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# Initiating Kinetic Plasma Simulation

Prof. Dr. Mohammed Shihab

Plasma Technology



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# Charged Particle in Electric Field

□ Equation of motion:

$$m \frac{d^2x}{dt^2} = q E(x) \quad \& \quad E(x) = - k x$$

**Result: Harmonic Oscillator**

□ Equivalent to Equation of motion:

$$\frac{d^2x}{dt^2} + \omega^2 x = 0 \quad \& \quad \omega = \text{sqrt}(q k/m)$$

□ Represents oscillatory motion



# Exact and Euler Solution

□ **Exact Solution: Motion is periodic**  
 $x(t) = x_0 \cos(\omega t) + (v_0/\omega) \sin(\omega t)$

□ **Euler: Convert to first-order system:**  
 $dx/dt = v,$   
 $dv/dt = a,$

## Discretization

$$v(n+1) = v(n) + a(n) dt$$
$$x(n+1) = x(n) + v(n+1) dt$$

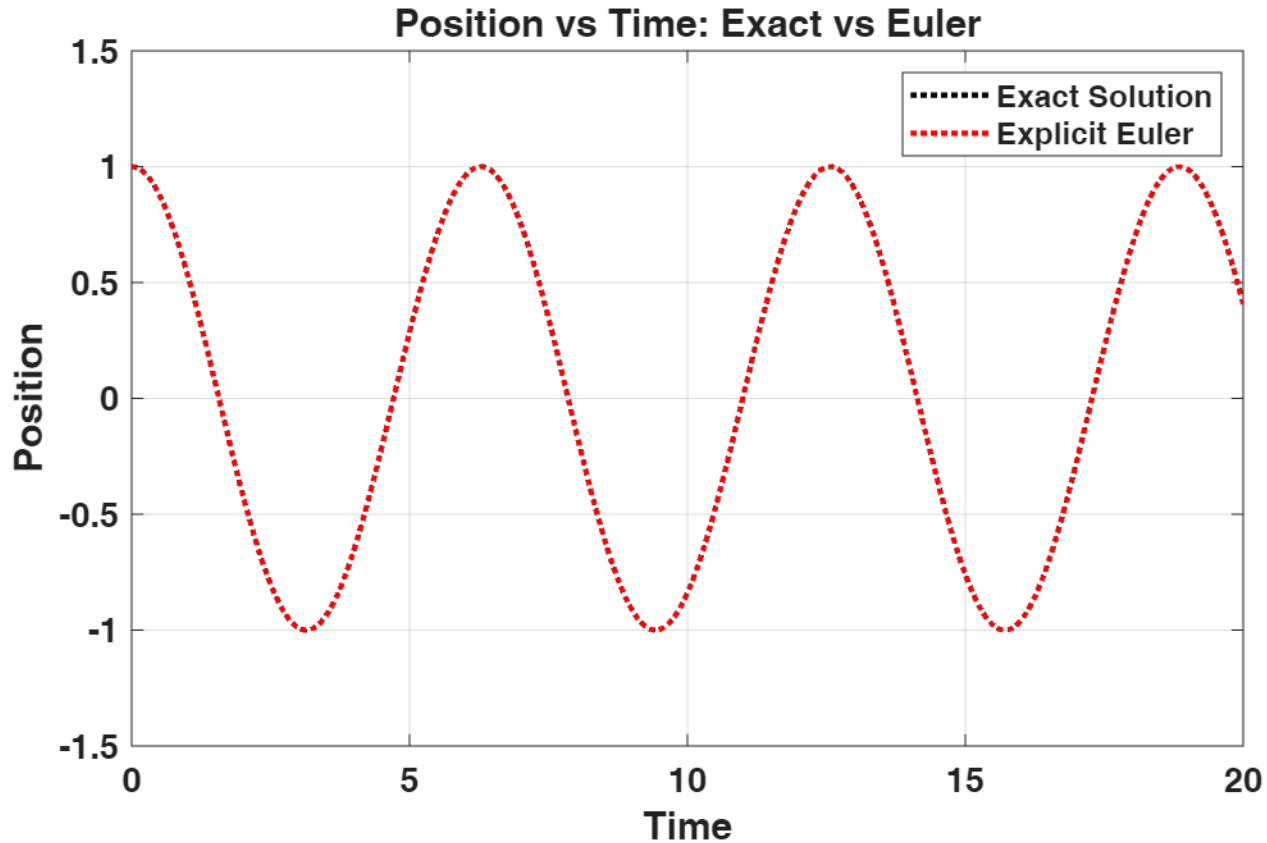
(semi-explicit)

Where:  $a(n) = -(q k/ m) x(n)$





# Exact Solution Vs Numerical Solution



□ **Position of the particle:  $dt=0.001$**

Exact: sinusoidal motion

Euler: very close to the exact solution



# Exact Solution Vs Numerical Solution

## ❑ Advantage:

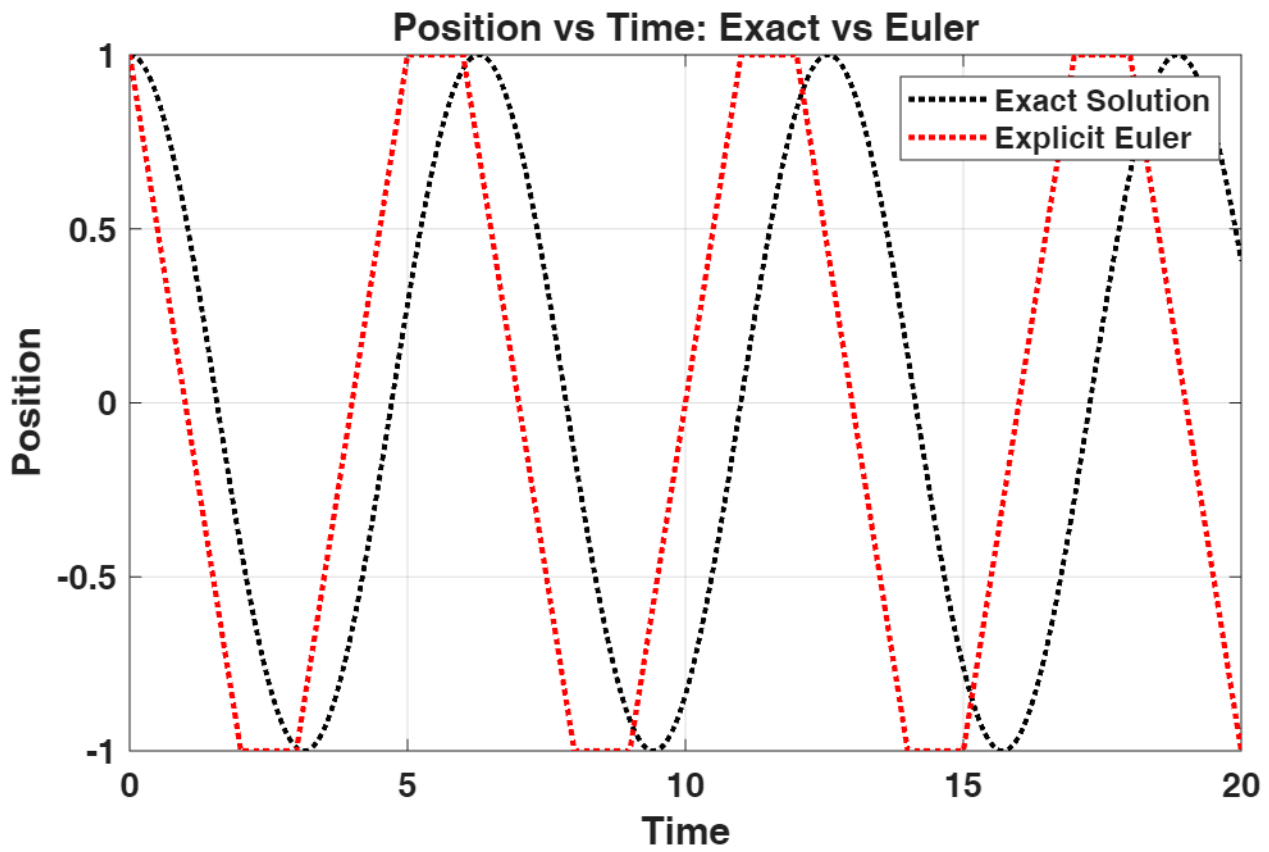
- Simple and easy
- Low computational cost
- Good for basic problems

## ❑ Disadvantage:

- Low accuracy
- Energy not conserved (numerical heating)
- Unstable for oscillations



# Exact Solution Vs Numerical Solution



❑ Larger dt:  $dt=1$

Euler: phase and amplitude errors

Deviation increases with time



# Numerical Instability

## ❑ Advantage:

- Condition:  $\omega dt \ll 1$
- Large  $dt$  leads to instability
- Causes artificial heating

## ❑ Disadvantage:

- Low accuracy
- Energy not conserved
- Unstable for oscillations





# Numerical Errors

## □ Taylor Expansion

$$x(t+dt) = x(t) + dt x'(t) + (dt^2/2) x''(t) + O(dt^3)$$

$$\text{Remember: } v(n+1) = v(n) + a(n) dt$$
$$x(n+1) = x(n) + v(n+1) dt$$

Euler keeps only first two terms

## □ Local Truncation Error

Ignored term:  $(dt^2/2) x''(t)$

Local error per step  $\sim O(dt^2)$

## □ Global Error

Number of steps  $N = T/dt$

Global error =  $N \times \text{local error} = (T/dt)(dt^2) = O(dt)$



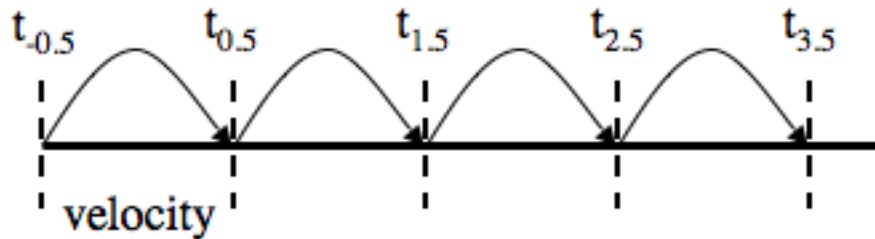
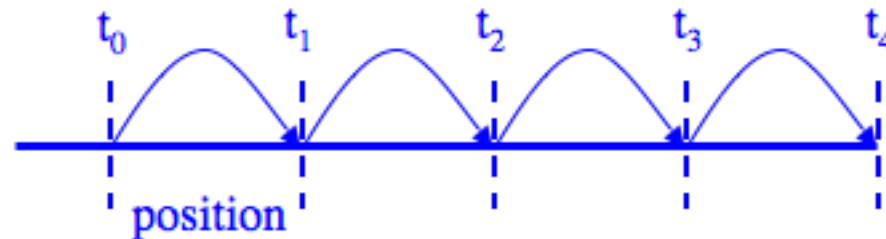
# Leapfrog Method

- ❑ Explicit time integration method
- ❑ Widely used for charged particle dynamics
- ❑ Better than Euler for oscillatory systems
- ❑ Time Staggering

$$x^n = x(n\Delta t)$$

$$v^{n+1/2} = v((n+1/2)\Delta t)$$

Improves stability and accuracy.





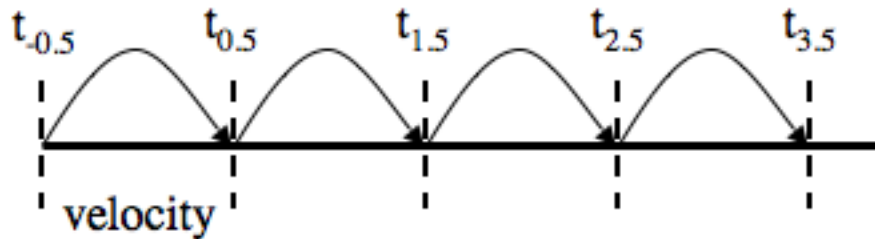
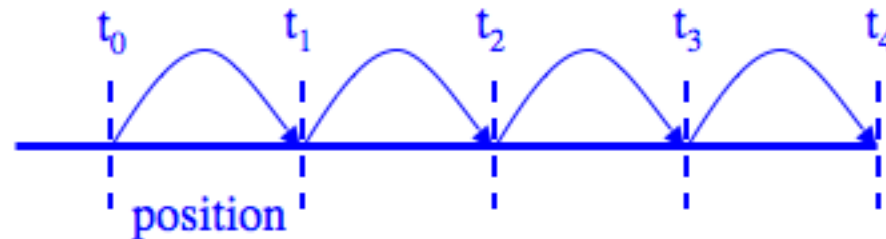
# Leapfrog Method

$$\mathbf{v}^{n+1/2} = \mathbf{v}^{n-1/2} + \mathbf{a}(\mathbf{x}^n) \Delta t$$

$$\mathbf{x}^{n+1} = \mathbf{x}^n + \mathbf{v}^{n+1/2} \Delta t$$

➤ Need initial half-step velocity

$$\mathbf{v}^{1/2} = \mathbf{v}^0 + (\Delta t/2) \mathbf{a}(\mathbf{x}^0)$$



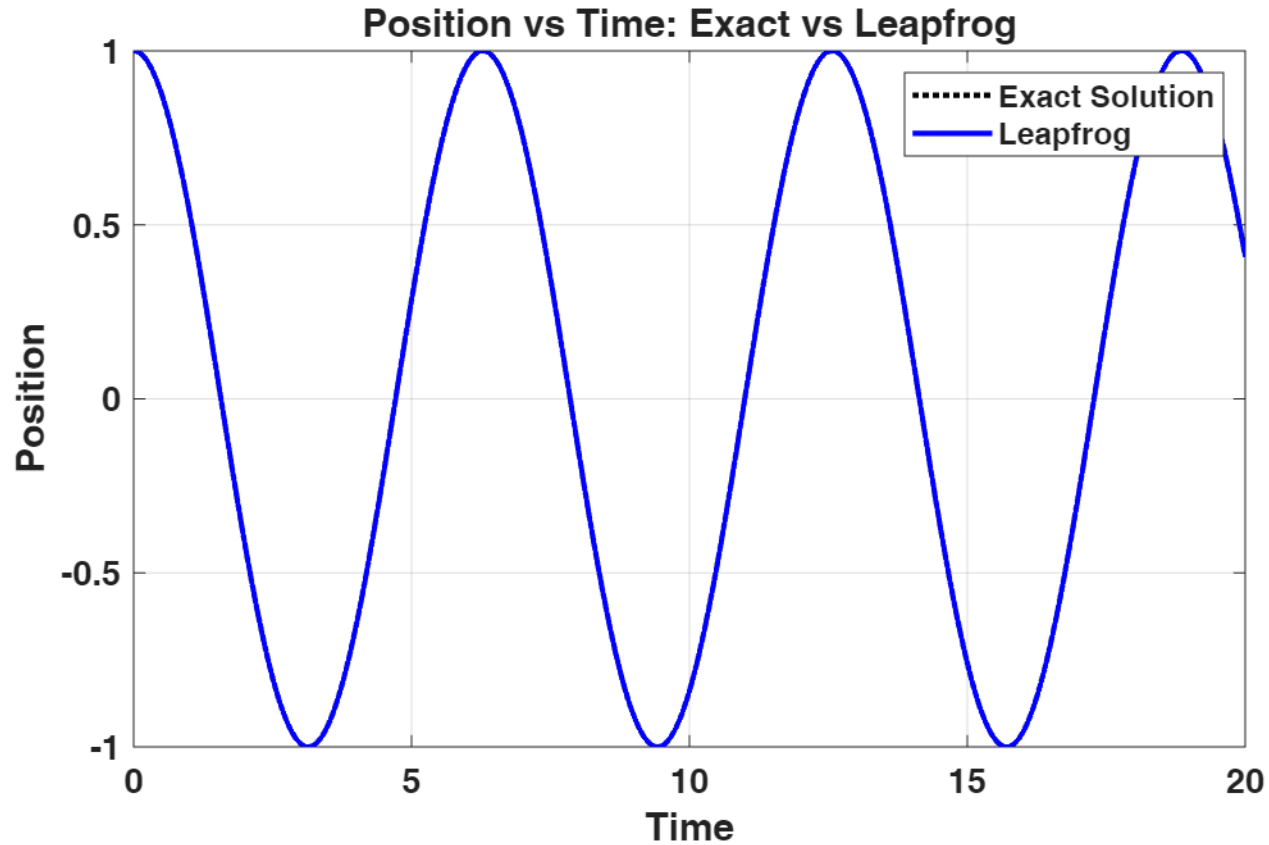


# Leapfrog Method

- ❑ Second-order accuracy
  - ❑ Better energy behavior
  - ❑ Time-centered scheme
  - ❑ Stable for oscillatory motion
- 
- Euler: first-order, energy drift
  - Leapfrog: second-order, stable
  - Much better for plasma simulations
- 
- ❑ For harmonic oscillator:  $\omega\Delta t < 2$
  - ❑ Too large  $\Delta t \rightarrow$  instability



