{\nu_T}{\sigma_k} \frac{\partial k}{\partial x_j} \right) + \nu_T \left(\frac{\partial \langle u_i \rangle}{\partial x_j} + \frac{\partial \langle u_j \rangle}{\partial x_i} \right) \frac{\partial \langle u_i \rangle}{\partial x_j} - C_D \frac{k^{3/2}}{l}$$

$l = l_m \rightarrow$ limitation!

ex) Spalart-Allmaras model

$$\nu_T = \tilde{\nu} f_{\nu_1}$$

$$\frac{D\hat{\nu}}{Dt} = C_{b1} \tilde{S} \tilde{\nu} + \frac{1}{\sigma} \left\{ \nabla \cdot [(\nu + \hat{\nu}) \nabla \hat{\nu}] + C_{b2} (\nabla \tilde{\nu})^2 \right\} - C_{w1} f_w \left[\frac{\tilde{\nu}}{d} \right]^2$$

d : wall distance

$$f_{\nu_1} = \frac{\kappa^3}{\kappa^3 + C_{\nu_1}^3}, \quad \kappa = \frac{\tilde{\nu}^2}{\nu}$$

$$v_t = v \times l$$

- Two equation model : two transport eqs

$$u \sim \sqrt{k}, \quad l? \rightarrow \varepsilon = 2\nu \langle s_{ij} s_{ij} \rangle \sim \frac{k^{3/2}}{l}$$

$$\nu_T = C_\mu \frac{k^2}{\varepsilon}$$

$$s_{ij} = \frac{1}{2} \left(\frac{\partial u'_i}{\partial x_j} + \frac{\partial u'_j}{\partial x_i} \right)$$

$$k\text{-eq} : \frac{\partial k}{\partial t} + \langle u_j \rangle \frac{\partial k}{\partial x_j} = \frac{\partial}{\partial x_j} \left(\frac{\nu_t}{\sigma_k} \frac{\partial k}{\partial x_j} \right) + \nu_t \left(\frac{\partial \langle u_i \rangle}{\partial x_j} + \frac{\partial \langle u_j \rangle}{\partial x_i} \right) \frac{\partial \langle u_i \rangle}{\partial x_j} - \varepsilon$$

$$\varepsilon\text{-eq} : \frac{\partial \varepsilon}{\partial t} + \langle u_j \rangle \frac{\partial \varepsilon}{\partial x_j} = \frac{\partial}{\partial x_j} \left(\frac{\nu_t}{\sigma_\varepsilon} \frac{\partial \varepsilon}{\partial x_j} \right) + C_{1\varepsilon} \frac{\varepsilon}{k} \nu_t \left(\frac{\partial \langle u_i \rangle}{\partial x_j} + \frac{\partial \langle u_j \rangle}{\partial x_i} \right)$$

b.c's : high Re # model : $\leftarrow C_{2\varepsilon} \frac{\varepsilon^2}{k}$

DNS - low Re #

low Re # model : $k = 0$

$$\frac{\partial \varepsilon}{\partial n} = 0$$

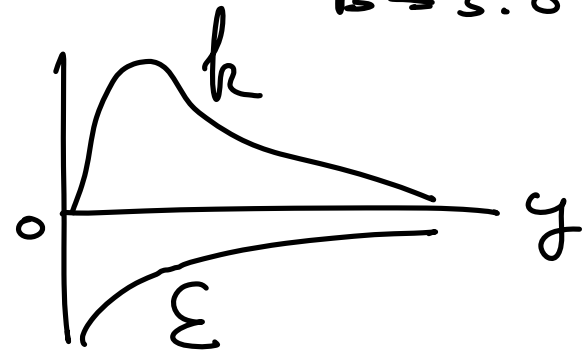
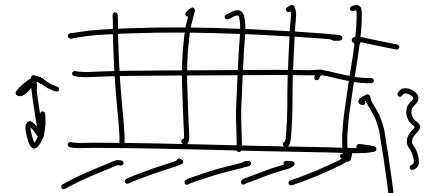
$$\frac{\partial k}{\partial n} = 0$$

