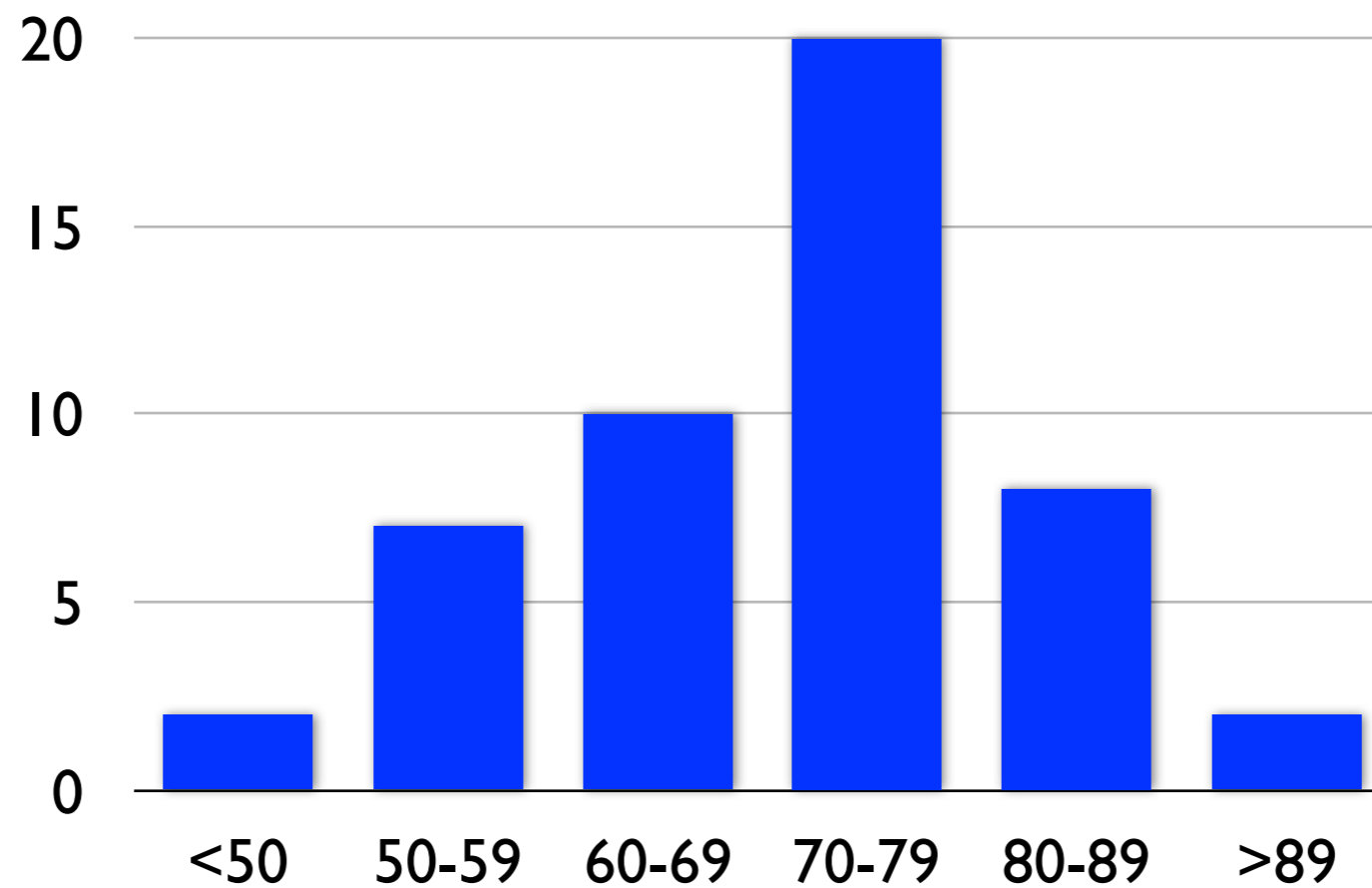


10.34: Numerical Methods Applied to Chemical Engineering

Lecture 16:
ODE-IVP and Numerical Integration

Quiz I Results

- Mean: 70.6
- Standard deviation: 11.0



Recap

- Implicit methods for ODE-IVPs

Recap

- Example:
 - Use implicit Euler to solve:

$$\frac{dx}{dt} = \lambda x, x(0) = x_0$$

Give a closed form formula for the numerical solution

Recap

- Example:
 - Use implicit Euler to solve:

$$\frac{dx}{dt} = \lambda x, x(0) = x_0$$

- Let:

$$x_k = x(k\Delta t)$$

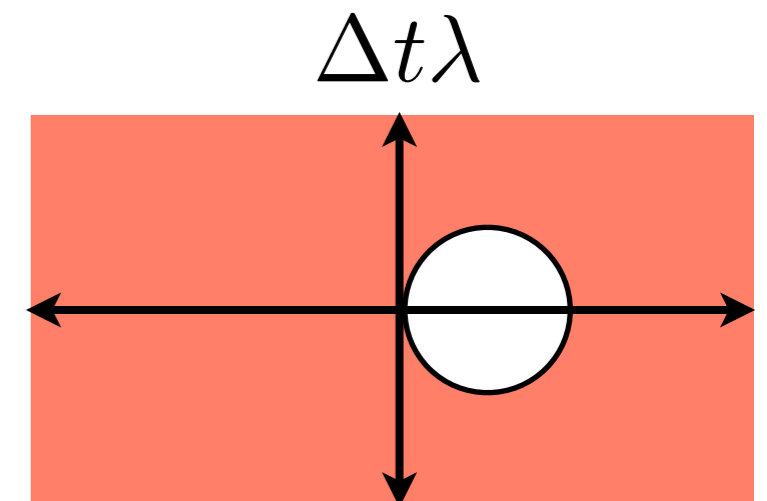
$$x_{k+1} = x_k + \Delta t \lambda x_{k+1}$$

$$x_{k+1} = \frac{1}{1 - \Delta t \lambda} x_k$$

$$x_k = \left(\frac{1}{1 - \Delta t \lambda} \right)^k x_0$$

- Stability:

$$|1 - \Delta t \lambda| \geq 1 \Rightarrow (1 - \Delta t \operatorname{Re} \lambda)^2 + (\Delta t \operatorname{Im} \lambda)^2 \geq 1$$



Recap

- Example:
 - Use implicit Euler to solve:

$$\frac{dx}{dt} = \lambda x, x(0) = x_0$$

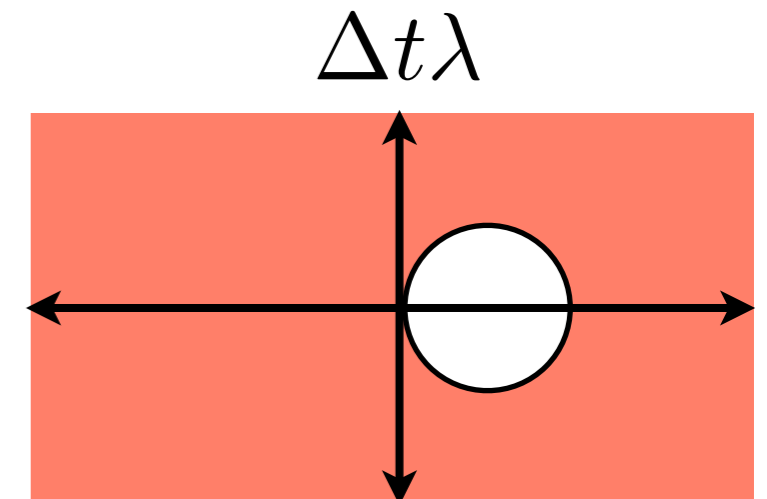
- Numerical solution:

$$x_k = \left(\frac{1}{1 - \Delta t \lambda} \right)^k x_0$$

- Exact solution:

$$x_k = x_0 e^{k\lambda\Delta t}$$

- Stability and accuracy do not correlate!