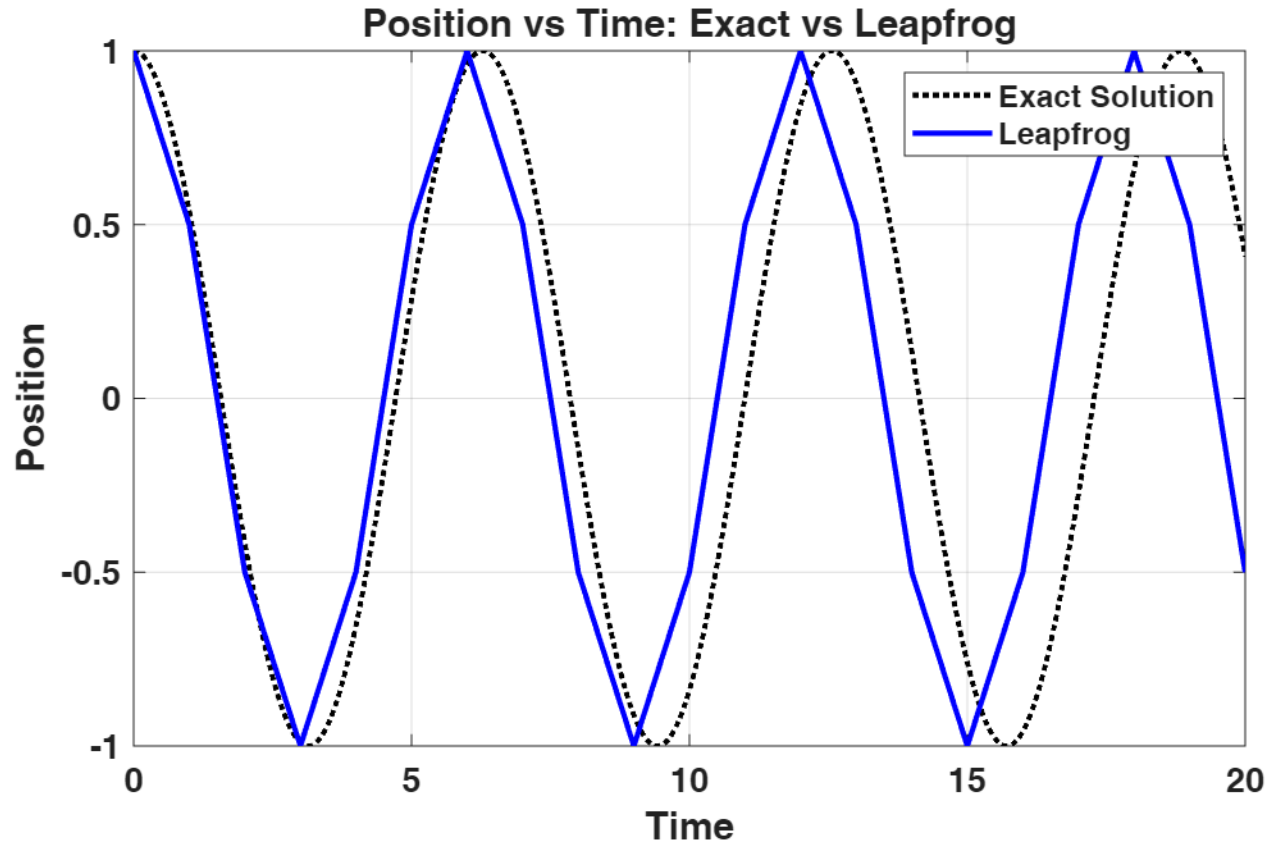
# Leapfrog Method



□  $\Delta t \rightarrow 0.001$



# Leapfrog Method



□  $\Delta t \rightarrow 1$



# Leapfrog Method

$$x(t+\Delta t) = x + v\Delta t + (\Delta t^2/2) a + O(\Delta t^3)$$

- ❑ Difference appears at  $O(\Delta t^3)$ 
  - Local error per step  $\sim O(\Delta t^3)$
- ❑ Number of steps  $N = T/\Delta t$ 
  - Global error =  $N \times$  local error  
 $= (T/\Delta t)(\Delta t^3) = O(\Delta t^2)$

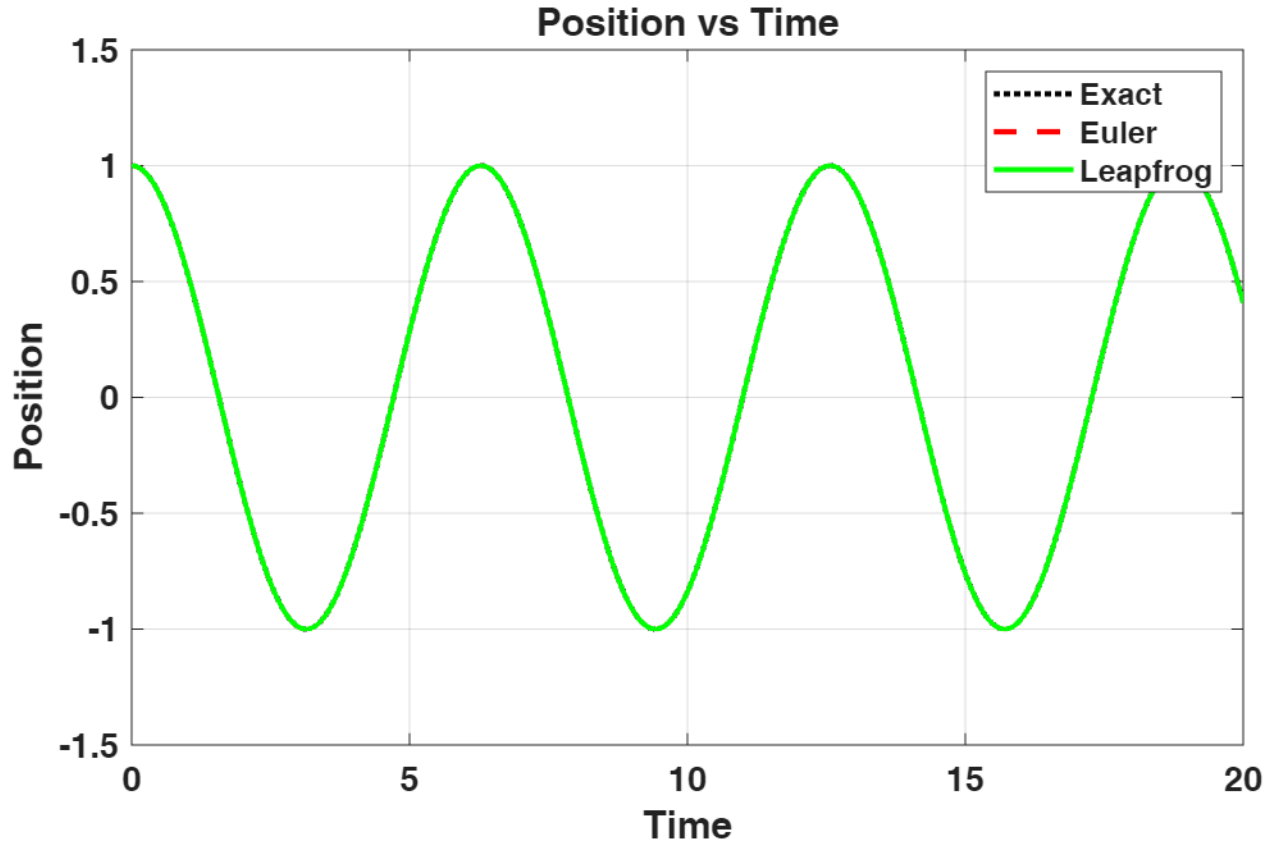
Euler: uses slope at start  $\rightarrow O(\Delta t)$

Leapfrog: uses midpoint  $\rightarrow O(\Delta t^2)$





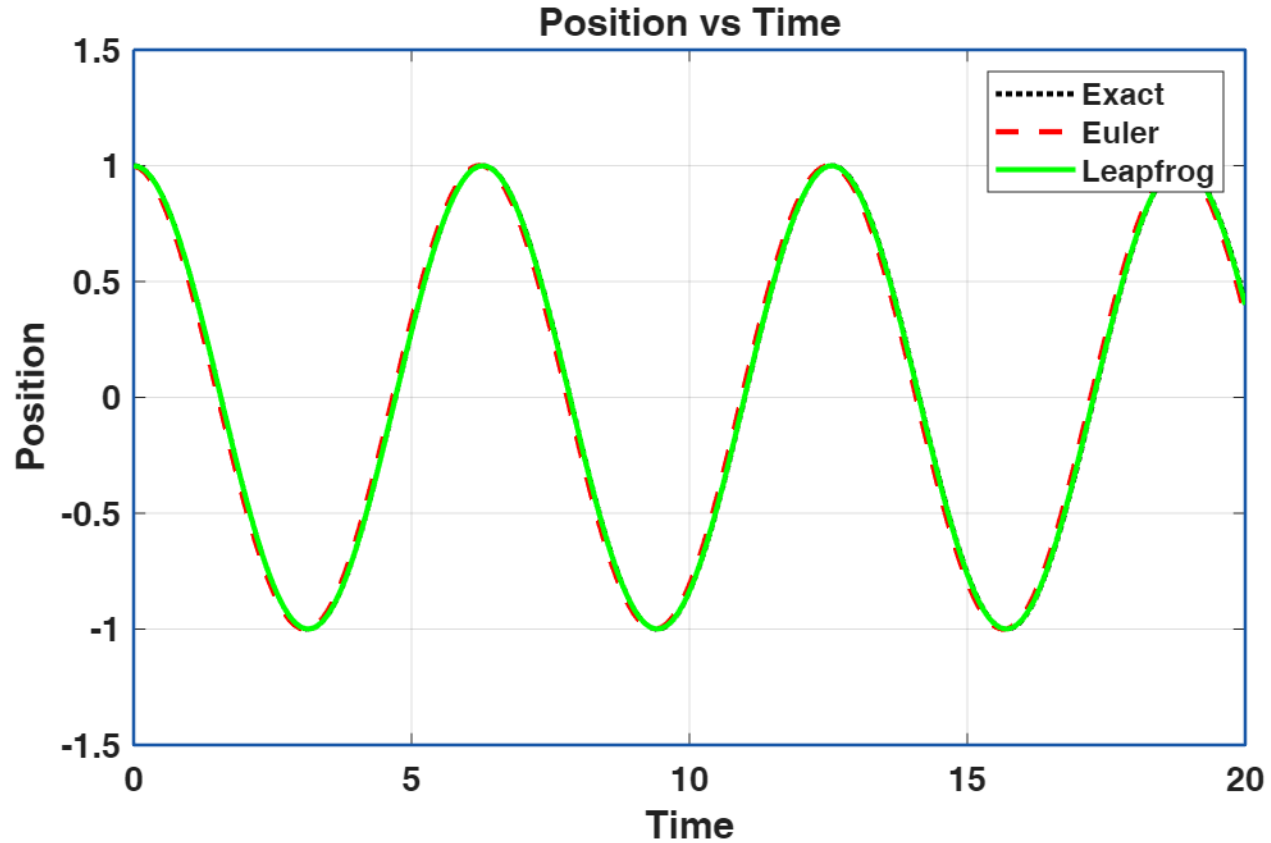
# Euler vs Leapfrog



□  $dt=0.001$



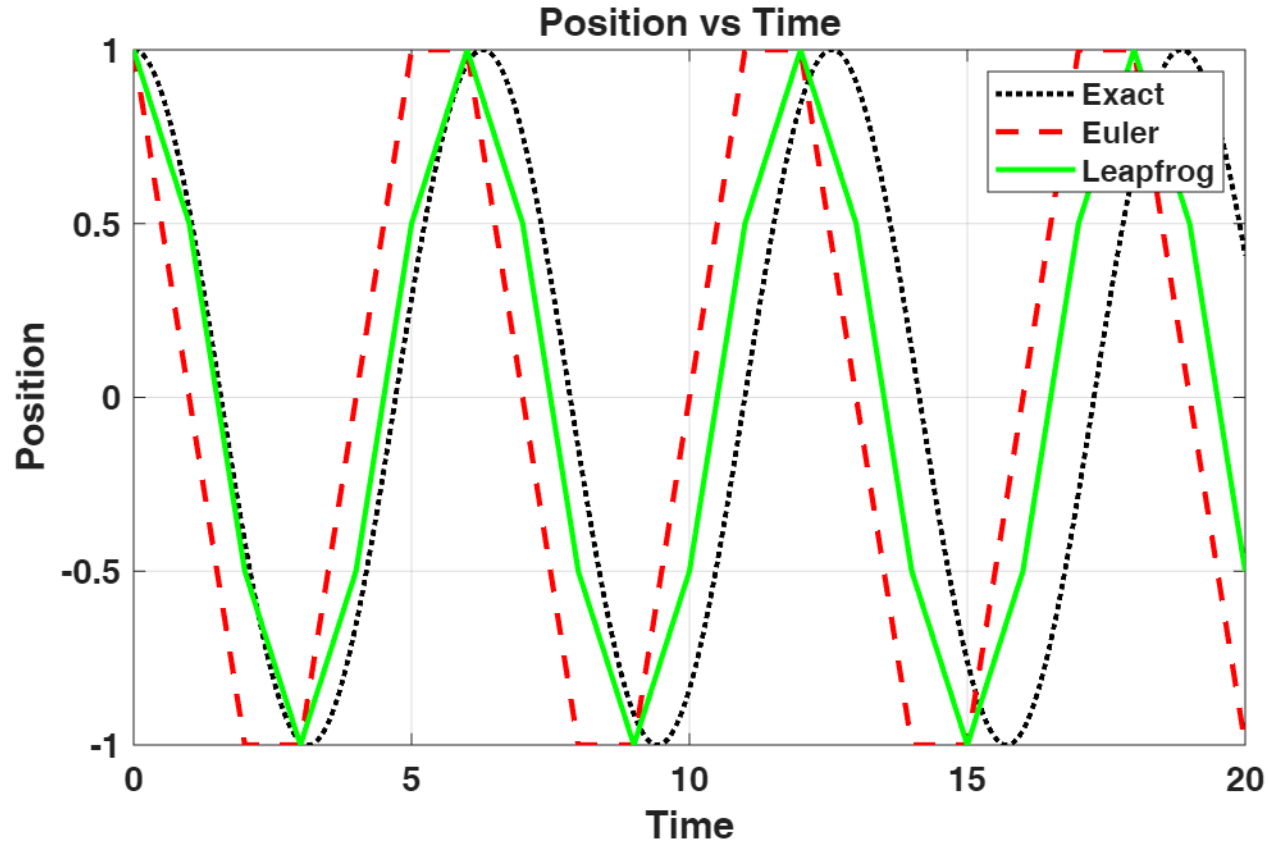
# Euler vs Leapfrog



□  $dt=0.1$



# Euler vs Leapfrog (explicit)



□  $dt=1$



# Explicit vs implicit Euler

□ Explicit Euler (Forward):

$$v(n+1) = v(n) + a(n) \Delta t$$

$$x(n+1) = x(n) + v(n) \Delta t$$

Uses current state only (no solve)