$$\varepsilon = C_{\mu}^{1/4} k^{3/4} / (k y)$$

$$\tau_w = \rho C_{\mu}^{1/4} k \sqrt{k} \bar{u} / \ln(y^+ \varepsilon)$$

$$\varepsilon = e^{KB}$$

$$B = 5.0$$



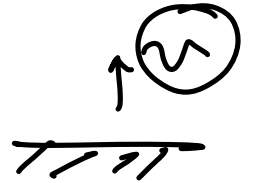
Other turbulence models

✓ SST $k-\omega$ model (Menter, 1993 or 1994)

$k-\varepsilon-v^2-f$ model (Durbin, —)

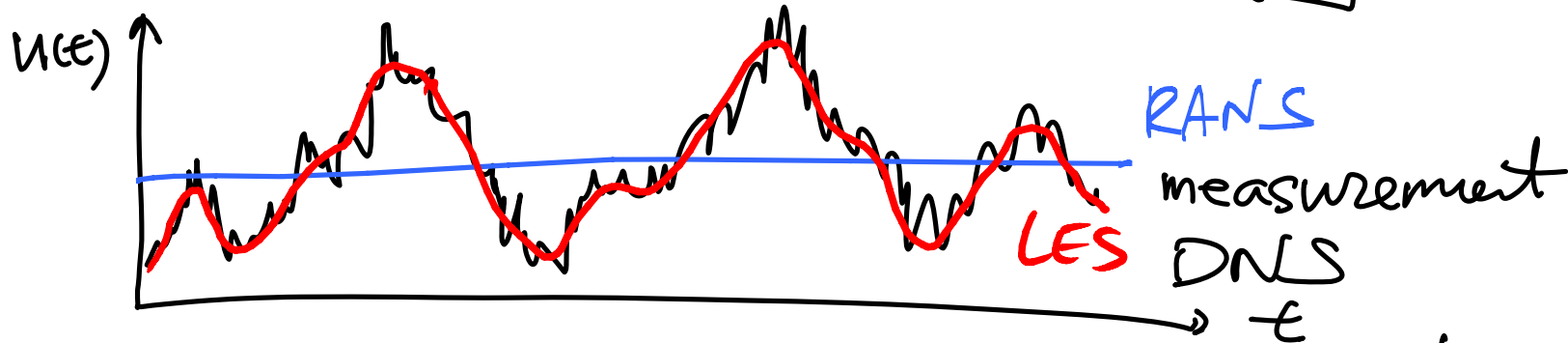
Reynolds stress model (Launder, Reece & Rodi, —)

$$\overline{uu}, \overline{vw}, \overline{v^2}, \overline{uv}, \overline{uw}, \overline{vw}$$



⇒ RANS models have been widely used for the prediction of turbulent flows in engineering applications. RANS, however, fails to predict massively separated flow and flows with high curvatures

