

Intro to CFD (Fluent)

1. Fluent instructions / tutorial
2. General idea of gaseous combustion model
3. Governing equations
 - overall continuity
 - species continuity
 - energy
4. Elliptic vs. Parabolic Eqs.
5. Pressure

General Idea of Comprehensive Model

1. Pick geometry

2. Pick variables

Turbulent vs laminar
 gaseous vs particle-laden
 adiabatic vs heat losses
 radiation?
 reacting vs non-reacting
 steady-state vs transient

3. Set up grid system to solve basic equations

- cartesian ($x-y-z$) vs cylindrical axi-symmetric ($r-z-\theta$)
- staggered vs non-staggered (collocated)
- "stair-step" boundary vs. more complicated coordinate mapping
(like body-fitted coordinates, ^{unstructured grid} etc.)
- finite difference vs finite volume vs finite element

GOAL: Solve all conservation equations for continuum phase as painlessly as possible with one algorithm!

2. GOVERNING EQUATIONS

A. Overall Continuity

$$\frac{\partial \rho}{\partial t} + \vec{\nabla} \cdot \rho \vec{v} = S_{pm}$$

@ steady-state

S_{pm} = mass source term from condensed phase
= 0 for gaseous combustion only

B. Species Continuity

$$\frac{\partial \rho_i}{\partial t} + \vec{\nabla} \cdot \vec{n}_i = r_i \quad \text{BSL 18.1-6}$$

$$\rho_i = \rho \omega_i \quad \omega_i = \text{mass fraction}$$

r_i = net production rate of species i , including source (if any) from condensed phase

$$\vec{n}_i = -\rho D_{ij} \vec{\nabla} \omega_i + \omega_i \vec{n}_T$$

$$\vec{n}_T = \rho \vec{v} \quad \omega_i \vec{n}_T = \rho_i \vec{v}$$

$$\frac{\partial \rho_i}{\partial t} + \vec{\nabla} \cdot \rho_i \vec{v} = \vec{\nabla} \cdot (\rho D_{ij} \vec{\nabla} \omega_i) + r_i \quad (\text{BSL 18.1-14})$$

transient

convection

diffusion

production
from reaction

Momentum

$$\frac{\partial \rho \vec{v}}{\partial t} = - \left[\vec{\nabla} \cdot \rho \vec{v} \vec{v} \right] - \vec{\nabla} P - \left[\vec{\nabla} \cdot \vec{\tau} \right] + \rho \vec{g} + F_p$$

momentum exchange
pressure
viscous forces
gravity
interaction with solid phase

BS;L 3.2-10

(Table 10.4-1)

$$\tau_{ij} = \left[-\mu \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right) \right] + \sum_{j,k} \mu \frac{\partial u_k}{\partial x_j}$$

laminar ↑

BS;L 3.2-11
-16

Kuo 3-57

Energy pick h or T (usually h in large systems)

$$\rho \frac{D\hat{H}}{Dt} = -\nabla \cdot \vec{q} + \frac{Dp}{Dt} - \vec{\tau} : \nabla \vec{v} + \epsilon_{jij} g_i$$

BS;L 18.3-1E

$$\rho \left[\frac{\partial \hat{H}}{\partial t} + \vec{v} \cdot \nabla \hat{H} \right] = -\nabla \cdot \vec{q} + \left[\frac{\partial p}{\partial t} + \vec{v} \cdot \nabla p \right] - \vec{\tau} : \nabla \vec{v}$$

multiply continuity eqn. by \hat{H}

$$\hat{H} \frac{\partial \rho}{\partial t} + \hat{H} \rho \vec{v} \cdot \nabla = \hat{H} S_{pm}$$

Now add to energy eqn.