Implicit Euler (Backward):

$$v(n+1) = v(n) + a(n+1) \Delta t$$

$$x(n+1) = x(n) + v(n+1) \Delta t$$

Requires solving for future state





# Explicit vs implicit Euler

Both methods: first-order accuracy (global error  $O(\Delta t)$ )

Explicit Euler: Unstable for oscillations

- Energy increases (numerical heating)
- Phase error grows

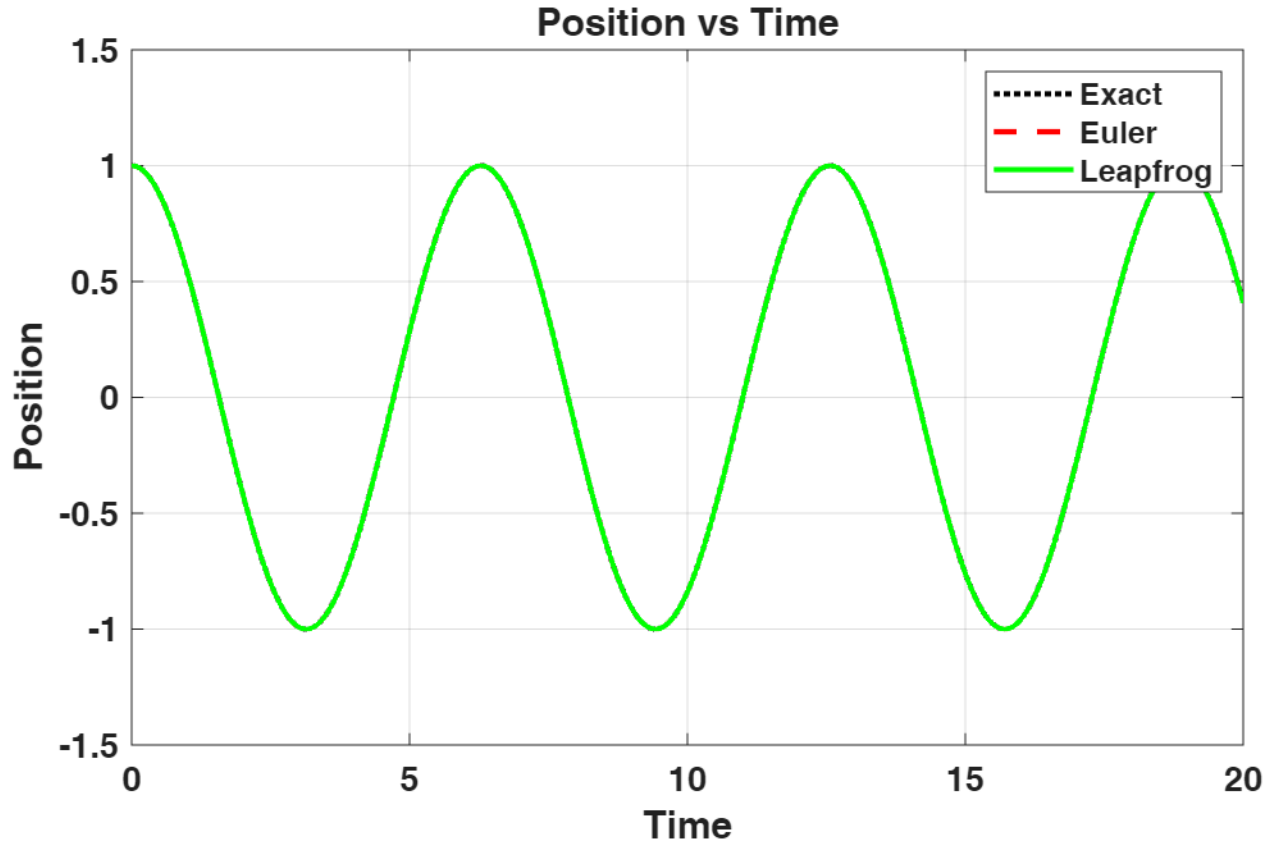
Implicit Euler: Unconditionally stable

- Energy decreases (numerical damping)
- Oscillations decay

Conclusion: same order, opposite energy behavior



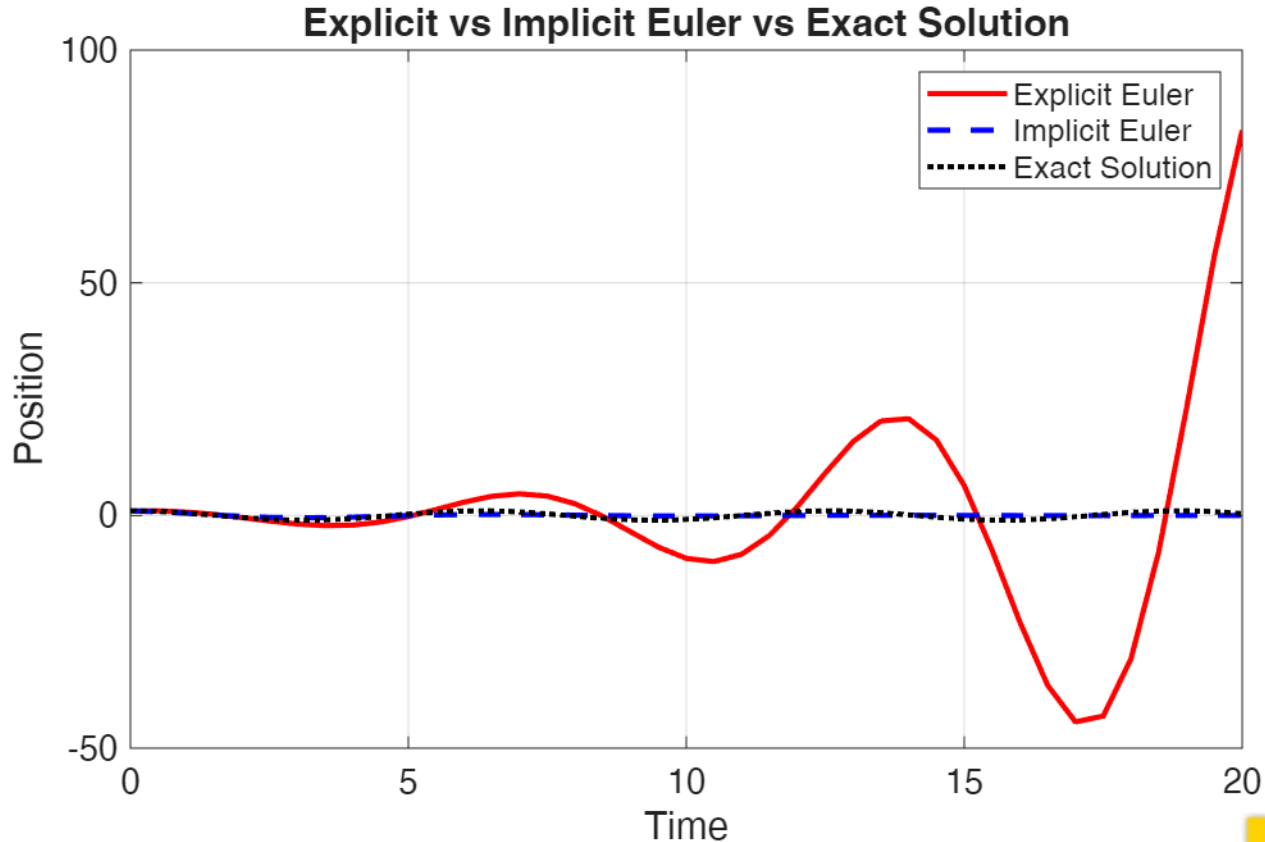
# Implicit vs explicit Euler schemes



□  $dt=0.001$



# Implicit vs explicit Euler schemes

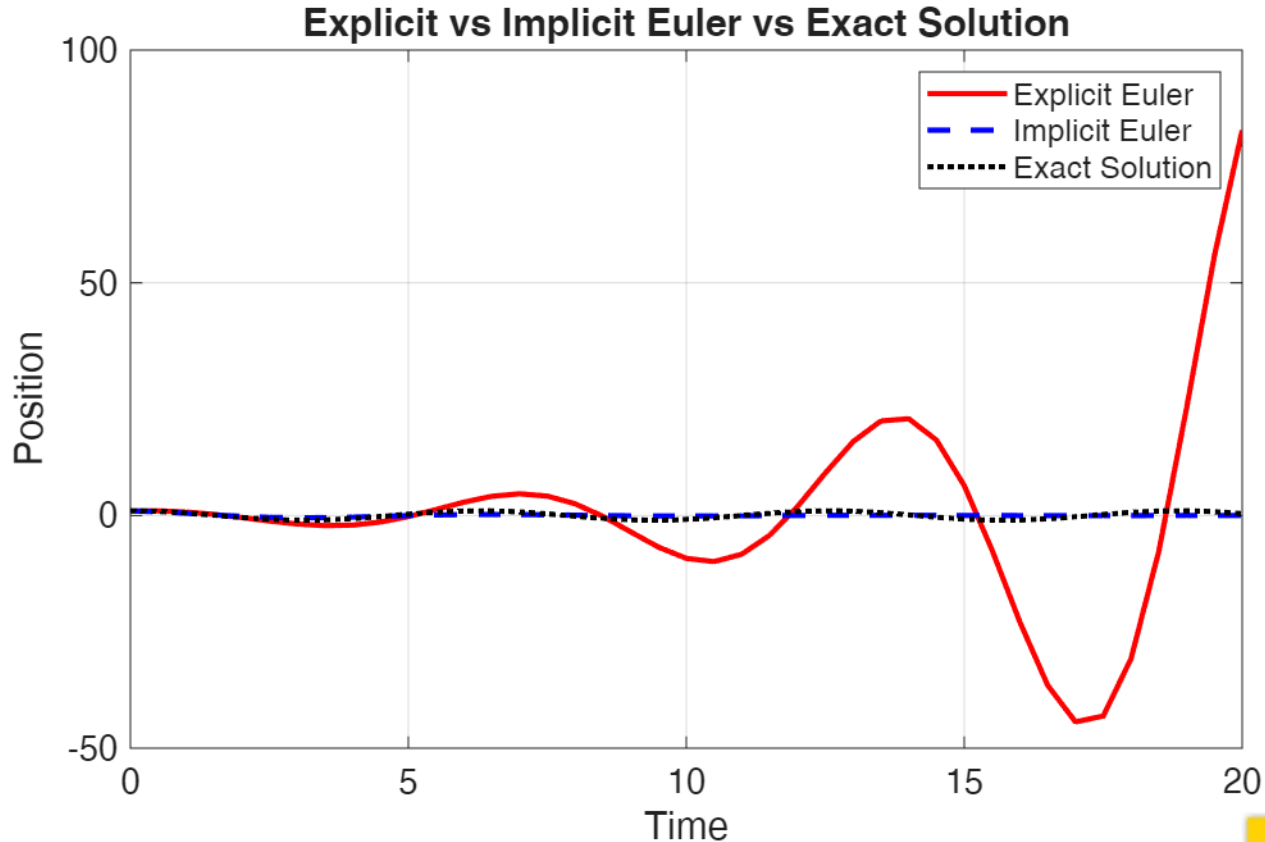


- $dt=0.5$
- Do not do simulation without experiments or analytical models





# Implicit vs explicit Euler schemes



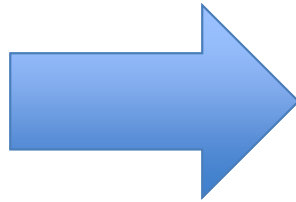
- $dt=0.5$
- Do not do simulation without experiments or analytical models



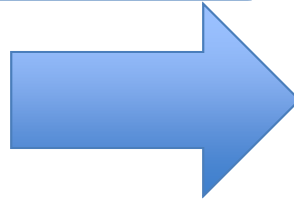


# Newton Method Flow (Implicit Scheme)

Initial Guess  
 $x^{(k)}$

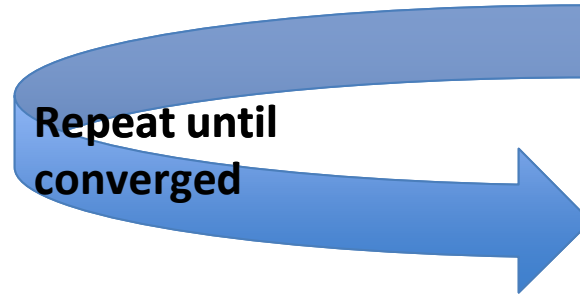


Compute Residual  
 $R(x^{(k)})$



Correction Step  
 $x^{(k+1)} = x^{(k)} - R/R'$

Repeat until  
converged





# Explicit vs implicit Euler

□ Implicit Euler:  $y(n+1) = y(n) + \Delta t f(y(n+1))$

Unknown  $y(n+1)$  appears inside  $f(\cdot)$

Rearrange

$$\rightarrow R(y(n+1)) = y(n+1) - y(n) - \Delta t f(y(n+1)) = 0$$

□ This is a nonlinear equation

We must solve for  $y(n+1)$  at every time step



# Newton-Raphson method

- ❑ Newton iteration:  $y(k+1) = y(k) - R(y(k))/R'(y(k))$
- ❑ Efficient for nonlinear equations
- ❑ Fast (quadratic) convergence near solution
- ❑ Works for general nonlinear forces (plasma physics)





