

# CE 530 Molecular Simulation

## Lecture 11

### Molecular Dynamics Simulation

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# Review and Preview

## ○ MD of hard disks

- *intuitive*
- *collision detection and impulsive dynamics*

## ○ Monte Carlo

- *convenient sampling of ensembles*
- *no dynamics*
- *biasing possible to improve performance*

## ○ Molecular dynamics

- *equations of motion*
- *integration schemes*
- *evaluation of dynamical properties*
- *extensions to other ensembles*
- *focus on atomic systems for now*

# Classical Equations of Motion

- Several formulations are in use
  - *Newtonian*
  - *Lagrangian*
  - *Hamiltonian*
- Advantages of non-Newtonian formulations
  - *more general, no need for “fictitious” forces*
  - *better suited for multiparticle systems*
  - *better handling of constraints*
  - *can be formulated from more basic postulates*
- Assume conservative forces

$$\vec{\mathbf{F}} = -\vec{\nabla}U \quad \text{Gradient of a scalar potential energy}$$

# Newtonian Formulation

- Cartesian spatial coordinates  $\mathbf{r}_i = (x_i, y_i, z_i)$  are primary variables
  - for  $N$  atoms, system of  $N$  2nd-order differential equations

$$m \frac{d^2 \mathbf{r}_i}{dt^2} \equiv m \ddot{\mathbf{r}}_i = \mathbf{F}_i$$

- Sample application: 2D motion in central force field

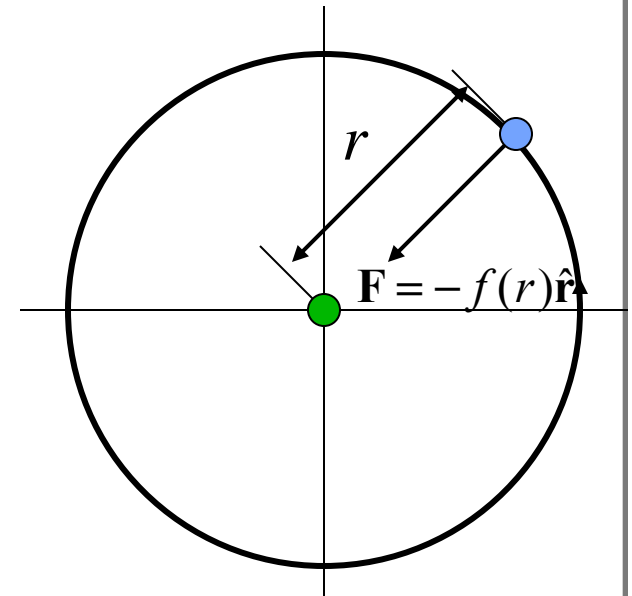
$$m\ddot{x} = \mathbf{F} \cdot \hat{\mathbf{e}}_x = -f(r) \hat{\mathbf{r}} \cdot \hat{\mathbf{e}}_x = -xf \left( \sqrt{x^2 + y^2} \right)$$

$$m\ddot{y} = \mathbf{F} \cdot \hat{\mathbf{e}}_y = -f(r) \hat{\mathbf{r}} \cdot \hat{\mathbf{e}}_y = -yf \left( \sqrt{x^2 + y^2} \right)$$

- Polar coordinates are more natural and convenient

$$mr^2 \dot{\theta} = \ell \quad \text{constant angular momentum}$$

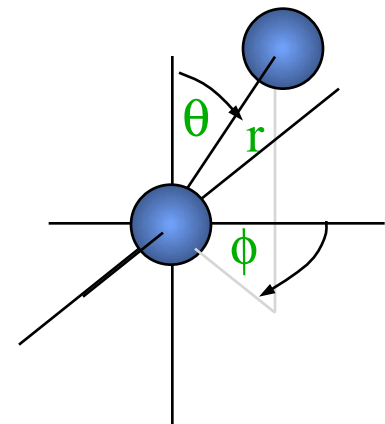
$$m\ddot{r} = -f(r) + \frac{\ell^2}{mr^3} \quad \text{fictitious (centrifugal) force}$$



# Generalized Coordinates

## ○ Any convenient coordinates for description of particular system

- use  $q_i$  as symbol for general coordinate
- examples
  - diatomic  $\{q_1, \dots, q_6\} = \{x_{\text{com}}, y_{\text{com}}, z_{\text{com}}, r_{12}, \theta, \phi\}$
  - 2-D motion in central field  $\{q_1, q_2\} = \{r, \theta\}$



## ○ Kinetic energy

- general quadratic form

$$K = \underbrace{M_0(\mathbf{q}) + \sum M_j(\mathbf{q})\dot{q}_j}_{\text{usually vanish}} + \frac{1}{2} \sum \sum M_{jk}(\mathbf{q})\dot{q}_j\dot{q}_k$$

- examples

→ rotating diatomic  $K = \frac{1}{2} m (\dot{q}_1^2 + \dot{q}_2^2 + \dot{q}_3^2) + \frac{1}{8} m \left[ \dot{r}^2 + r^2 \dot{\theta}^2 + (r \sin \theta)^2 \dot{\phi}^2 \right]$

→ 2-D central motion  $K = \frac{1}{2} m (\dot{r}^2 + r^2 \dot{\theta}^2)$

# Lagrangian Formulation

- Independent of coordinate system
- Define the Lagrangian
  - $L(\mathbf{q}, \dot{\mathbf{q}}) \equiv K(\mathbf{q}, \dot{\mathbf{q}}) - U(\mathbf{q})$
- Equations of motion

$$\frac{d}{dt} \left( \frac{\partial L}{\partial \dot{q}_j} \right) - \frac{\partial L}{\partial q_j} = 0 \quad j = 1 \dots N$$

- $N$  second-order differential equations

- Central-force example

$$L = \frac{1}{2} m (\dot{r}^2 + r^2 \dot{\theta}^2) - U(r)$$

$$\frac{d}{dt} \left( \frac{\partial L}{\partial \dot{r}} \right) = \frac{\partial L}{\partial r} \Rightarrow \boxed{m\ddot{r} = mr\dot{\theta}^2 - f(r)} \quad \frac{d}{dt} \left( \frac{\partial L}{\partial \dot{\theta}} \right) = \frac{\partial L}{\partial \theta} \Rightarrow \boxed{\frac{d}{dt} (mr^2\dot{\theta}) = 0}$$

$$\vec{\mathbf{F}}_r = -\vec{\nabla}_r U = -f(r)$$

# Hamiltonian Formulation 1. Motivation

- Appropriate for application to statistical mechanics and quantum mechanics
- Newtonian and Lagrangian viewpoints take the  $q_i$  as the fundamental variables
  - *$N$ -variable configuration space*
  - *$\dot{q}_i$  appears only as a convenient shorthand for  $dq/dt$*
  - *working formulas are 2nd-order differential equations*
- Hamiltonian formulation seeks to work with 1st-order differential equations
  - *$2N$  variables*
  - *treat the coordinate and its time derivative as independent variables*
  - *appropriate quantum-mechanically*

# Hamiltonian Formulation 2. Preparation

## ○ Mathematically, Lagrangian treats $q$ and $\dot{q}$ as distinct

- $L(q_j, \dot{q}_j, t)$

- *identify the generalized momentum as*

$$p_j = \frac{\partial L}{\partial \dot{q}_j}$$

- *e.g. if  $L = K - U = \frac{1}{2}m\dot{q}^2 - U(q)$ ;  $p = \partial L / \partial \dot{q} = m\dot{q}$*

- *Lagrangian equations of motion  $\frac{dp_j}{dt} = \frac{\partial L}{\partial q_j}$*

## ○ We would like a formulation in which $p$ is an independent variable

- *$p_i$  is the derivative of the Lagrangian with respect to  $\dot{q}_i$ , and we're looking to replace  $\dot{q}_i$  with  $p_i$*
- *we need ...?*



## Hamiltonian Formulation 3. Definition

○ ...a Legendre transform!

