

Particle Systems

Lesson 03

Lesson 03 Outline

- * Newton dynamics of particles
- * Ordinary differential equation (ODE) solver
- * Particle - obstacle collision detection
- * Practical design of particle system
- * Demos / tools / libs

Newton Dynamics



Newton's Dynamics

- * Three fundamental Newton's laws of motion
 - (1) Every body remains in a state of rest or uniform motion (constant velocity) unless it is acted upon by an external unbalanced force.
 - (2) A body of mass m subject to a force f undergoes an acceleration a that has the same direction as the force and a magnitude that is directly proportional to the force and inversely proportional to the mass: $f = ma$.
 - (3) The mutual forces of action and reaction between two bodies are equal, opposite and collinear.

Particle Dynamics

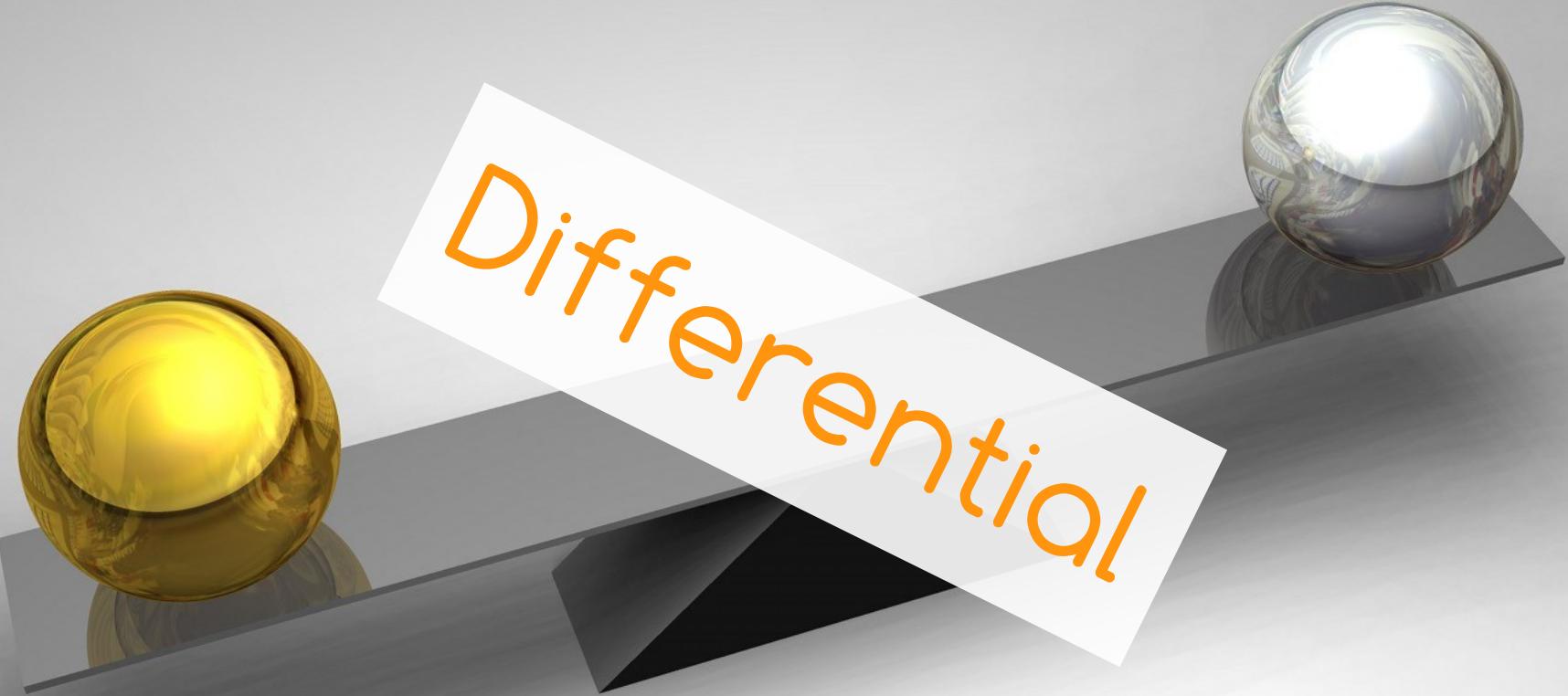
- * Dynamical properties of Particles

- Mass (m) in [kg]: parameter
- Position (ρ) in [m]: $d\rho = v$
- Velocity (v) in [m/s]: $dv = a$
- Momentum (L) in [kgm/s]: $L=mv$
- Acceleration (a) in (m/s²): $a = m^{-1}F$; gravity, wind, user...
- Force (F) in [kgm/s²]: $F = ma = dL$

- * The equation of unconstrained motion (ODE)

- $d(\rho, v) = (v, a)$

Ordinary



Differential

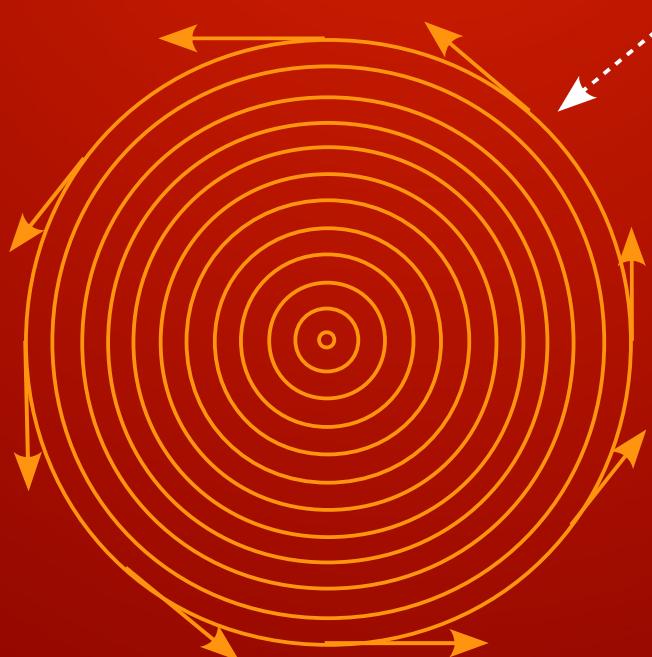
Equations

Ordinary Differential Equations

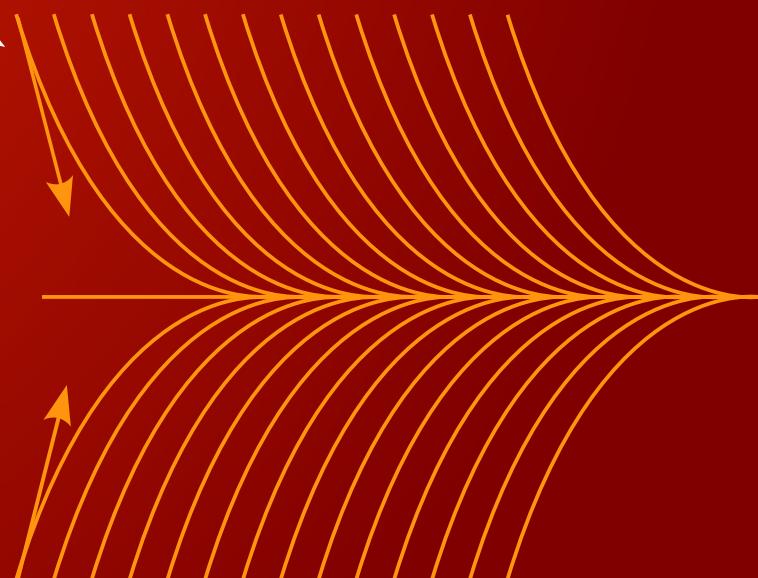
- * **Definition:** An ordinary differential equation (ODE) is a relation that contains functions of only one independent variable, and one or more of their derivatives with respect to that variable.
- * **Problem:** How to evaluate (in time) position $\rho(t)$ of a particle, when we only know its change in time $\rho'(t)$ is a function of position and time: $\rho'(t) = F(\rho(t), t)$
- * **Examples**
 - $\rho'(t) = -10\rho(t)$
 - $\rho'(t) = t^2\rho(t) - 3\rho^2(t) + 7$
- * **Objective:** Given function $F(\rho, t)$ and the value $\rho(t)$ at some time t , we can compute $\rho'(t) = F(\rho(t), t)$