Multistep Methods

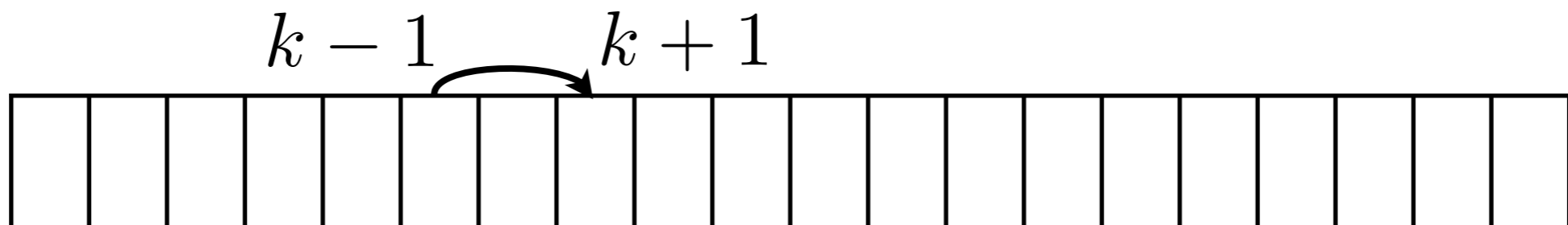
- Multistep methods utilize information over multiple time steps to approximate the solution of an ODE.
- These can be designed for higher accuracy, larger stability bounds or both.
- Example: Leapfrog method

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}(t), t)$$

- Approximate derivative with central difference:

$$\frac{1}{2\Delta t} (\mathbf{x}(t + \Delta t) - \mathbf{x}(t - \Delta t)) = \mathbf{f}(\mathbf{x}(t), t)$$

$$\mathbf{x}(t_{k+1}) = \mathbf{x}(t_{k-1}) + 2\Delta t \mathbf{f}(\mathbf{x}(t_k), t_k)$$



Multistep Methods

- Local accuracy of the leap frog method:

$$\frac{d\mathbf{x}}{dt} = \frac{1}{2\Delta t} (\mathbf{x}(t_{k+1}) - \mathbf{x}(t_{k-1})) + O((\Delta t)^2) = \mathbf{f}(\mathbf{x}(t_k), t_k)$$

$$\mathbf{x}(t_{k+1}) = \mathbf{x}(t_{k-1}) + 2\Delta t \mathbf{f}(\mathbf{x}(t_k), t_k) + O((\Delta t)^3)$$

- Stability of the leap frog method:

$$\frac{dx}{dt} = \lambda x$$

$$x_{k+1} = x_{k-1} + 2\Delta t \lambda x_k$$

$$\begin{pmatrix} x_{k+1} \\ x_k \end{pmatrix} = \begin{pmatrix} 2\Delta t \lambda & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x_k \\ x_{k-1} \end{pmatrix}$$

$$\begin{pmatrix} x_{k+1} \\ x_k \end{pmatrix} = \mathbf{C}^k \begin{pmatrix} x_1 \\ x_0 \end{pmatrix}$$

Multistep Methods

- Local accuracy of the leap frog method:

$$\frac{d\mathbf{x}}{dt} = \frac{1}{2\Delta t} (\mathbf{x}(t_{k+1}) - \mathbf{x}(t_{k-1})) + O((\Delta t)^2) = \mathbf{f}(\mathbf{x}(t_k), t_k)$$

$$\mathbf{x}(t_{k+1}) = \mathbf{x}(t_{k-1}) + 2\Delta t \mathbf{f}(\mathbf{x}(t_k), t_k) + O((\Delta t)^3)$$

- Stability of the leap frog method: $\frac{dx}{dt} = \lambda x$

$$x_{k+1} = x_{k-1} + 2\Delta t \lambda x_k$$

$$\begin{pmatrix} x_{k+1} \\ x_k \end{pmatrix} = \mathbf{C}^k \begin{pmatrix} x_1 \\ x_0 \end{pmatrix}$$

$$\mathbf{C} = \begin{pmatrix} 2\Delta t \lambda & 1 \\ 1 & 0 \end{pmatrix}$$

What are the eigenvalues of this matrix?

Multistep Methods

- Local accuracy of the leap frog method:

$$\frac{d\mathbf{x}}{dt} = \frac{1}{2\Delta t} (\mathbf{x}(t_{k+1}) - \mathbf{x}(t_{k-1})) + O((\Delta t)^2) = \mathbf{f}(\mathbf{x}(t_k), t_k)$$

$$\mathbf{x}(t_{k+1}) = \mathbf{x}(t_{k-1}) + 2\Delta t \mathbf{f}(\mathbf{x}(t_k), t_k) + O((\Delta t)^3)$$