# Newton-Raphson method

- ❑ Newton iteration:  $y(k+1) = y(k) - R(y(k))/R'(y(k))$
- ❑ Efficient for nonlinear equations
- ❑ Fast (quadratic) convergence near solution
- ❑ Works for general nonlinear forces (plasma physics)





# Implicit Midpoint Method: Concept

- ❑ Time-centered implicit integration method
- ❑ Evaluates system at midpoint of time step
- ❑ High accuracy for oscillatory systems
- ❑ Widely used in physics simulations

➤  $dx/dt = v, dv/dt = a(x)$

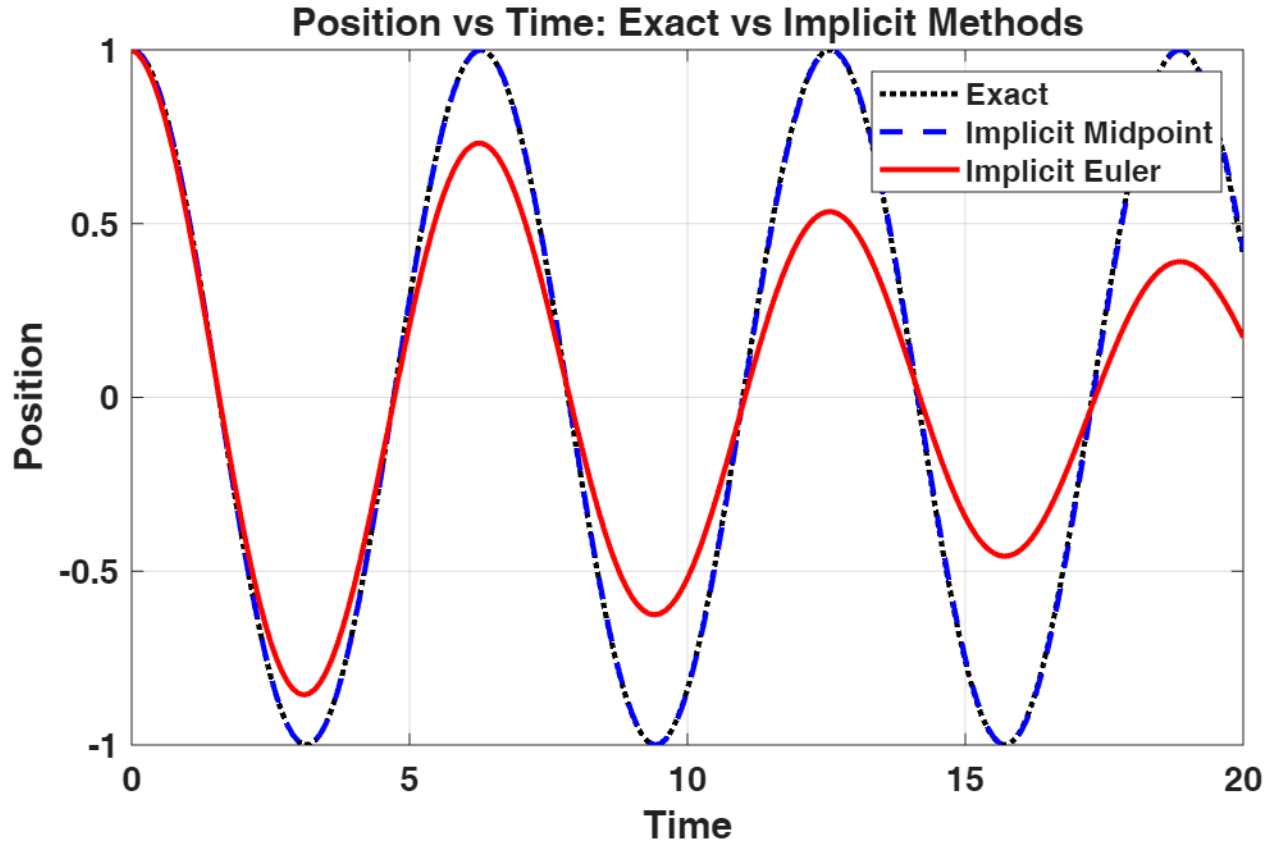
➤  $x(n+1) = x(n) + \Delta t (v(n) + v(n+1))/2$

➤  $v(n+1) = v(n) + \Delta t a((x(n) + x(n+1))/2)$

Solve nonlinear system (Newton method)



# Implicit Midpoint Vs Euler



- $dt=0.1$
- Midpoint: Second-order accurate ( $O(\Delta t^2)$ )
- Good energy conservation



# Implicit Leapfrog Method: Concept

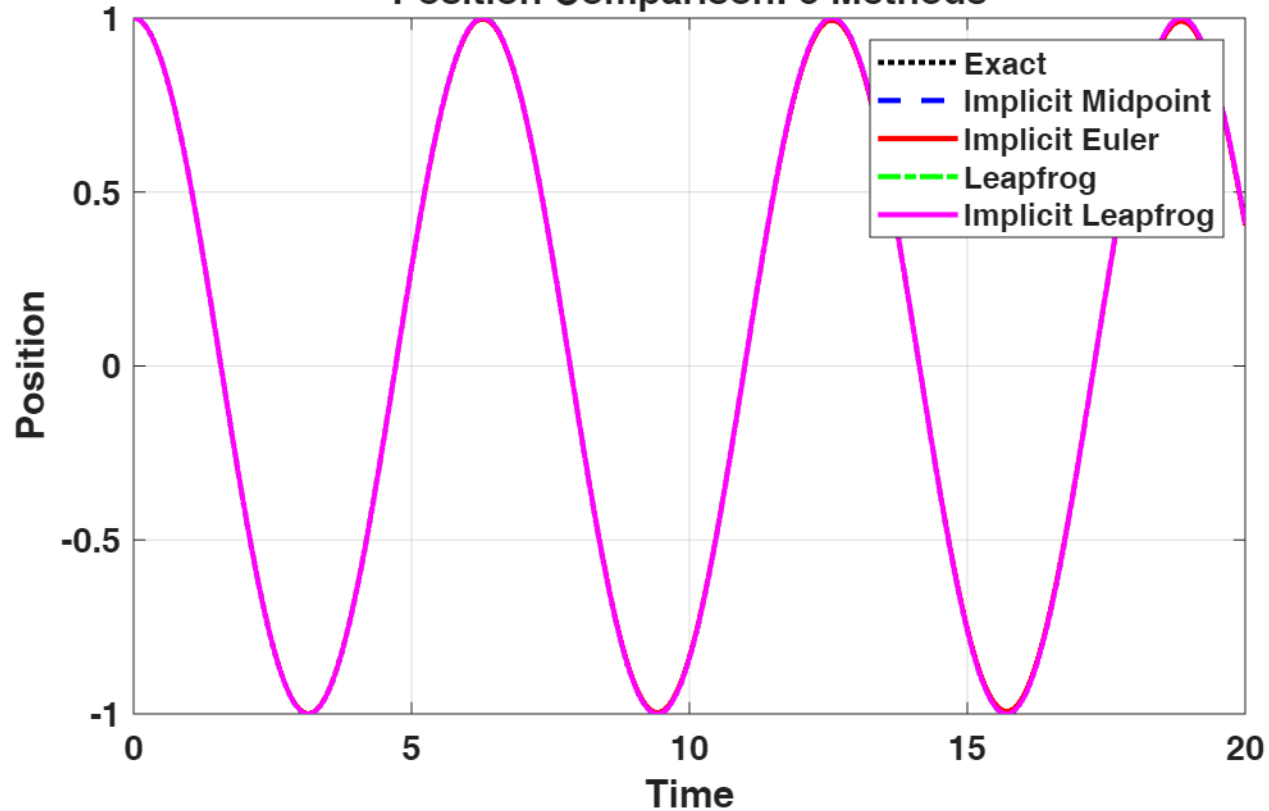
- Time-centered integration scheme, Force evaluated at midpoint position, Improves accuracy for oscillatory motion, Implicit because future position appears in force
- $V(n+1/2) = v(n) + (\Delta t/2) a((x(n) + x(n+1))/2)$
- $X(n+1) = x(n) + \Delta t v^{(n+1/2)}$
- $v^{n+1} = v(n) + \Delta t a((x(n) + x(n+1))/2)$
- Second-order accurate ( $O(\Delta t^2)$ ) and More stable than explicit Euler