ODE - Numerical problems

- * Inaccuracy Problem



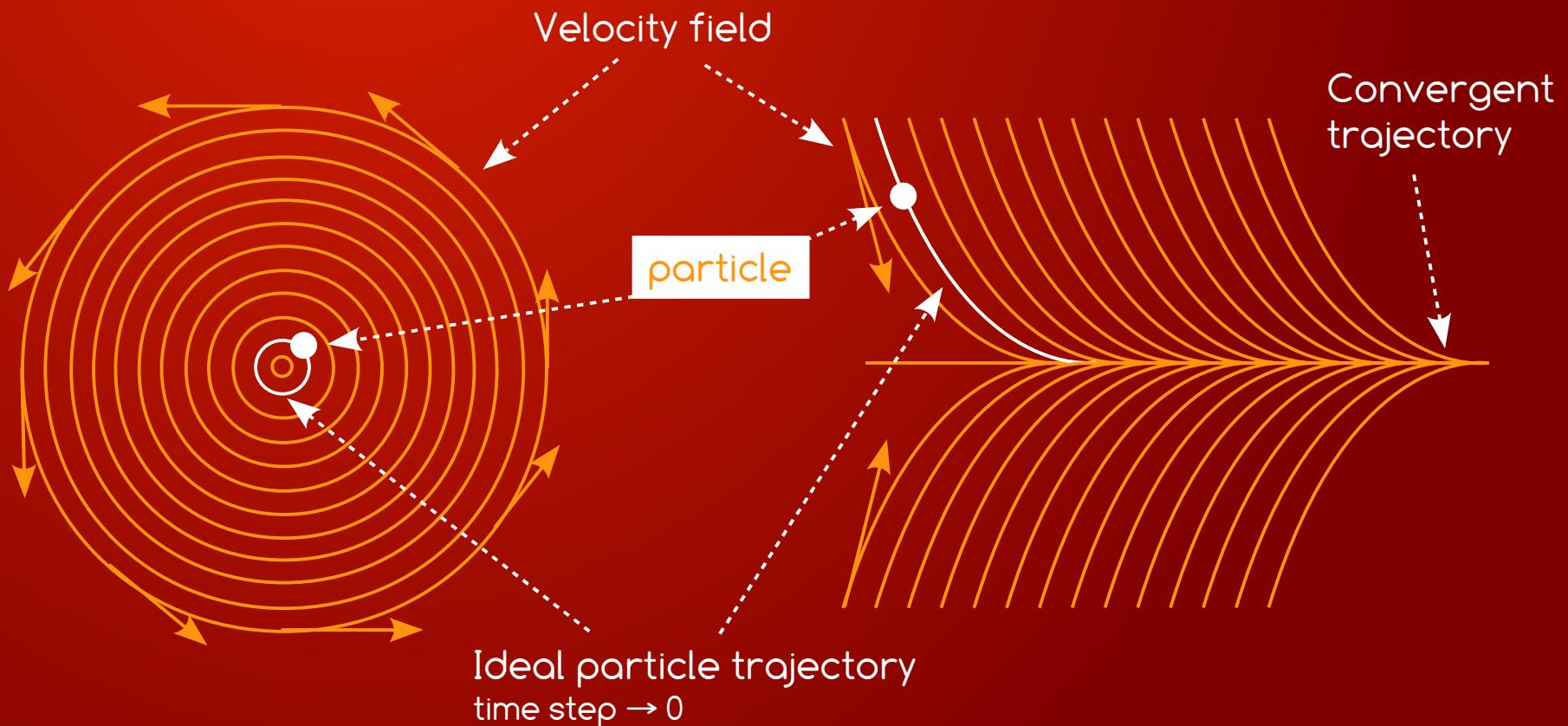
- * Instability Problem



ODE - Numerical problems

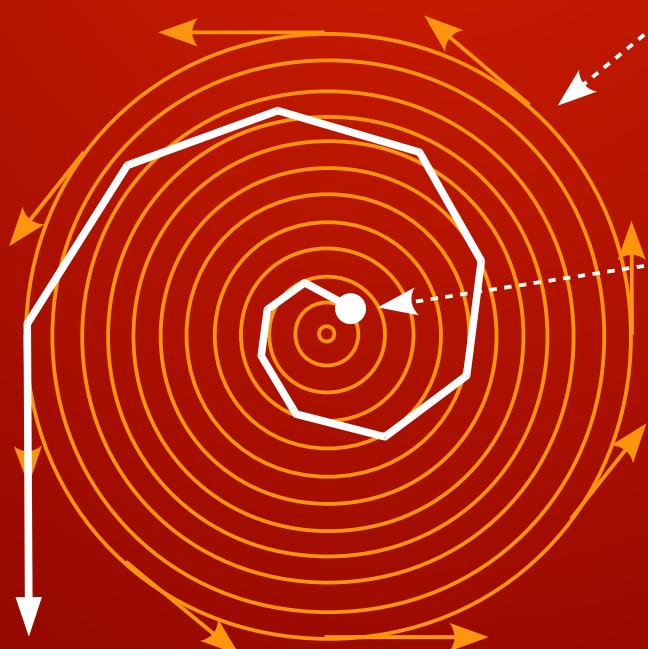
* Inaccuracy Problem

* Instability Problem

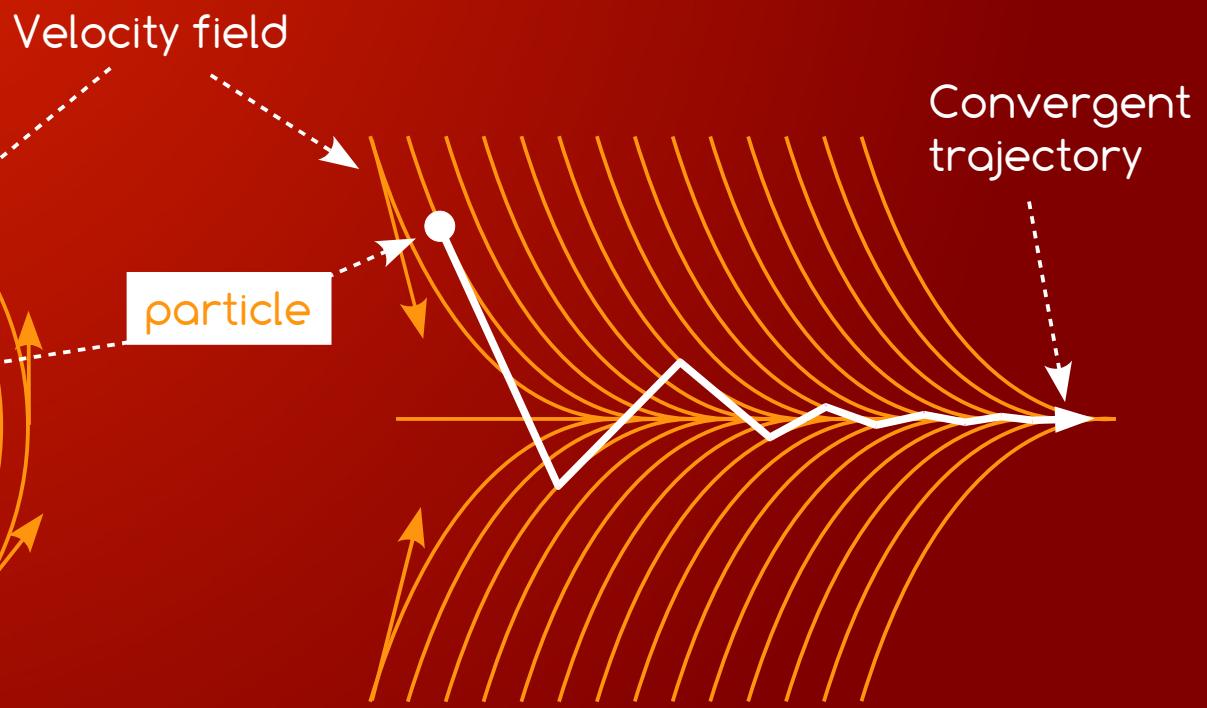


ODE - Numerical problems

* Inaccuracy Problem



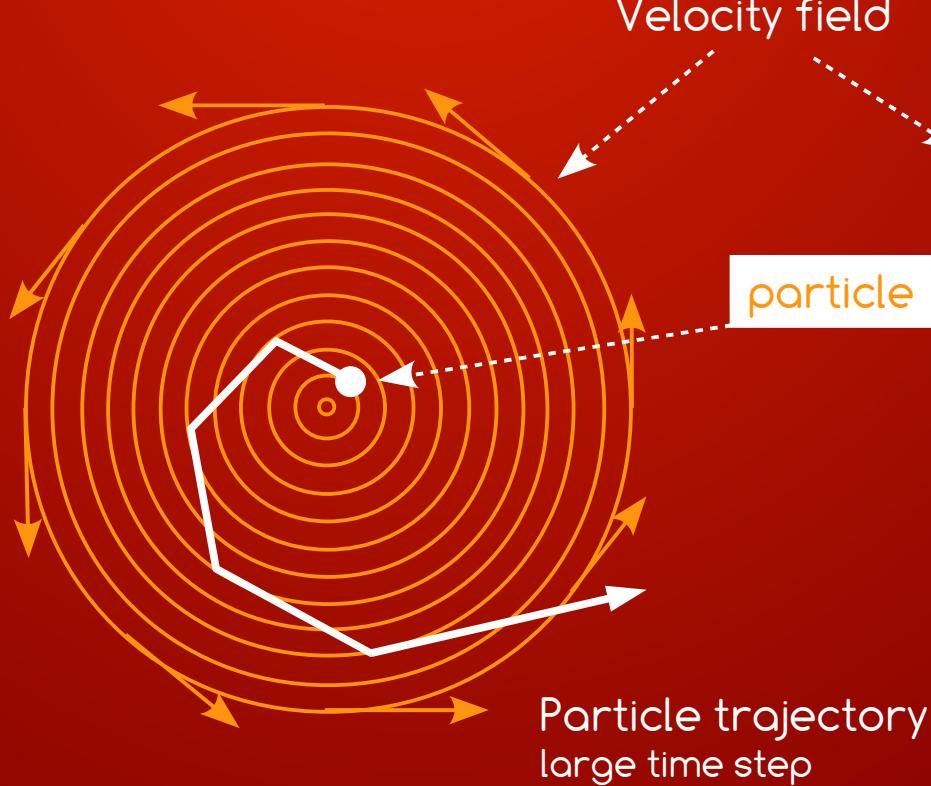
* Instability Problem



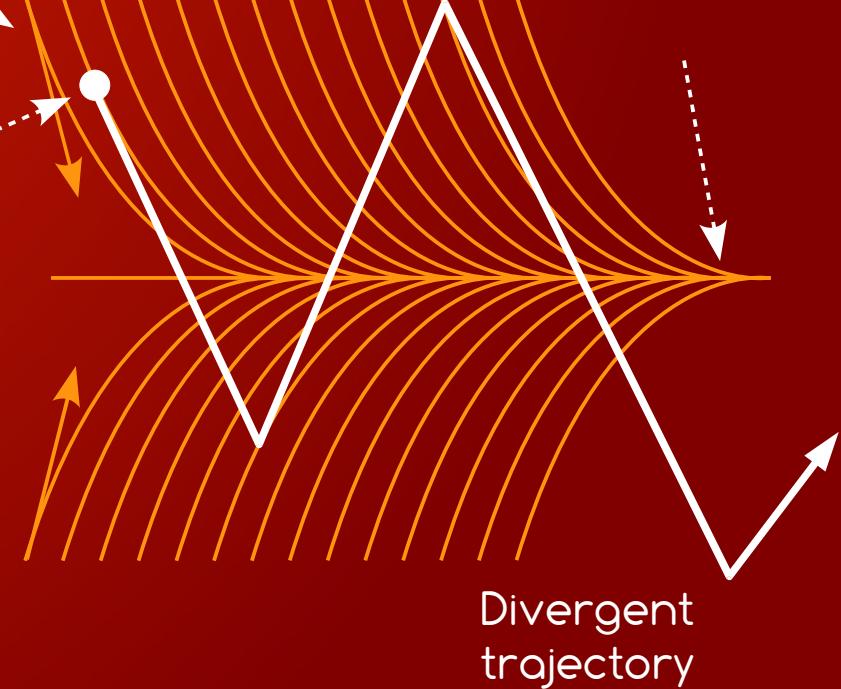
Particle trajectory
small time step

ODE - Numerical problems

* Inaccuracy Problem

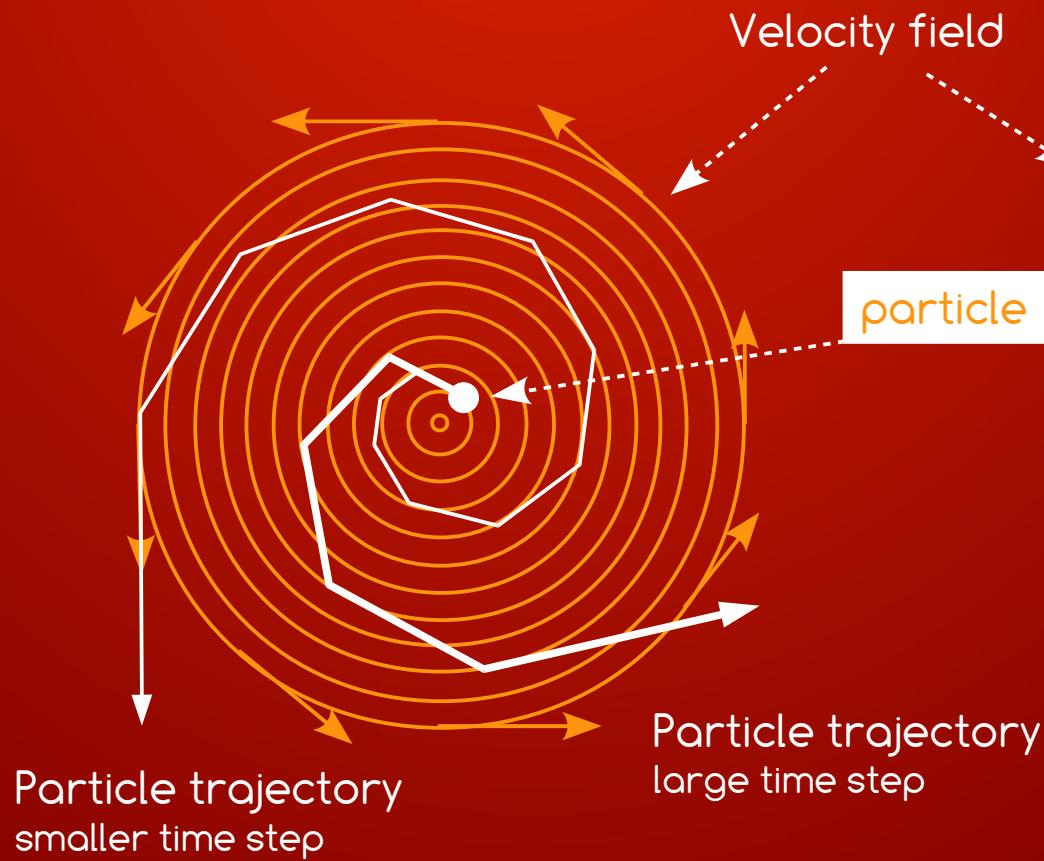


* Instability Problem

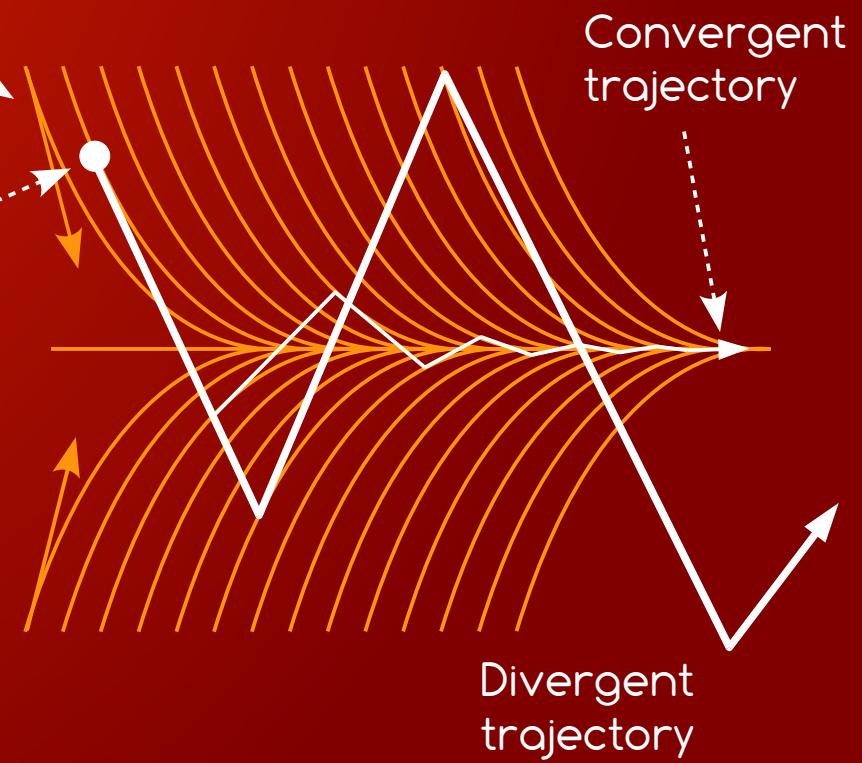


ODE - Numerical problems

* Inaccuracy Problem



* Instability Problem



ODE Solvers

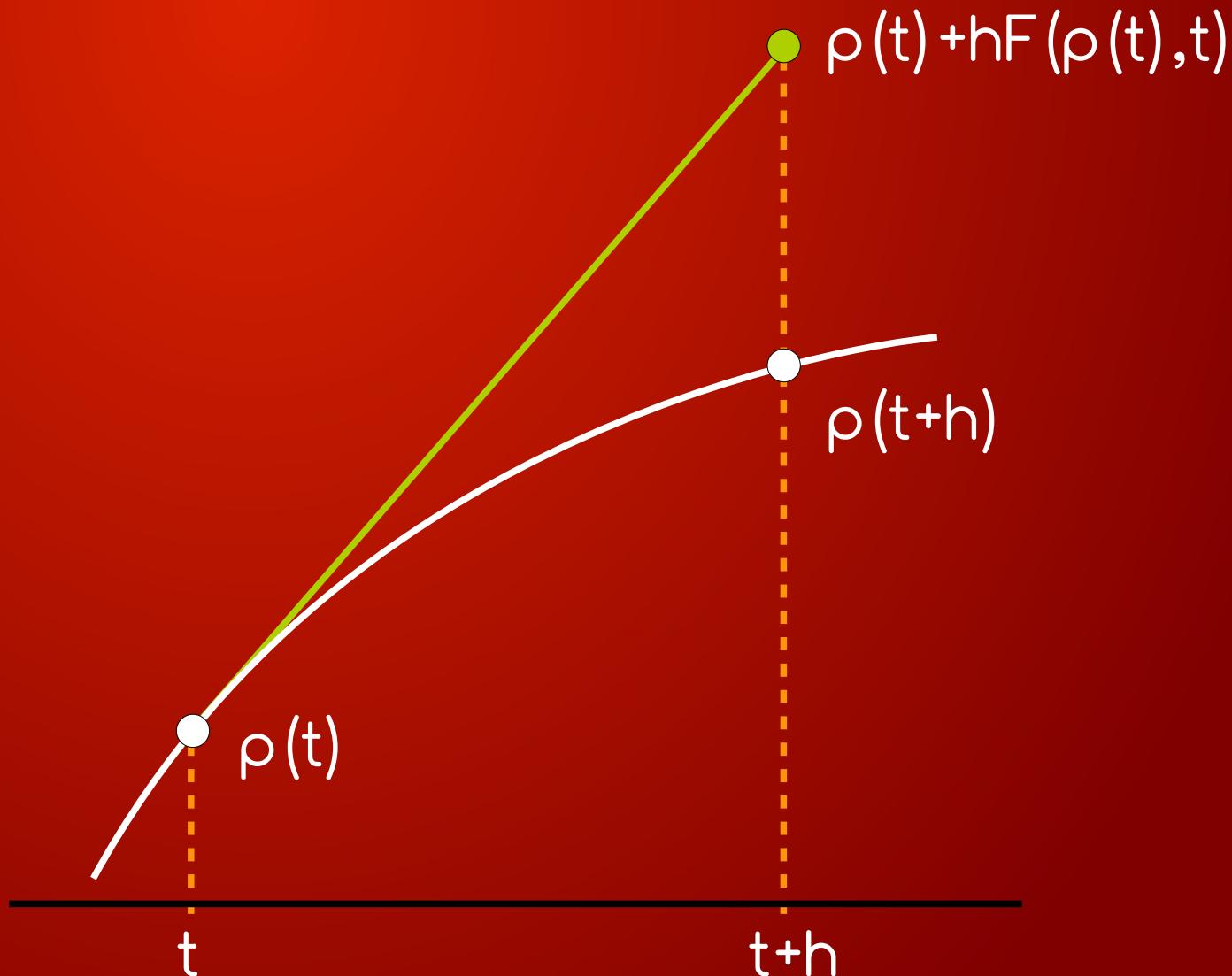
- * Explicit Schemes

- Euler
- Mid Point
- Runge Kutta 4
- Verlet

- * Implicit Schemes

- Implicit Euler

Explicit Euler Scheme



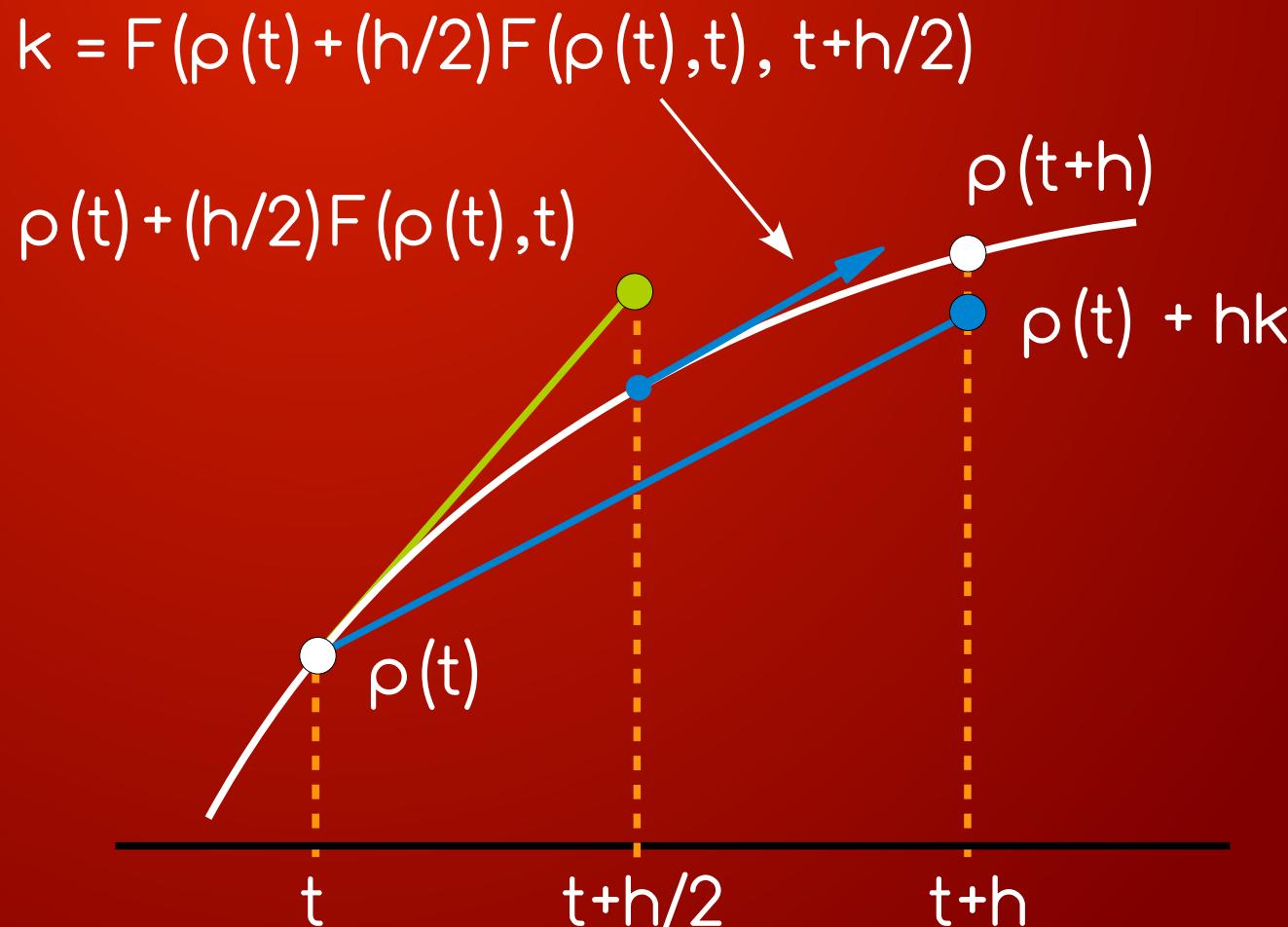
Explicit Euler Scheme

- * **Idea:** Given initial value $\rho(t_0)$ of function ρ at time t_0 we can find $\rho(t_0+h)$ using Taylor expansion as
- * $\rho(t_0+h) = \rho(t_0) + h\rho'(t_0) + O(h^2)$; $\rho'(t_0) = F(\rho(t_0), t_0)$
