

# Particle Systems & Time Integration

---



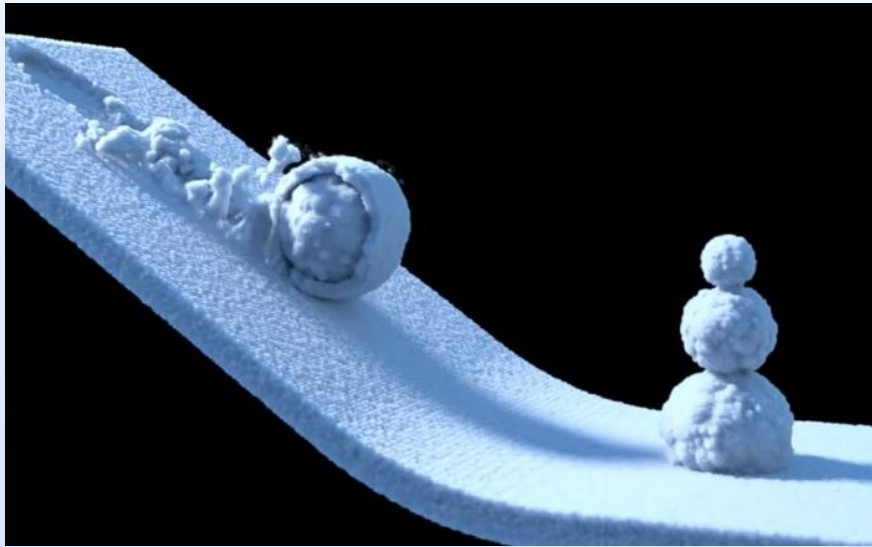
# But first...

---

Office hours Wed @ 3pm, Smith Hall 232

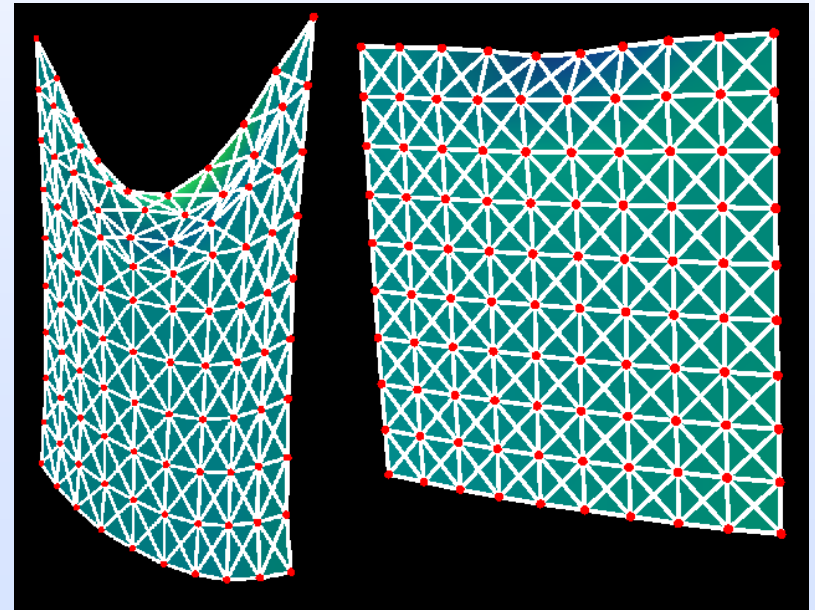
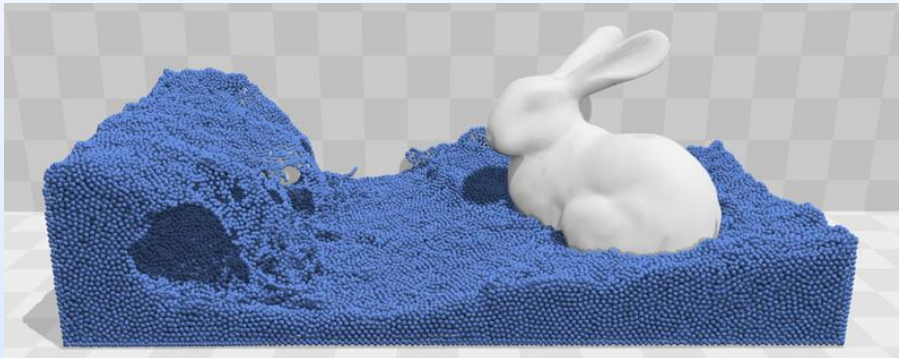
# Particle Systems