$$\hat{H} \frac{\partial \rho}{\partial t} + \rho \frac{\partial \hat{H}}{\partial t} + \hat{H} \nabla \cdot \rho \vec{v} + \rho \vec{v} \cdot \nabla \hat{H} = -\nabla \cdot \vec{q} + \frac{\partial p}{\partial t} + \vec{v} \cdot \nabla p - \vec{\tau} : \nabla \vec{v} + \hat{H} S_{pm}$$

$$\frac{\partial (\rho \hat{H})}{\partial t} + \vec{\nabla} \cdot (\rho \hat{H} \vec{v}) = -\nabla \cdot \vec{q} + \frac{\partial p}{\partial t} + \vec{v} \cdot \nabla p - \vec{\tau} : \nabla \vec{v} + \hat{H} S_{pm}$$

Also, remember that $q = q^{(c)} + q^{(d)}$

$$= -k \nabla T + \sum \hat{H}_j \vec{j}_j$$

BS;L 18.4-2

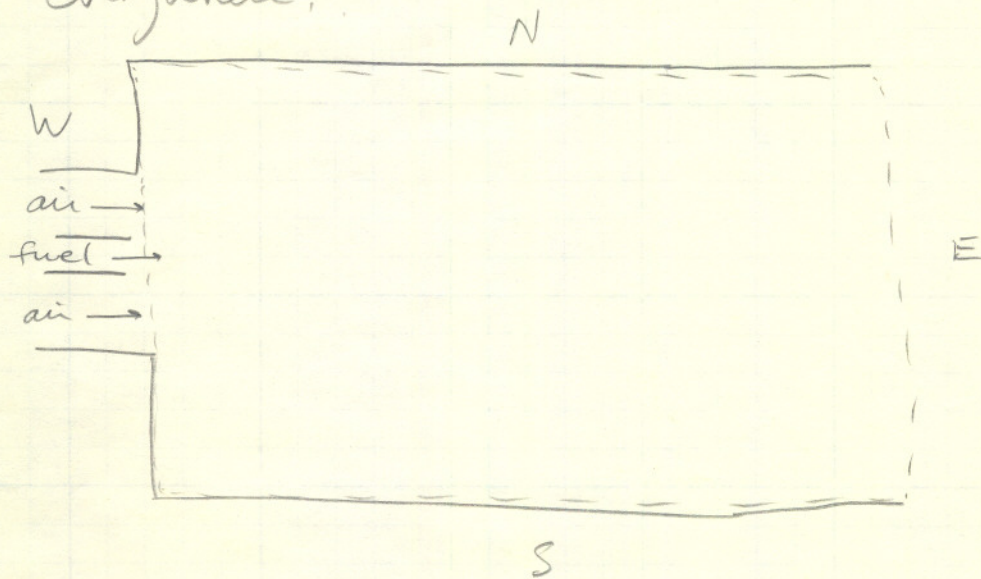
$$A \frac{\partial^2 u}{\partial \xi^2} + B \frac{\partial^2 u}{\partial \xi \partial \eta} + C \frac{\partial^2 u}{\partial \eta^2} + \dots = 0$$

only in
viscous
terms $\Rightarrow B \rightarrow 0$

$$B^2 - 4AC \approx -4AC < 0$$

elliptic eqns!

In other words, we need boundary conditions everywhere!!



Boundary Conditions needed

- inlets
- walls
- outlets
- symmetry axis (if any)

flow (flux)
 ↙ set boundary
 ↘ extrapolated boundary

Pressure (Makes problem difficult)

→ What equation do we solve for pressure?

- we have to monkey around, using the Continuity and Momentum equations!

- Is it important?

