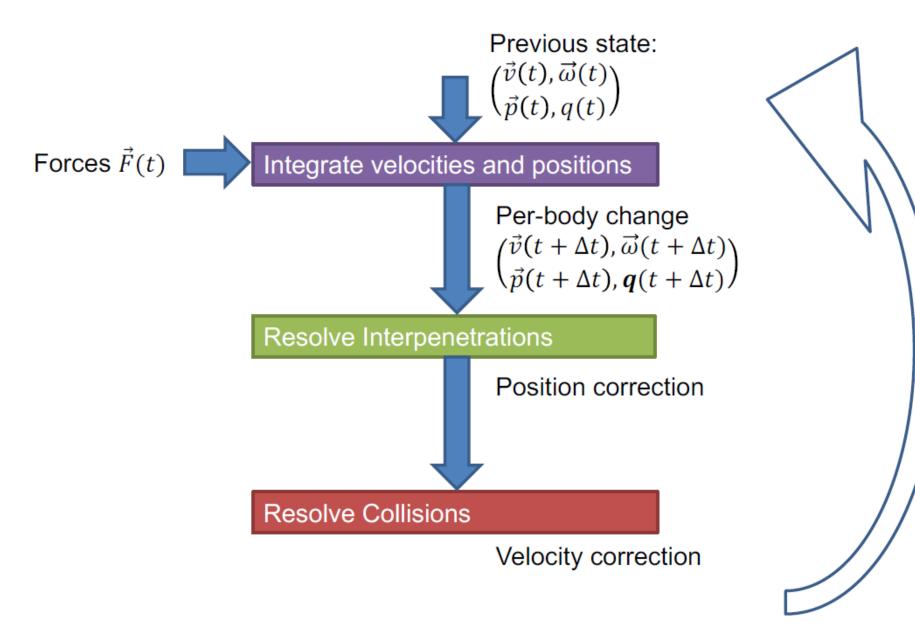
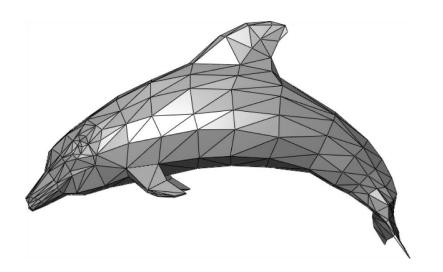
The game-engine loop & Time Integration

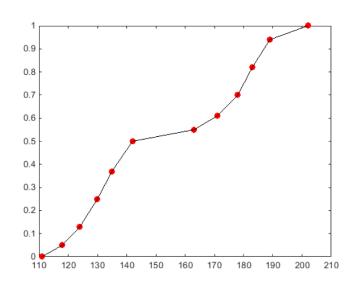
The Basic Game-Engine Loop



Challenges

- Kinematics: continuous motion in continuous time.
 - Events are local and always valid.
- Computer simulation:
 - Discrete time steps Δt.
 - Discrete Space (mesh, particles, grids)





Updating Position

- Force induces acceleration.
- When mass is constant:

$$F(p,t) = m \cdot a(p,t)$$

- Derivatives: v'(t) = a(t) and p'(t) = v(t)
- Thus: $F(p,t) = m \cdot p''(t)$

- A differential equation.
 - Often impossible to solve analytically.
 - More often, a = a(v, t) (velocity dependence).
- Discretization: stability and convergence issues.

Taylor Approximation

• A function at $t + \Delta t$ can be approximated by a polynomial centered at t with arbitrary precision:

$$p(t + \Delta t)$$

$$\approx p(t) + \Delta t \cdot p'(t) + \frac{\Delta t^2}{2} p''(t) + \dots + \frac{\Delta t^n}{n!} p^{(n)}(t)$$

We do not usually use (or have) more than 2nd derivatives.