○ Define the *Hamiltonian*,  $H$

$$\begin{aligned}
 H(\mathbf{q}, \mathbf{p}) &= -\left[ L(\mathbf{q}, \dot{\mathbf{q}}) - \sum p_j \dot{q}_j \right] \\
 &= -K(\mathbf{q}, \dot{\mathbf{q}}) + U(\mathbf{q}) + \sum \frac{\partial K}{\partial \dot{q}_j} \dot{q}_j \\
 &= -\sum a_j \dot{q}_j^2 + U(\mathbf{q}) + \sum (2a_j \dot{q}_j) \dot{q}_j \\
 &= +\sum a_j \dot{q}_j^2 + U(\mathbf{q}) \\
 &= K + U
 \end{aligned}$$

○  $H$  equals the total energy (kinetic plus potential)

# Hamiltonian Formulation 4. Dynamics

## ○ Hamilton's equations of motion

- *From Lagrangian equations, written in terms of momentum*

Differential change in L

$$\begin{aligned} dL &= \frac{\partial L}{\partial q} dq + \frac{\partial L}{\partial \dot{q}} d\dot{q} \\ &= \dot{p}dq + pd\dot{q} \end{aligned}$$

Legendre transform

$$\begin{aligned} H &= -(L - p\dot{q}) \\ dH &= -(pdq - \dot{q}dp) \\ dH &= -\dot{p}dq + \dot{q}dp \end{aligned}$$

$$\left. \begin{aligned} \dot{q} &= +\frac{\partial H}{\partial p} \\ \dot{p} &= -\frac{\partial H}{\partial q} \end{aligned} \right\}$$

Hamilton's equations of motion

$$\frac{dp}{dt} = \dot{p} = \frac{\partial L}{\partial q}$$

Lagrange's equation  
of motion

$$p = \frac{\partial L}{\partial \dot{q}}$$

Definition of momentum

Conservation of energy

$$\frac{dH}{dt} = -\dot{p} \frac{dq}{dt} + \dot{q} \frac{dp}{dt} = -\dot{p}\dot{q} + \dot{q}\dot{p} = 0$$

# Hamiltonian Formulation 5. Example

## ○ Particle motion in central force field

$$H = K + U$$

$$= \frac{p_r^2}{2m} + \frac{p_\theta^2}{2mr^2} + U(r)$$

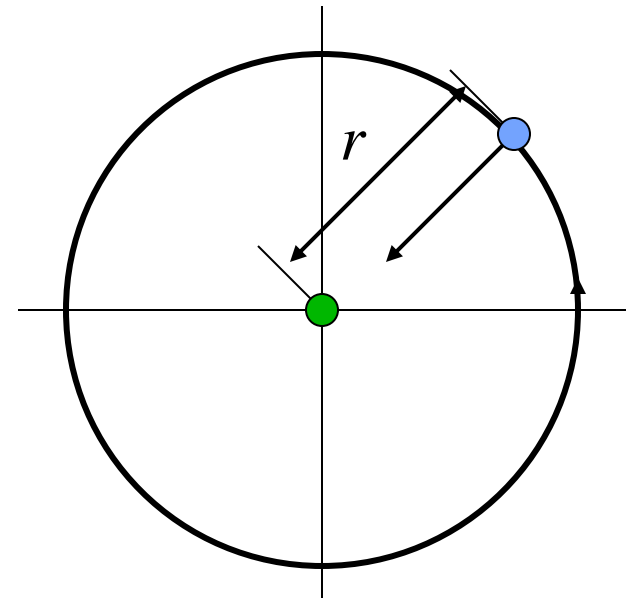
$$\dot{q} = + \frac{\partial H}{\partial p}$$

$$\dot{p} = - \frac{\partial H}{\partial q}$$

$$(1) \frac{dr}{dt} = \frac{p_r}{m} \quad (2) \frac{d\theta}{dt} = \frac{p_\theta}{mr^2}$$

$$(3) \frac{dp_r}{dt} = \frac{p_\theta^2}{mr^3} - f(r) \quad (4) \frac{dp_\theta}{dt} = 0$$

$$\vec{F}_r = -\vec{\nabla}_r U = -f(r)$$



Lagrange's equations

$$m\ddot{r} = mr\dot{\theta}^2 - f(r)$$

$$\frac{d}{dt}(mr^2\dot{\theta}) = 0$$

## ○ Equations no simpler, but theoretical basis is better

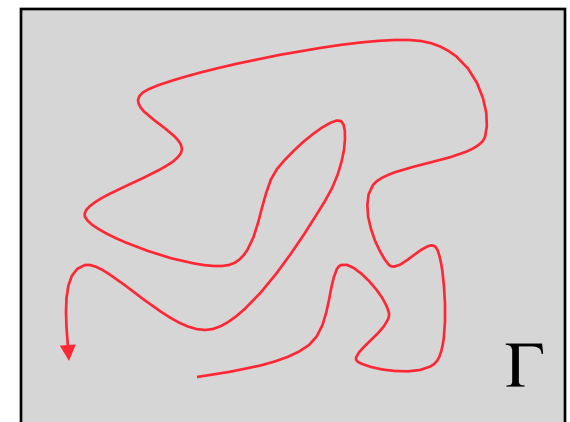
# Phase Space (again)

## ○ Return to the complete picture of phase space

- *full specification of microstate of the system is given by the values of all positions and all momenta of all atoms*
  - $\Gamma = (\mathbf{p}^N, \mathbf{r}^N)$
- *view positions and momenta as completely independent coordinates*
  - connection between them comes only through equation of motion

## ○ Motion through phase space

- *helpful to think of dynamics as “simple” movement through the high-dimensional phase space*
  - facilitate connection to quantum mechanics
  - basis for theoretical treatments of dynamics
  - understanding of integrators