How much velocity can a $\Delta P = 1$ pascal drive?

$$-\Delta P = \frac{\rho \Delta v^2}{2} \quad (\text{Bernoulli's eqn.})$$

$$1 \text{ Pascal} = 0.987 \times 10^{-5} \text{ atm} = 1 \frac{\text{kg}}{\text{m}^2 \text{s}^2}$$

$$\rho = \text{air @ } 1500 \text{ K} = .2322 \frac{\text{kg}}{\text{m}^3}$$

$$\frac{1 \frac{\text{kg}}{\text{m}^2 \text{s}^2}}{.2322 \frac{\text{kg}}{\text{m}^3}} = 4.3 \frac{\text{m}}{\text{s}} = \Delta(v^2) = v_2^2 - v_1^2$$

$$v_1 = 0$$

$$v_1 = 5 \text{ m/s}$$

$$v_1 = 20 \text{ m/s}$$

$$v_2 = 2.07 \text{ m/s}$$

$$v_2 = 5.4 \text{ m/s}$$

$$v_2 = 20.1 \text{ m/s}$$

$$\Delta v = 207 \text{ cm/s}$$

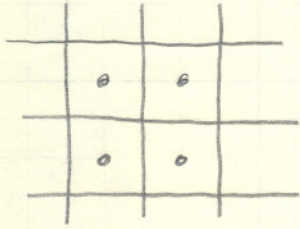
$$\Delta v = 40 \text{ cm/s}$$

$$\Delta v = .1 \text{ cm/s}$$

message: small ΔP can cause large Δv , especially when v is small to begin with!

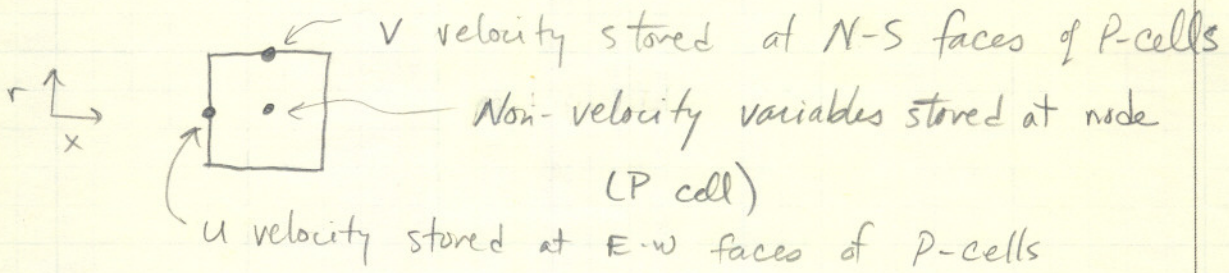
GRIDDING CONCEPTS

A. Collocated grid



- All variables are stored at grid centers (nodes)
- Interpolation necessary when properties at cell interfaces are needed.

B. Staggered grid (used in PCGC-3, Fluent)



Reason: When you integrate the PDE's, velocities are needed at the faces of P-cells, so why not store them there instead of interpolating all the time

Advantages: - Code ^{numerical} stability (better than standard collocated grid)

Disadvantages: - complexity, especially w/ non-uniformly spaced grids
- hard to implement for BFC, advanced gridding concepts