⑤ Large eddy simulation



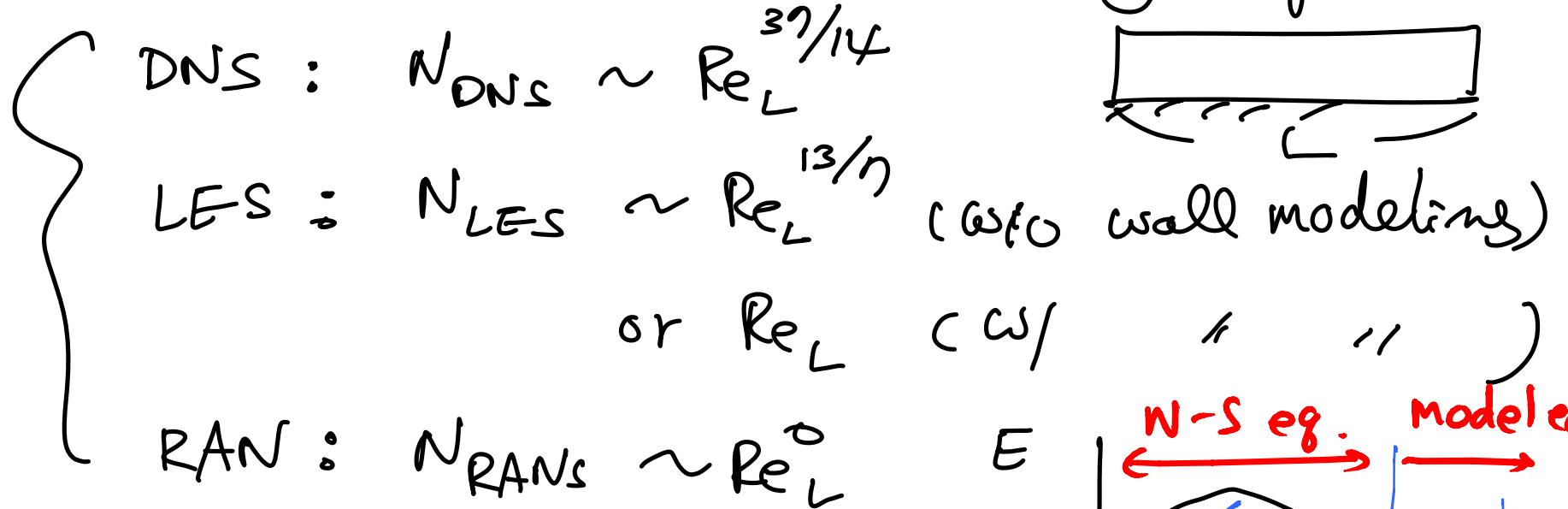
The premise of LES is that the motions that are resolved are the dynamically important ones and the errors introduced by modeling the small scale motions are significantly smaller than those incurred in RANS where the entire turbulence stresses are modeled.

Kim, M M (1987)

LES - unsteady & 3D simulation

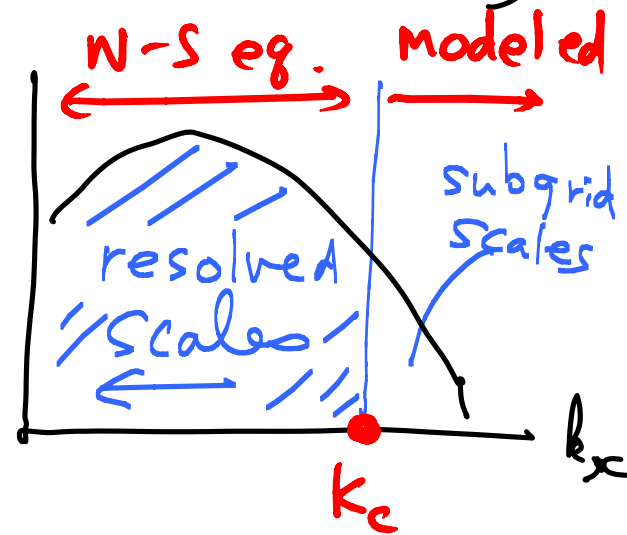
Choi & Moin (2012, POF)

turb. boundary layer

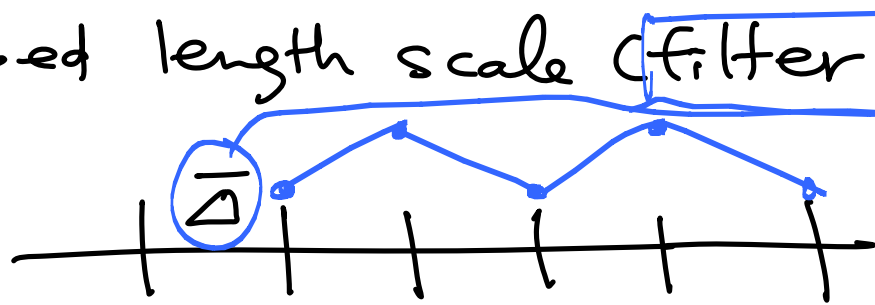


Filtering:

an operation which damps out all the spatial fluctuations of flow variables smaller than a