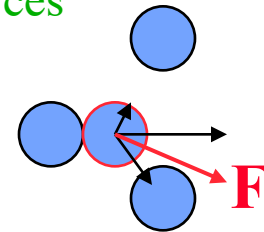
# Integration Algorithms

## ○ Equations of motion in cartesian coordinates

$$\begin{aligned} \frac{d\mathbf{r}_j}{dt} &= \frac{\mathbf{p}_j}{m} \\ \frac{d\mathbf{p}_j}{dt} &= \mathbf{F}_j \end{aligned}$$

$$\left. \begin{aligned} \mathbf{r} &= (r_x, r_y) \\ \mathbf{p} &= (p_x, p_y) \end{aligned} \right\} \text{2-dimensional space (for example)}$$

$$\mathbf{F}_j = \sum_{\substack{i=1 \\ i \neq j}}^N \mathbf{F}_{ij} \quad \text{pairwise additive forces}$$



## ○ Desirable features of an integrator

- *minimal need to compute forces (a very expensive calculation)*
- *good stability for large time steps*
- *good accuracy*
- *conserves energy and momentum*
- *time-reversible*
- *area-preserving (symplectic)*

More on these later

# Verlet Algorithm

## 1. Equations

- Very simple, very good, very popular algorithm
- Consider expansion of coordinate forward and backward in time

$$\mathbf{r}(t + \delta t) = \mathbf{r}(t) + \frac{1}{m} \mathbf{p}(t) \delta t + \frac{1}{2m} \mathbf{F}(t) \delta t^2 + \frac{1}{3!} \ddot{\mathbf{r}}(t) \delta t^3 + O(\delta t^4)$$

$$\mathbf{r}(t - \delta t) = \mathbf{r}(t) - \frac{1}{m} \mathbf{p}(t) \delta t + \frac{1}{2m} \mathbf{F}(t) \delta t^2 - \frac{1}{3!} \ddot{\mathbf{r}}(t) \delta t^3 + O(\delta t^4)$$

- Add these together

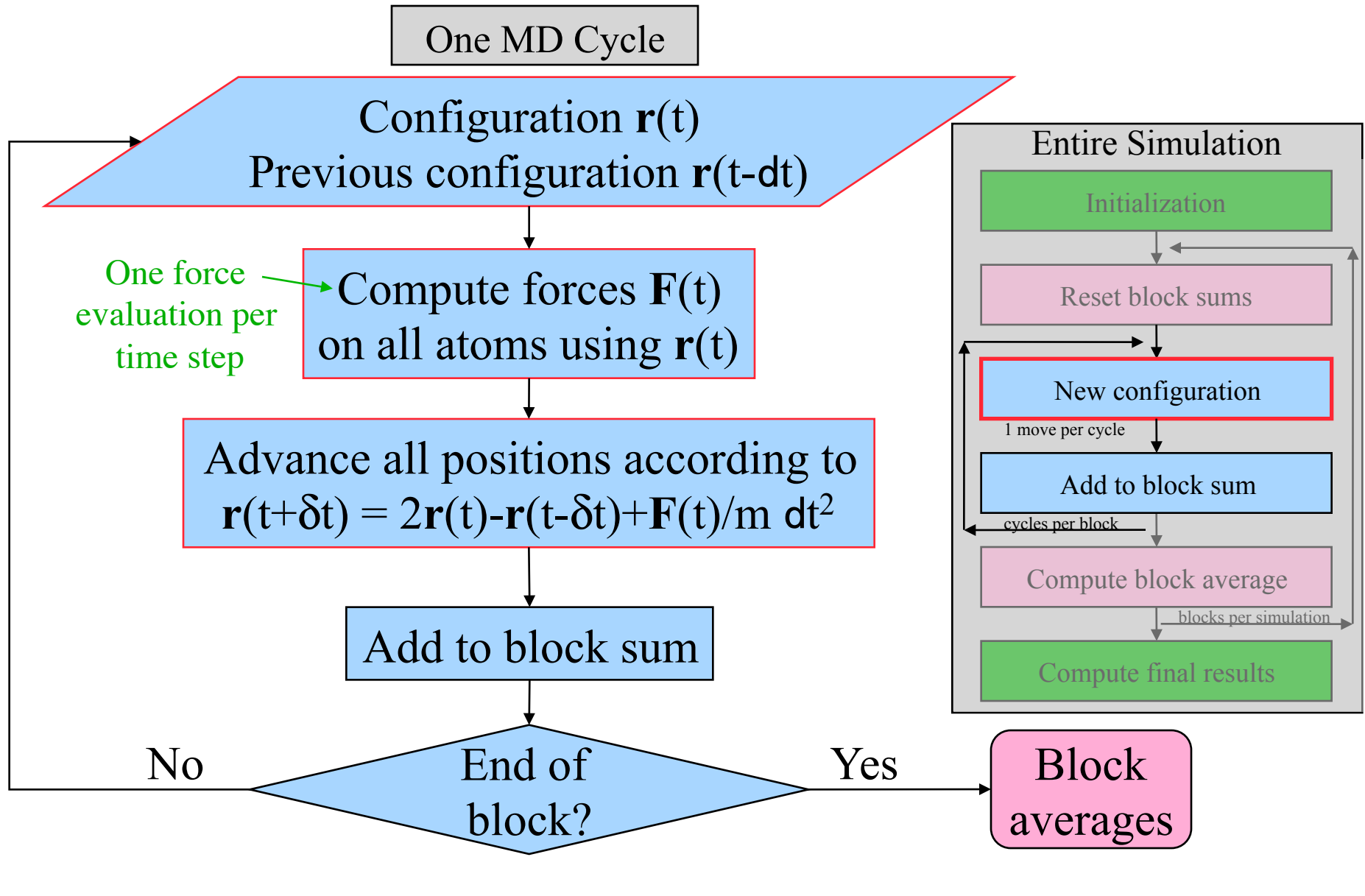
$$\mathbf{r}(t + \delta t) + \mathbf{r}(t - \delta t) = 2\mathbf{r}(t) + \frac{1}{m} \mathbf{F}(t) \delta t^2 + O(\delta t^4)$$

- Rearrange

$$\mathbf{r}(t + \delta t) = 2\mathbf{r}(t) - \mathbf{r}(t - \delta t) + \frac{1}{m} \mathbf{F}(t) \delta t^2 + O(\delta t^4)$$