# Comparison

Position Comparison: 5 Methods

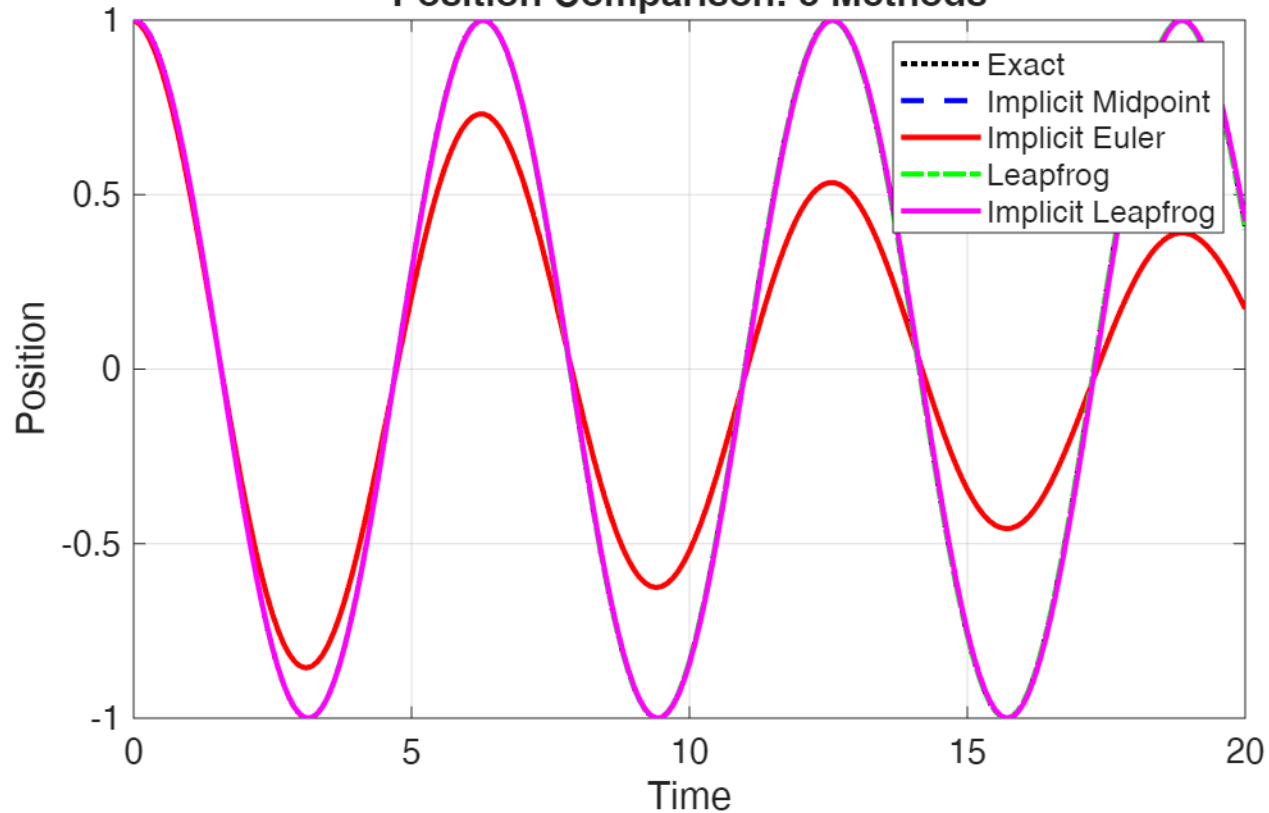


□ dt=0.001



# Comparison

Position Comparison: 5 Methods

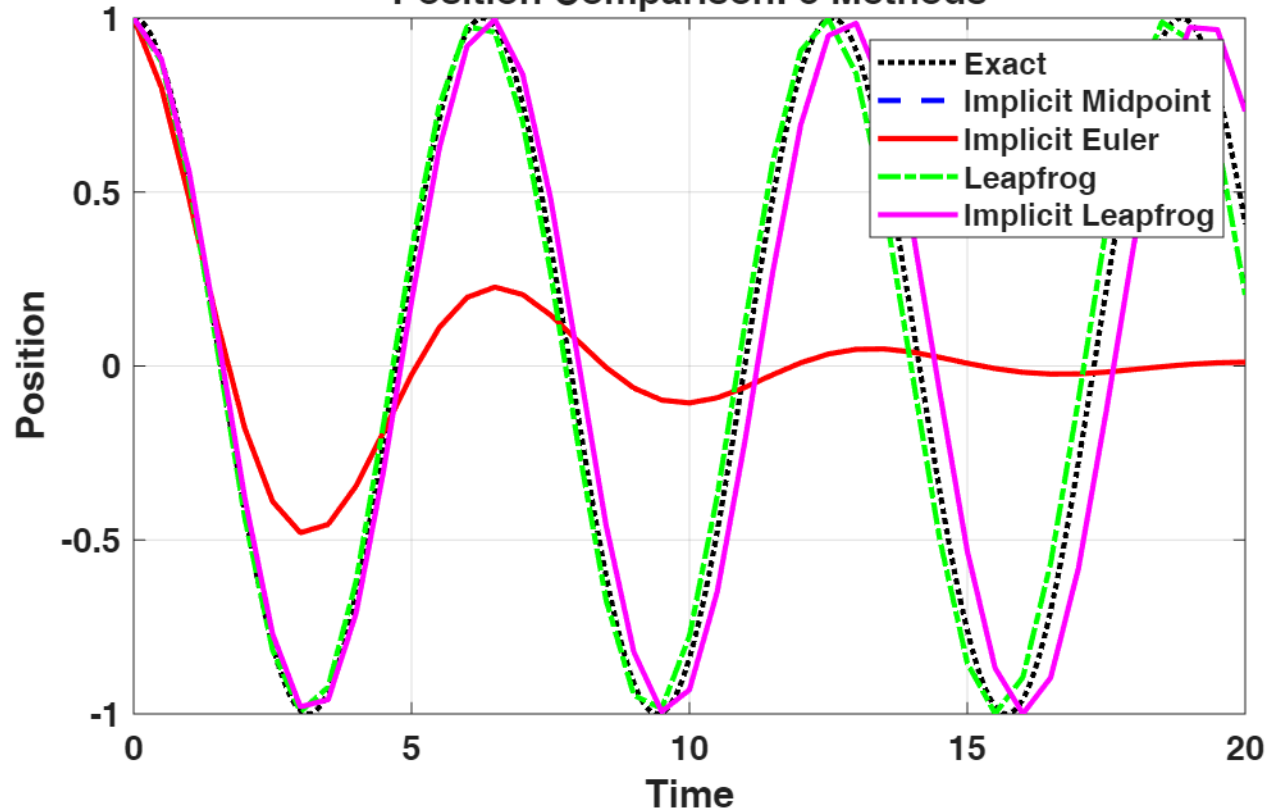


□ dt=0.01



# Comparison

Position Comparison: 5 Methods

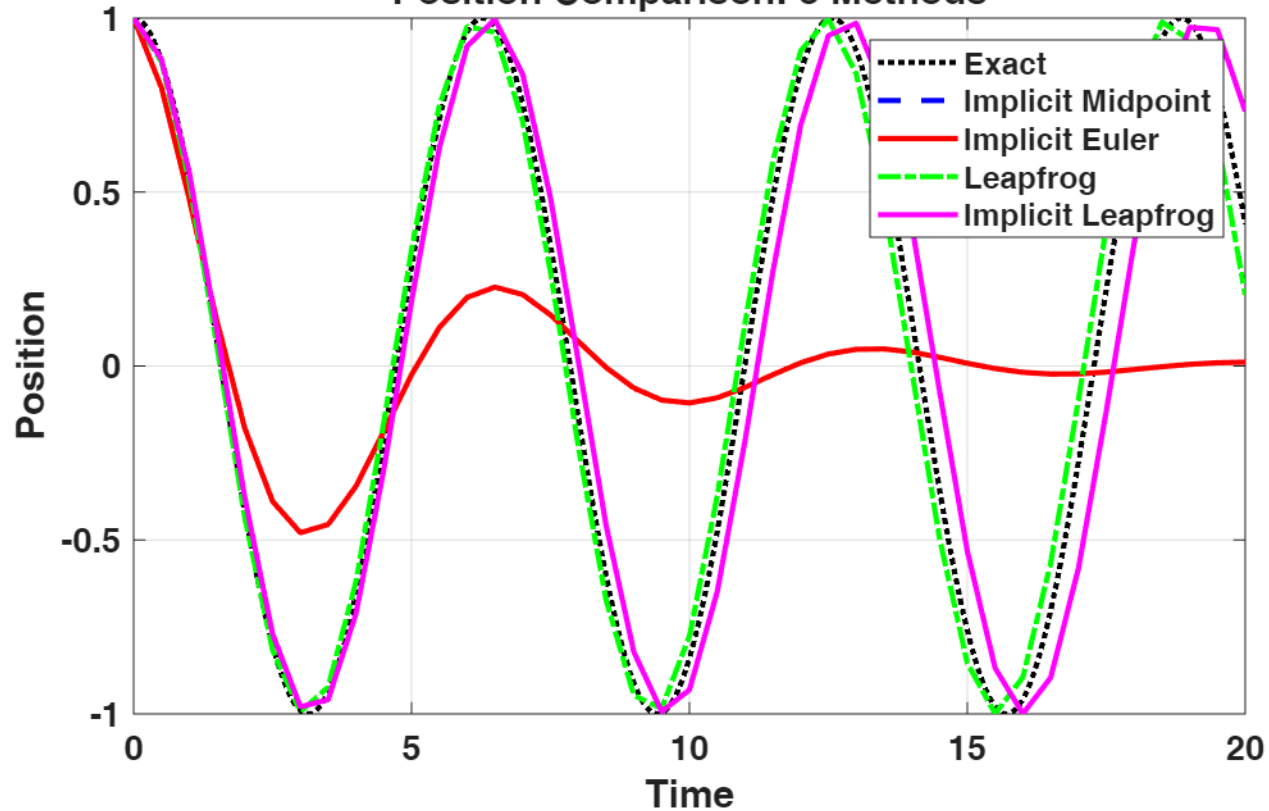


□  $dt=0.01$



# Comparison

Position Comparison: 5 Methods

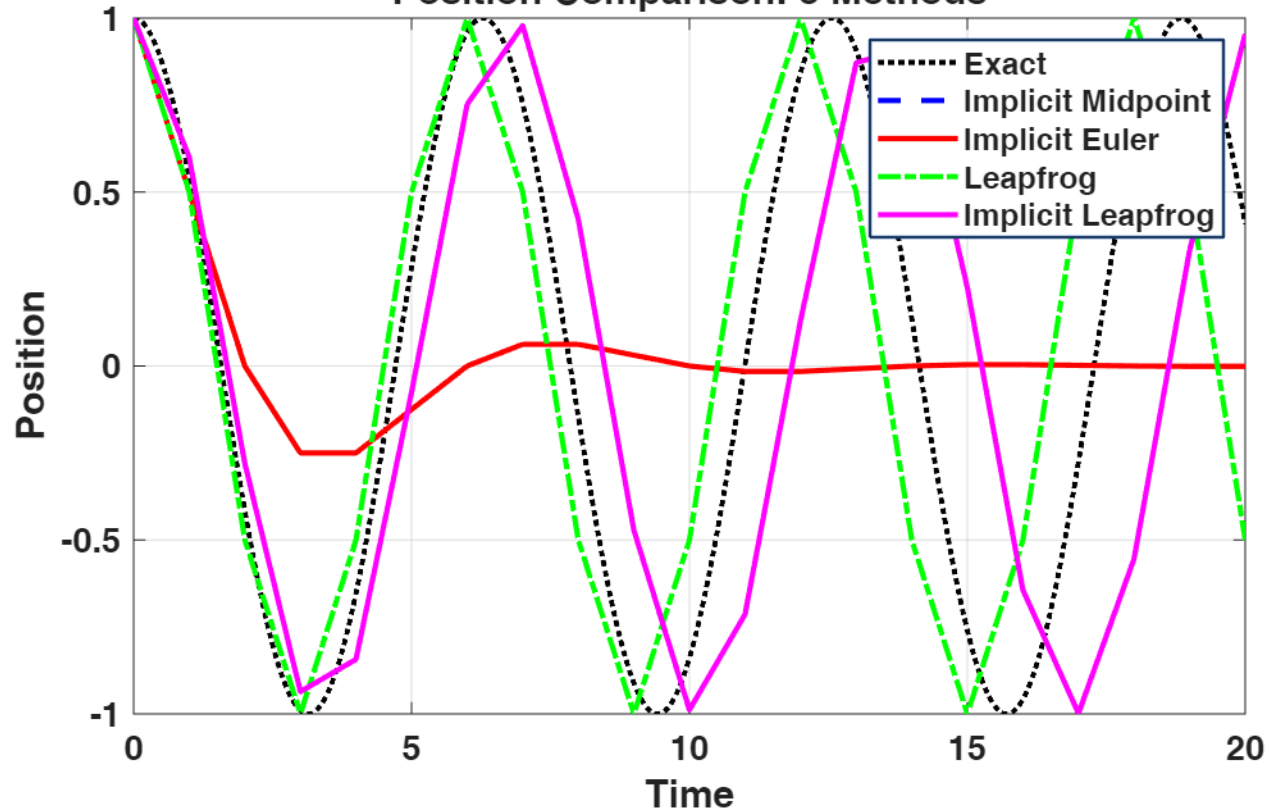


□ dt=0.5



# Comparison

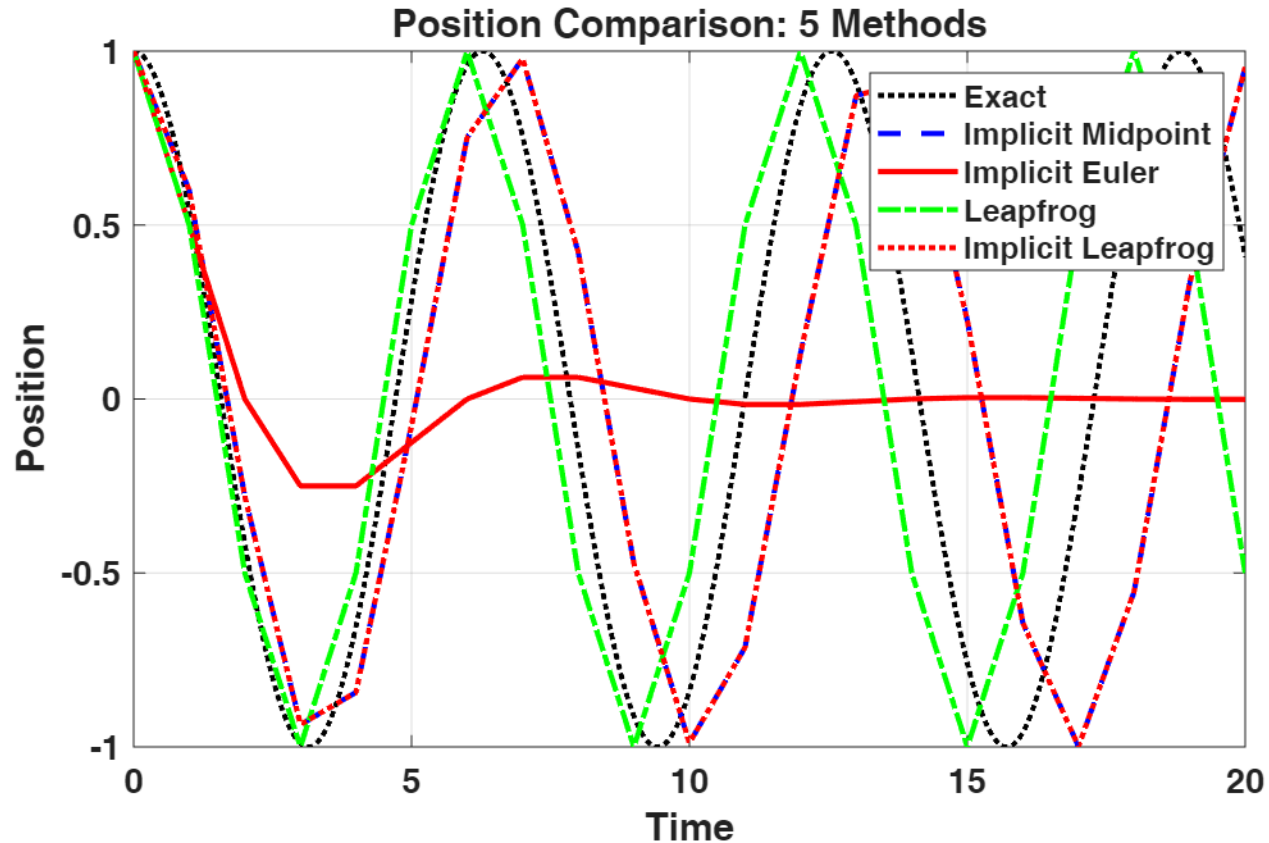
Position Comparison: 5 Methods



- $dt=1$
- We have to consider  $\omega dt \ll 1$



# Comparison



- $dt=1$
- We have to consider  $\omega dt \ll 1$
- Leapfrog is the best one.

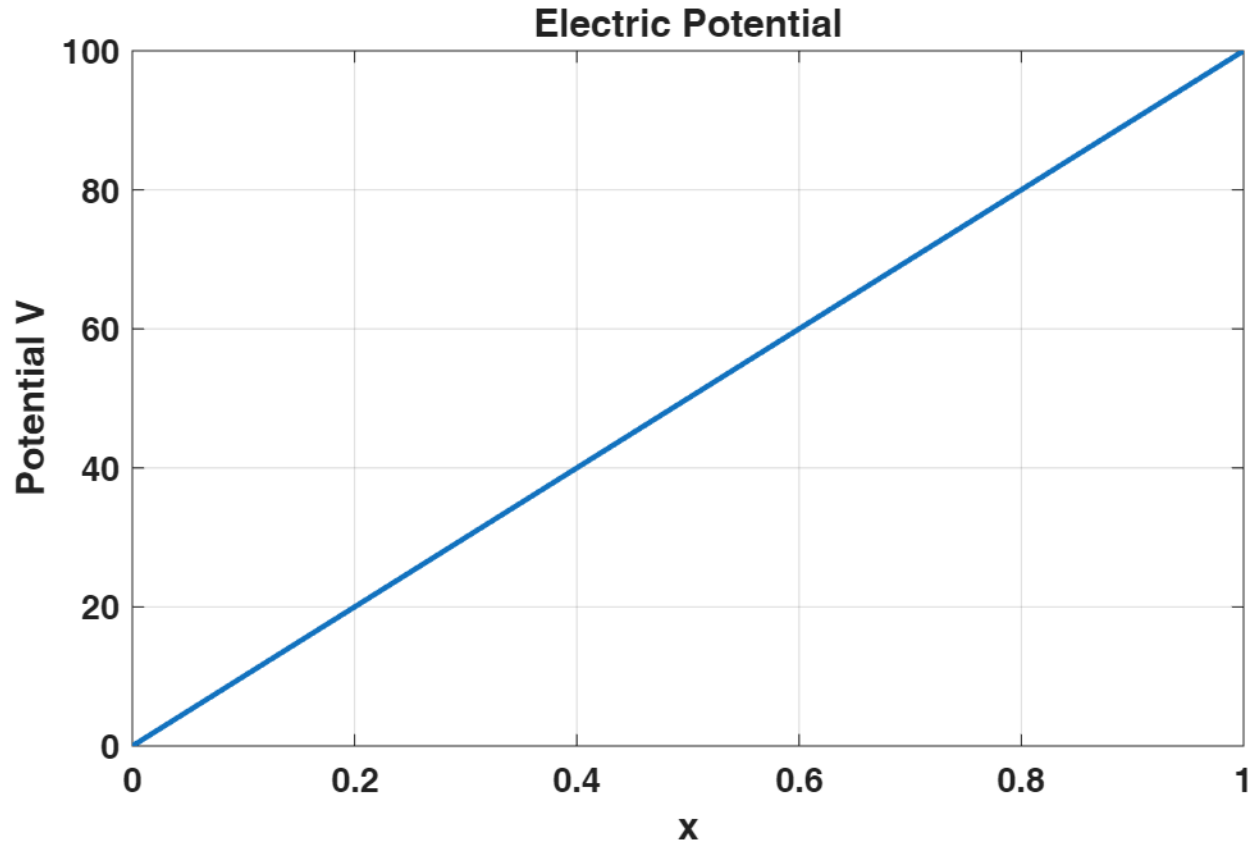


# 1D Electrostatic Problem

- Equation:  $d^2V/dx^2 = 0$
- Exact solution:  $V(x) = 100 * x / L$
- Electric field:  $E = -dV/dx = \text{constant}$
- Finite difference:  $V(i) = (V(i+1) + V(i-1))) / 2$