- * **Numerical algorithm:**
- * $\rho_{n+1} = \rho_n + h * F(\rho_n, t_n)$ where ρ_0 = some initial value
- * **Pros / Cons:**
 - Very simple, fast and easy to implement
 - Huge error = $O(h^2)$
 - Can be unstable – cumulated error increases to infinity

Explicit Midpoint Scheme

- * **Idea:** Use approximate derivative $\rho'(t+h/2)$ of $\rho(t)$ at time $t+h/2$ instead of the simple $\rho'(t)$
- * $\rho(t+h) = \rho(t) + hF(\rho(t+h/2), t+h/2) + O(h^3)$
- * **Problem:** We do not know function $\rho(t)$ or its derivative at time $t+h/2$.
- * Knowing $\rho'(t+h/2) = F(\rho(t+h/2), t+h/2)$ we need to estimate only $\rho(t+h/2)$
- * **Solution:** Estimate it using Taylor expansion
- * $\rho(t+h/2) = \rho(t) + (h/2)\rho'(t) + O(h^2)$
- * **Finally:**
- * $\rho(t+h) = \rho(t) + hF(\rho(t) + (h/2)\rho'(t), t+h/2) + O(h^3)$

Explicit Midpoint Scheme



Explicit Midpoint Scheme

- * Numerical algorithm:

$$\rho_{n+1} = \rho_n + hF(\rho_n + (h/2)F(\rho_n, t_n), t_n + h/2)$$

- * Pros / cons

- Very simple, fast and easy to implement
- Smaller error = $O(h^3)$
- Need to evaluate F two times - more computation