- *update without ever consulting velocities!*

# Verlet Algorithm 2. Flow diagram



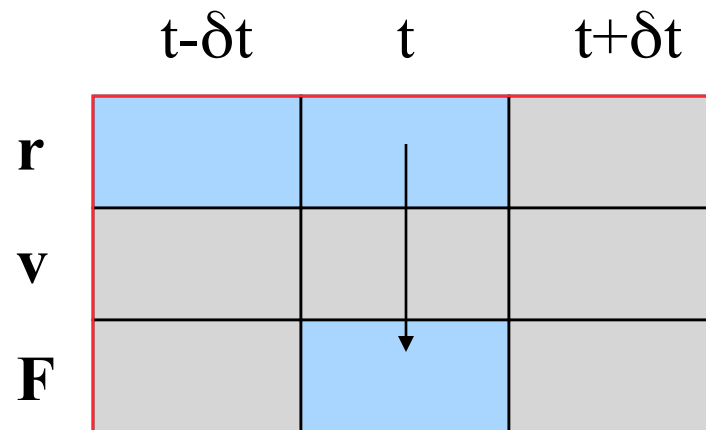
## Verlet Algorithm 2. Flow Diagram

	$t-\delta t$	$t$	$t+\delta t$
<b>r</b>			
<b>v</b>			
<b>F</b>			

Given current position and position at end of previous time step

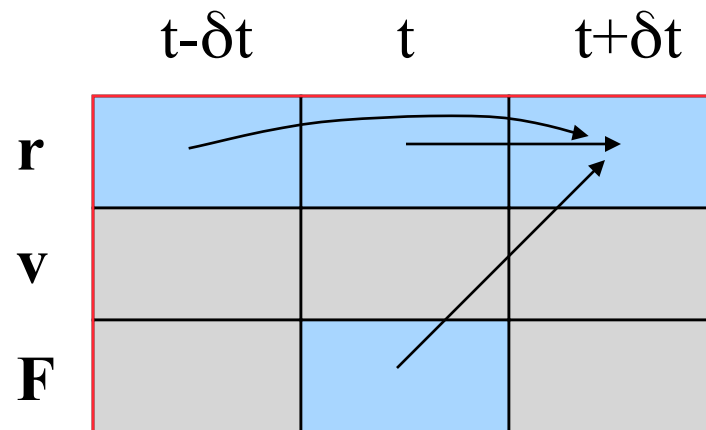


## Verlet Algorithm 2. Flow Diagram



Compute the force at the current position

## Verlet Algorithm 2. Flow Diagram



Compute new position from present and previous positions, and present force

## Verlet Algorithm 2. Flow Diagram

	$t-2\delta t$	$t-\delta t$	$t$	$t+\delta t$
<b>r</b>				
<b>v</b>				
<b>F</b>				

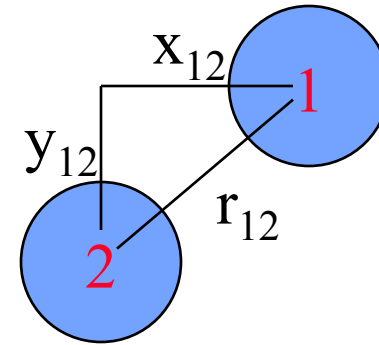
Advance to next time step,  
repeat

# Forces 1. Formalism

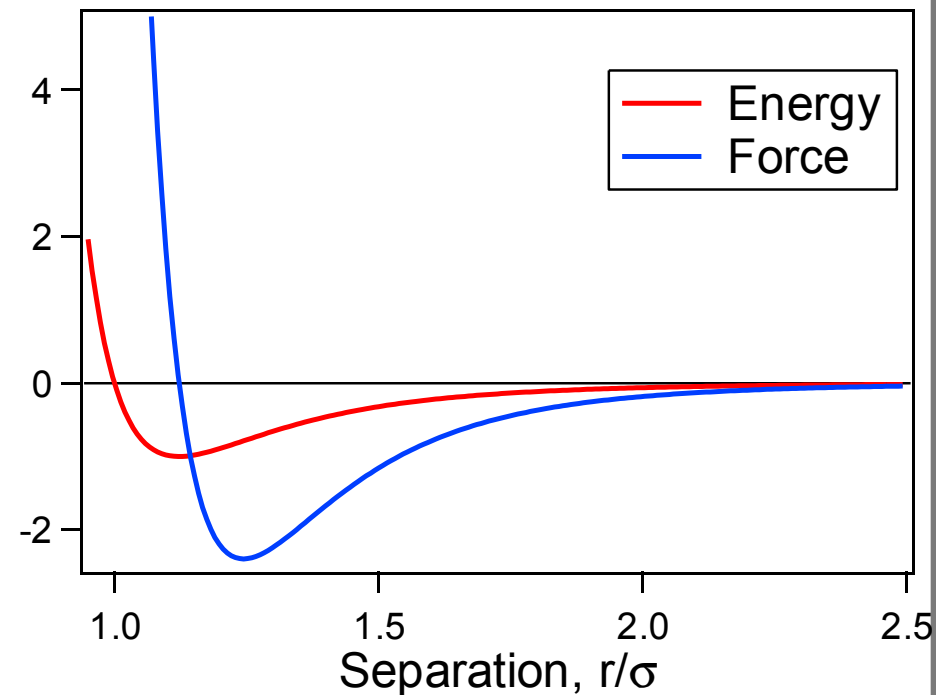
○ Force is the gradient of the potential

$$\begin{aligned}
 \mathbf{F}_{2 \rightarrow 1} &= -\nabla u(r_{12}) \\
 \text{Force on 1,} &= -\frac{\partial u(r_{12})}{\partial x_1} \mathbf{e}_x - \frac{\partial u(r_{12})}{\partial y_1} \mathbf{e}_y \\
 \text{due to 2} & \\
 &= -\frac{du(r_{12})}{dr_{12}} \left[ \frac{\partial r_{12}}{\partial x_1} \mathbf{e}_x + \frac{\partial r_{12}}{\partial y_1} \mathbf{e}_y \right] \\
 &= -\frac{f(r_{12})}{r_{12}} \left[ x_{12} \mathbf{e}_x + y_{12} \mathbf{e}_y \right]
 \end{aligned}$$

$$\mathbf{F}_{2 \rightarrow 1} = -\mathbf{F}_{1 \rightarrow 2}$$



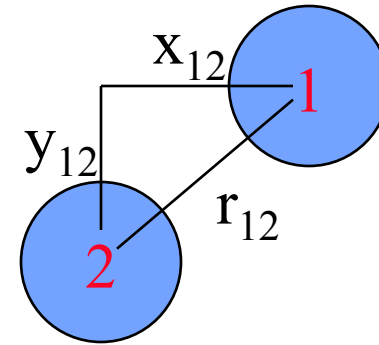
$$r_{12} = \left[ (x_2 - x_1)^2 + (y_2 - y_1)^2 \right]^{1/2}$$