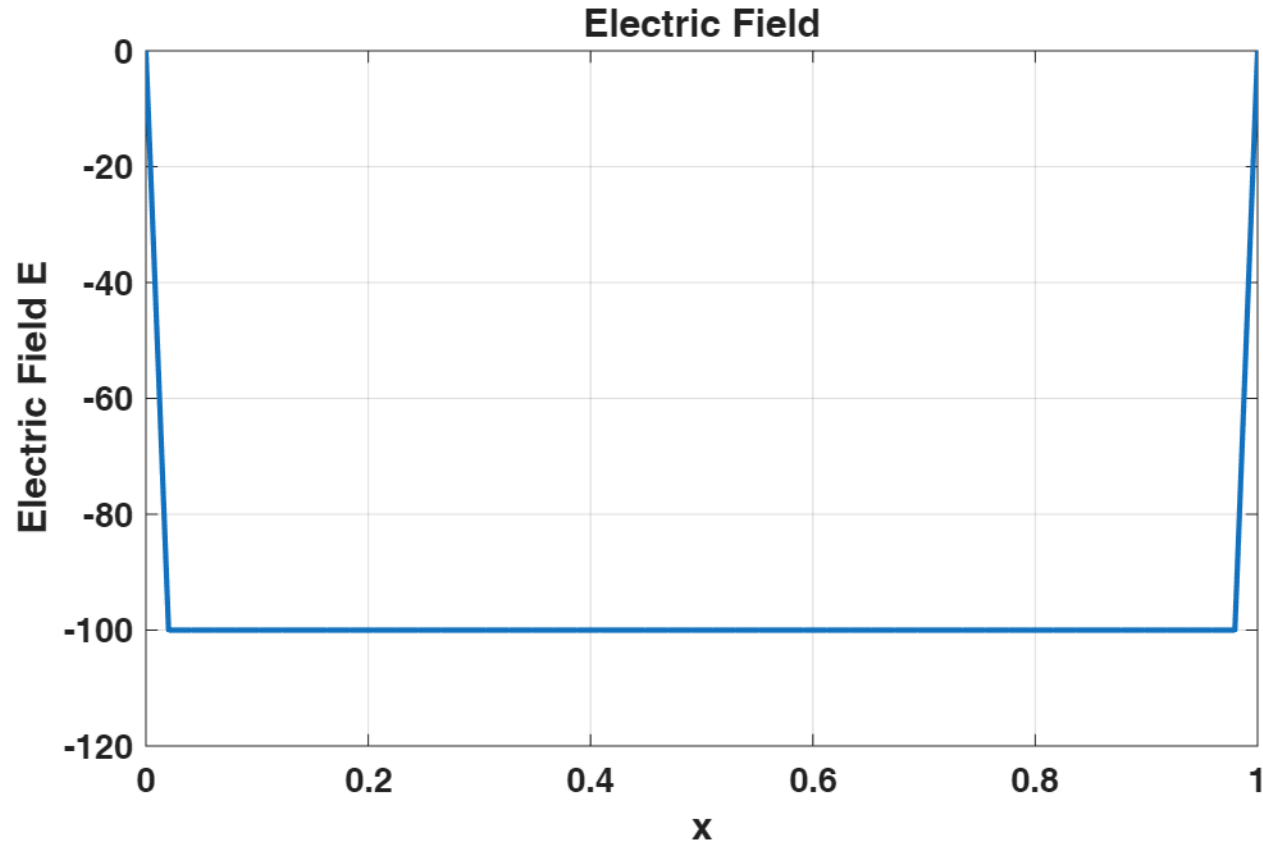


# 1D Electrostatic Problem





# 1D Electrostatic Problem





# Electrostatic Problem Setup

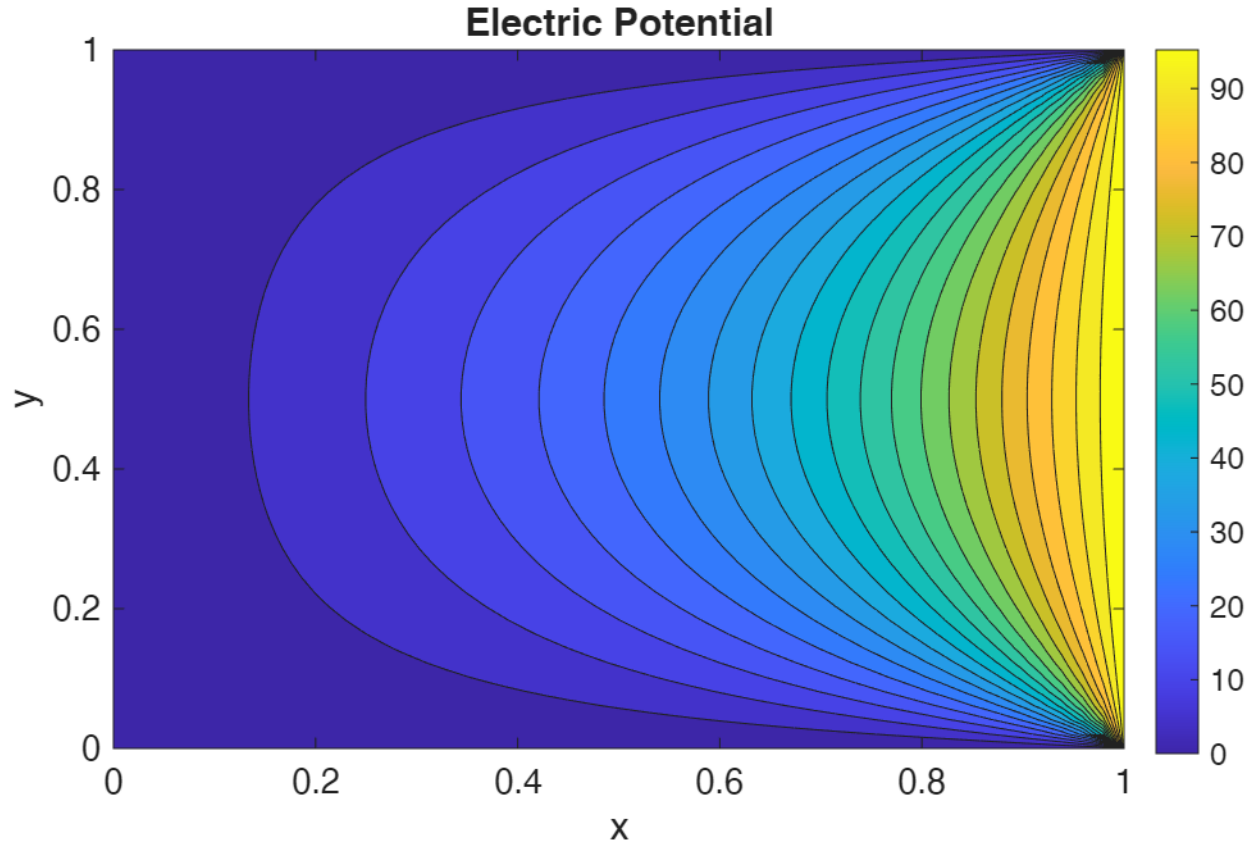
Solve Laplace equation:  $\nabla^2 V = 0$

- Two electrodes: 0 V (left), 100 V (right)
- Discretize Laplace equation:
- $V(i,j) = (V(i+1,j)+V(i-1,j)+V(i,j+1)+V(i,j-1))/4$
- Use Gauss-Seidel iteration, update grid until convergence (error < tolerance)
- Electric field:  $E = -\nabla V$
- $E_x = -(V(i,j+1)-V(i,j-1))/(2dx)$
- $E_y = -(V(i+1,j)-V(i-1,j))/(2dy)$





# 2D





# 1D Poisson Equation: Implicit

- Solve:  $d^2V/dx^2 = -\rho(x)/\epsilon$
- $V(x)$ : electric potential,  $\rho(x)$ : charge density
- Boundary conditions:  $V(0)$ ,  $V(L)$
- Approximate second derivative:
- $(V(i+1) - 2V(i) + V(i-1))) / dx^2 = -\rho(i)/\epsilon$
- Rearranged:
- $-V(i-1) + 2V(i) - V(i+1) = \rho(i) dx^2 / \epsilon$
- One equation per grid point





# Implicit Matrix Formulation

- System written as:  $A V = b$
- $A$ : tridiagonal matrix (couples all points)
- $V$ : unknown potential vector
- $b$ : includes charge density + boundary conditions
- Solve using:  $V = A \setminus b$





# Electric Field & Physical Meaning

- Electric field:  $E = -dV/dx$
- Numerically:  $E(i) = -(V(i+1)-V(i-1))/(2dx)$
- $\rho = 0 \rightarrow$  linear potential, constant field
- $\rho \neq 0 \rightarrow$  curved potential, varying field
- Implicit method solves entire system at once

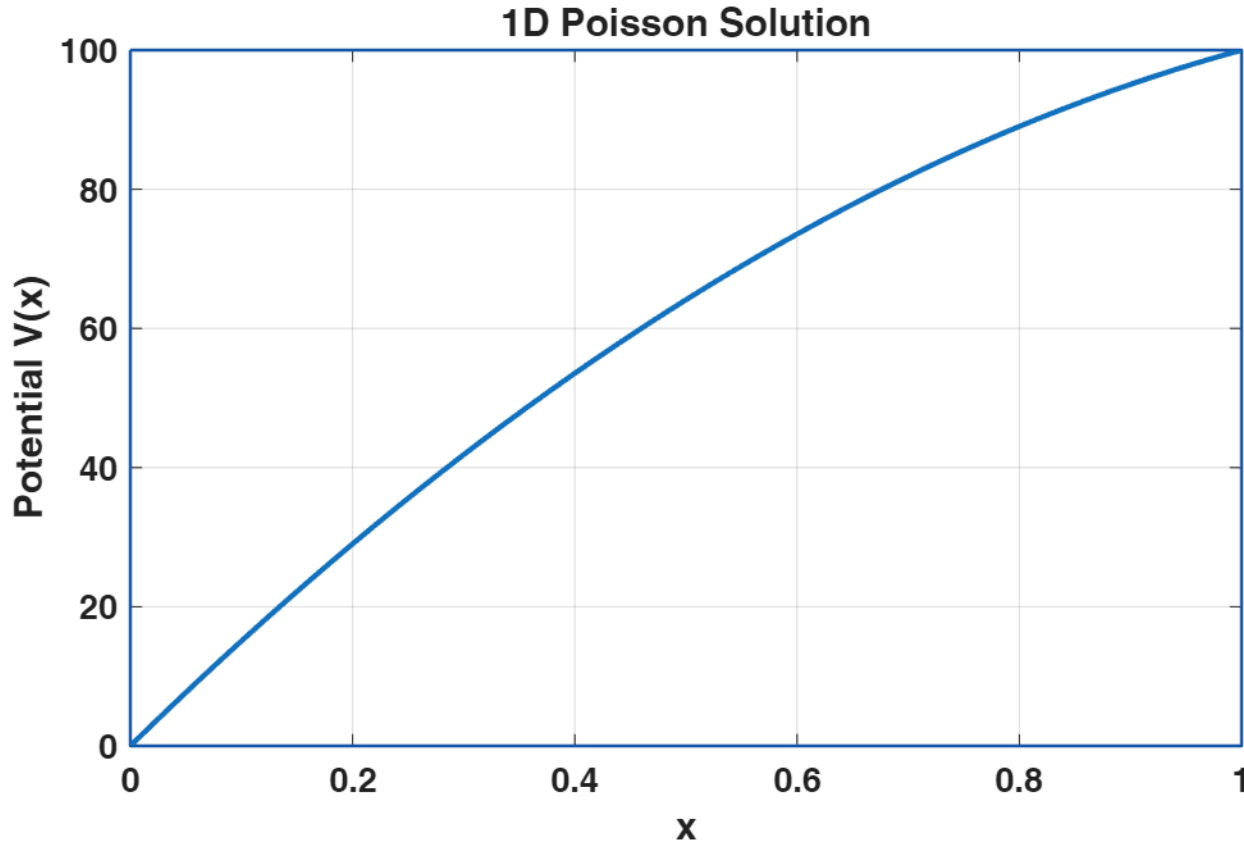


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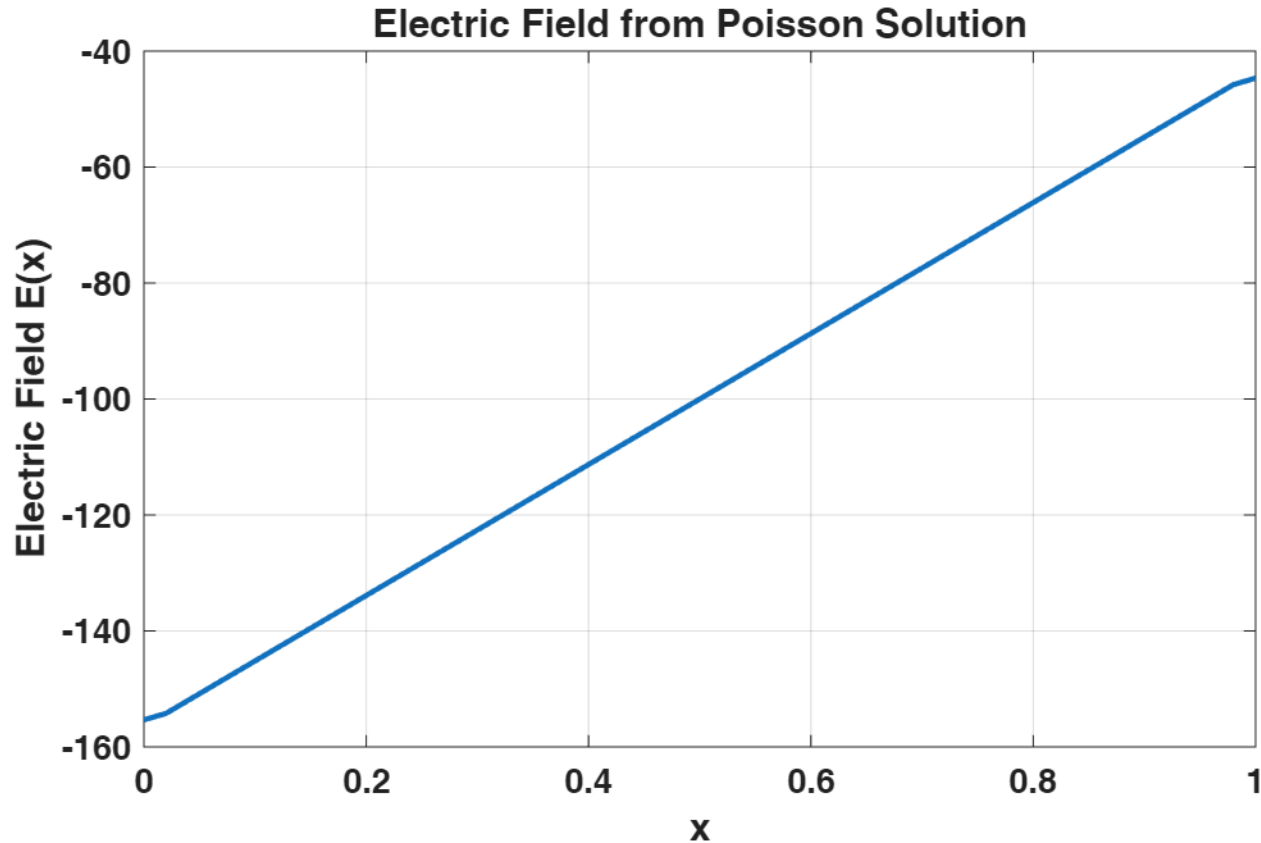