Runge-Kutta Scheme

* Numerical algorithm

$$k_1 = h F(\mathbf{p}(t_0), t_0)$$

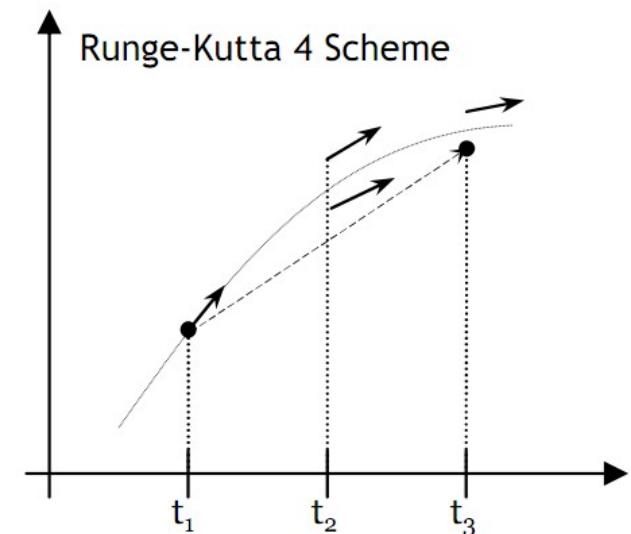
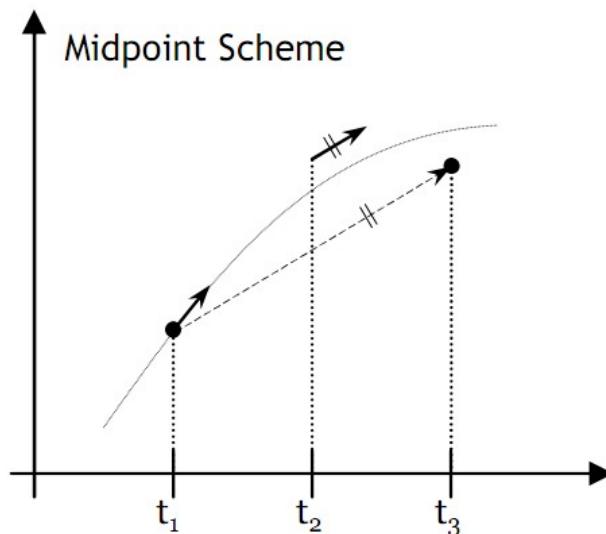
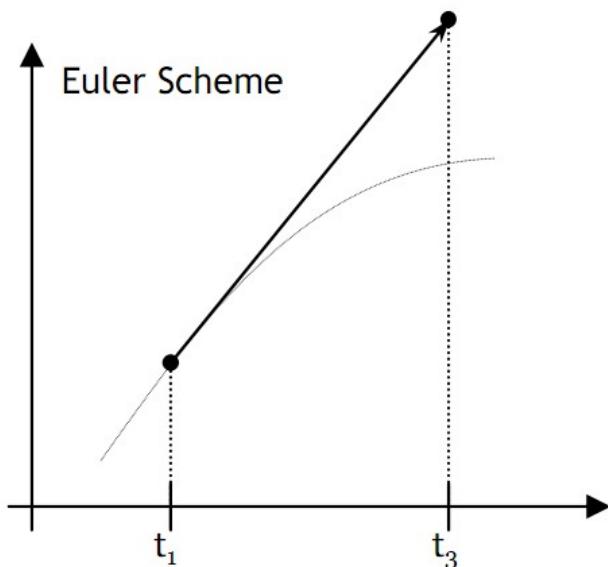
$$k_2 = h F\left(\mathbf{p}\left(t_0\right)+\frac{k_1}{2}, t_0+\frac{h}{2}\right)$$

$$k_3 = h F\left(\mathbf{p}\left(t_0\right)+\frac{k_2}{2}, t_0+\frac{h}{2}\right)$$

$$k_4 = h F(\mathbf{p}(t_0) + k_3, t_0 + h)$$

$$\mathbf{p}(t_0+h) = \mathbf{p}(t_0) + \frac{k_1}{6} + \frac{k_2}{3} + \frac{k_3}{3} + \frac{k_4}{6} + O(h^5)$$

Explicit Integration schemes



Verlet Scheme

- * **Preconditions:** Equations are pure 2-order ODEs.
- * **Idea:** Taylor expand $\rho(t)$ at $\rho(t+h)$ and $\rho(t-h)$ and subtract / add equations

$$\mathbf{p}(t+h) = \mathbf{p}(t) + h \dot{\mathbf{p}}(t) + \frac{h^2}{2} \ddot{\mathbf{p}}(t) + \frac{h^3}{6} \dddot{\mathbf{p}}(t) + O(h^4)$$

$$\mathbf{p}(t-h) = \mathbf{p}(t) - h \dot{\mathbf{p}}(t) + \frac{h^2}{2} \ddot{\mathbf{p}}(t) - \frac{h^3}{6} \dddot{\mathbf{p}}(t) + O(h^4)$$

$$\mathbf{p}(t+h) = 2\mathbf{p}(t) - \cancel{h}\mathbf{p}(t-h) + h^2 \ddot{\mathbf{p}}(t) + O(h^4)$$

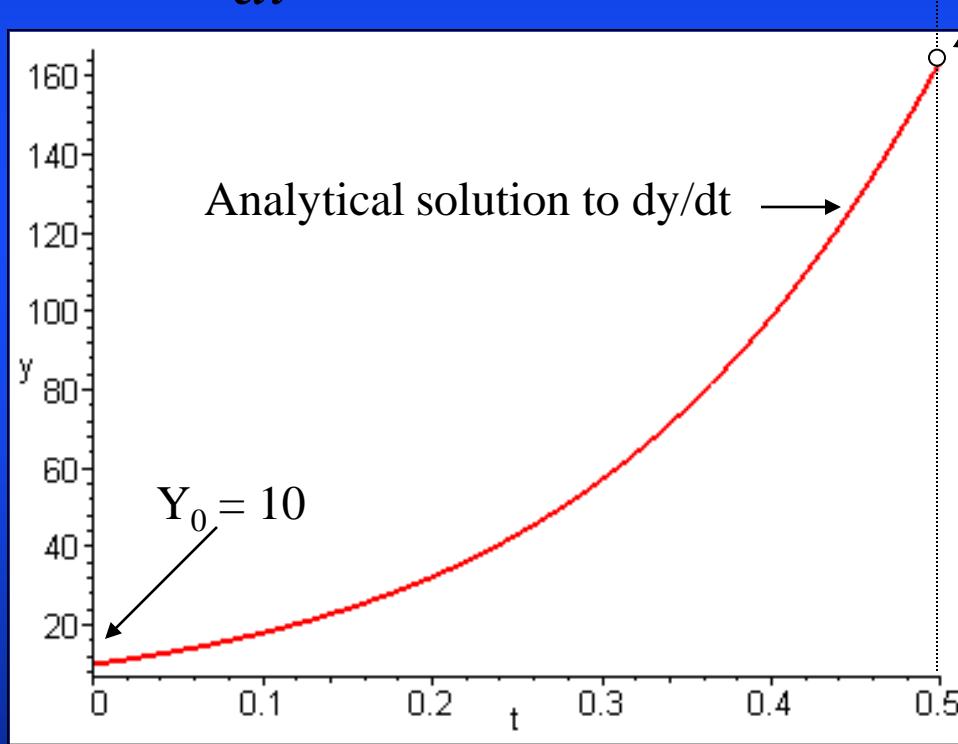
$$\dot{\mathbf{p}}(t+h) = \frac{1}{2h} \mathbf{p}(t+h) - \frac{1}{2h} \mathbf{p}(t-h) + O(h^2)$$

Implicit Euler

- * **Explicit Euler:** $\rho(t+h) = \rho(t) + hF(\rho(t), t) + O(h^2)$
- * **Implicit Euler:** $\rho(t+h) = \rho(t) + hF(\rho(t+h), t+h) + O(h^2)$
- * **Problem:** We need to solve for $\rho(t+h)$
- * **Solution:** Taylor expand $F(\rho, t)$ in ρ
- * $F(\rho + \Delta\rho, t) = F(\rho, t) + \Delta\rho F'(\rho, t) + O(\Delta\rho^2)$
- * **Set:** $\Delta\rho$ as $hF(\rho + \Delta\rho, t)$
- * $F(\rho + \Delta\rho, t) = F(\rho, t) + hF(\rho + \Delta\rho, t)F'(\rho, t) + O(\Delta\rho^2)$
- * $F(\rho + \Delta\rho, t) = (1 - hF'(\rho, t))^{-1}F(\rho, t) + O(\Delta\rho^2)$
- * **Problem:** $F'(\rho, t)$ (Jacobian) must be known
- * More on cloth modeling