## Forces 2. LJ Model

○ Force is the gradient of the potential

$$\mathbf{F}_{2 \rightarrow 1} = -\frac{f(r_{12})}{r_{12}} [x_{12}\mathbf{e}_x + y_{12}\mathbf{e}_y]$$



$$r_{12} = \left[ (x_2 - x_1)^2 + (y_2 - y_1)^2 \right]^{1/2}$$

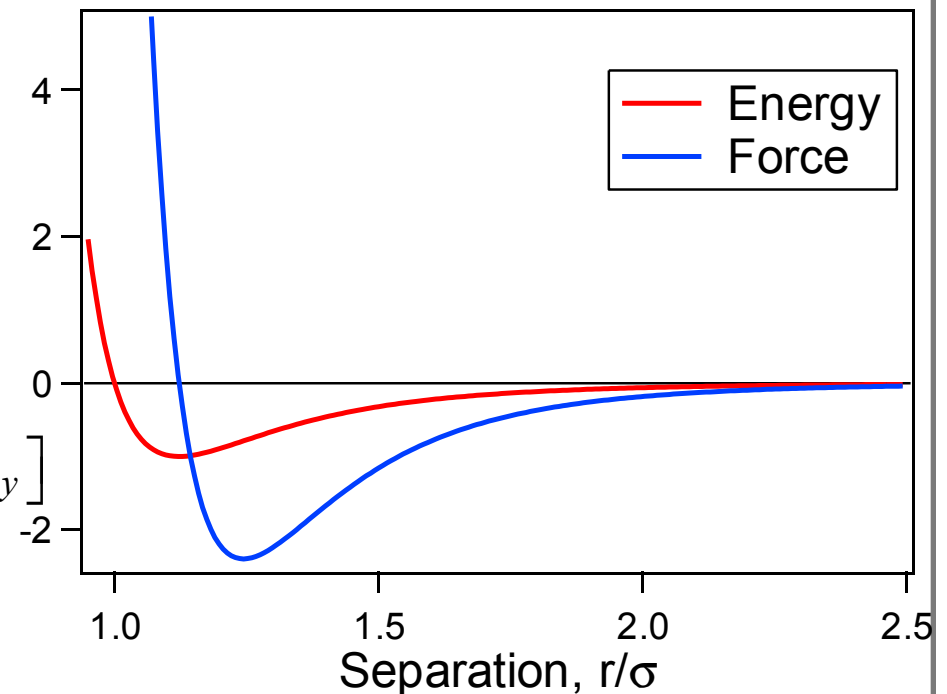
e.g., Lennard-Jones model

$$u(r) = 4\epsilon \left[ \left( \frac{\sigma}{r} \right)^{12} - \left( \frac{\sigma}{r} \right)^6 \right]$$

$$f(r) = -\frac{du}{dr}$$

$$= +\frac{48\epsilon}{\sigma} \left[ \left( \frac{\sigma}{r} \right)^{13} - \frac{1}{2} \left( \frac{\sigma}{r} \right)^7 \right]$$

$$\mathbf{F}_{2 \rightarrow 1} = -\frac{48\epsilon}{\sigma^2} \left[ \left( \frac{\sigma}{r_{12}} \right)^{14} - \frac{1}{2} \left( \frac{\sigma}{r_{12}} \right)^8 \right] [x_{12}\mathbf{e}_x + y_{12}\mathbf{e}_y]$$



# Verlet Algorithm. 4. Loose Ends

## ○ Initialization

- *how to get position at “previous time step” when starting out?*
- *simple approximation*

$$\mathbf{r}(t_0 - \delta t) = \mathbf{r}(t_0) - \mathbf{v}(t_0)\delta t$$

## ○ Obtaining the velocities

- *not evaluated during normal course of algorithm*
- *needed to compute some properties, e.g.*
  - *temperature*
  - *diffusion constant*
- *finite difference*

$$\mathbf{v}(t) = \frac{1}{2\delta t}[\mathbf{r}(t + \delta t) - \mathbf{r}(t - \delta t)] + O(\delta t^2)$$

# Verlet Algorithm 5. Performance Issues

## ○ Time reversible

- *forward time step*

$$\mathbf{r}(t + \delta t) = 2\mathbf{r}(t) - \mathbf{r}(t - \delta t) + \frac{1}{m}\mathbf{F}(t)\delta t^2$$

- *replace  $\delta t$  with  $-\delta t$*

$$\mathbf{r}(t + (-\delta t)) = 2\mathbf{r}(t) - \mathbf{r}(t - (-\delta t)) + \frac{1}{m}\mathbf{F}(t)(-\delta t)^2$$

$$\mathbf{r}(t - \delta t) = 2\mathbf{r}(t) - \mathbf{r}(t + \delta t) + \frac{1}{m}\mathbf{F}(t)\delta t^2$$

- *same algorithm, with same positions and forces, moves system backward in time*

## ○ Numerical imprecision of adding large/small numbers

$$\mathbf{r}(t + \delta t) - \mathbf{r}(t) = \mathbf{r}(t) - \mathbf{r}(t - \delta t) + \frac{1}{m}\mathbf{F}(t)\delta t^2$$

Diagram illustrating numerical imprecision in the Verlet algorithm. The equation is shown with boxes around terms and labels indicating their order of magnitude:

- $\mathbf{r}(t + \delta t) - \mathbf{r}(t)$  is labeled  $O(\delta t^1)$ .
- $\mathbf{r}(t)$  is labeled  $O(\delta t^0)$ .
- $\mathbf{r}(t - \delta t)$  is labeled  $O(\delta t^0)$ .
- $\frac{1}{m}\mathbf{F}(t)\delta t^2$  is labeled  $O(\delta t^2)$ .

# Initial Velocities

(from Lecture 3)

## ○ Random direction

- *randomize each component independently*
- *randomize direction by choosing point on spherical surface*

## ○ Magnitude consistent with desired temperature. Choices:

- *Maxwell-Boltzmann:  $\text{prob}(v_x) \propto \exp(-\frac{1}{2}mv_x^2 / kT)$*
- *Uniform over  $(-1/2, +1/2)$ , then scale so that  $\frac{1}{N} \sum v_{i,x}^2 = kT / m$*
- *Constant at  $v_x = \pm\sqrt{kT / m}$*
- *Same for y, z components*

## ○ Be sure to shift so center-of-mass momentum is zero

$$P_x \equiv \frac{1}{N} \sum p_{i,x}$$

$$p_{i,x} \rightarrow p_{i,x} - P_x$$



# Leapfrog Algorithm

- Eliminates addition of small numbers  $O(\delta t^2)$  to differences in large ones  $O(\delta t^0)$
- Algorithm

$$\mathbf{r}(t + \delta t) = \mathbf{r}(t) + \mathbf{v}(t + \frac{1}{2} \delta t) \delta t$$

$$\mathbf{v}(t + \frac{1}{2} \delta t) = \mathbf{v}(t - \frac{1}{2} \delta t) + \frac{1}{m} \mathbf{F}(t) \delta t$$

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- Mathematically equivalent to Verlet algorithm

$$\mathbf{r}(t + \delta t) = \mathbf{r}(t) + \left[ \mathbf{v}(t - \frac{1}{2} \delta t) + \frac{1}{m} \mathbf{F}(t) \delta t \right] \delta t$$

# Leapfrog Algorithm

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$\mathbf{r}(t)$  as evaluated from  
previous time step

$$\mathbf{r}(t) = \mathbf{r}(t - \delta t) + \mathbf{v}(t - \frac{1}{2} \delta t) \delta t$$

# Leapfrog Algorithm

- Eliminates addition of small numbers  $O(\delta t^2)$  to differences in large ones  $O(\delta t^0)$
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$$\mathbf{r}(t + \delta t) = \mathbf{r}(t) + \mathbf{v}(t + \frac{1}{2} \delta t) \delta t$$