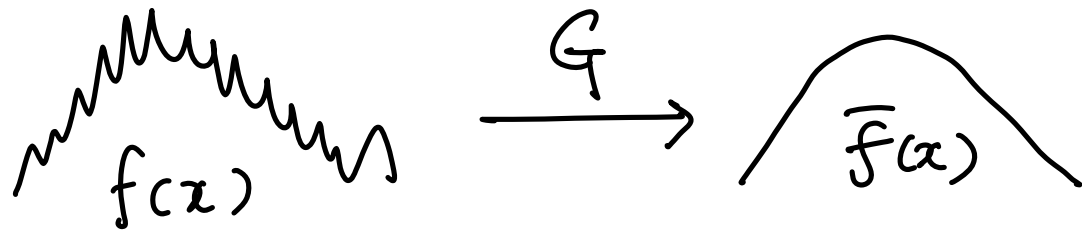
prescribed length scale (filter width)



↓
cutoff wavenumber

$$k_c = \pi / \Delta$$

$$k_c \cdot 2\bar{\Delta} = 2\pi$$



$$\bar{f}(x) = \int f(x') G(x, x') dx'$$

FVM
FDM
FEM
Spectral method

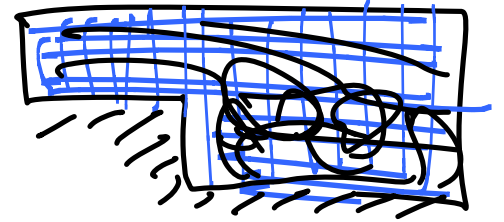
- Box filter: $G(x, x') = \begin{cases} 1 & \text{for } x_i - \bar{\Delta}/2 < x'_i < x_i + \bar{\Delta}/2 \\ 0 & \text{otherwise} \end{cases}$
- Gaussian filter: $G(x, x') = \left(\frac{6}{\pi\bar{\Delta}}\right)^{3/2} \exp\left[\frac{-6(x_i - x'_i)^2}{\bar{\Delta}^2}\right]$
- Sharp cutoff filter: $G(x, x') = 2 \sin[\pi(x_i - x'_i)/\bar{\Delta}] / \pi(x_i - x'_i)$

removes modes at $k > k_c = \pi/\delta$.

- Governing equations for LES

Filtered (spatially filtered)

continuity & Navier-Stokes eqs.



$$\int \left(\frac{\partial u_i}{\partial x_i} = 0 \right) \delta \, dx' \longrightarrow \frac{\partial \bar{u}_i}{\partial x_i} = 0 \quad \bar{u}_i: \text{filtered velocity}$$

$$\frac{\partial \bar{u}_i}{\partial t} + \frac{\partial}{\partial x_j} (\bar{u}_i \bar{u}_j) = -\frac{1}{\rho} \frac{\partial \bar{p}}{\partial x_i} + \nu \frac{\partial^2 \bar{u}_i}{\partial x_j \partial x_j} - \frac{\partial}{\partial x_j} \tau_{ij}$$

$\tau_{ij} = \overline{u_i u_j} - \bar{u}_i \bar{u}_j$: subgrid-scale (SGS) stresses

↳ modeling SGS stresses!