First-Order Approximation

• If Δt is small enough, we approximate linearly:

$$p(t + \Delta t) \approx p(t) + \Delta t. p'(t)$$

 Euler's Method: approximating forward both velocity and position within the same step:

$$v(t + \Delta t) = v(t) + a(t)\Delta t = v(t) + \frac{F(t)}{m}\Delta t$$

$$p(t + \Delta t) = p(t) + v(t)\Delta t$$
 Unknown for next time step Assumed known for this time step

Euler's Method

- Note: we approximate the velocity as constant between frames.
 - We compute the acceleration of the object from the net force applied on it:

$$a(t) = \frac{F(t)}{m}$$

• We compute the velocity from the acceleration:

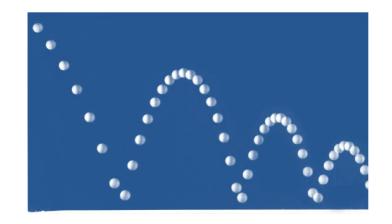
$$v(t + \Delta t) = v(t) + a(t)\Delta t$$

We compute the position from the velocity:

$$p(t + \Delta t) = p(t) + v(t)\Delta t$$

Issues with Linear Dynamics

- A mere sequence of instants.
 - Without the precise instant of bouncing.



- Trajectories are piecewiselinear.
 - Constant velocity and acceleration in-between frames.



Time Step

When $\Delta t \to 0$, we converge to $p(t) = \int_0^t v(s) ds$

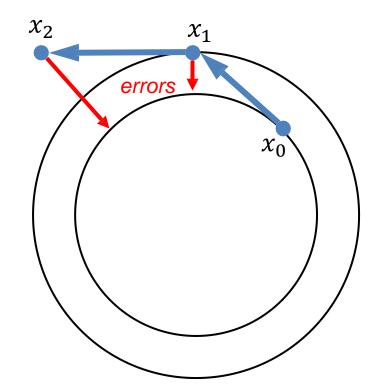
 (Im)possible solution: reducing Δt to convergence levels?

- First-order method, piecewise-constant velocity: not very stable.
- Our objective: make the most with every Δt you get.

Time Step

- First-order assumption: the slope at t as a good estimate for the slope over the entire interval Δt
- The approximation can drift off the function.
- Farther drifting

 tangent approximation worse.



Time step

```
void takeStep(ParticleSystem* ps, float h)
      velocities = ps->getStateVelocities()
      positions = ps->getStatePositions()
      forces = ps->getForces(positions, velocities)
      masses = ps->getMasses()
      accelerations = forces / masses
      newPositions = positions + h*velocities
      newVelocities = velocities + h*accelerations
      ps->setStatePositions(newPositions)
      ps->setStateVelocities(newVelocities)
```

Midpoint Method

Estimating tangent in Half step:

$$v(t + \frac{\Delta t}{2}) = v(t) + a(t, v) \frac{\Delta t}{2}$$

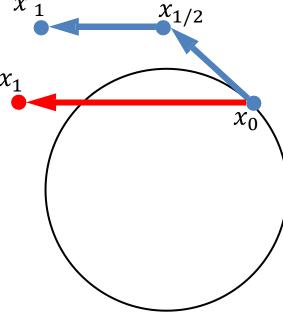
Full step:

$$v(t + \Delta t) = v(t) + a(t + \frac{\Delta t}{2}, t + \frac{\Delta t}{2})\Delta t$$

- 2nd-order approximation.
- Compute position similarly with v.

Midpoint Method

- Approximating the tangent in mid-interval.
- Applying it to initial point across the entire interval.
- Error order: the square of the time step $O(\Delta t^2)$. Better than Euler's method $(O(\Delta t))$ when $\Delta t < 1$.
- Approximating with a quadratic curve instead of a line. x_1'
- ...can still drift off the function.