- Stability of the leap frog method: $\frac{dx}{dt} = \lambda x$

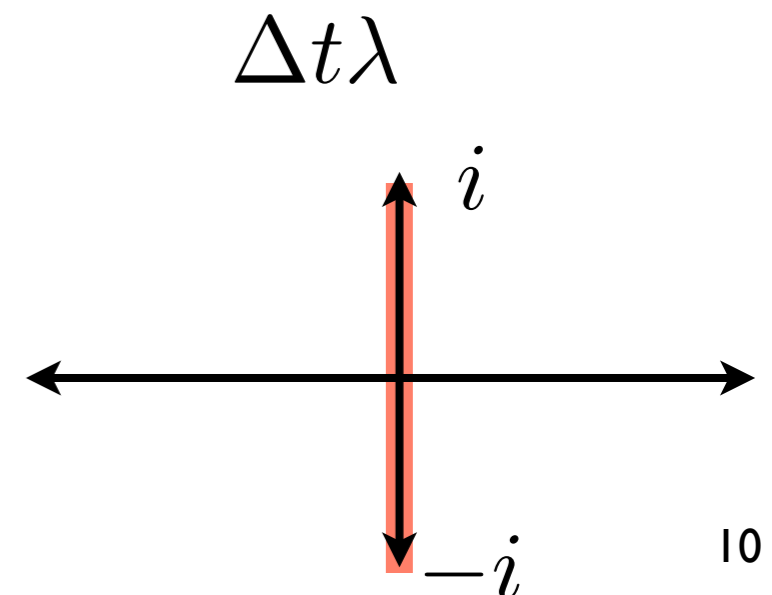
$$x_{k+1} = x_{k-1} + 2\Delta t \lambda x_k$$

- Both eigenvalues of \mathbf{C} must be bounded:

$$|\Delta t \lambda \pm \sqrt{(\Delta t \lambda)^2 + 1}| \leq 1$$

consider when $|\Delta t \lambda| \ll 1$:

$$|\Delta t \lambda \pm 1| \leq 1$$



Multistep Methods

- Exercise:
 - Should I use the leap frog method to integrate the equations of motion for a mass-spring system?

$$m \frac{d^2 x}{dt^2} = -kx$$

- If so, what time steps should I limit myself to?
- If not, what other integrator could I use?

Multistep Methods

- Exercise:
 - Should I use the leap frog method to integrate the equations of motion for a mass-spring system?

$$m \frac{d^2 x}{dt^2} = -kx$$

- Transform to system of first order ODEs:

$$\frac{d}{dt} \begin{pmatrix} v \\ x \end{pmatrix} = \begin{pmatrix} 0 & -k/m \\ 1 & 0 \end{pmatrix} \begin{pmatrix} v \\ x \end{pmatrix}$$

- Eigenvalues of matrix: $\lambda = \pm i \sqrt{\frac{k}{m}}$
- Since eigenvalues are imaginary, leap frog is stable when:

$$\Delta t < \sqrt{\frac{m}{k}}$$

Multistep Methods

- Multistep methods can be implicit as well such as the backward differentiation formulas or Adams-Moulton integrators.
- Example: Backwards differentiation

$$\mathbf{x}_{k+1} = \frac{4}{3}\mathbf{x}_k - \frac{1}{3}\mathbf{x}_{k-1} + \frac{2}{3}\Delta t \mathbf{f}(\mathbf{x}_{k+1}, t_{k+1})$$

- Second order accurate.
- How would you identify the stability bounds?

Numerical Integration

- Consider the definite integral:

$$\int_{t_0}^{t_f} \mathbf{f}(\tau) d\tau$$

- We can define a variable:

$$\mathbf{x}(t) = \int_{t_0}^t \mathbf{f}(\tau) d\tau$$

- which, if $\mathbf{f}(t)$ is continuous, satisfies the differential equation:

$$\frac{d}{dt} \mathbf{x}(t) = \mathbf{f}(t), \quad \mathbf{x}(t_0) = \mathbf{0}$$

- Thus, a definite integral of a known, continuous function can be determined using methods for ODE-IVPs to compute:

$$\mathbf{x}(t_f)$$

Numerical Integration

- Consider the definite integral:

$$\int_{t_0}^{t_f} \mathbf{f}(\tau) d\tau$$

- If the discontinuities in $\mathbf{f}(t)$ are known, then ODE-IVP solvers can be used in the domain between the discontinuities too!
- If the discontinuities in $\mathbf{f}(t)$ are unknown, then Monte-Carlo methods (discussed later are a better option).
- This approach is efficient with adaptive time stepping methods because an appropriate spacing between points can be chosen when t changes more or less rapidly with $\mathbf{f}(t)$
- For multi-dimensional integrals, this approach is not as straightforward, however.