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$$\mathbf{r}(t + \delta t) = \mathbf{r}(t) + \left[ \mathbf{v}(t - \frac{1}{2} \delta t) + \frac{1}{m} \mathbf{F}(t) \delta t \right] \delta t$$

$\mathbf{r}(t)$  as evaluated from  
previous time step

$$\mathbf{r}(t) = \mathbf{r}(t - \delta t) + \mathbf{v}(t - \frac{1}{2} \delta t) \delta t$$

$$\mathbf{r}(t + \delta t) = \mathbf{r}(t) + \left[ (\mathbf{r}(t) - \mathbf{r}(t - \delta t)) + \frac{1}{m} \mathbf{F}(t) \delta t^2 \right]$$

# Leapfrog Algorithm

- Eliminates addition of small numbers  $O(\delta t^2)$  to differences in large ones  $O(\delta t^0)$
- Algorithm

$$\mathbf{r}(t + \delta t) = \mathbf{r}(t) + \mathbf{v}(t + \frac{1}{2} \delta t) \delta t$$

$$\mathbf{v}(t + \frac{1}{2} \delta t) = \mathbf{v}(t - \frac{1}{2} \delta t) + \frac{1}{m} \mathbf{F}(t) \delta t$$

- Mathematically equivalent to Verlet algorithm

$$\mathbf{r}(t + \delta t) = \mathbf{r}(t) + \left[ \mathbf{v}(t - \frac{1}{2} \delta t) + \frac{1}{m} \mathbf{F}(t) \delta t \right] \delta t$$

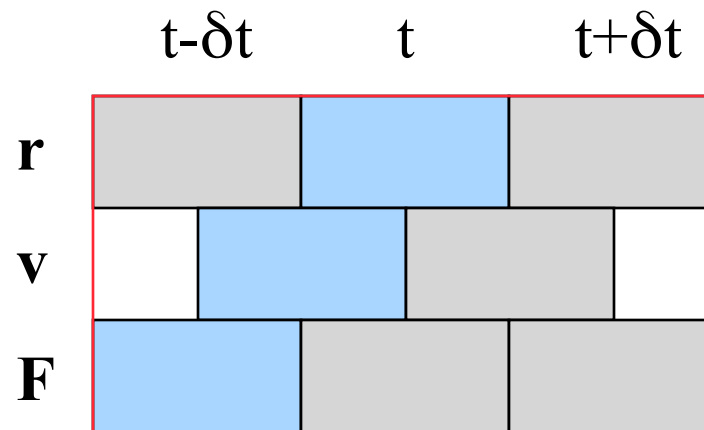
$\mathbf{r}(t)$  as evaluated from  
previous time step

$$\mathbf{r}(t) = \mathbf{r}(t - \delta t) + \mathbf{v}(t - \frac{1}{2} \delta t) \delta t$$

$$\mathbf{r}(t + \delta t) = \mathbf{r}(t) + \left[ (\mathbf{r}(t) - \mathbf{r}(t - \delta t)) + \frac{1}{m} \mathbf{F}(t) \delta t^2 \right]$$

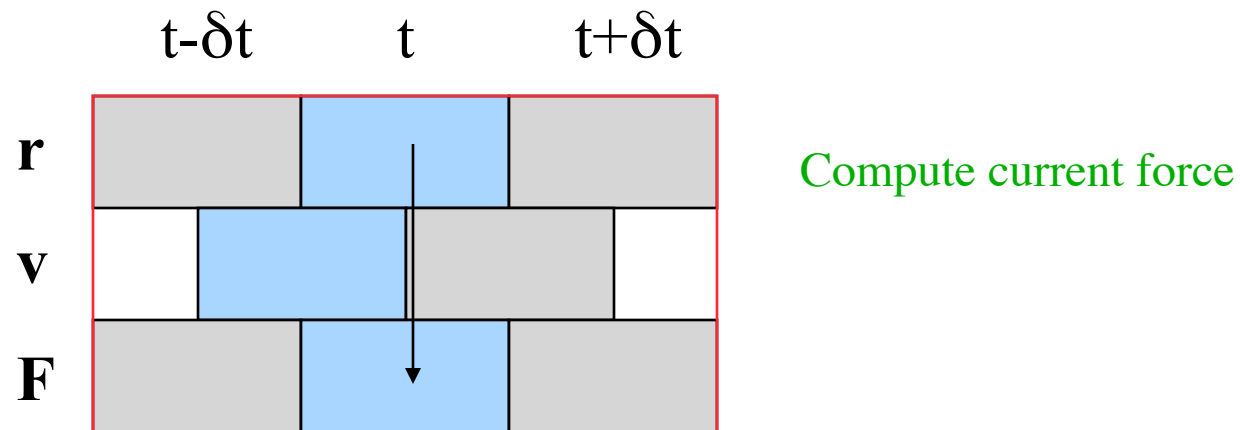
$$\mathbf{r}(t + \delta t) = 2\mathbf{r}(t) - \mathbf{r}(t - \delta t) + \frac{1}{m} \mathbf{F}(t) \delta t^2 \quad \text{original algorithm}$$