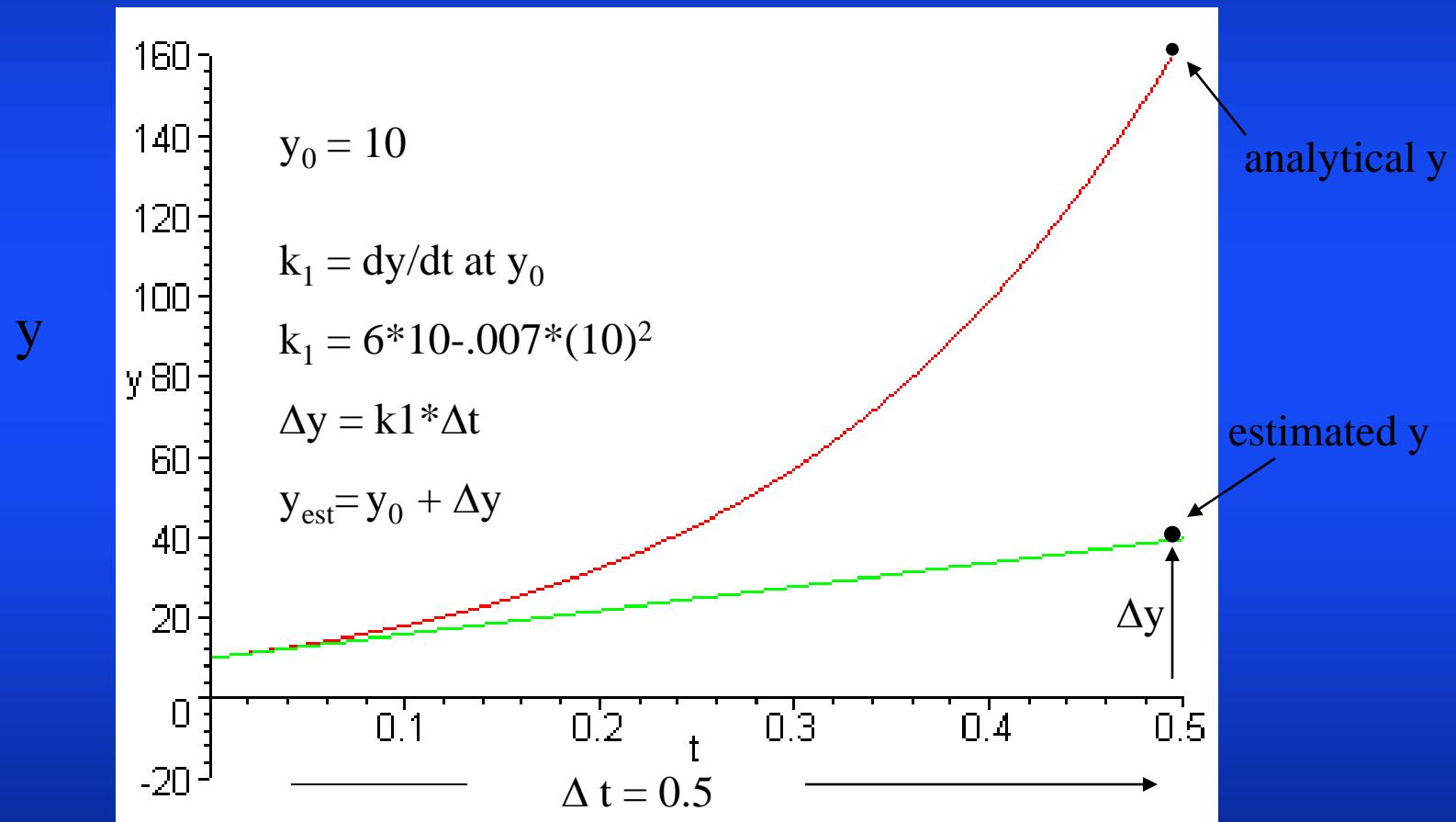
Example

$$\frac{dy}{dt} = 6y - .007y^2$$

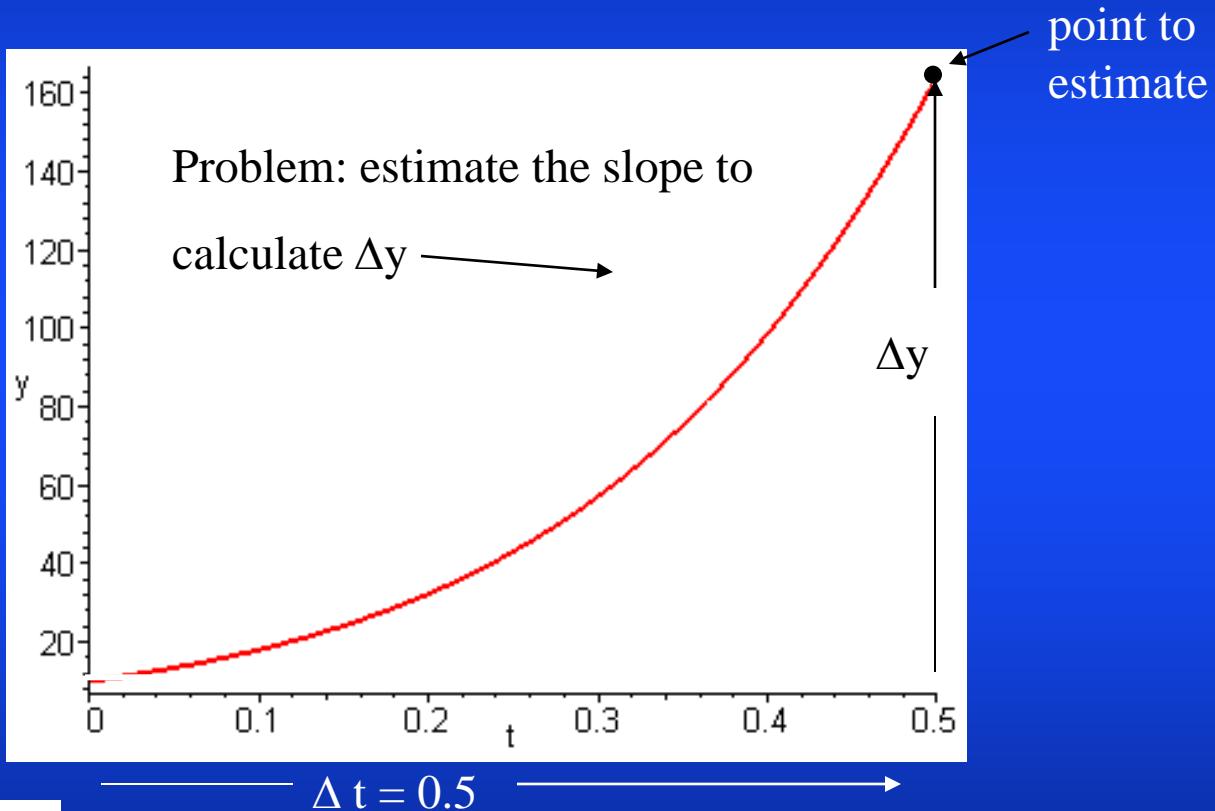


point to
estimate

Euler (pronounced “oiler”)



Runge-Kutta (pronounced Run-gah Kut-tah)



$$\frac{dy}{dt} = 6y - .007y^2$$

Runge-Kutta (4th order)

$f'(t, y) = \text{derivative at } (t, y)$

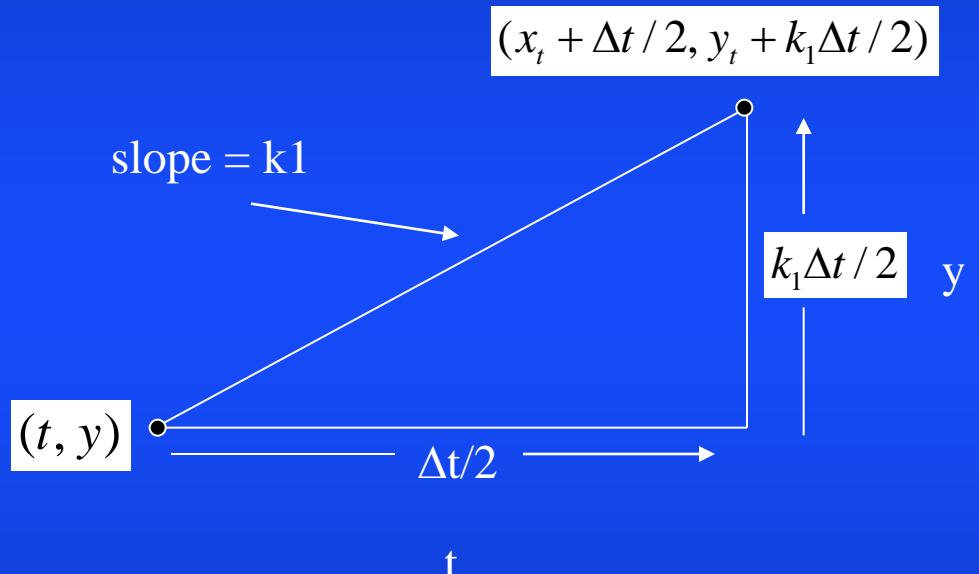
$$k_1 = f'(t, y)$$

$$k_2 = f'(t + \Delta t / 2, y + k_1 \Delta t / 2)$$

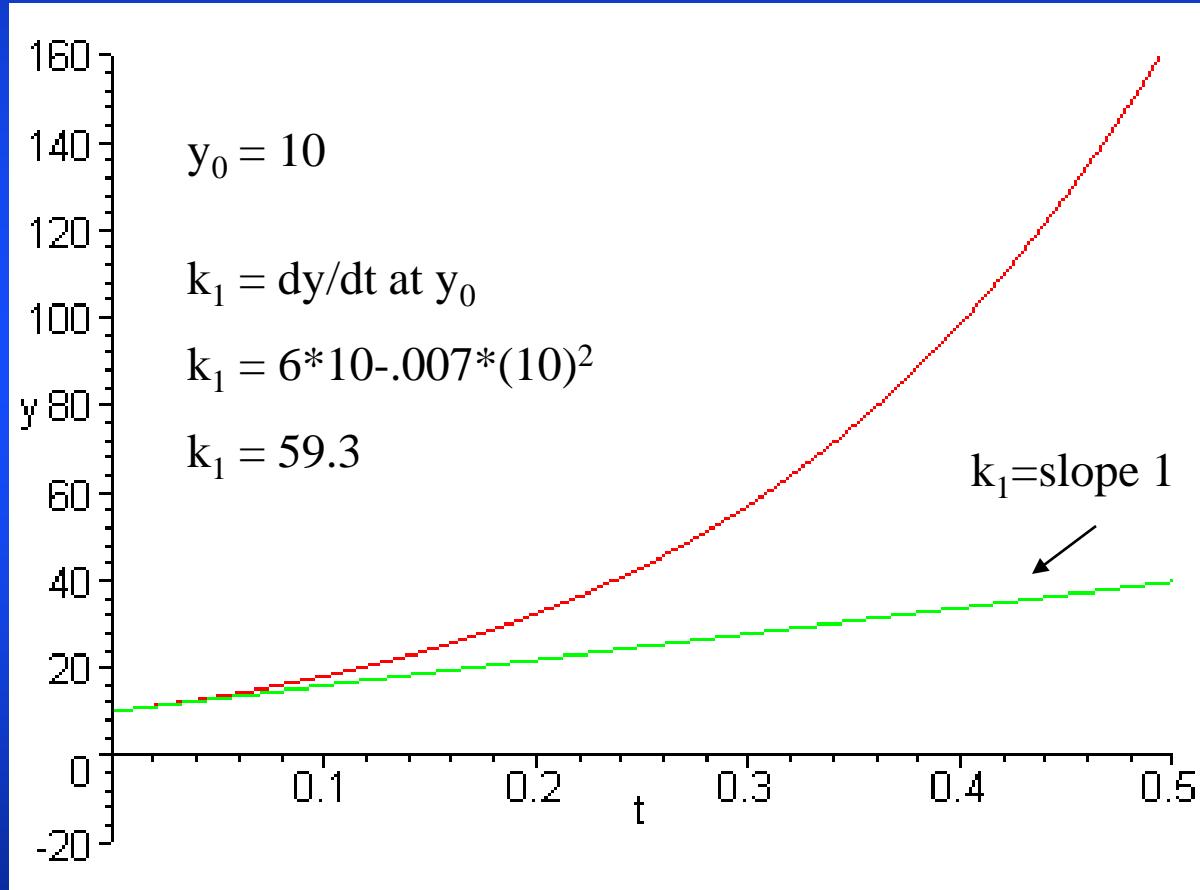
$$k_3 = f'(t + \Delta t / 2, y + k_2 \Delta t / 2)$$

$$k_4 = f'(t + \Delta t, y + k_3 \Delta t)$$

$$y_{t+\Delta} = y_t + \Delta t \frac{1}{6} (k_1 + 2k_2 + 2k_3 + k_4)$$

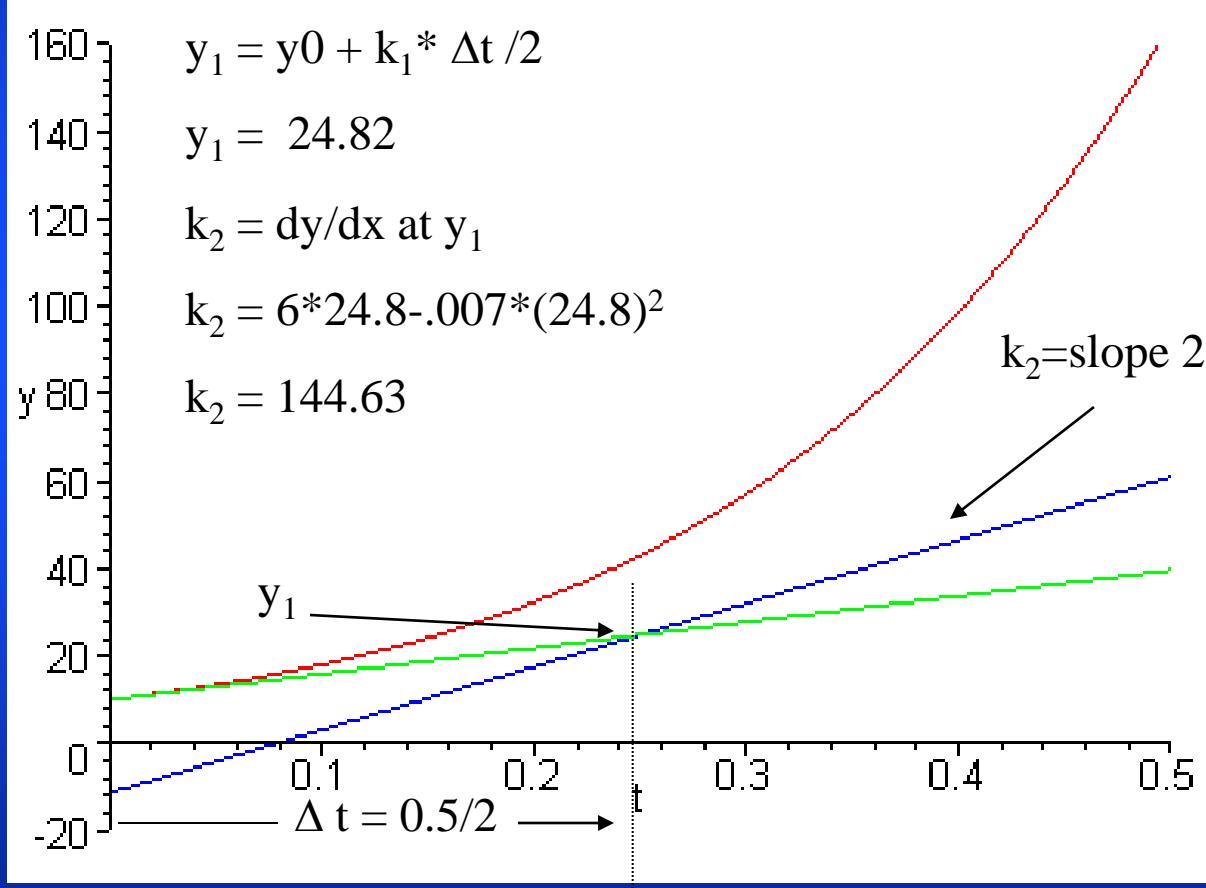


Step 1: Evaluate slope at current value of state variable.