- ◆ Some dynamical systems can be naturally described as many small particles



# Particle Systems

- ◆ Others are more difficult to get a handle on, but may still be approximated through particle systems



# Particle Basics

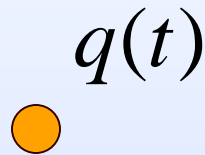
---

- ◆ Each particle has a position
  - perhaps other attributes too: orientation, age, color, velocity, temperature, radius, ...
  - Call the state  $\mathbf{q}$  (aka generalized coordinates)
- ◆ Seed randomly somewhere at start
  - Maybe some created each frame
- ◆ Move (evolve state  $\mathbf{q}$ ) each frame according to some formula
- ◆ Eventually “die” when some condition met

# Particle Basics

---

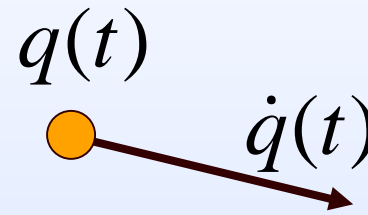
◆ But let's start with a 1D particle...



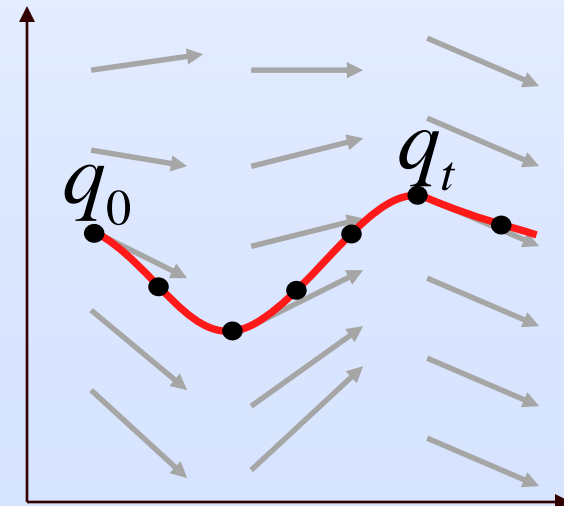
# Motion

- ◆ An ordinary differential equation (ODE) describes how the particle's state changes through time:

$$\dot{\mathbf{q}}(t) = \frac{d\mathbf{q}}{dt}(t) = \mathbf{v}(\mathbf{q}, t)$$



- ODE defines a vector field in state-space. The function we seek,  $\mathbf{q}(t)$ , must be solved for
- In general, analytic solutions hopeless
- Need to solve numerically starting from an initial configuration  $\mathbf{q}_0 (= \mathbf{q}(t=0))$
- Main idea: approximate the derivative and discretize in time



# Forward Euler

---

- ◆ Simplest method:

$$q_{n+1} = q_n + \Delta t v(q_n, t_n)$$

- ◆ First order accurate:
  - Global error accumulated over fixed time interval is  $O(\Delta t)$
  - Thus it converges to the right answer
- ◆ But want error to be small



# Forward Euler Accuracy

---

- ◆ Obvious approach: make  $\Delta t$  small
- ◆ But then need more time steps - expensive
  - also note -  $O(1)$  error made in modeling
  - even if numerical error was 0, still wrong!
  - need to validate against experiments
- ◆ Smaller time steps == better, but if we wanted fastest sim possible, how large could we go?

# Forward Euler Stability

---

## ◆ The Test Equation

$$\frac{dq}{dt} = aq, a \in \mathbb{C}$$

- ◆ Linear ODE, known analytic solution  $q(t) = e^{at}$ 
  - What does it look like?

# The Test Equation

---

- ◆ Gives a rough picture of the stability of a method
  - ‘a’ will in general represent eigenvalues of Jacobian (first-order approximation to nonlinear ODEs)

$$v(q, t) \approx v(q^*, t^*) + \underbrace{\frac{\partial v}{\partial q} \cdot (q - q^*)}_{Aq} + \frac{\partial v}{\partial t} \cdot (t - t^*)$$

- Nonlinear effects can definitely cause problems
- Even with linear problems, what follows assumes constant time steps - varying (but supposedly stable) steps can induce instability
  - see J. P. Wright, “Numerical instability due to varying time steps...”, JCP 1998