- Smagorinsky eddy viscosity model (1963; SM)

$$\tau_{ij} - \frac{1}{3} \tau_{kk} \delta_{ij} = -2 \nu_T \bar{S}_{ij} \quad (\text{eddy viscosity hypothesis})$$

$$\nu_T = (C_s \bar{\Delta})^2 |\bar{S}|, \quad |\bar{S}| = \sqrt{2 \bar{S}_{ij} \bar{S}_{ij}}, \quad \bar{S}_{ij} = \frac{1}{2} \left(\frac{\partial \bar{u}_i}{\partial x_j} + \frac{\partial \bar{u}_j}{\partial x_i} \right)$$

↑ Smagorinsky constant coefficient 0.1 ~ 0.3

However, C_s is not universal

and requires damping ft. near the wall.

1982 channel flow LES (Moin & Kim) $C_s (1 - e^{-0.4})$ ↑ γ

→ C_s should be given a priori depending on the flow field.

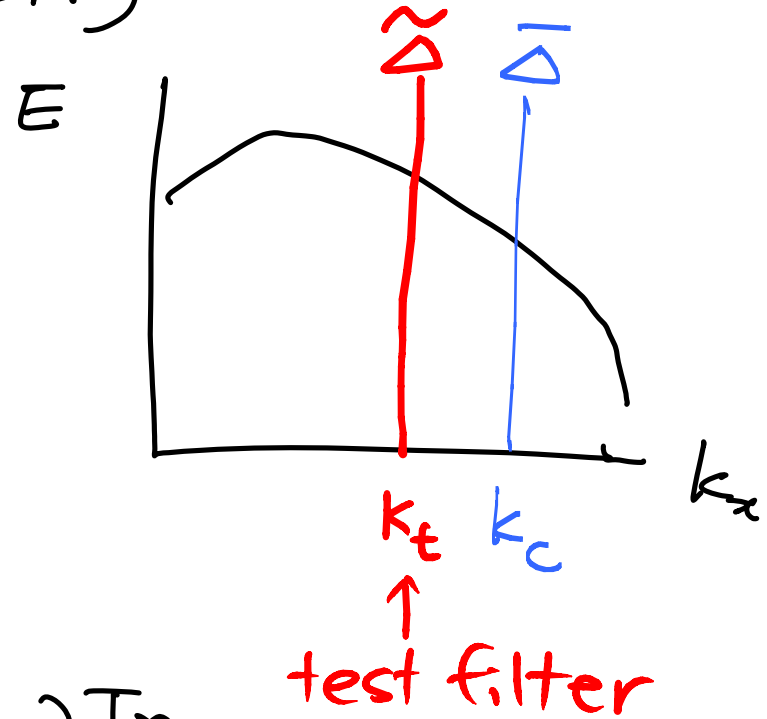
• Dynamic Smagorinsky model (DSM)

Germano et al. (1991)
 ↑ Moin

Apply another (test) filter: \sim

$$\frac{\partial \bar{u}_i}{\partial x_i} = 0$$

$$\frac{\partial \bar{u}_i}{\partial t} + \frac{\partial \bar{u}_i \bar{u}_j}{\partial x_j} = -\frac{1}{\rho} \frac{\partial \bar{p}}{\partial x_i} + \nu \frac{\partial^2 \bar{u}_i}{\partial x_j \partial x_j} - \frac{\partial T_{ij}}{\partial x_j}$$



$$\begin{aligned}
 T_{ij} &= \overline{u_i u_j} - \overline{u_i} \overline{u_j} \\
 \tilde{T}_{ij} &= \widetilde{u_i u_j} - \widetilde{\overline{u_i} \overline{u_j}}
 \end{aligned}
 \quad \Bigg] \quad
 \underbrace{T_{ij} - \tilde{T}_{ij}}_{\text{Germano identity}} = \overline{u_i u_j} - \overline{\overline{u_i} \overline{u_j}} \equiv \underline{L_{ij}}$$

$$\tilde{T}_{ij} - \frac{1}{3} \delta_{ij} \tilde{T}_{kk} = -2 (c_s \overline{\Delta})^2 |\overline{S}| \overline{S}_{ij}$$