Midpoint Method

```
void takeStep(ParticleSystem* ps, float h)
{
      velocities = ps->getStateVelocities()
      positions = ps->getStatePositions()
      forces = ps->getForces(positions, velocities)
      masses = ps->getMasses()
      accelerations = forces / masses
      midPositions = positions + 0.5*h*velocities
      midVelocities = velocities + 0.5*h*accelerations
      midForces = ps->getForces(midPositions, midVelocities)
      midAccelerations = midForces / masses
      newPositions = positions + h*midVelocities
      newVelocities = velocities + h*midAccelerations
      ps->setStatePositions(newPositions)
      ps->setStateVelocities(newVelocities)
```

Improved Euler's Method

- Considers the tangent lines to the solution curve at both ends of the interval.
- Velocity to the first point (Euler's prediction):

$$v_1 = v(t) + \Delta t. a(t, v)$$

Velocity to the second point (correction point):

$$v_2 = v(t) + \Delta t. a(t + \Delta t, v_1)$$

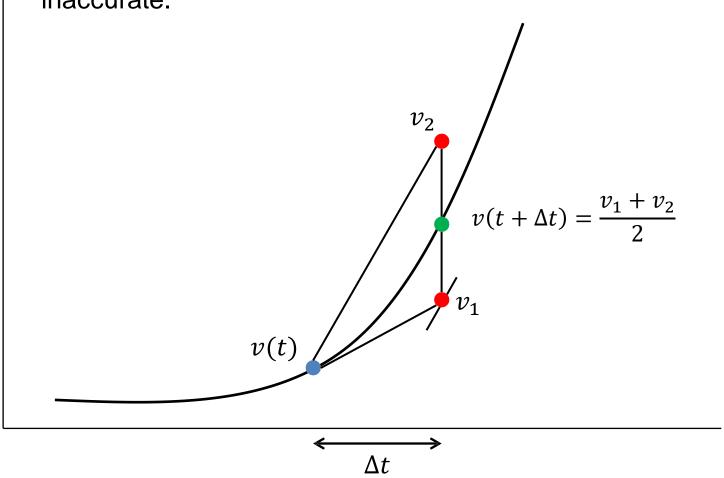
Improved Euler's velocity

$$v(t + \Delta t) = \frac{v_1 + v_2}{2}$$

• Compute position similarly with v_1, v_2 , instead of a.

Improved Euler's Method

The order of the error is $O(\Delta t^2)$. The final derivative is still inaccurate.



Runge-Kutta Method

 There are methods that provide better than quadratic error.

• The Runge-Kutta order-four method (RK4) is $O(\Delta t^4)$.

 A combination of the midpoint and modified Euler's methods, with higher weights to the midpoint tangents than to the endpoints tangents.

RK4

 Computing the four following tangents (note dependence of acceleration on velocity):

$$v_1 = \Delta t \cdot a(t, v(t))$$

$$v_2 = \Delta t \cdot a\left(t + \frac{\Delta t}{2}, v(t) + \frac{1}{2}v_1\right)$$

$$v_3 = \Delta t \cdot a\left(t + \frac{\Delta t}{2}, v(t) + \frac{1}{2}v_2\right)$$

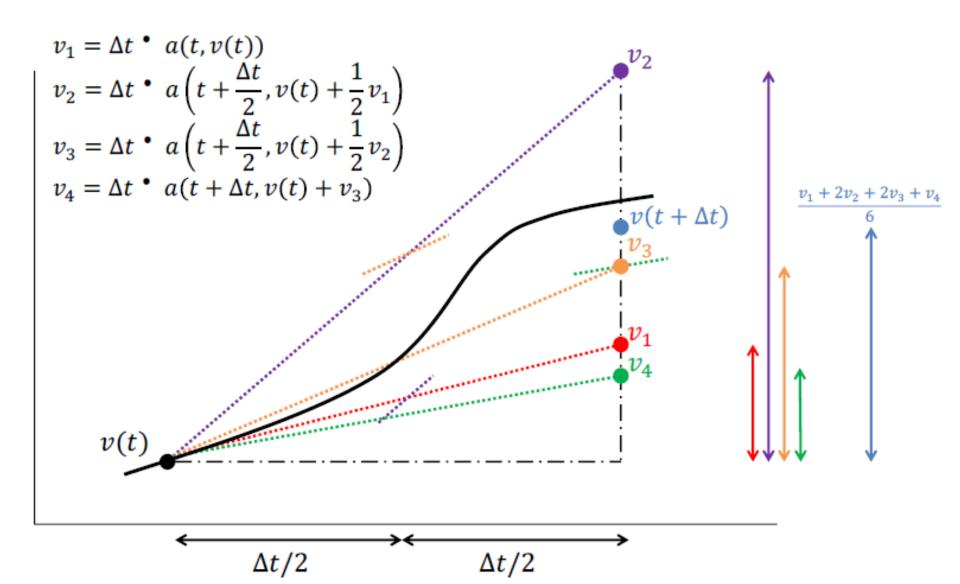
$$v_4 = \Delta t \cdot a\left(t + \Delta t, v(t) + v_3\right)$$