u_i, v_i cells now constitute whole new grid cell systems

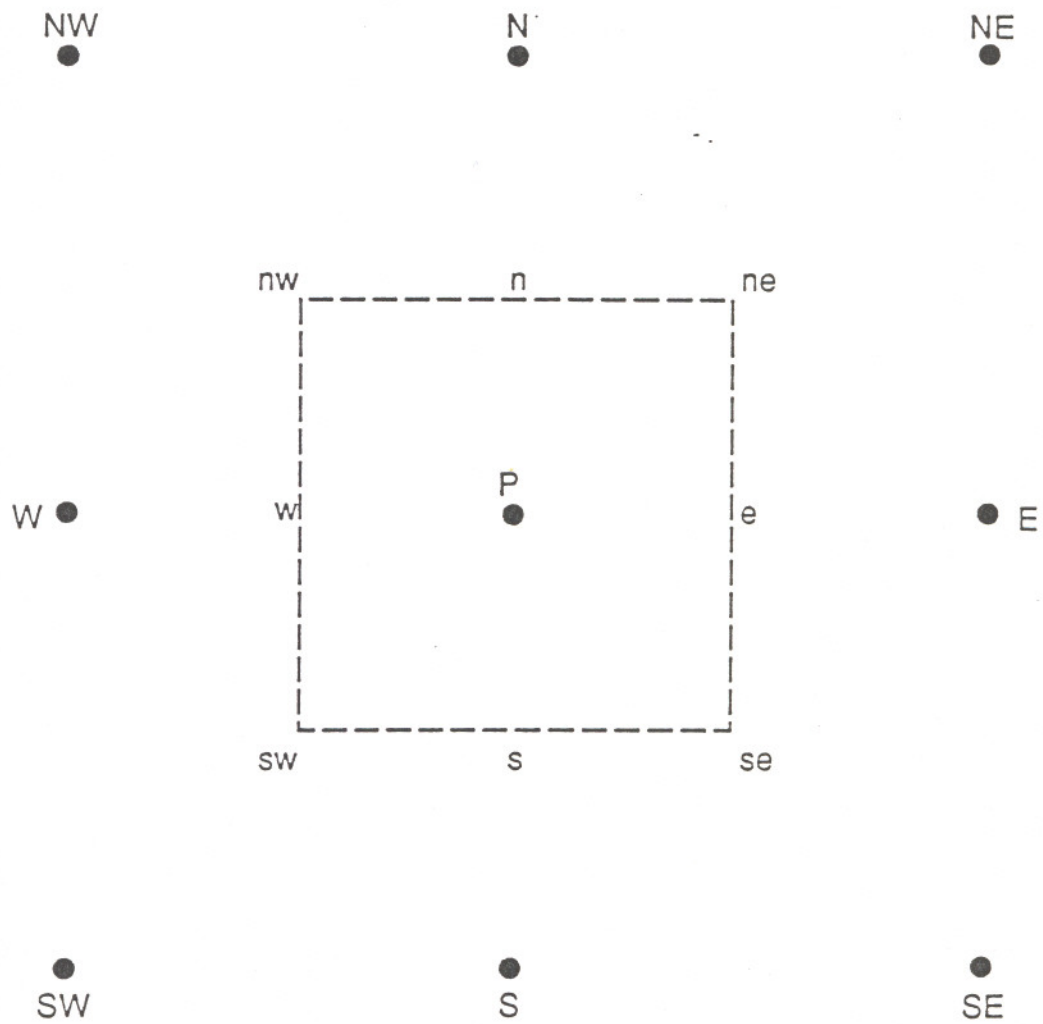


Figure 10. Illustration of the grid symbols for a computational cell.