$$T_{ij} - \frac{1}{3} \delta_{ij} T_{kk} = -2 (c_s \hat{\Delta})^2 |\hat{S}| \hat{S}_{ij}$$

$$\Rightarrow \underbrace{-2 (c_s \hat{\Delta})^2 |\hat{S}| \hat{S}_{ij} + 2 (c_s \overline{\Delta})^2 |\overline{S}| \overline{S}_{ij}}_{\text{Germano identity}} = L_{ij} - \frac{1}{3} \delta_{ij} L_{kk}$$

$$-2 (c_s \overline{\Delta})^2 \left[\left(\frac{\hat{\Delta}}{\overline{\Delta}} \right)^2 |\hat{S}| \hat{S}_{ij} - |\overline{S}| \overline{S}_{ij} \right] = L_{ij} - \frac{1}{3} \delta_{ij} L_{kk}$$

$$\overbrace{M_{ij}}$$

$$\rightarrow L_{ij} - \frac{1}{3} \delta_{ij} L_{kk} + 2 (C_s \bar{\Delta})^2 M_{ij} = 0$$

6 eqs but 1 C_s . \Rightarrow least square error method

$$Q = \left[L_{ij} - \frac{1}{3} \delta_{ij} L_{kk} + 2 (C_s \bar{\Delta})^2 M_{ij} \right]^2 \quad \text{Lilly et al. (1992)}$$

$$\frac{\partial Q}{\partial C_s^2} = 0 \Rightarrow (C_s \bar{\Delta})^2 = -\frac{1}{2} \frac{L_{ij} M_{ij}}{M_{ij} M_{ij}}$$

C_s is a fn of space and time, and is obtained at each time step during computation

→ called, dynamic Smagorinsky model.

Dynamic Smagorinsky model (DSM) - continued

$$\tau_{ij} - \frac{1}{3} \delta_{ij} \tau_{kk} = -2 \underbrace{(C_s \bar{\Delta})^2}_{\text{blue underline}} |\bar{S}| \bar{s}_{ij}$$

$$\rightarrow (C_s \bar{\Delta})^2 = -\frac{1}{2} \frac{L_{ij} M_{ij}}{M_{ij} M_{ij}}$$

$$L_{ij} = \overbrace{u_i u_j} - \overbrace{u_i} \overbrace{u_j}$$

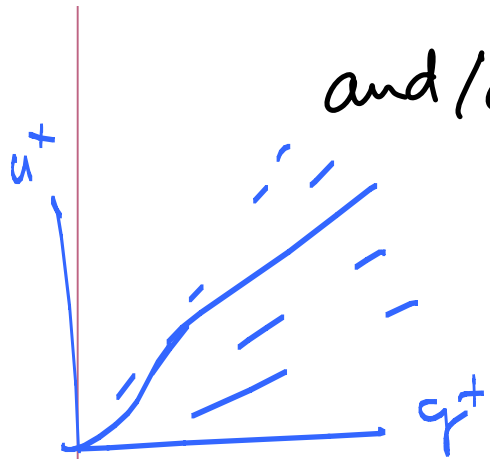
$$M_{ij} = \left(\frac{\hat{\Delta}}{\bar{\Delta}} \right)^2 \overbrace{|\bar{S}| \bar{s}_{ij}} - \overbrace{|\bar{S}| \bar{s}_{ij}}$$

→ do computation!

⇒ $(C_s)^2 < 0$ at some grids.