# Leapfrog Algorithm 2. Flow Diagram

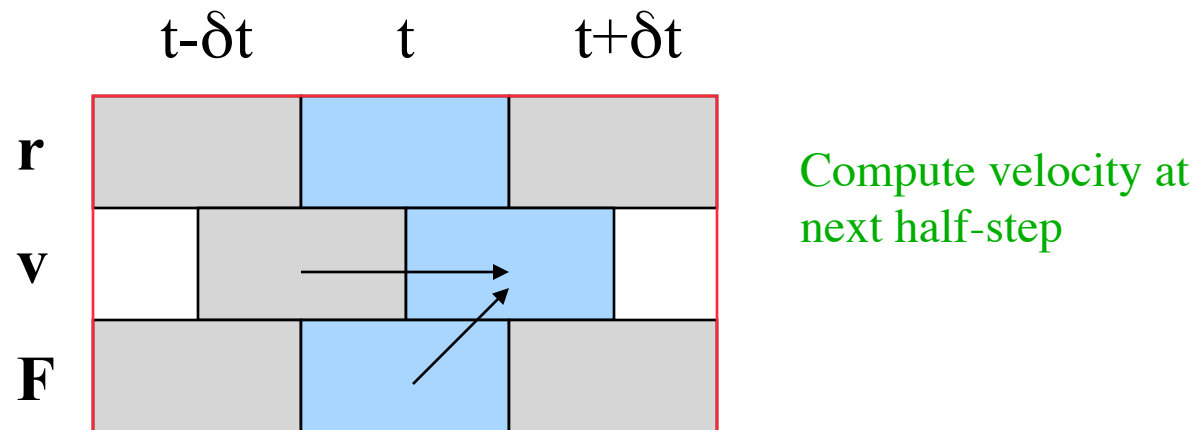


Given current position, and  
velocity at last half-step

# Leapfrog Algorithm 2. Flow Diagram

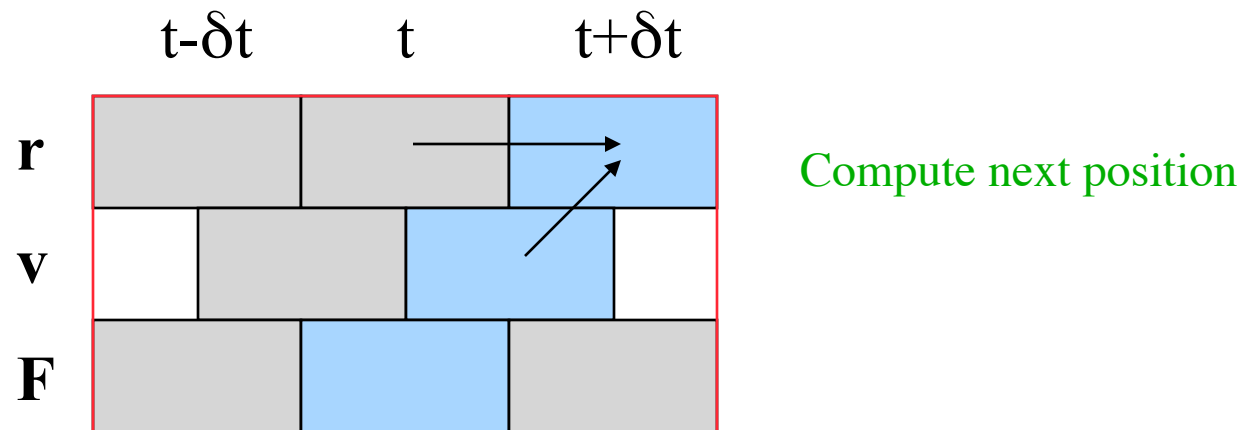


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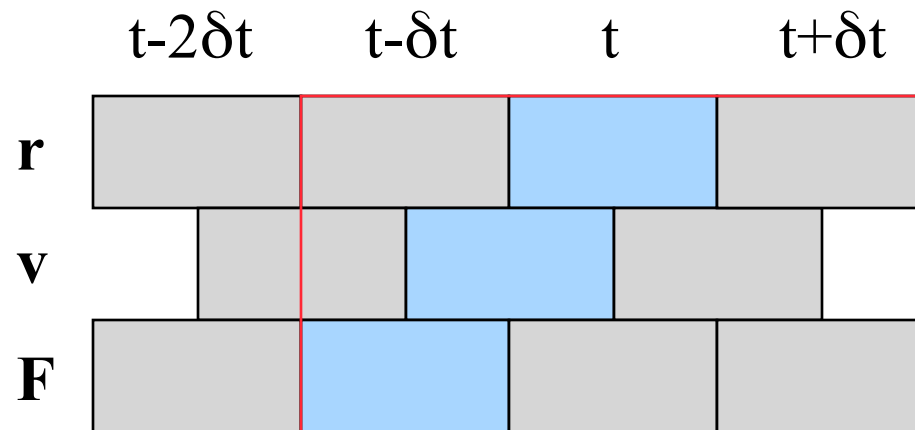




# Leapfrog Algorithm 2. Flow Diagram



# Leapfrog Algorithm 2. Flow Diagram



Advance to next time step,  
repeat

## Leapfrog Algorithm. 3. Loose Ends

### ○ Initialization

- *how to get velocity at “previous time step” when starting out?*
- *simple approximation*

$$\mathbf{v}(t_0 - \frac{1}{2} \delta t) = \mathbf{v}(t_0) - \frac{1}{m} \mathbf{F}(t_0) \frac{1}{2} \delta t$$

### ○ Obtaining the velocities

- *interpolate*

$$\mathbf{v}(t) = \frac{1}{2} \left[ \mathbf{v}(t + \frac{1}{2} \delta t) + \mathbf{v}(t - \frac{1}{2} \delta t) \right]$$

# Velocity Verlet Algorithm

- Roundoff advantage of leapfrog, but better treatment of velocities
- Algorithm

$$\mathbf{r}(t + \delta t) = \mathbf{r}(t) + \mathbf{v}(t)\delta t + \frac{1}{2m}\mathbf{F}(t)\delta t^2$$

$$\mathbf{v}(t + \delta t) = \mathbf{v}(t) + \frac{1}{2m}[\mathbf{F}(t) + \mathbf{F}(t + \delta t)]\delta t$$

- Implemented in stages
  - *given current force*
  - *compute  $\mathbf{r}$  at new time*
  - *add current-force term to velocity (gives  $\mathbf{v}$  at half-time step)*
  - *compute new force*
  - *add new-force term to velocity*
- Also mathematically equivalent to Verlet algorithm (in giving values of  $\mathbf{r}$ )