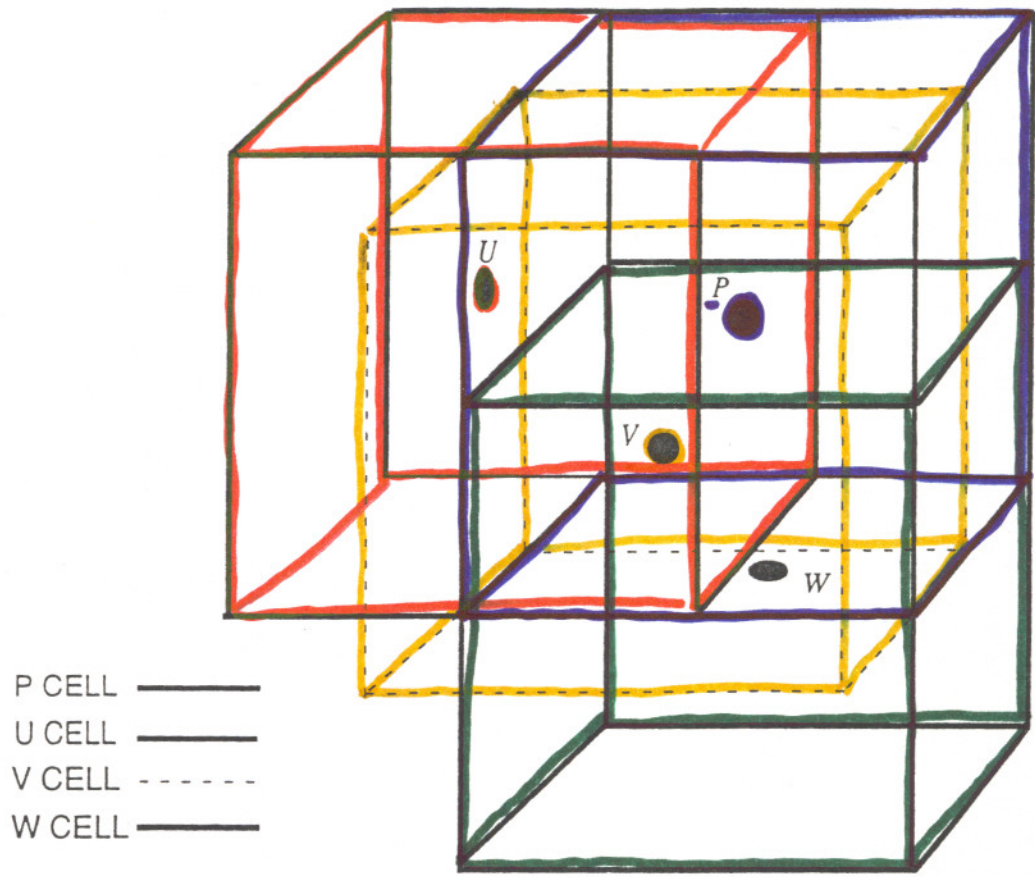


Figure 3.12: Side View of the Computational Cell Types Used in PCGC-3

Basic TEACH algorithm

(method for getting algebraic equations from PDE's based on finite-volume integration of PDE terms)

A. Put all equations in a standard form (2-d r-z coordinates shown here)

$$0 = \frac{\partial}{\partial x}(\rho u r \phi) + \frac{\partial}{\partial r}(\rho v r \phi) - \frac{\partial}{\partial x} \left(r \Gamma \frac{\partial \phi}{\partial x} \right) - \frac{\partial}{\partial r} \left(r \Gamma \frac{\partial \phi}{\partial r} \right) - r S \phi$$

B. Integrate these eqns. over the P cell, one term at a time.

$$\int_{x_w}^{x_e} \int_{r_s}^{r_n} \int_0^{2\pi} [\text{eqn.}] dx dr d\theta = 0 = \int_{x_w}^{x_e} \int_{r_s}^{r_n} [\quad] dx dr$$

due to symmetry, multiplying by 2π .

1st convection term

$$= \int_{x_w}^{x_e} \int_{r_s}^{r_n} \frac{\partial}{\partial x}(\rho u r \phi) dx dr = \int_{r_s}^{r_n} \left\{ \rho u r \phi \right\}_w^e dr = \int_{r_s}^{r_n} \left\{ \rho u \phi \right\}_w^e r dr$$

(integrate a derivative)

using mean value,

$$= \left(\left[\overline{\rho u \phi} \right]_e - \left[\overline{\rho u \phi} \right]_w \right) \left(\frac{r_n^2}{2} - \frac{r_s^2}{2} \right)$$

Because of staggered grid, $u_e = u_E$ ← node of u-cell on East side, using capital 'E'
 $u_w = u_W$

$$= \left(\rho_e u_E \phi_e - \rho_w u_W \phi_w \right) A_{ew}$$

↑ interpolate to get these

Define:

$$C_E = \rho_e u_E A_{ew} \quad (\text{convection coefficients})$$

$$C_W = \rho_w u_W A_{ew}$$

Similar stuff to other terms (radial convection, also diffusion terms)