Blend as follows:

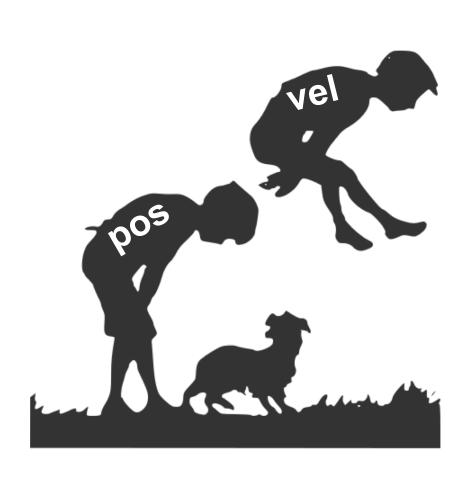
$$v(t + \Delta t) = v(t) + \frac{v_1 + 2v_2 + 2v_3 + v_4}{6}$$

Compute position similarly with v values.

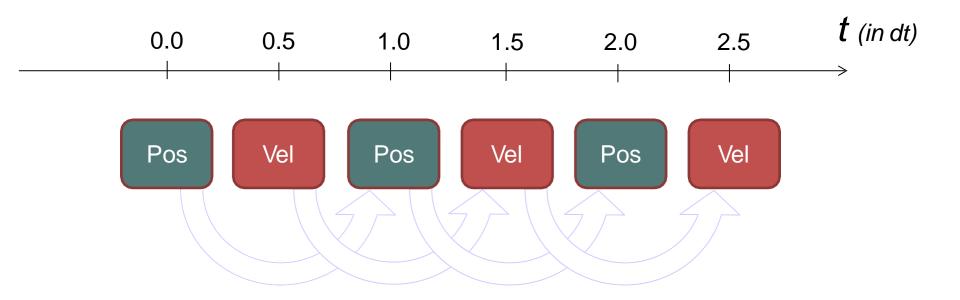
RK4



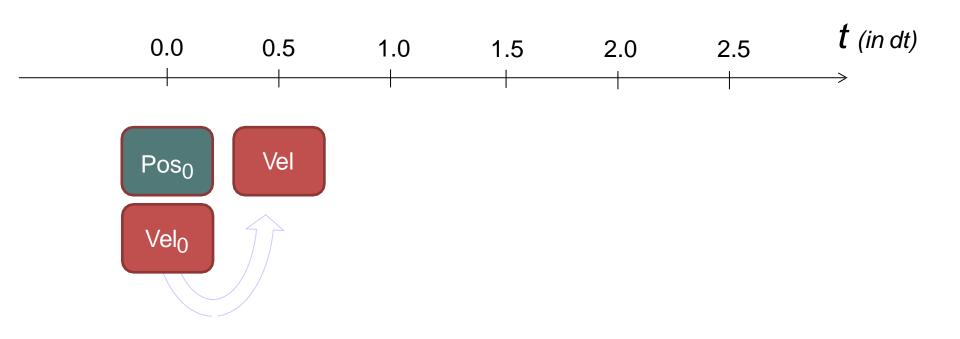
Leapfrog Integration ("a cavallina")