# Using the Test Equation

---

- ◆ Forward Euler on test equation is

$$q_{n+1} = q_n + \Delta t a q_n$$

- ◆ Solving gives

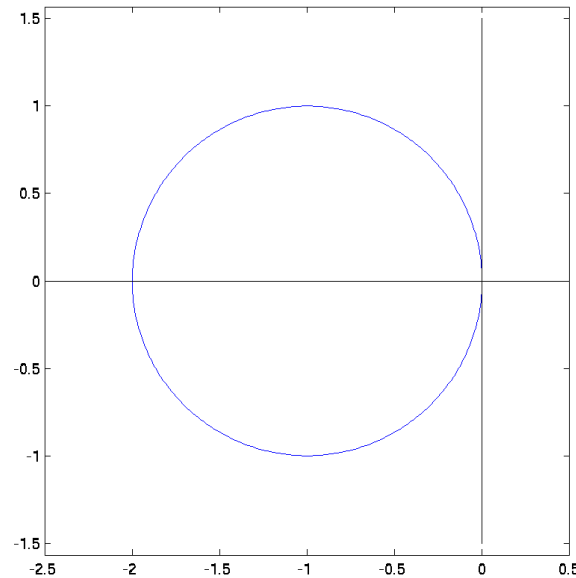
$$q_n = (1 + a\Delta t)^n q_0$$

- ◆ So for stability, need

$$|1 + a\Delta t| < 1$$

# Stability Region

- ◆ Can plot all the values of  $a\Delta t$  on the complex plane where F.E. is stable:



- ◆ Big problem with Forward Euler: not very stable

# Real Eigenvalue

---

- ◆ Say eigenvalue is real (and negative)
  - Corresponds to a damping motion, smoothly coming to a halt
- ◆ Then need:

$$\Delta t < \frac{2}{|a|}$$

- ◆ Is this bad?
  - If 'a' is big, could mean small time steps needed for stability (aka stiff problem)

# Imaginary Eigenvalue

---

- ◆ If eigenvalue is pure imaginary (oscillatory or rotational motion), cannot make  $\Delta t$  small enough
- ◆ Forward Euler unconditionally unstable for these kinds of problems!
- ◆ Need to look at other methods

# Runge-Kutta Methods

---

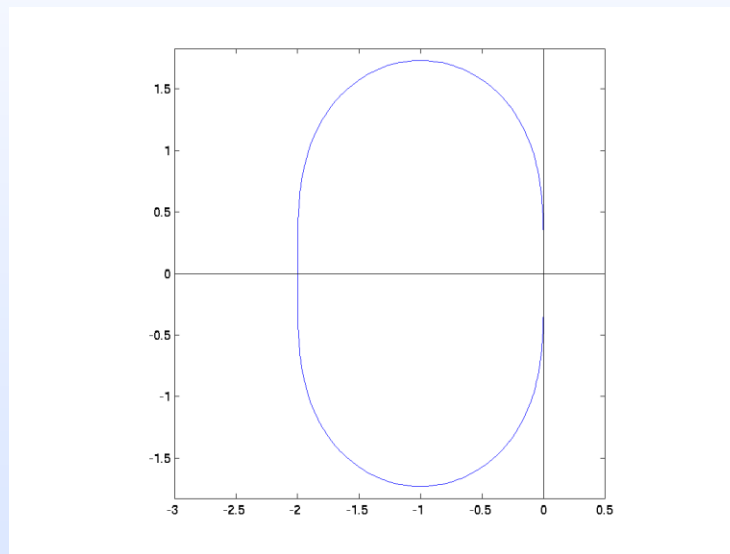
- ◆ Also “explicit”
  - next  $q$  is an explicit function of previous
- ◆ But evaluate  $v$  at a few locations to get a better estimate of next  $q$
- ◆ E.g. midpoint method (RK2)

$$q_{n+\frac{1}{2}} = q_n + \frac{1}{2} \Delta t v(q_n, t_n)$$
$$q_{n+1} = q_n + \Delta t v(q_{n+\frac{1}{2}}, t_{n+\frac{1}{2}})$$



# Midpoint RK2

- ◆ Second order: error is  $O(\Delta t^2)$
- ◆ Larger stability region:



- ◆ But still not stable on imaginary axis

# RK4

---

- ◆ Often most bang for the buck
- ◆ Combination of Forward Euler steps and averaging

$$v_1 = v(q_n, t_n)$$

$$v_2 = v\left(q_n + \frac{1}{2} \Delta t v_1, t_{n+1/2}\right)$$

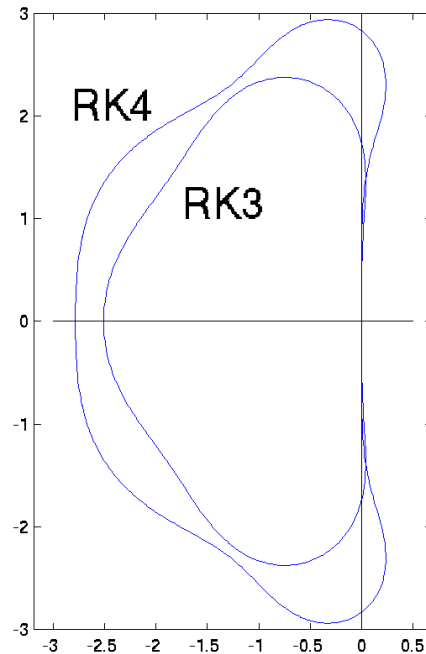
$$v_3 = v\left(q_n + \frac{1}{2} \Delta t v_2, t_{n+1/2}\right)$$

$$v_4 = v(q_n + \Delta t v_3, t_{n+1})$$

$$q_{n+1} = q_n + \Delta t \left( \frac{1}{6} v_1 + \frac{2}{6} v_2 + \frac{2}{6} v_3 + \frac{1}{6} v_4 \right)$$

# Higher Order Runge-Kutta

- ◆ RK3 and up naturally include part of the imaginary axis



# Implicit Methods

---

# Backward Euler

---

- ◆ The simplest implicit method:

$$q_{n+1} = q_n + \Delta t v(q_{n+1}, t_{n+1})$$

- ◆ the next  $x$  **implicitly** defined since it appears in derivative
  - Need to solve equations to figure it out
- ◆ First order accurate
- ◆ We'll come back to it later for a general formulation!

# Backward Euler Stability

---

- ◆ Test equation shows stable when  $|1 - a\Delta t| > 1$
- ◆ This includes everything except a circle in the positive real-part half-plane. *Unconditionally stable* for linear ODEs.
- ◆ It's stable even when the physics is unstable!
- ◆ This is the biggest problem: damps out motion unrealistically

# Trapezoidal Rule

---

- ◆ Can improve by going to second order:

$$q_{n+1} = q_n + \Delta t \left( \frac{1}{2} v(q_n, t_n) + \frac{1}{2} v(q_{n+1}, t_{n+1}) \right)$$

- ◆ This is actually just a half step of F.E., followed by a half step of B.E.
  - F.E. is under-stable, B.E. is over-stable, the combination is **just right**
- ◆ Stability region is the left half of the plane: **exactly** the same as the underlying ODE!
- ◆ Really good for pure rotation (doesn't amplify or damp)

# What to ask for from a numerical integrator?

---

- ◆ No one “best” integrator – pick the right tool for the job!
- ◆ Many different integrators because there are many notions of “good”
  - Convergence/accuracy
  - Stability
  - Computational Efficiency
  - Monotonicity
  - ...



# Monotonicity

---

- ◆ Test equation with real, negative  $\lambda$ 
  - True solution is  $x(t)=x_0e^{\lambda t}$ , which smoothly decays to zero, doesn't change sign (**monotone**)
- ◆ Forward Euler at stability limit:
  - $x=x_0, -x_0, x_0, -x_0, \dots$
- ◆ Not smooth, oscillating sign, no good!
- ◆ So monotonicity limit stricter than stability

# Monotonicity

---

- ◆ Backward Euler is unconditionally monotone
  - No problems with oscillation, just too much damping
- ◆ Trapezoidal Rule can suffer though, because of that half-step of F.E.
  - could get ugly oscillation instead of smooth damping
  - for some nonlinear problems, possible to hit instability

# Summary 1

---

- ◆ Need to move particles in velocity field according to underlying ODE
- ◆ Forward Euler
  - Simple, first choice unless problem has oscillations/rotations
- ◆ Runge-Kutta is better, but requires more evaluations
  - RK4 general purpose workhorse

# Summary 2

---

- ◆ If stability limit is a problem, look at implicit methods
  - e.g. explicit time steps are way too small
- ◆ Trapezoidal Rule
  - If monotonicity isn't a problem
- ◆ Backward Euler
  - Almost always works, but may over-damp!