# Velocity Verlet Algorithm

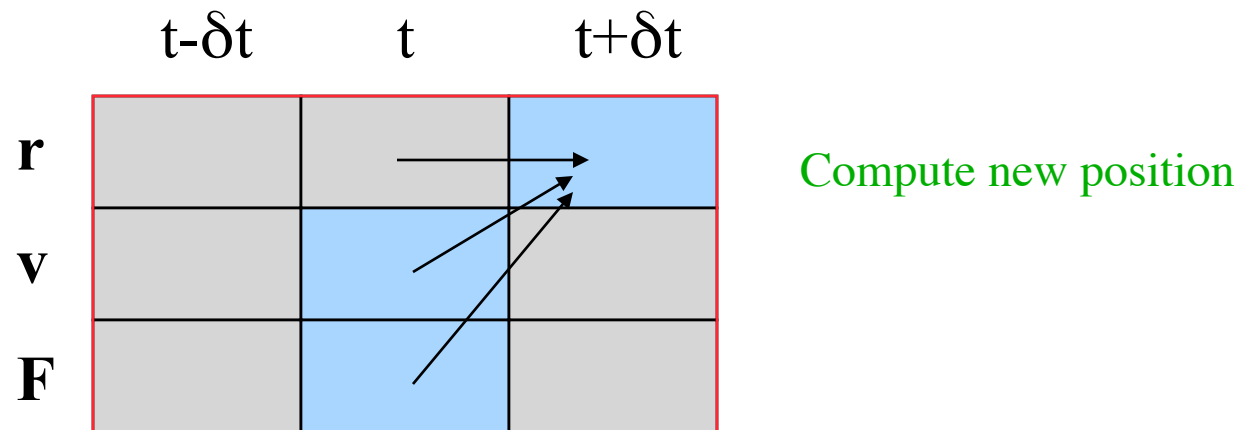
## 2. Flow Diagram

	$t-\delta t$	$t$	$t+\delta t$
<b>r</b>			
<b>v</b>			
<b>F</b>			

Given current position,  
velocity, and force

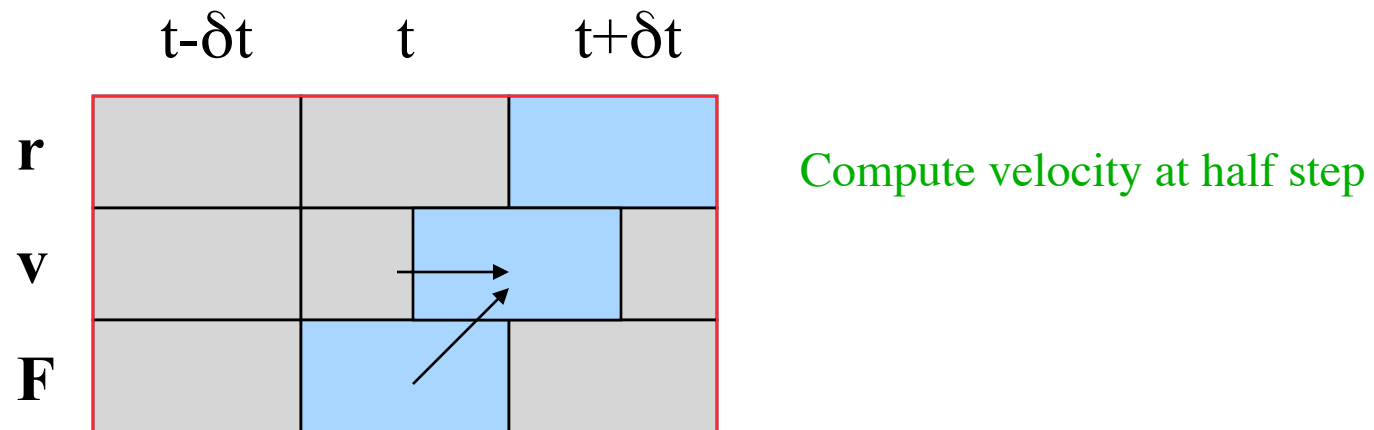
# Velocity Verlet Algorithm

## 2. Flow Diagram



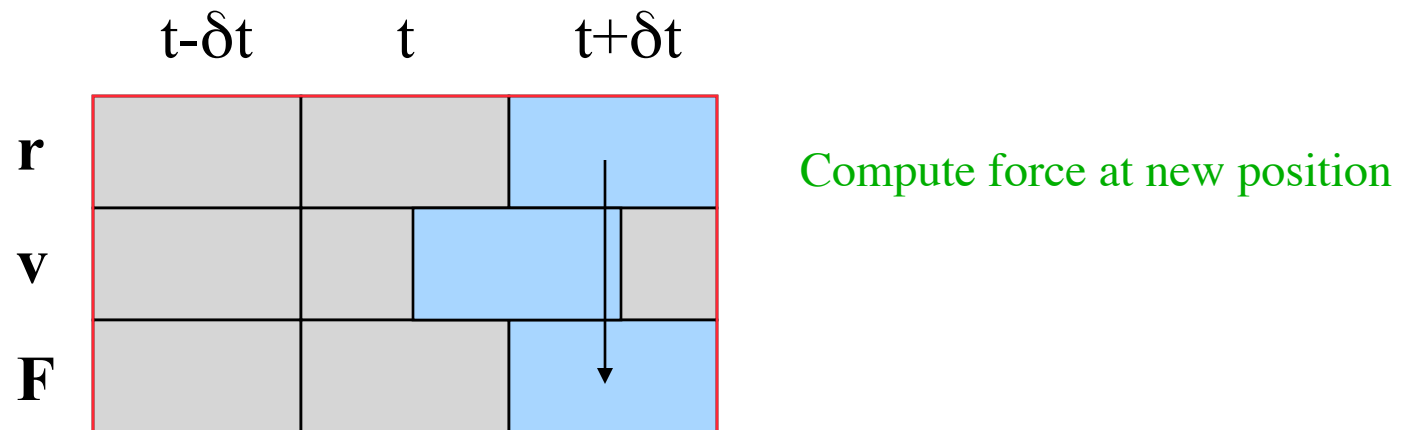
# Velocity Verlet Algorithm

## 2. Flow Diagram



# Velocity Verlet Algorithm

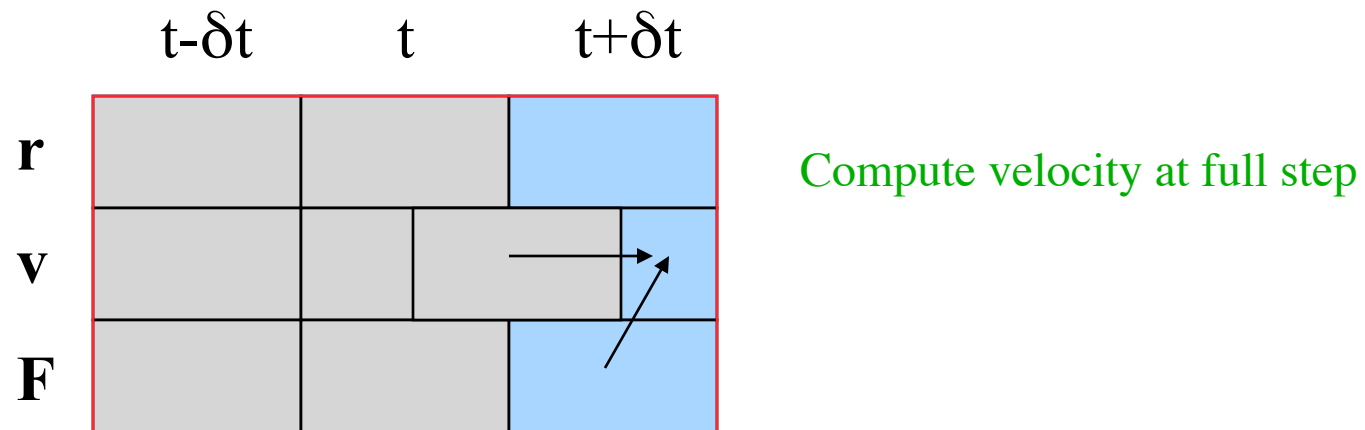
## 2. Flow Diagram





# Velocity Verlet Algorithm

## 2. Flow Diagram



# Velocity Verlet Algorithm

## 2. Flow Diagram

	$t-2\delta t$	$t-\delta t$	$t$	$t+\delta t$
<b>r</b>				
<b>v</b>				
<b>F</b>				

Advance to next time step,  
repeat

# Other Algorithms

## ○ Predictor-Corrector

- *not time reversible*
- *easier to apply in some instances*
  - constraints
  - rigid rotations

## ○ Beeman

- *better treatment of velocities*

## ○ Velocity-corrected Verlet

# Summary

## ○ Several formulations of mechanics

- *Hamiltonian preferred*
  - independence of choice of coordinates
  - emphasis on phase space

## ○ Integration algorithms

- *Calculation of forces*
- *Simple Verlet algorithms*
  - Verlet
  - Leapfrog
  - Velocity Verlet

## ○ Next up: Calculation of dynamical properties