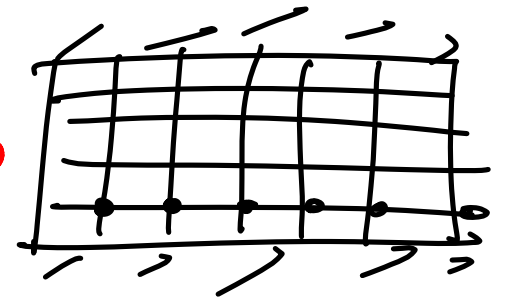
↳ numerical instability

⇒ requires an averaging over homogeneous direction _(CS)

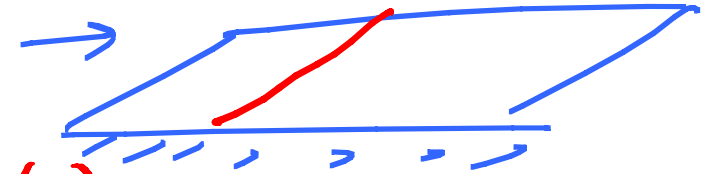
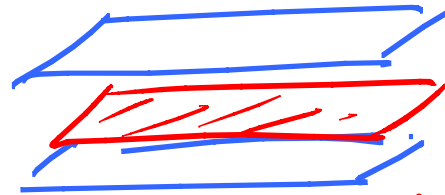
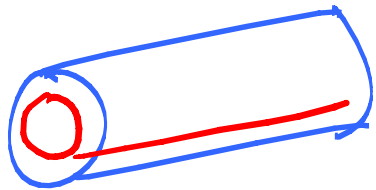


and/or ad hoc clipping: $C_s = 0$ if $C_s < 0$

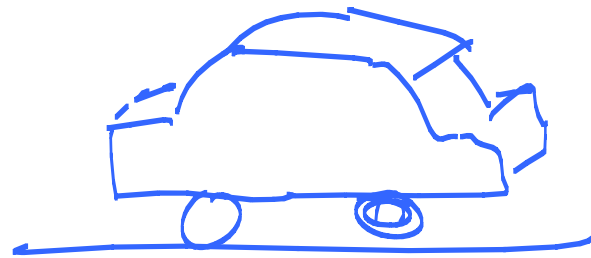
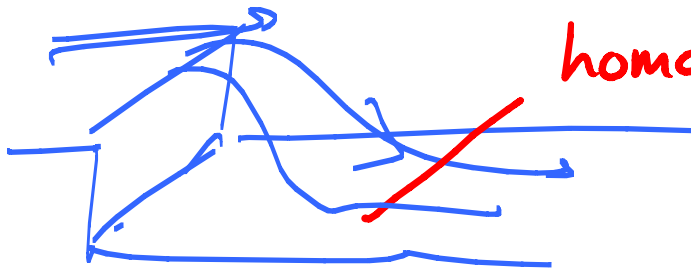
$$(C_s \bar{\Delta})^2 = -\frac{1}{2} \frac{\langle L_{ij} M_{ij} \rangle_n}{\langle M_{ij} M_{ij} \rangle_n}$$



↳ numerical stability



homo. direction(s)



no
homo.
direction

↳ dynamic Smag. modelx

- Vreman model (2004, PoF)

$$\nu_T = C_v \sqrt{\frac{\overline{\Pi_\beta}}{\overline{\alpha_{ij}} \overline{\alpha_{ij}}}}, \quad \overline{\alpha_{ij}} = \frac{\partial \overline{u_j}}{\partial x_i},$$

Vreman coefficient

$$C_v = 0.07$$

isotropic flow

$$\overline{\Pi_\beta} = \beta_{11} \beta_{22} - \beta_{12}^2 + \beta_{11} \beta_{33} - \beta_{13}^2 + \beta_{22} \beta_{33} - \beta_{23}^2,$$

$$\beta_{ij} = \sum_{m=1}^3 \overline{\Delta_m}^2 \overline{\alpha_{mi}} \overline{\alpha_{mj}}$$

$$\nu_T = 2(C_s \overline{\Delta})^2 |\overline{s_{ij}}|$$

laminar $|\overline{s_{ij}}| \neq 0$

Park, Lee & Choi (2006, PoF)