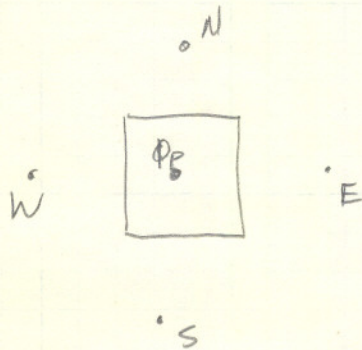
$$A_E = D_E - \frac{1}{2} C_E$$

\uparrow diffusion coefficient \uparrow convection coefficient

Final form

$$[A_E + A_W + A_N + A_S] \phi_P = A_E \phi_E + A_W \phi_W + A_N \phi_N + A_S \phi_S + S_u + S_\phi \phi_P$$

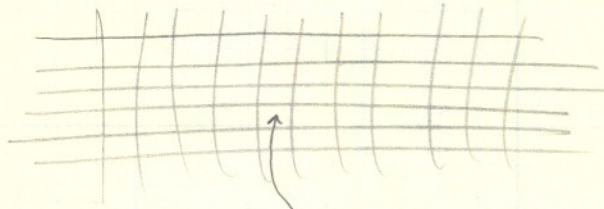
(some tricks then played for convergence) leftover terms



* This says that ϕ_P depends on contributions from surrounding neighbors.

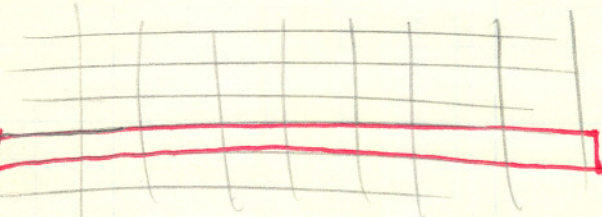
- note: upwind, power-law & hybrid differencing incorporated.

Matrix Solution Algorithm



for each cell i, j , we have $\underbrace{\left(\sum_{d \in \text{NEFW}} A_d - S \right)}_{A_P} \phi_P = \sum_{d \in \text{NEFW}} (A_d \phi_d) + S_u$

Therefore, use successive tri-diagonal algorithms to iterate



Solve \rightarrow this row, holding every thing else constant, then solve by column, & continue

Tridiagonal matrix

$$\begin{bmatrix} \circ & \circ & \circ & & \\ \circ & \circ & \circ & & \\ \circ & \circ & \circ & & \\ \circ & \circ & \circ & & \\ \circ & \circ & \circ & & \\ \circ & \circ & \circ & & \\ \circ & \circ & \circ & & \\ \circ & \circ & \circ & & \end{bmatrix} \begin{bmatrix} \phi_1 \\ \phi_2 \\ \phi_3 \\ \vdots \\ \phi_n \end{bmatrix} = \begin{bmatrix} \text{RHS} \\ \vdots \\ \text{RHS} \end{bmatrix}$$

or

$$D_i \phi_i = A_i \phi_{i+1} + B_i \phi_{i-1} + C_i$$

\Downarrow

gets row (or column) of ϕ 's calculated

why? \Rightarrow large matrices are hard to invert (time; memory)

Convergence tricks

A. Flow chart

1. Pick one variable (i.e., u, v, ψ, h , etc.)

2. Calculate A_p, A_d 's, S_u for that variable based on latest guesses for other variables at each cell!

3. Do row by row & then column by column TDMA for that variable until residual for that variable is low.

$$RES_{ij} = (A_p \phi_p - \sum A_d \phi_d - S_u)_{ij}$$

$$Res_{tot} = \sum RES_{ij}$$

4. Go to next variable until all variables are solved

5. Start over with first variable, based on newest updates, until all residuals are low

B. Under-relaxation

Convention says we need to under-relax to try to get convergence

$$\phi^{new} = \lambda \phi^{new} + (1-\lambda) \phi^{old}$$

λ = under-relaxation factor between 0 and 1

1 = no under-relaxation

0 = no change (i.e., no solution other than 1st guess)

In practice, the source terms are under-relaxed

$$A_p^{new} = \frac{A_p^{old}}{\lambda}$$

$$S_u^{new} = S_u^{old} + \frac{(1-\lambda) \phi_p^{old} A_p^{new}}{\lambda} \quad \text{or something similar}$$