# Electric Field & Physical Meaning



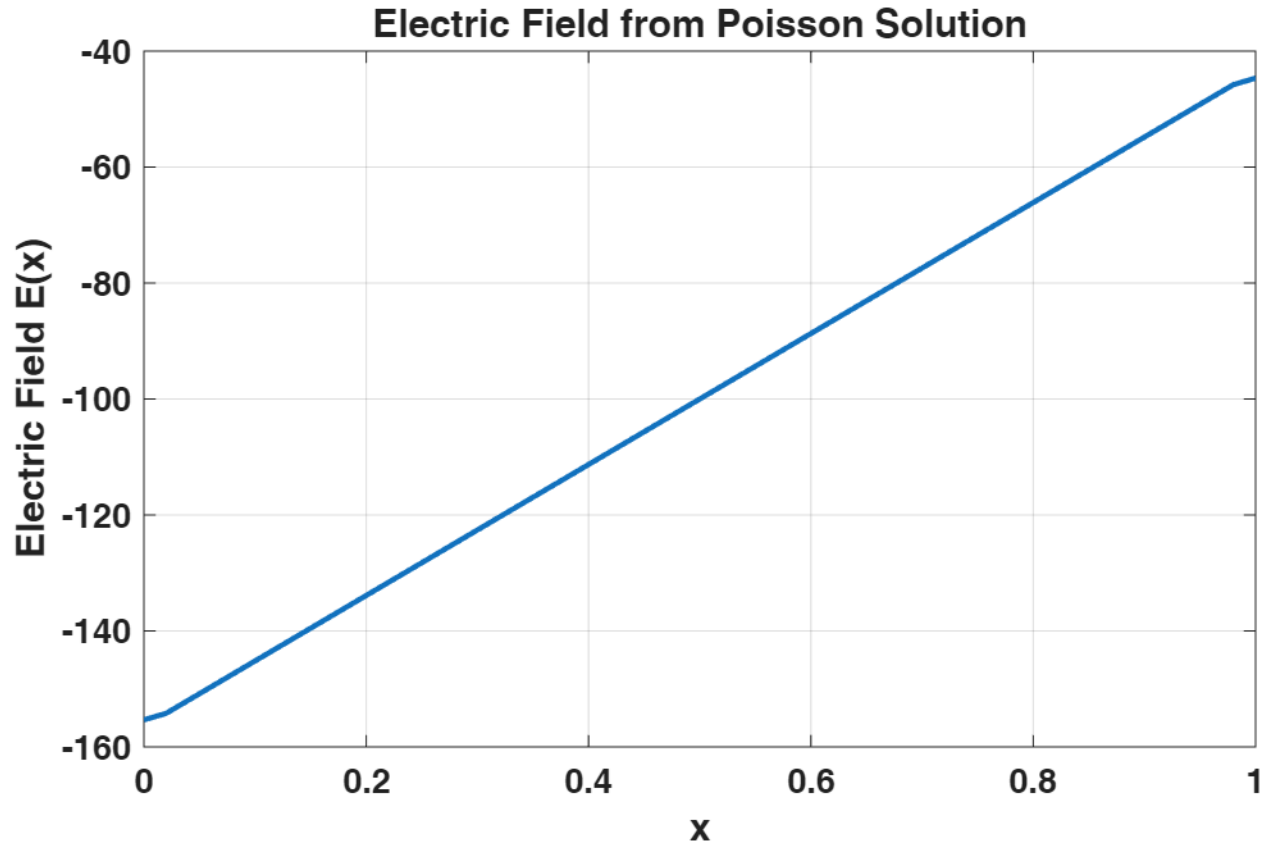


# Electric Field & Physical Meaning



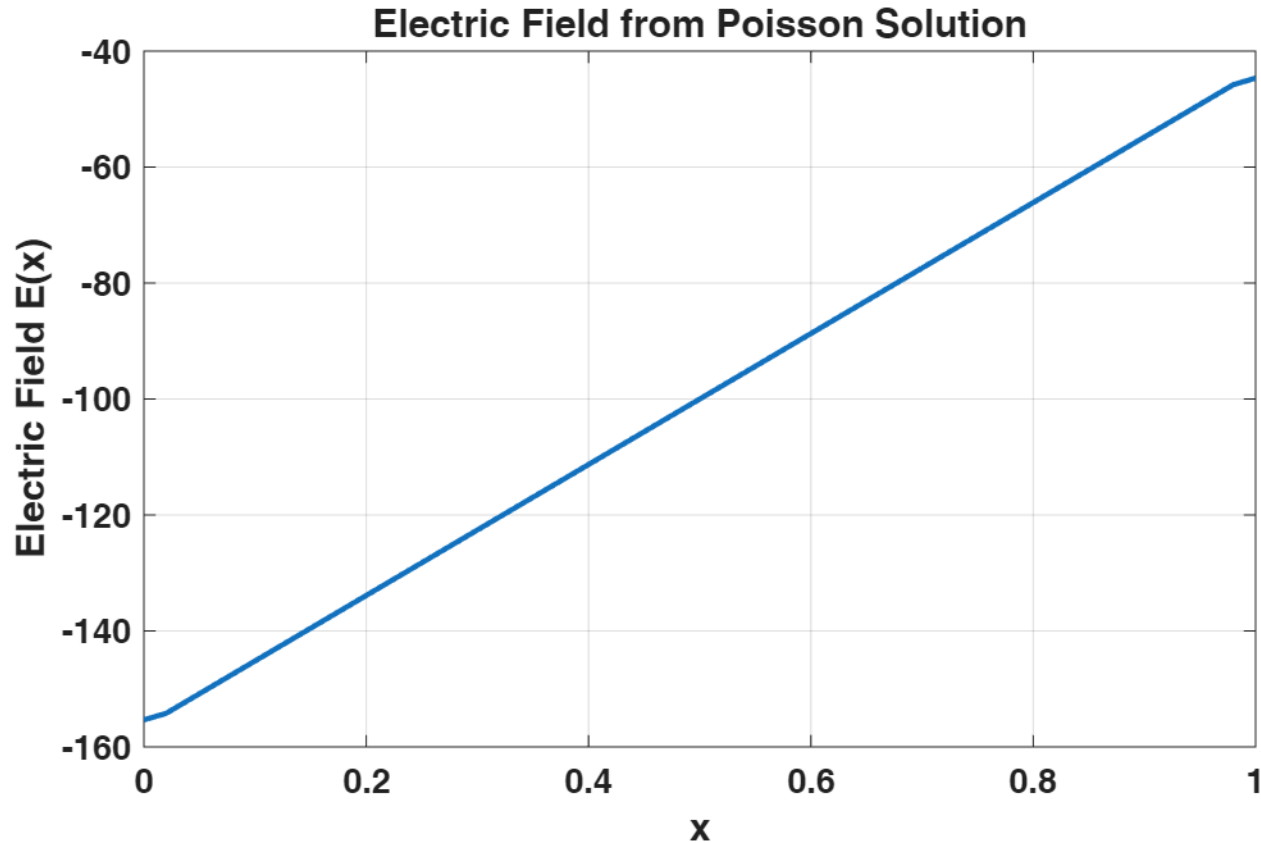


# Electric Field & Physical Meaning





# Electric Field & Physical Meaning



# 1D Poisson Equation (Fourier Method)



- Solve:  $d^2\varphi/dx^2 = -\rho(x)/\epsilon$
- $\varphi$ : potential,  $\rho$ : charge density
- Goal: compute  $\varphi$  and electric field  $E$
- Assume periodic boundary conditions



# Key Idea of Fourier Method

- Transform equation into Fourier space
- Derivative becomes algebraic operator
- $d/dx \rightarrow ik$ ,  $d^2/dx^2 \rightarrow -k^2$
- Convert differential equation  $\rightarrow$  algebraic equation.  $\varphi(x) \leftrightarrow \tilde{\varphi}(k)$ ,  $\rho(x) \leftrightarrow \tilde{\rho}(k)$
- Apply FFT to transform  $\rho(x)$
- Poisson equation becomes:
- $-k^2 \tilde{\varphi}(k) = -\tilde{\rho}(k)/\epsilon$





# Algorithm Steps

- 1. Define  $\rho(x)$  on grid
- 2. Compute FFT  $\rightarrow \hat{\rho}(k)$
- 3. Solve  $\hat{\varphi}(k) = \hat{\rho}(k)/(\epsilon k^2)$
- 4. Inverse FFT  $\rightarrow \varphi(x)$
- 5. Compute  $E(x)$  from  $\varphi$



# Advantages and Limitations

- Advantages: Highly accurate for smooth periodic functions, Fast:  $O(N \log N)$  using FFT, No iteration required, Handles derivatives naturally
- Limitations: Requires periodic boundary conditions,  $k = 0$  mode requires special treatment, Not suitable for fixed boundary potentials directly



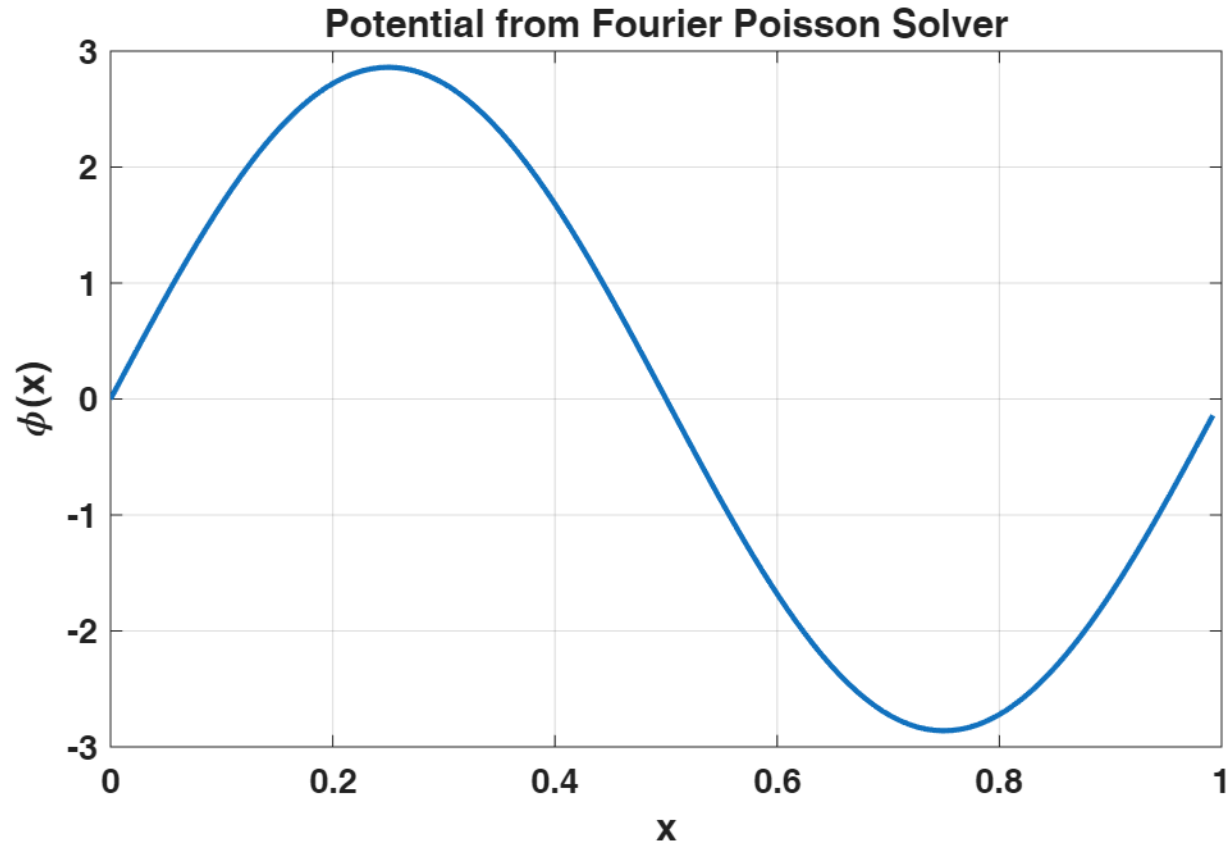


# Physical Interpretation

- Each Fourier mode represents a wave
- Poisson equation solved mode-by-mode
- Long wavelength modes  $\rightarrow$  larger potential
- Electric field derived from spatial variation