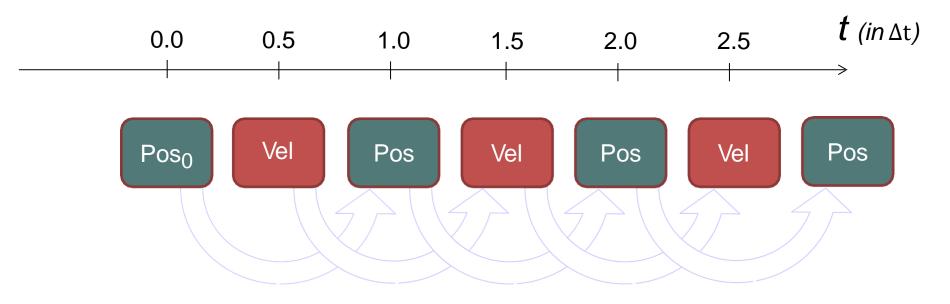
Leapfrog Integration



Leapfrog Integration first step



Leapfrog Integration



$$\vec{p}(1) = \vec{p}(0) + \vec{v}(0.5)\Delta t$$

$$\vec{p}(2) = \vec{p}(1) + \vec{v}(1.5)\Delta t$$

$$\vec{p}(3) = \vec{p}(2) + \vec{v}(2.5)\Delta t$$

$$\vec{v}(1.5) = \vec{v}(0.5) + \vec{a}\Delta t$$

 $\vec{v}(2.5) = \vec{v}(1.5) + \vec{a}\Delta t$

Leapfrog Method

- More accurate than Euler-based methods
 - Residue of $O(\Delta t^3)$
- But at the same cost as Euler's method!
- Major advantage: fully reversible!

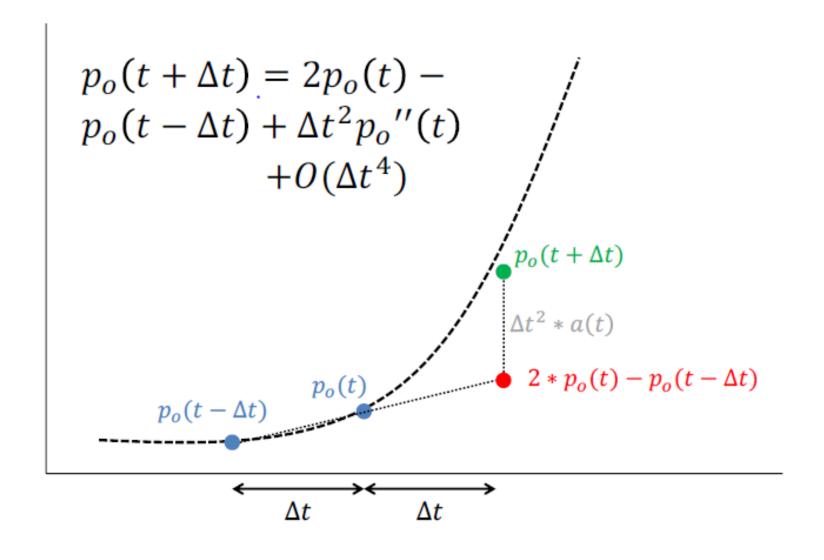
Verlet integration

 Based on the Taylor expansion series of the previous time step and the next one:

$$\begin{aligned} p_o(t + \Delta t) + p_o(t - \Delta t) \\ &\approx p_o(t) + \Delta t \cdot p_o{'}(t) + \frac{\Delta t^2}{2} \cdot p_o{''}(t) + \cdots \\ &+ p_o(t) - \Delta t \cdot p_o{'}(t) + \frac{\Delta t^2}{2} * p_o{''}(t) - \cdots \end{aligned}$$
Cancels out!

Verlet integration

Approximating without velocity:



Verlet integration

- An $O(\Delta t^2)$ order of error.
- Very stable and fast without the need to estimate velocities.
- We need an estimation of the first $p_0 = (t \Delta t)$
 - Usually obtained from one step of Euler's or RK4 method.
- Difficult to manage velocity related forces such as drag or collision.