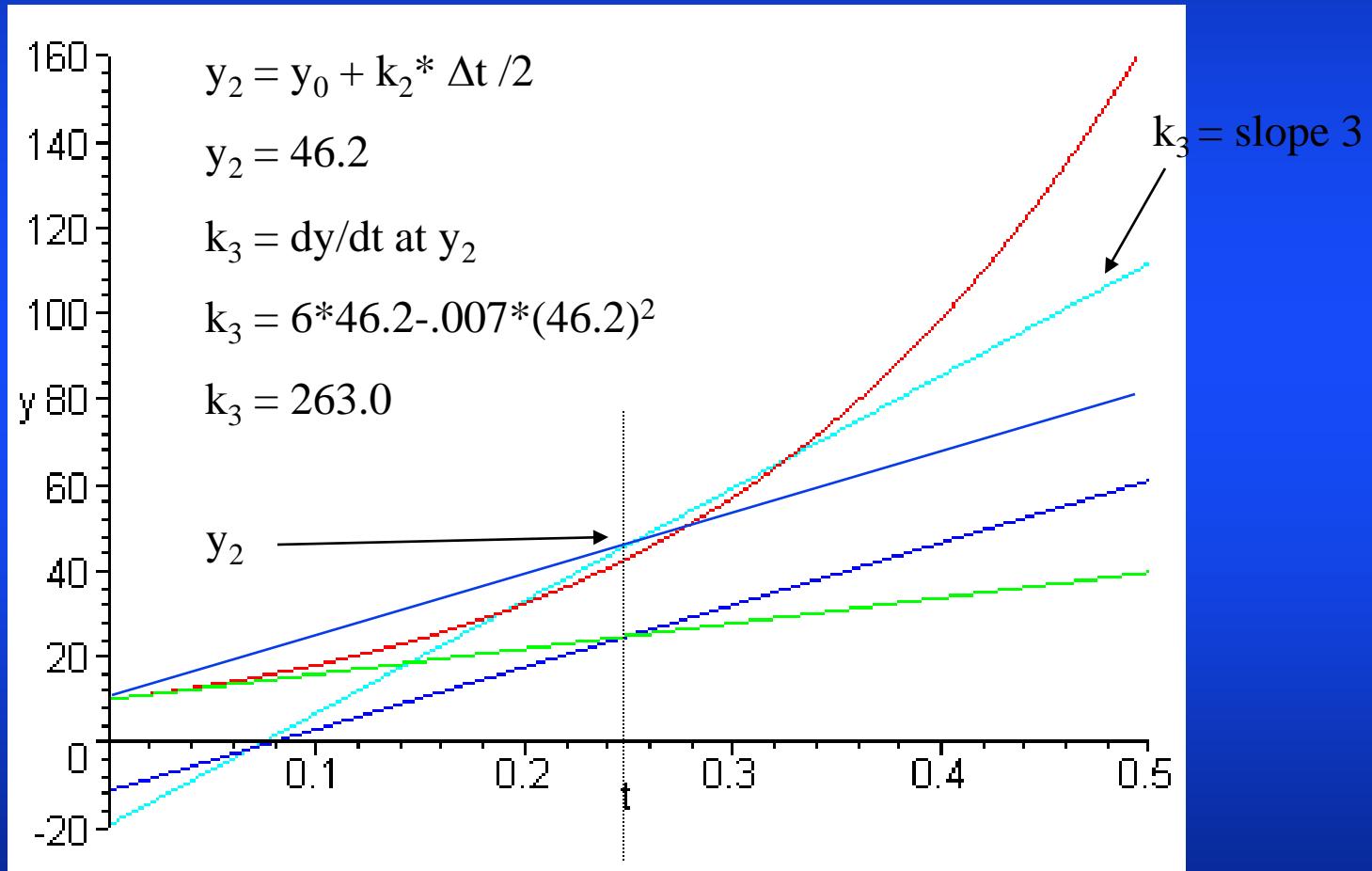
Step 2: Calculate y_1 at $t + \Delta t/2$ using k_1 .

Evaluate slope at y_1 .



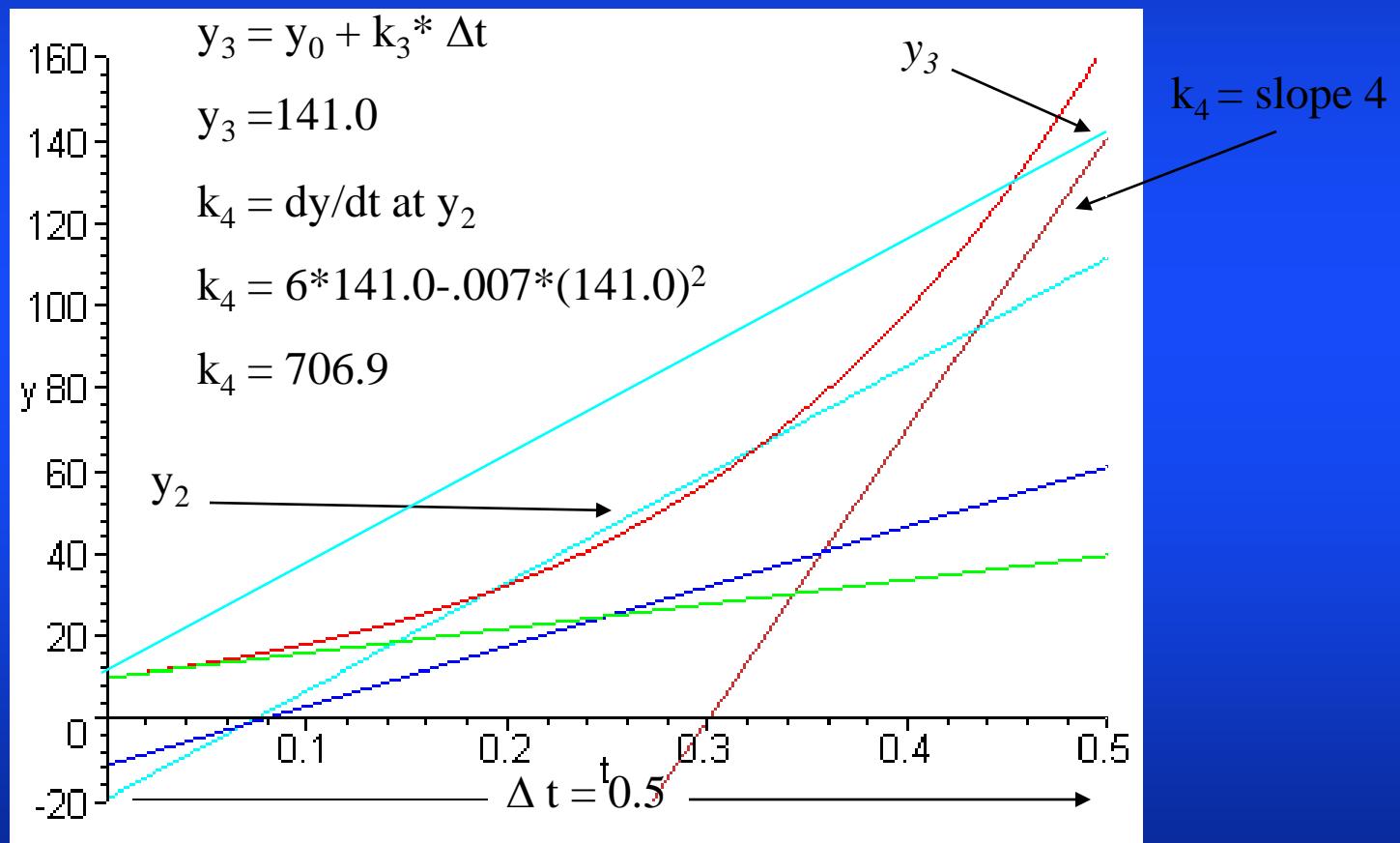
Step 3: Calculate y_2 at $t + \Delta t/2$ using k_2 .

Evaluate slope at y_2 .



Step 4: Calculate y_3 at $t + \Delta t$ using k_3 .

Evaluate slope at y_3 .

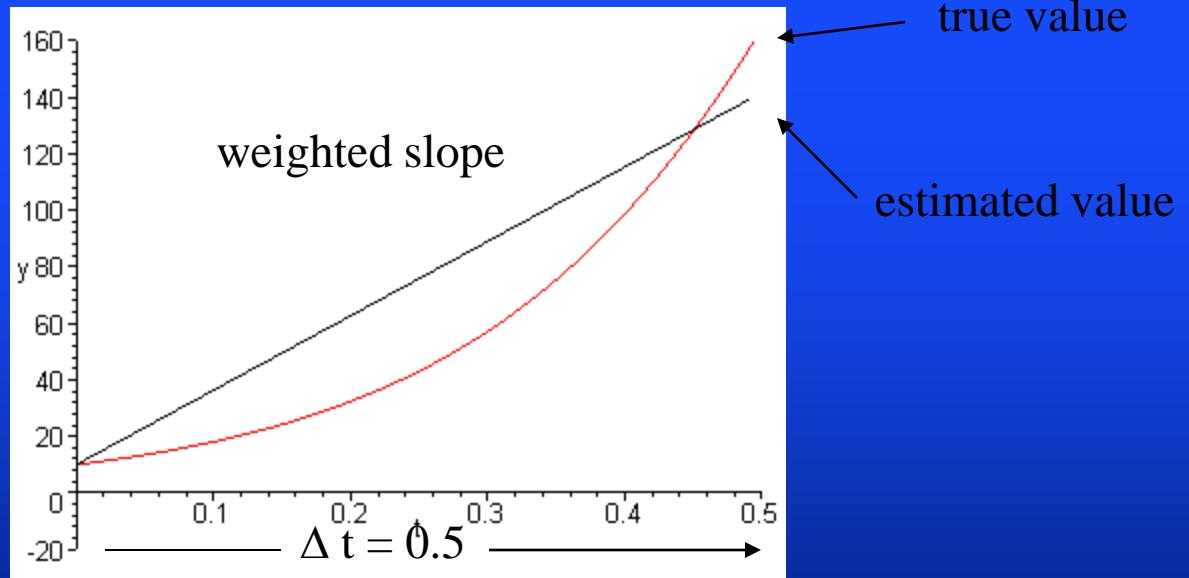


Step 5: Calculate weighted slope.