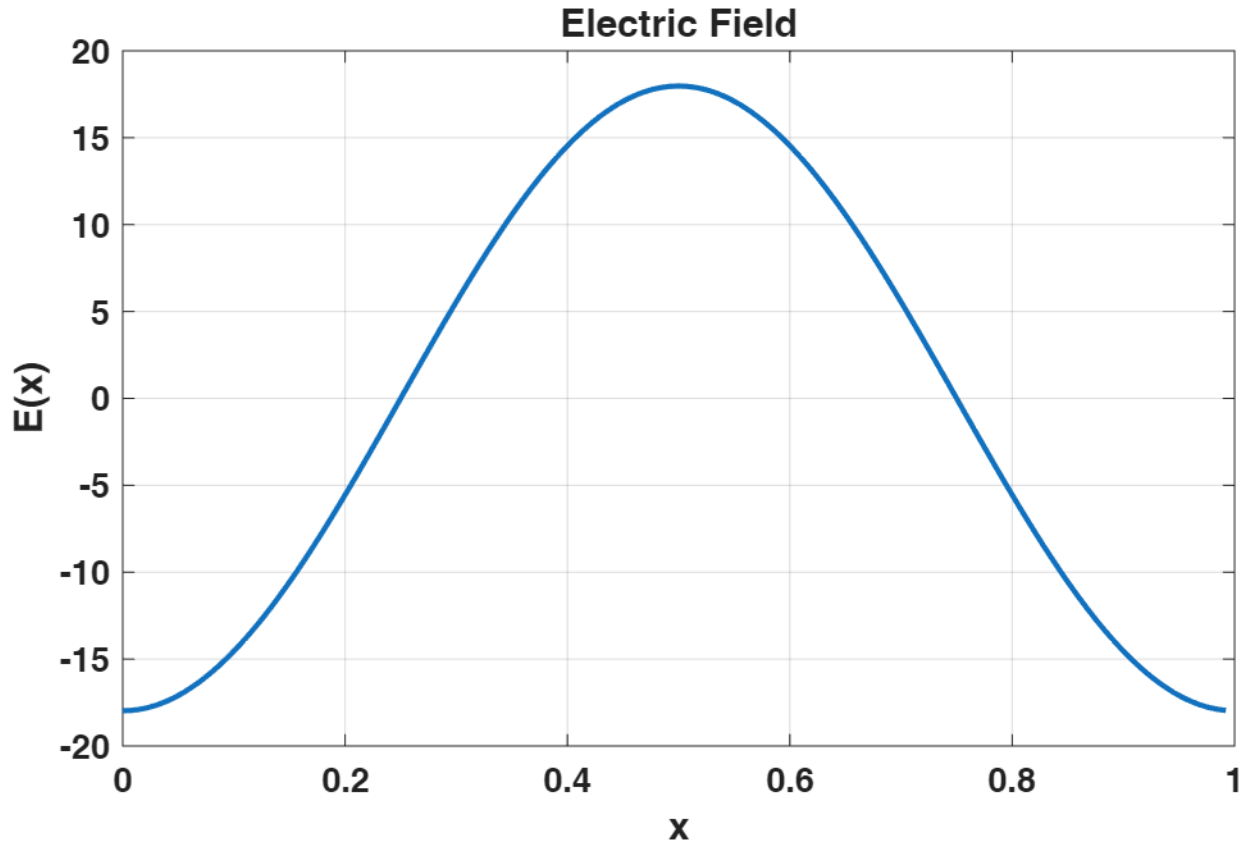


# Potential





# Field





**Thank You!**

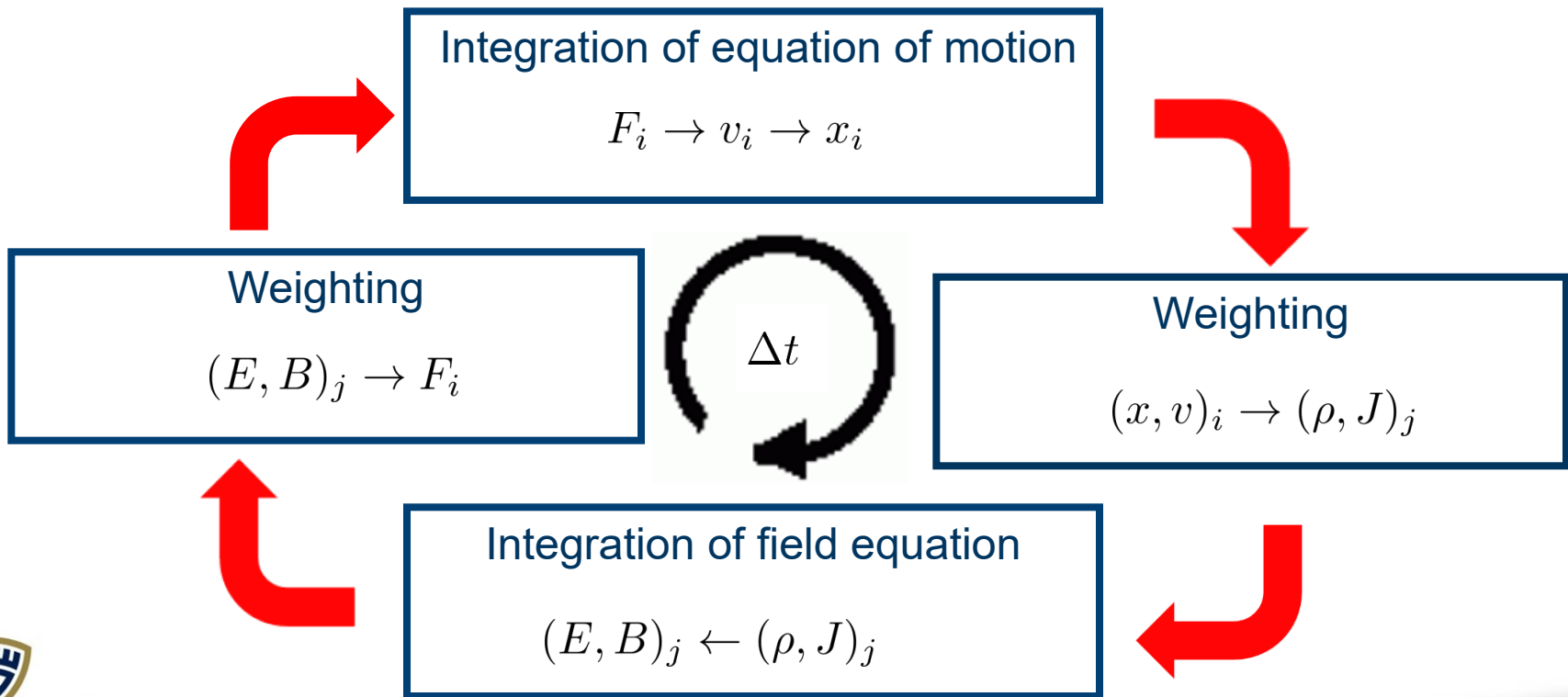
**We are ready to start PIC  
simulation**





# Particle in Cell Simulation

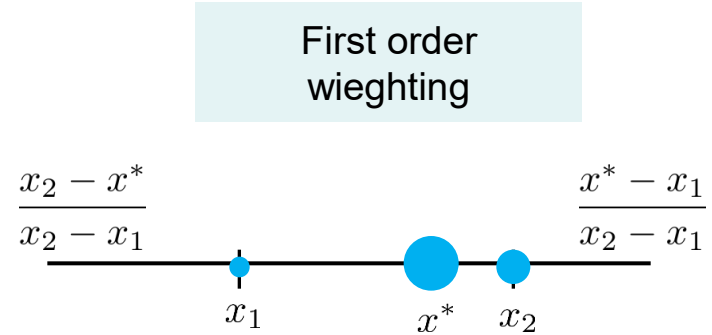
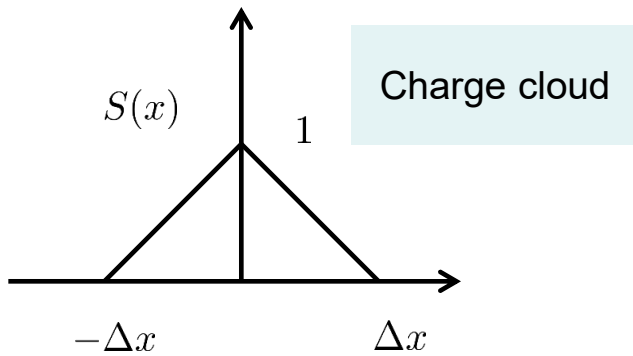
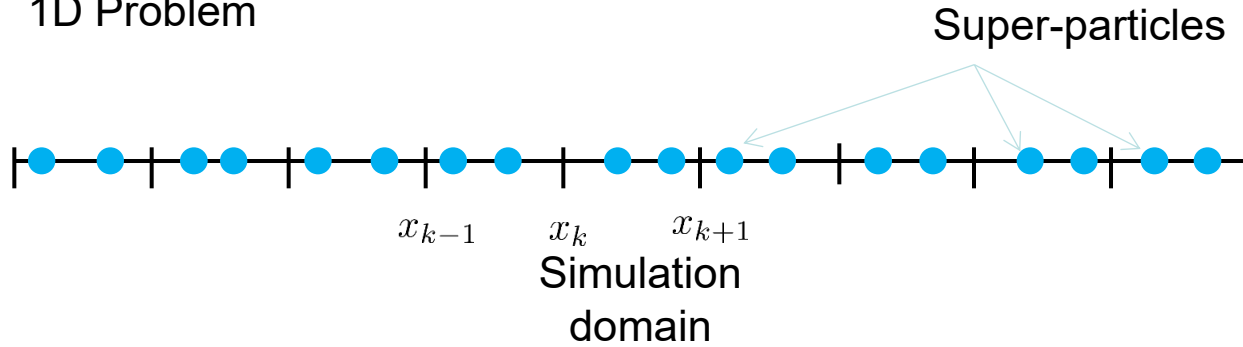
- Kinetic means „of or relating to motion“.
  - It is impractical to solve the equation of motion of all plasma particles.
  - Boltzman equation is an integro-differential equation.
- Particle-in-Cell : Super particle  $10^6 - 10^9$  real particles.





# Particles

1D Problem





# Fields

Poisson's eq.

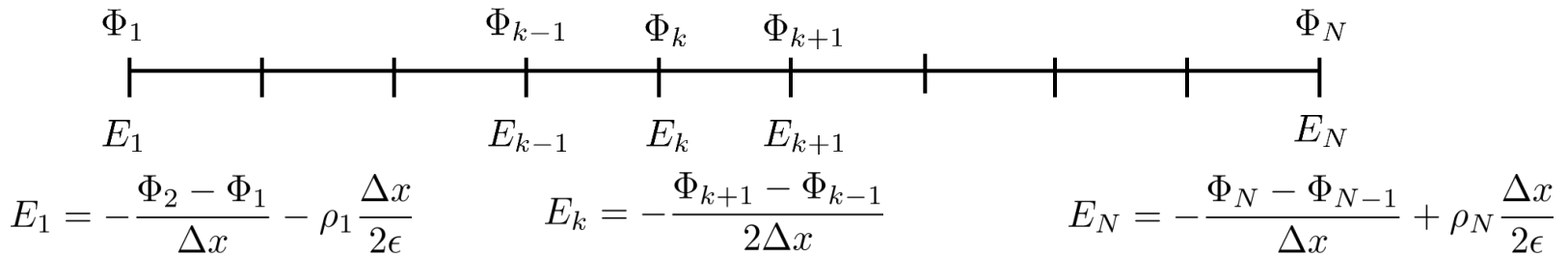
$$\nabla^2 \Phi = -\frac{\rho}{\epsilon}$$



$$\frac{\Phi_{k+1} - 2\Phi_k + \Phi_{k-1}}{\Delta x^2} = -\frac{\rho}{\epsilon}$$

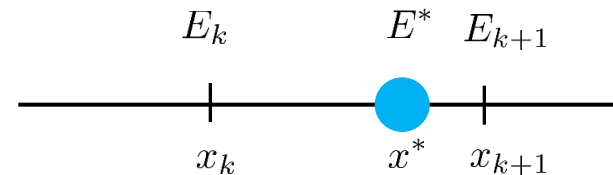
Boundary condition

Boundary condition



Interpolation of the fields to the particle positions

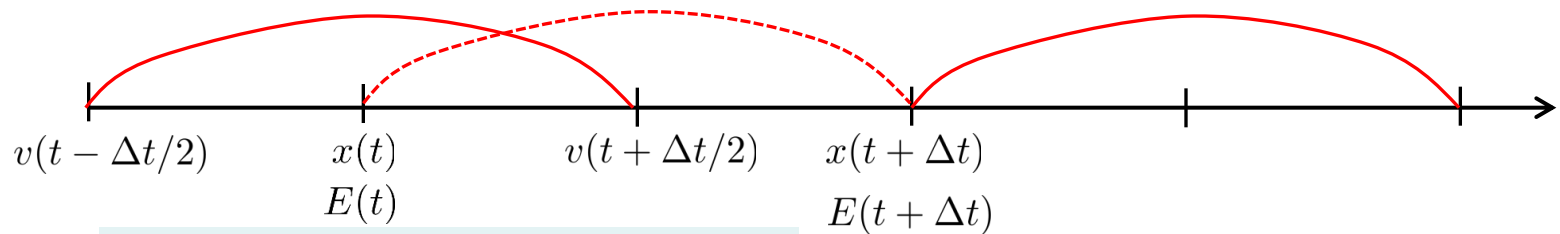
$$E^* = \frac{x^* - x_k}{\Delta x} E_{k+1} + \frac{x_{k+1} - x^*}{\Delta x} E_k$$





# Pushing particles

„Leapfrog“  
scheme



Descritization of equation of motions

$$\frac{v(t + \Delta t/2) - v(t - \Delta t/2)}{\Delta t} = \frac{q}{m} E(t)$$

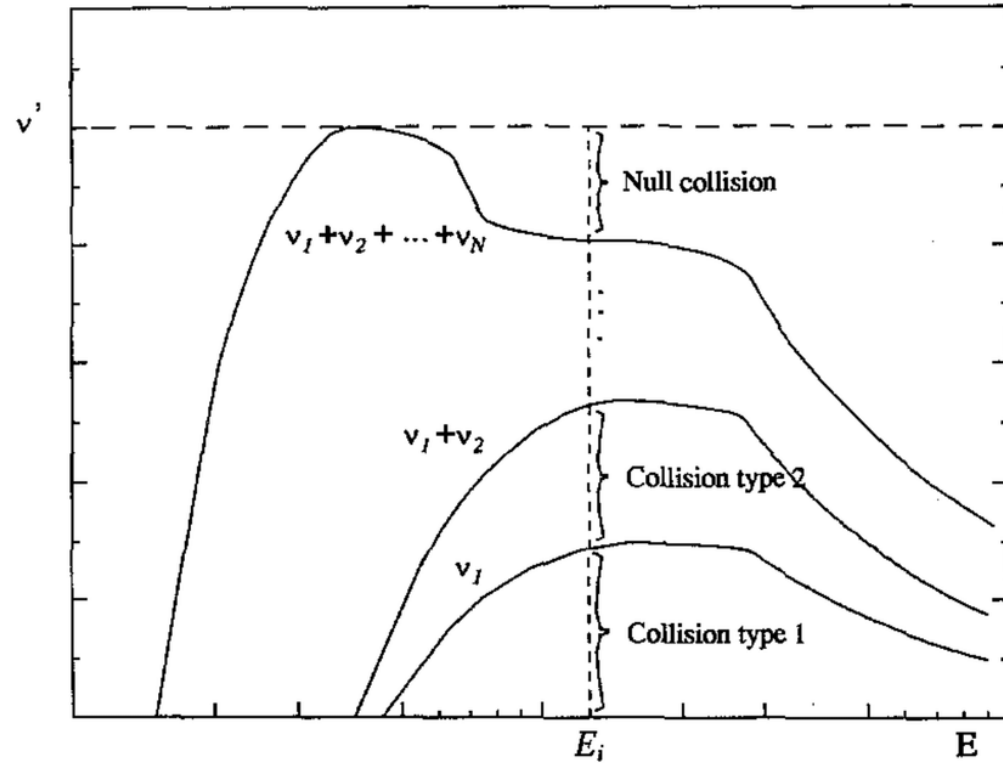
$$\frac{x(t + \Delta t) - x(t)}{\Delta t} = v(t + \Delta t/2)$$

Monte-Carlo Scheme is required for collisions



# Monte Carlo: null collision method

- Many collisions take place:  
impact ionization, charge exchange,  
hard-sphere, ...
- Let the probability of them  
 $P_1, P_2, P_3, P_4, \dots$
- Calculate the total probabilities  $P_T$
- Calculate relative probabilities  
 $P_1/P_T, P_2/P_T, P_3/P_T, P_4/P_T, \dots$



- Generate a random number between  $[0,1]$
- if  $P_1/P_T = < \text{The random number} < (P_1 + P_2)/P_T$
- Event 1 takes place
- If not

$(P_1 + P_2)/P_T = < \text{The random number} < (P_1 + P_2 + P_3)/P_T$





# Challenges of PIC simulation

- Numerical instabilities:
  - Accuracy criterion  $\omega_p \Delta t \leq 0.2$
  - Courant criterion  $v_{\max} \Delta t \leq \Delta x$
  - The computational grid has to resolve the Debye length  $\Delta x \leq \lambda_D$
- In order to have a good statistics, a reasonable high number of particles per Debye length must be used  $N_D \gg 1$
- Keep the probability for collisions small
$$P_{\text{coll}} = 1 - e^{-\nu t} \leq 0.1$$
- Alternatives:
  - Implicit schemes
  - Parallelization