Implicit methods

- So far: computing current position p(t) and velocity v(t) for the next position (forward).
 - Those are denoted as explicit methods.
- In implicit methods, we make use of the quantities from the next time step!

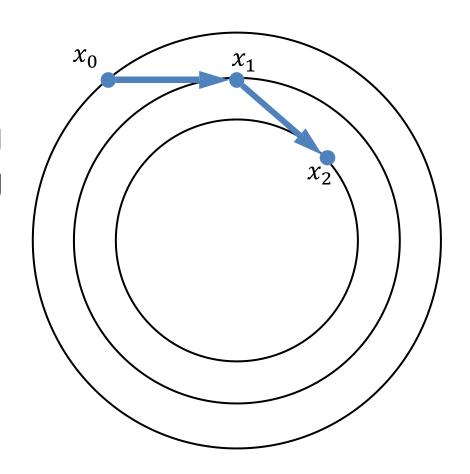
$$p(t) = p(t + \Delta t) - \Delta t \cdot v(t + \Delta t)$$

- This particular one: backward Euler.
- Computing in inverse:
- Finding position $p(t + \Delta t)$ which produces p(t) if simulation is run backwards.

Implicit methods

 Not more accurate than explicit methods, but more stable.

 Especially for a damping of the position (e.g. drag force or kinetic friction).



Backward Euler

How to compute the velocity from the future?

- Given the forces applied, extracting from the formula:
 - Example: a drag force $F_D = -b.v$ is applied:

$$\frac{v(t + \Delta t) - v(t)}{\Delta t} = -b \cdot v(t + \Delta t)$$

And therefore

$$v(t + \Delta t) = \frac{v(t)}{1 + \Delta t. b}$$

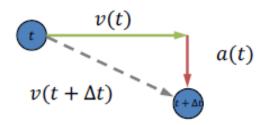
Backward Euler

- Often not knowing the forces in advance (likely case in a game).
- Or that the backward equation is not easy to solve.
- We use a predictor-corrector method:
 - one step of explicit Euler's method
 - use the predicted position to calculate $v(t + \Delta t)$
- More accurate than explicit method but twice the amount of computation.

Semi-Implicit Method

- Combines the simplicity of explicit Euler and stability of implicit Euler.
- Runs an explicit Euler step for velocity and then an implicit Euler step for position:

$$v(t + \Delta t) = v(t) + \Delta t * a(t) = v(t) + \Delta t * F(t)/m$$
$$p(t + \Delta t) = p(t) + \Delta t * \frac{v(t)}{t} = p(t) + \Delta t * v(t + \Delta t)$$



Semi-Implicit Method

- The position update in the second step uses the next velocity in the implicit scheme.
 - Good for position-dependent forces.
 - Conserves energy over time, and thus stable.

• Not as accurate as RK4 (order of error is still $O(\Delta t)$), but cheaper and yet stable.

Very popular choice for game physics engine.

Angular Integration

Examples: Forward Euler

$$\frac{dq}{dt} = \frac{1}{2}(0,\vec{\omega})q$$

becomes:

$$\frac{q(t + \Delta t) - q(t)}{\Delta t} = \frac{1}{2}(0, \vec{\omega}(t))q(t) \Rightarrow$$
$$q(t + \Delta t) = q(t) + \frac{1}{2}\Delta t(0, \vec{\omega}(t))q(t)$$

• The rest of the formulations follow suit for angular displacement θ , velocity ω and acceleration α .

Summary

- First-order methods
 - Implicit and Explicit Euler method, Semi-implicit Euler, Exponential Euler
- Second-order methods
 - Verlet integration, Velocity Verlet, Trapezoidal rule, Beeman's algorithm, Midpoint method, Improved Euler's method, Heun's method, Newmark-beta method, Leapfrog integration.
- Higher-order methods
 - Runge-Kutta family methods, Linear multistep method.
- Position-based methods
 - Leapfrog, Verlet.