Use weighted slope to estimate y at $t + \Delta t$

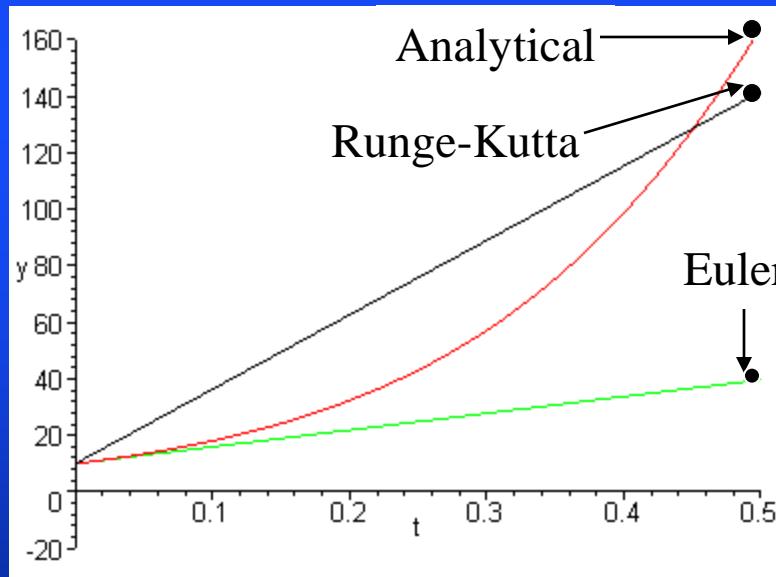
$$\text{weighted slope} = \frac{1}{6}(k_1 + 2k_2 + 2k_3 + k_4)$$

$$Y_{t+\Delta} = Y_t + \Delta t \frac{1}{6}(k_1 + 2k_2 + 2k_3 + k_4)$$

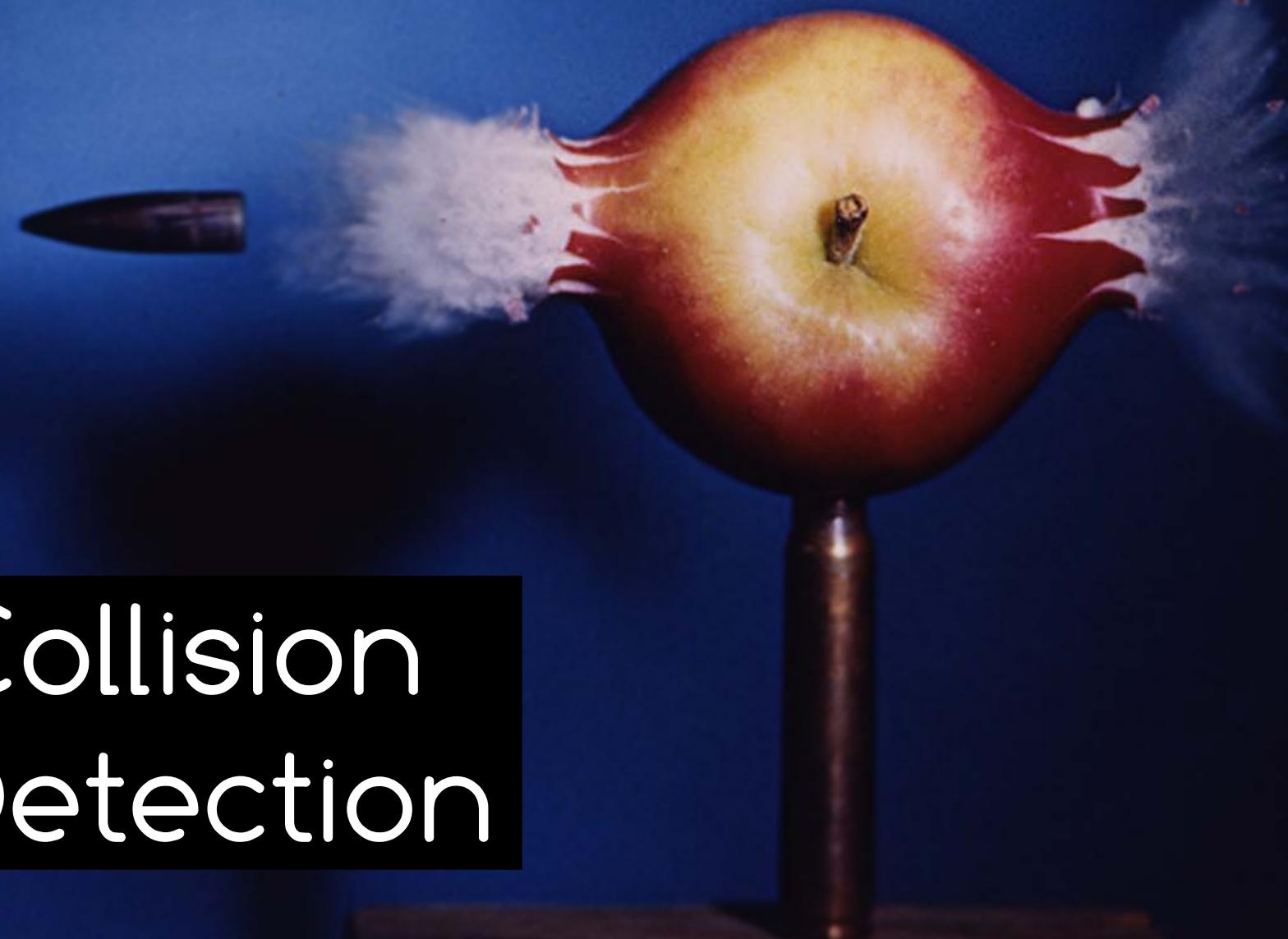


Conclusions

- 4th order Runge-Kutta offers substantial improvement over Eulers.
- Both techniques provide estimates, not “true” values.
- The accuracy of the estimate depends on the size of the step used in the algorithm.



Particle Obstacle

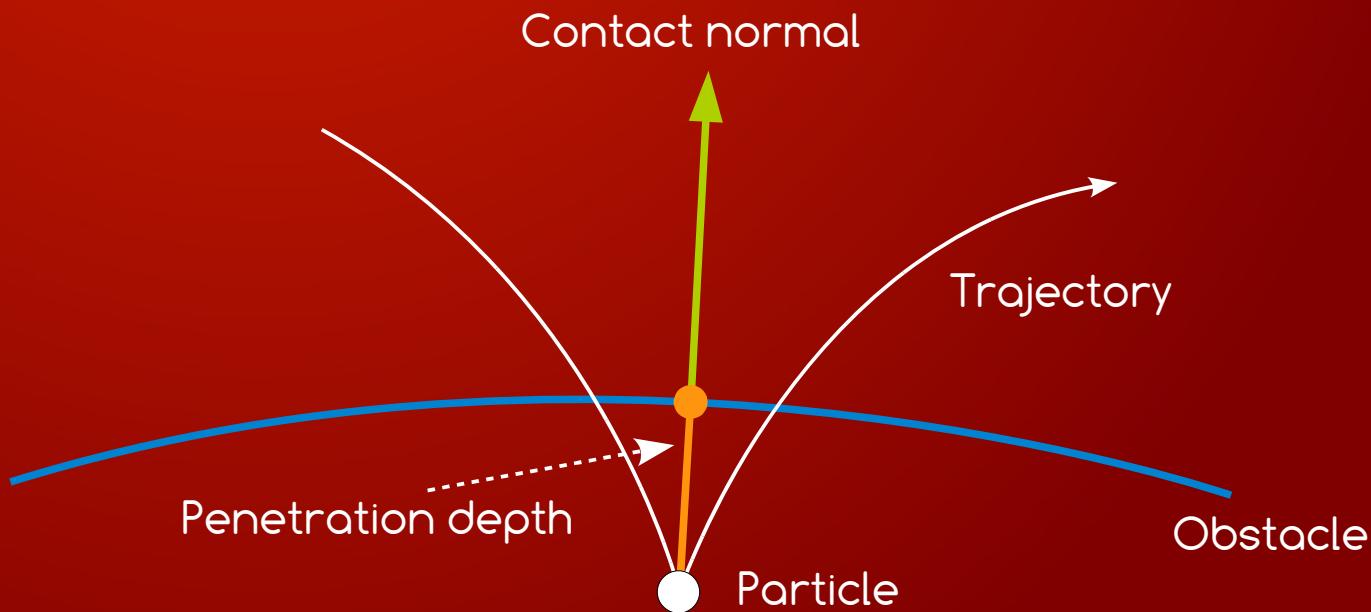


Collision Detection

Collision Scenario

- ★ Particle-obstacle contact info

- **Penetration depth (d)**: minimal distance to separate particle from obstacle
- **Contact normal (n)**: direction vector along which we can get particle out of obstacle (by moving about penetration depth)



Newton's Impact Model

$$u_n(t^+) = -e_n u_n(t^-)$$

- ★ Pre-collision relative normal velocity: $u_n(t^-)$
- ★ Post-collision relative normal velocity: $u_n(t^+)$
- ★ Coefficient of restitution: $0 \leq e_n \leq 1$
- ★ Plastic collisions: $e_n = 0$
- ★ Elastic collisions: $e_n = 1$

Impulse based Collision Resolution

- * Collision Impulse: Time integral of repulsive forces acting on bodies during collision

$$\mathbf{j}(t) = \int_t^{t+h} \mathbf{f}(a) da$$

- * Impulses cause direct change of velocity: $\Delta \mathbf{u} = M^{-1} \mathbf{j}$
- * $\Delta \mathbf{u} = \Delta \mathbf{u}_1 - \Delta \mathbf{u}_2 = M_1^{-1} \mathbf{j} - M_2^{-1} \mathbf{j} = (M_1^{-1} - M_2^{-1}) \mathbf{j} = K \mathbf{j}$
- * $\Delta \mathbf{u}_n = \mathbf{n}^T K \mathbf{j} = \mathbf{n}^T \mathbf{u}(t+h) - \mathbf{n}^T \mathbf{u}(t) = -e_n \mathbf{n}^T \mathbf{u}(t) - \mathbf{n}^T \mathbf{u}(t) = -(1+e_n) \mathbf{n}^T \mathbf{u}(t)$
- * $\mathbf{j} = -(1+e_n) \mathbf{n}^T \mathbf{u}(t) / \mathbf{n}^T (M_1^{-1} - M_2^{-1}) \mathbf{n}$
- * $\mathbf{u}_1 += j \mathbf{n}; \mathbf{u}_2 -= j \mathbf{n};$

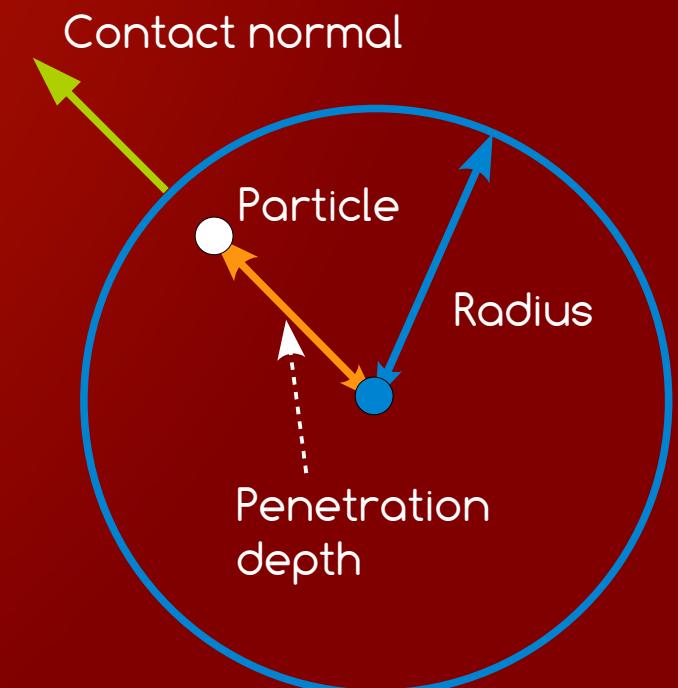
Particle – Sphere Collisions

- * Particle - Sphere Model

- Particle position: $\rho = (x, y, z)$
- Sphere Center: $c = (x, y, z)$
- Sphere Radius: r