Lee, Choi & Park (2010, PoF)

Dynamic global model to determine C_v

Germano identity $L_{ij} = \overline{T_{ij}} - \widetilde{T_{ij}} = \overline{u_i u_j} - \widetilde{u_i u_j}$

$$\overline{T_{ij}} - \frac{1}{3} \delta_{ij} T_{kk} = \overline{u_i u_j} - \overline{u_i} \overline{u_j} = -2C_v \sqrt{\frac{\overline{u_\beta u_\beta}}{\overline{d_{ij}} \overline{d_{ij}}}} \overline{S_{ij}}$$

$$\overline{T_{ij}} - \frac{1}{3} \delta_{ij} T_{kk} = \widetilde{u_i u_j} - \widetilde{u_i} \widetilde{u_j} = -2C_v \sqrt{\frac{\widetilde{u_\beta u_\beta}}{\widetilde{d_{ij}} \widetilde{d_{ij}}}} \widetilde{S_{ij}}$$

and

introduce least square error minimization to get

$$\Rightarrow C_v(t) = -\frac{1}{2} \frac{\langle L_{ij} M_{ij} \rangle_V}{\langle M_{ij} M_{ij} \rangle_V}$$

volume averaging C_v

$$L_{ij} = \overline{u_i u_j} - u_i \overline{u_j}, \quad M_{ij} = \sqrt{\frac{V\beta}{\alpha_{ij} \overline{\alpha_{ij}}}} \overline{S_{ij}} - \sqrt{\frac{V\beta}{\alpha_{ij} \overline{\alpha_{ij}}}} S_{ij}$$

C_v is constant in space but varies in time
 No clipping
 No homo. direction required

→ Can be applied to complex geometry.

Ch. 10. Compressible flow - skip

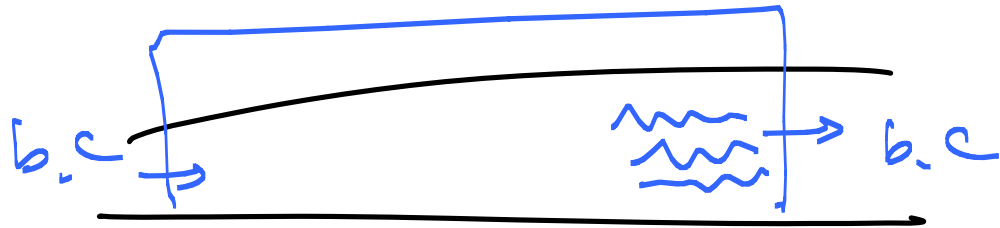
Ch. 11. Efficiency and accuracy improvement

- errors
 - modeling errors : turbulence
combustion
multi-phase flow
 - discretization errors
 - iteration errors $\leftarrow \underline{Ax} = b$ 10^{-12} $\underline{10^{-5}}$ $\underline{10^{-6}}$
 - programming and user errors X
- estimating errors — validation and uncertainty quantification

- grid quality and optimization
- multigrid methods for flow calculation
- adaptive grid methods and local grid refinement
- parallel computing in CFD

Ch. 12 Special Topics

1. Heat and mass transfer



$$\frac{\partial \theta}{\partial t} + u_j \frac{\partial \theta}{\partial x_j} = \frac{1}{PrRe} \nabla^2 \theta$$

$$\frac{\partial u_i}{\partial t} + u_j \frac{\partial u_i}{\partial x_j} = \frac{1}{Re} \nabla^2 u_i - \nabla p$$

2. Flows w/ variable fluid properties

$$\frac{\partial}{\partial x_j} \left(\mu \frac{\partial u_i}{\partial x_j} \right)$$

$\mu(x, t)$

$$\mu = \mu_0 + (\mu - \mu_0)$$

3. Moving grids
4. Free-surface flows
5. Meteorological and oceanographic applications
6. Multiphase flows
7. Combustion