Numerical Integration

- One alternative is integration by polynomial interpolation:

$$\int_{t_0}^{t_f} \mathbf{f}(\tau) d\tau = \sum_{k=1}^N \int_{t_{k-1}}^{t_k} \mathbf{f}(\tau) d\tau \approx \sum_{k=1}^N \int_{t_{k-1}}^{t_k} \mathbf{P}_k(\tau) d\tau$$

- where $\mathbf{P}_k(\tau)$ is a polynomial approximation of $\mathbf{f}(\tau)$ in the domain
$$\tau \in [t_{k-1}, t_k]$$
- If the size of the domains of integration and the order of the polynomial interpolant can be used to control the accuracy of the integration.
- Example: quadratic interpolation – Simpson's rule:

$$\mathbf{P}_k(\tau) = \mathbf{f}(t_{k-1}) + \frac{1}{t_k - t_{k-1}} (\mathbf{f}(t_k) - \mathbf{f}(t_{k-1})) (\tau - t_{k-1})$$

$$\int_{t_{k-1}}^{t_k} \mathbf{P}_k(\tau) d\tau = \frac{1}{2} (\mathbf{f}(t_k) + \mathbf{f}(t_{k-1})) (t_k - t_{k-1})$$

Numerical Integration

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$$\int_{t_0}^{t_f} \mathbf{f}(\tau) d\tau = \sum_{k=1}^N \int_{t_{k-1}}^{t_k} \mathbf{f}(\tau) d\tau \approx \sum_{k=1}^N \int_{t_{k-1}}^{t_k} \mathbf{P}_k(\tau) d\tau$$

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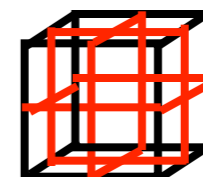
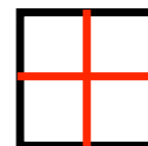
$$\int_{t_{k-1}}^{t_k} \mathbf{P}_k(\tau) d\tau = \frac{1}{6} (\mathbf{f}(t_k) + 4\mathbf{f}((t_k + t_{k-1})/2) + \mathbf{f}(t_{k-1})) (t_k - t_{k-1})$$

Numerical Integration

- Multidimensional integration:

- Of the sort:
$$\int_{y_L}^{y_U} \int_{z_L}^{z_U} \mathbf{f}(y, z) dy dz$$

- For any number of dimensions larger than 3, this is best handled with Monte Carlo methods
- For dimensions less than 3, this integration can be done with polynomial interpolation.
 - Fit the function to a polynomial of a prescribed degree within small regions of the domain of integration.
 - Sum integrals over the polynomial fits in each fit region.
 - This fails with higher dimensions because the number of fit regions grows exponentially with dimension.
- Example:



Numerical Integration

- Improper integrals:

- Of the sort:
$$\int_{t_0}^{\infty} \mathbf{f}(\tau) d\tau$$

- Can be split into two domains of integration

$$\int_{t_0}^{\infty} \mathbf{f}(\tau) d\tau = \int_{t_0}^{t_f} \mathbf{f}(\tau) d\tau + \int_{t_f}^{\infty} \mathbf{f}(\tau) d\tau$$

- The first integral can be handled with ODE-IVP methods or polynomial interpolation
- The second must be handled separately through either:
 - transformation onto a finite domain
 - or substitution of an asymptotic approximation
- This same idea applies to integrable singularities as well.

Numerical Integration

- Improper integrals:
 - Example:

$$\begin{aligned} & \int_0^{t_f} \frac{\cos \tau}{\sqrt{\tau}} d\tau \\ & \approx \int_0^{t_0} \frac{1 - \tau^2/2}{\sqrt{\tau}} d\tau + \int_{t_0}^{t_f} \frac{\cos \tau}{\sqrt{\tau}} d\tau \\ & \approx 2t_0^{1/2} - \frac{1}{5}t_0^{5/2} + \int_{t_0}^{t_f} \frac{\cos \tau}{\sqrt{\tau}} d\tau \end{aligned}$$

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