- * Penetration depth: $d = |\rho - c| - r$

- * Contact normal: $n = \text{norm}(\rho - c)$



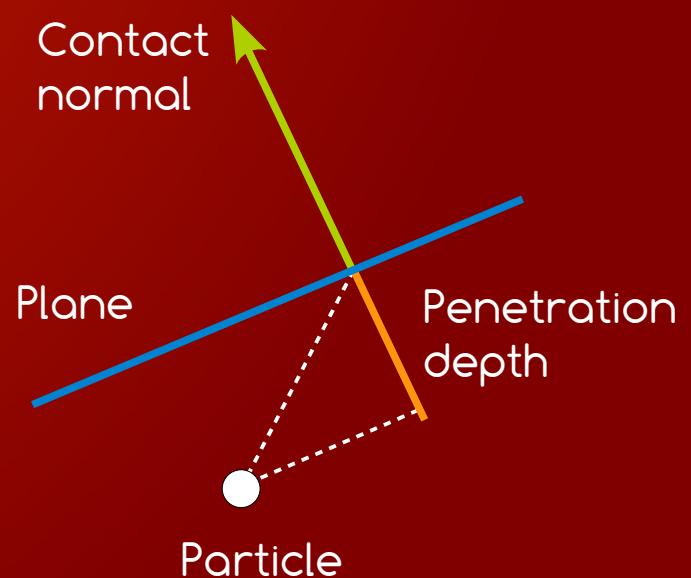
Particle – Plane Collisions

- * Particle - Plane Model

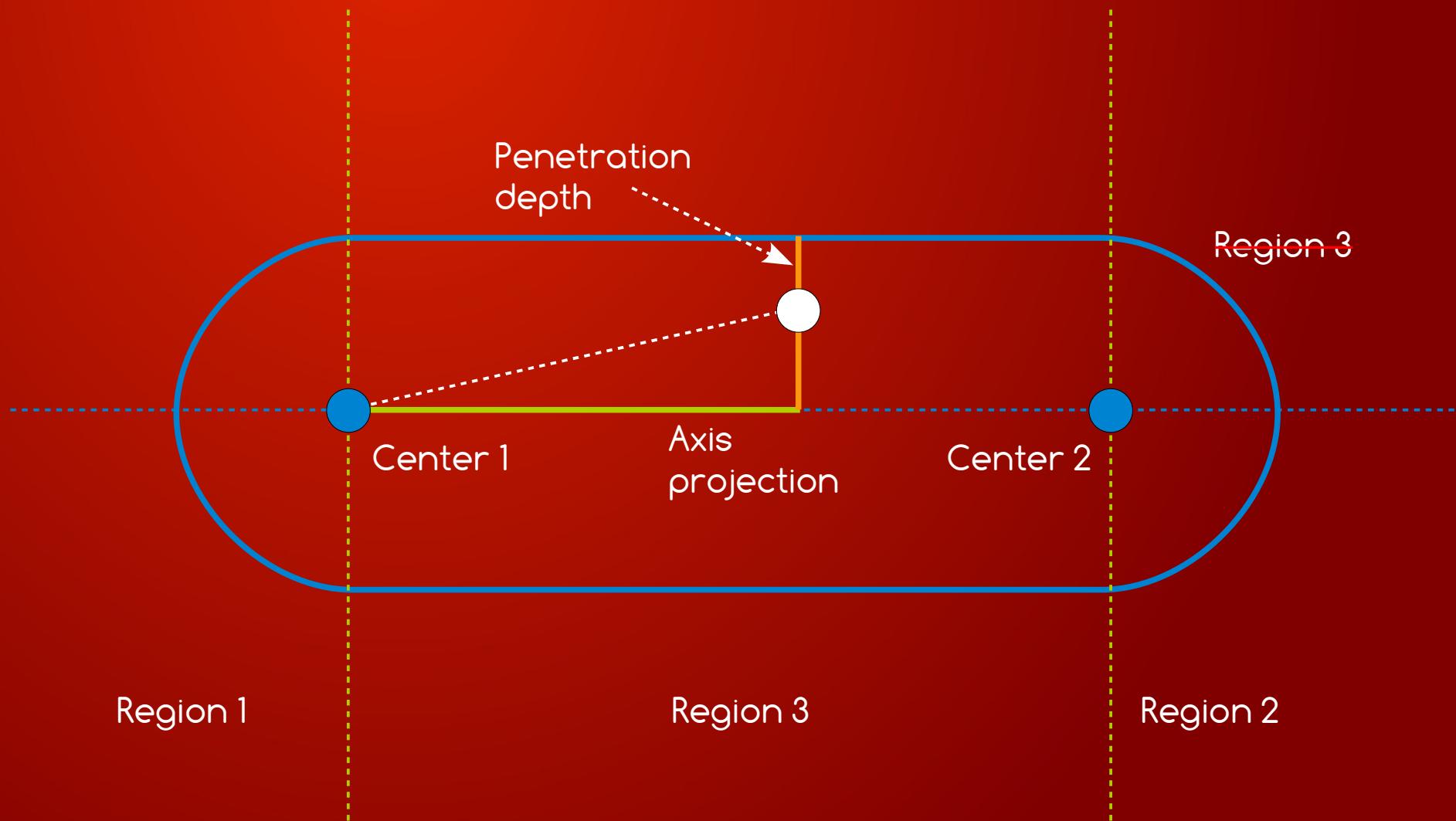
- Particle position: $\rho = (x, y, z)$
- Plane origin: $o = (x, y, z)$
- Plane normal: $m = (x, y, z); |m| == 1$

- * Penetration depth: $d = m^T (\rho - c)$

- * Contact normal: $n = m$



Particle - Capsule Collisions



Particle - Capsule Collisions

- * Particle - Capsule Model

- Particle position: $\rho = (x, y, z)$
 - Center1/2: $c1/2 = (x, y, z)$
 - Radius: r

- * Algorithm:

- Detect Voronoi Region (1,2,3)
 - In region 1/2: Compute sphere penetration
 - In region 3: Compute point-line distance

- * Voronoi detection: Project $(\rho - c1)$ onto $(c2 - c1)$

- $f = (c2 - c1)^\top (\rho - c1)$
 - Region1 ($f < 0$); Region2 ($0 < f < F$); Region3 ($f >= F$)
 - $F = (c2 - c1)^2$

Particle - Capsule Collisions

* Point - Center1 Case

- Penetration depth: $d = |\rho - c_1| - r$
- Contact normal: $n = \text{norm}(\rho - c_1)$

* Point - Center2 Case

- Penetration depth: $d = |\rho - c_2| - r$
- Contact normal: $n = \text{norm}(\rho - c_2)$

* Point – Axis Case

- $u = \text{norm}(c_2 - c_1); v = (\rho - c_1); e = u^T v; f = v^T v; g^2 = f - e^2$
- Penetration depth: $d = r - g$
- Penetration normal: $n = \text{norm}(v - eu)$

A close-up, colorful portrait of a woman's face. Her skin is painted with a variety of bright colors including red, yellow, green, and blue. She has blue eyes and is looking directly at the viewer. Her hair is blonde and visible on the right side of the frame. The overall effect is artistic and vibrant.

Practical design of Particle System

Particle System

* Particle System

- A set of similar particles – e.g. rendered with similar material
- Store in array bag structure

* Particle

- Has lifetime, physical and material properties
- During simulation lifetime is decremented until $< 0 \rightarrow$ dead
- Dead particles are reused for newly emitted particles

* Obstacles: Objects in the scene used as colliders

- Sphere, boxes, planes, capsules...

Emitters

- * Particle emitter: Creates new particles

- Particle emit rate: How many particles are emitted per sec
 - Particle initial values: Particle initialization before emission.
 - Custom (physical) and geometrical properties

- * Common emitters

- Point emitter: Emit particle from point
 - Sphere emitter: Emit particles inside volume (on surface)
 - Box emitter: Emit particles inside generic box
 - Cone emitter: Emit particles inside a cone
 - And many more...

Attractors

- * Particle attractor:

- Is a generic description of forces attracting close particles

- * Common attractors:

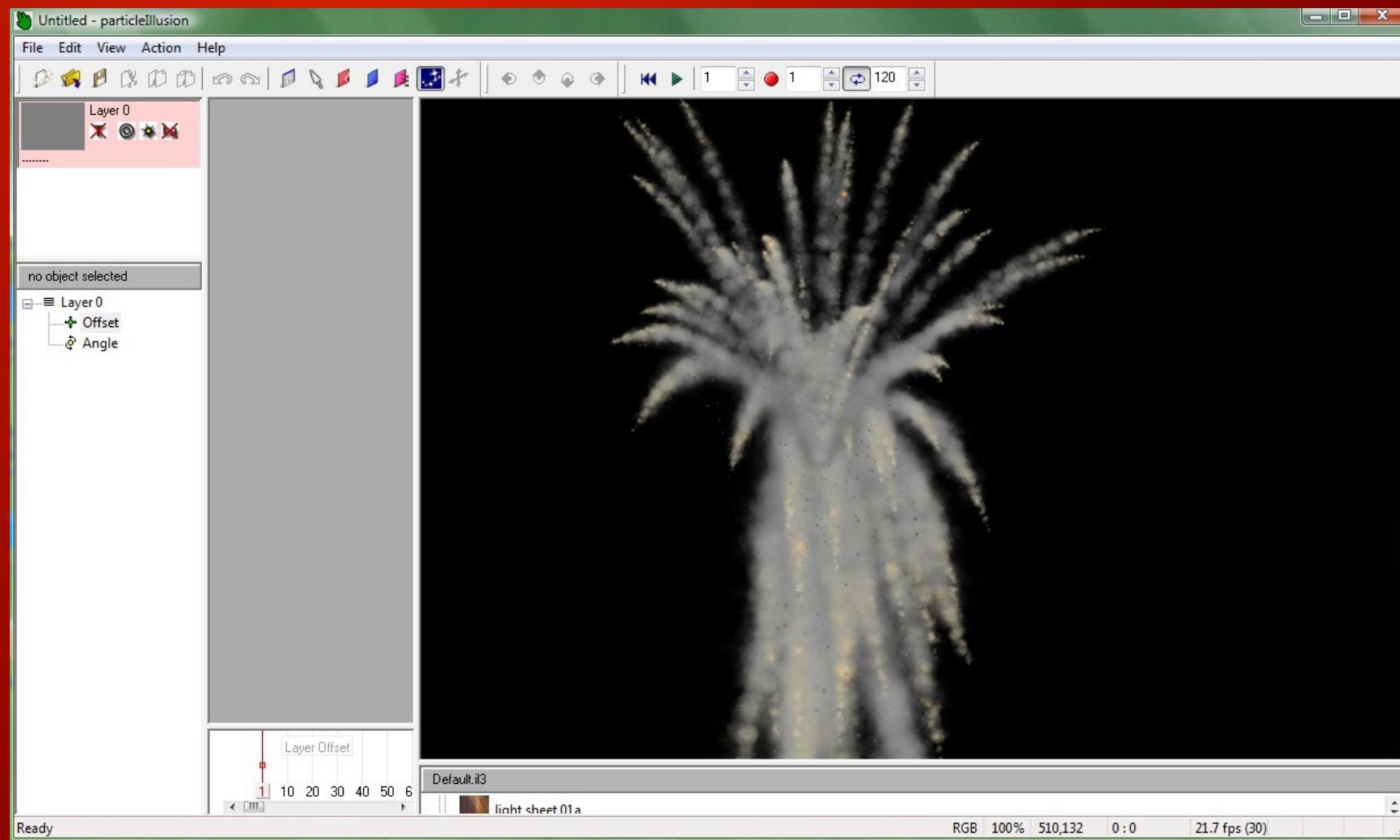
- Linear drag: wind, gravity, user drag
 - Vortex drag: rotational force field
 - Distance magnets: obstacles acts like magnets



Demos / tools / libs

Demos / Tools / Libs

* Particle Illusion (<http://www.wondertouch.com/>)





The End