

MASSACHUSETTS INSTITUTE OF TECHNOLOGY  
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 CAMBRIDGE, MASSACHUSETTS 02139  
**2.29 NUMERICAL FLUID MECHANICS — SPRING 2015**

## EQUATION SHEET – Quiz 1

### **Number Representation**

- Floating Number Representation:  $x = m b^e$ ,  $b^{-1} \leq m < b^0$

### **Truncation Errors and Error Analysis** $y = f(x_1, x_2, x_3, \dots, x_n)$

$$f(x_{i+1}) = f(x_i) + \Delta x f'(x_i) + \frac{\Delta x^2}{2!} f''(x_i) + \frac{\Delta x^3}{3!} f'''(x_i) + \dots + \frac{\Delta x^n}{n!} f^n(x_i) + R_n$$

- Taylor Series:

$$R_n = \frac{\Delta x^{n+1}}{n+1!} f^{(n+1)}(\xi)$$

- The Differential Error (general error propagation) Formula:  $\varepsilon_y \leq \sum_{i=1}^n \left| \frac{\partial f(x_1, \dots, x_n)}{\partial x_i} \right| \varepsilon_i$
- The Standard Error (statistical formula):  $E(\Delta_s y) \approx \sqrt{\sum_{i=1}^n \left( \frac{\partial f}{\partial x_i} \right)^2 \varepsilon_i^2}$
- Condition Number of  $f(x)$ :  $K_p = \left| \frac{\bar{x} f'(\bar{x})}{f(\bar{x})} \right|$

### **Roots of nonlinear equations** ( $x_{n+1} = x_n - h(x_n)f(x_n)$ )

- Bisection: successive division of bracket in half, next bracket based on sign of  $f(x_1^{n+1})f(x_{\text{mid-point}}^{n+1})$
- False-Position (Regula Falsi):  $x_r = x_U - \frac{f(x_U)(x_L - x_U)}{f(x_L) - f(x_U)}$
- Fixed Point Iteration (General Method or Picard Iteration):  
 $x_{n+1} = g(x_n)$  or  $x_{n+1} = x_n - h(x_n)f(x_n)$
- Newton Raphson:  $x_{n+1} = x_n - \frac{1}{f'(x_n)} f(x_n)$
- Secant Method:  $x_{n+1} = x_n - \frac{(x_n - x_{n-1})}{f(x_n) - f(x_{n-1})} f(x_n)$
- Order of convergence  $p$ : Defining  $e_n = x_n - x^*$ , the order of convergence  $p$  exists if there exist a constant  $C \neq 0$  such that:  $\lim_{n \rightarrow \infty} \frac{|e_{n+1}|}{|e_n|^p} = C$

### Conservation Law for a scalar $\phi$ , in integral and differential forms:

$$- \left\{ \frac{d}{dt} \int_{CM} \rho \phi dV \right\} = \frac{d}{dt} \int_{CV_{\text{fixed}}} \rho \phi dV + \underbrace{\int_{CS} \rho \phi (\vec{v} \cdot \vec{n}) dA}_{\substack{\text{Advection fluxes} \\ (\text{Adv. \& diff. = "convection" fluxes})}} = - \underbrace{\int_{CS} \vec{q}_\phi \cdot \vec{n} dA}_{\substack{\text{Other transports} \\ (\text{diffusion, etc})}} + \underbrace{\sum \int_{CV_{\text{fixed}}} s_\phi dV}_{\substack{\text{Sum of sources and} \\ \text{sinks terms (reactions, etc)}}}$$

$$- \frac{\partial \rho \phi}{\partial t} + \nabla \cdot (\rho \phi \vec{v}) = -\nabla \cdot \vec{q}_\phi + s_\phi$$

### Linear Algebraic Systems:

- Gauss Elimination: reduction,  $m_{ik} = \frac{a_{ik}^{(k)}}{a_{kk}^{(k)}}$ ,  $a_{ij}^{(k+1)} = a_{ij}^{(k)} - m_{ik} a_{kj}^{(k)}$ ,  $b_i^{(k+1)} = b_i^{(k)} - m_{ik} b_k^{(k)}$ , followed by a back-substitution.  $x_k = \left( b_k - \sum_{j=k+1}^n a_{kj}^{(k)} x_j \right) / a_{kk}^{(k)}$
- LU decomposition:  $\mathbf{A} = \mathbf{L}\mathbf{U}$ ,  $a_{ij} = \sum_{k=1}^{\min(i,j)} m_{ik} a_{kj}^{(k)}$
- Choleski Factorization:  $\mathbf{A} = \mathbf{R}^* \mathbf{R}$ , where  $\mathbf{R}$  is upper triangular and  $\mathbf{R}^*$  its conjugate transpose.
- Condition number of a linear algebraic system:  $K(\mathbf{A}) = \|\mathbf{A}^{-1}\| \|\mathbf{A}\|$
- A banded matrix of  $p$  super-diagonals and  $q$  sub-diagonals has a bandwidth  $w = p + q + 1$
- $\sum_{i=1}^n i = \frac{n(n+1)}{2}$  and  $\sum_{i=1}^n i^2 = \frac{n(n+1)(2n+1)}{6}$
- Eigendecomposition:  $\mathbf{A}\mathbf{x} = \lambda\mathbf{x}$  and  $\det(\mathbf{A} - \lambda\mathbf{I}) = 0$
- Norms:

$$\|\mathbf{A}\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^m |a_{ij}| \quad \text{"Maximum Column Sum"}$$

$$\|\mathbf{A}\|_\infty = \max_{1 \leq i \leq m} \sum_{j=1}^n |a_{ij}| \quad \text{"Maximum Row Sum"}$$

$$\|\mathbf{A}\|_F = \sqrt{\left( \sum_{i=1}^m \sum_{j=1}^n |a_{ij}|^2 \right)^2} \quad \text{"Frobenius norm" (or "Euclidean norm")}$$

$$\|\mathbf{A}\|_2 = \sqrt{\lambda_{\max} \{ \mathbf{A}^* \mathbf{A} \}} \quad \text{"L-2 norm" (or "spectral norm")}$$

### Iterative Methods for solving linear algebraic systems: $\mathbf{x}^{k+1} = \mathbf{B} \mathbf{x}^k + \mathbf{c} \quad k = 0, 1, 2, \dots$

- Necessary and sufficient condition for convergence:  
 $\rho(\mathbf{B}) = \max_{i=1 \dots n} |\lambda_i| < 1$ , where  $\lambda_i = \text{eigenvalue}(\mathbf{B}_{n \times n})$
- Jacobi's method:  $\mathbf{x}^{k+1} = -\mathbf{D}^{-1}(\mathbf{L} + \mathbf{U}) \mathbf{x}^k + \mathbf{D}^{-1} \mathbf{b}$
- Gauss-Seidel method:  $\mathbf{x}^{k+1} = -(\mathbf{D} + \mathbf{L})^{-1} \mathbf{U} \mathbf{x}^k + (\mathbf{D} + \mathbf{L})^{-1} \mathbf{b}$
- SOR Method:  $\mathbf{x}^{k+1} = (\mathbf{D} + \omega \mathbf{L})^{-1} [-\omega \mathbf{U} + (1 - \omega) \mathbf{D}] \mathbf{x}^k + \omega (\mathbf{D} + \omega \mathbf{L})^{-1} \mathbf{b}$
- Steepest Descent Gradient Method:  $\mathbf{x}_{i+1} = \mathbf{x}_i + \frac{\mathbf{r}_i^T \mathbf{r}_i}{\mathbf{r}_i^T \mathbf{A} \mathbf{r}_i} \mathbf{r}_i, \quad \mathbf{r}_i = \mathbf{b} - \mathbf{A} \mathbf{x}_i$
- Conjugate Gradient:  $\mathbf{x}_{i+1} = \mathbf{x}_i + \alpha_i \mathbf{v}_i$  ( $\alpha_i$  such that each  $\mathbf{v}_i$  are generated by orthogonalization of residuum vectors and such that search directions are  $\mathbf{A}$ -conjugate).

**Finite Differences – PDE types** (2nd order, 2D):  $A\phi_{xx} + B\phi_{xy} + C\phi_{yy} = F(x, y, \phi, \phi_x, \phi_y)$

$B^2 - AC > 0$ : hyperbolic;  $B^2 - AC = 0$ : parabolic;  $B^2 - AC < 0$ : elliptic

**Finite Differences – Error Types and Discretization Properties** ( $\mathcal{L}(\phi) = 0$ ,  $\hat{\mathcal{L}}_{\Delta x}(\hat{\phi}) = 0$ )

- Consistency:  $|\mathcal{L}(\phi) - \hat{\mathcal{L}}_{\Delta x}(\phi)| \rightarrow 0$  when  $\Delta x \rightarrow 0$
- Truncation error:  $\tau_{\Delta x} = \mathcal{L}(\phi) - \hat{\mathcal{L}}_{\Delta x}(\phi) \rightarrow O(\Delta x^p)$  for  $\Delta x \rightarrow 0$
- Error equation:  $\tau_{\Delta x} = \mathcal{L}(\phi) - \hat{\mathcal{L}}_{\Delta x}(\hat{\phi} + \varepsilon) = -\hat{\mathcal{L}}_{\Delta x}(\varepsilon)$  (for linear systems)
- Stability:  $\|\hat{\mathcal{L}}_{\Delta x}^{-1}\| < \text{Const.}$  (for linear systems)
- Convergence:  $\|\varepsilon\| \leq \|\hat{\mathcal{L}}_{\Delta x}^{-1}\| \|\tau_{\Delta x}\| \leq \alpha O(\Delta x^p)$

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