

Addenda to last class

- K. Sims, “Particle animation...”, SIGGRAPH’90
 - Ignore the parallel computing stuff
- There was an inconsistency in assignment #1 (y-axis vs. z-axis)
 - Updated PDF on the web
 - Vertical is now z-axis, horizontal is x-y plane
 - Welcome to a continual problem of axis labeling... (we’re not even looking at right-handed vs. left-handed)

Time Stepping Algorithm

- Set done = false
- While not done
 - Find good Δt
 - If $t + \Delta t \geq t_{\text{frame}}$
 - Set $\Delta t = t_{\text{frame}} - t$
 - Set done = true
 - Else if $t + 1.5\Delta t \geq t_{\text{frame}}$
 - Set $\Delta t = 0.5(t_{\text{frame}} - t)$
 - ...process time step...
 - Set $t = t + \Delta t$
- Write out frame data, continue to next frame

Time Stepping

- Sometimes can pick constant Δt
 - One frame, or 1/8th of a frame, or ...
- Often need to allow for variable Δt
 - Changing stability limit due to changing Jacobian
 - Difficulty in Newton converging
 - ...
- But need to land at the exact frame time
 - So clamp Δt so you can’t overshoot the frame
- Some algorithms behave oddly if time step changes dramatically...
 - Be careful that last time step isn’t much smaller

Another Word of Caution

- Even for linear problems, stability analysis still not bulletproof
 - Assumes constant time step
 - If time step varies, even under official stability limit, can actually go unstable!
 - See J. P. Wright, “Numerical instability due to varying time steps...”, JCP 1998
 - Safety margin really is a good idea!

1st order vs. 2nd order

- If particle state is just position (and colour, size, ...) then 1st order motion
 - No inertia
 - Good for very light particles that stay suspended in air: smoke, dust, ...
 - Good for some special cases (hacks)
- But most often, want inertia
 - State includes velocity, specify acceleration
 - Can then do parabolic arcs due to gravity, etc.

What's New?

- If $\mathbf{x}=(x,v)$ this is just a special form of $d\mathbf{x}/dt=\mathbf{v}(\mathbf{x},t)$
- But since we know the special structure, can we take advantage of it? (i.e. better time integration algorithms)
 - More stability for less cost?
 - Handle position and velocity differently to better control error?

Second Order Particle Motion

- This puts us in the realm of standard Newtonian physics
 - $F=ma$
- Alternatively put:
 - $dx/dt=v$
 - $dv/dt=F(x,v,t)/m$ (i.e. $a(x,v,t)$)

Linear Analysis

- Approximate acceleration:

$$a(x,v) \approx a_0 + \frac{\partial a}{\partial x} x + \frac{\partial a}{\partial v} v$$

- Split up analysis into different cases
- Begin with first term dominating: constant acceleration
 - e.g. gravity is most important

Constant Acceleration

- Solution is $v(t) = v_0 + a_0 t$
 $x(t) = x_0 + v_0 t + \frac{1}{2} a_0 t^2$
- No problem to get $v(t)$ right:
just need 1st order accuracy
- But $x(t)$ demands 2nd order accuracy
- So we can look at mixed methods:
 - 1st order in v
 - 2nd order in x

More Approximations...

- Typically K and D are symmetric semi-definite (there are good reasons)
 - What does this mean about their eigenvalues?
- Often, D is a linear combination of K and I (“Rayleigh damping”), or at least close to it
 - Then K and D have the same eigenvectors (but different eigenvalues)
 - Then the eigenvectors of the Jacobian are of the form $(u, \alpha u)^T$
 - [work out what α is in terms of λ_K and λ_D]

Linear Acceleration

- Dependence on x and v dominates:
 $a(x,v) = -Kx - Dv$
- Do the analysis from last class:

$$\frac{d}{dt} \begin{pmatrix} x \\ v \end{pmatrix} = \begin{pmatrix} 0 & I \\ -K & -D \end{pmatrix} \begin{pmatrix} x \\ v \end{pmatrix}$$

- Eigenvalues of this matrix?

Simplification

- α is the eigenvalue of the Jacobian, and
$$\alpha = -\frac{1}{2} \lambda_D \pm \sqrt{\left(\frac{1}{2} \lambda_D\right)^2 - \lambda_K}$$
- Same as eigenvalues of $\begin{pmatrix} 0 & 1 \\ -\lambda_K & -\lambda_D \end{pmatrix}$
- Can replace K and D (matrices) with corresponding eigenvalues (scalars)
 - Just have to analyze 2x2 system

Two Regimes

- Still messy! Simplify further
- If D dominates (e.g. air drag, damping)

$$\alpha \approx \{-\lambda_D, 0\}$$

- Exponential decay and constant
- If K dominates (e.g. spring force)

$$\alpha \approx \pm i\sqrt{\lambda_K}$$

Three Test Equations

- Constant acceleration (e.g. gravity)
 - $a(x,v,t)=g$
 - Want exact (2nd order accurate) position
- Position dependence (e.g. spring force)
 - $a(x,v,t)=-Kx$
 - Want stability but low damping
 - Look at imaginary axis
- Velocity dependence (e.g. damping)
 - $a(x,v,t)=-Dv$
 - Want stability, smooth decay
 - Look at negative real axis

Explicit methods from before

- Forward Euler
 - Constant acceleration: bad (1st order)
 - Position dependence: very bad (unstable)
 - Velocity dependence: ok (conditionally monotone/stable)
- RK3 and RK4
 - Constant acceleration: great (high order)
 - Position dependence: ok (conditionally stable, but damps out oscillation)
 - Velocity dependence: ok (conditionally monotone/stable)

Implicit methods from before

- Backward Euler
 - Constant acceleration: bad (1st order)
 - Position dependence: ok (stable, but damps)
 - Velocity dependence: good (monotone, 1st order)
- Trapezoidal Rule
 - Constant acceleration: great (2nd order)
 - Position dependence: great (stable, no damping)
 - Velocity dependence: good (stable but only conditionally monotone --- though maybe fixable)

New methods!

- This is again a big subject
- Again look at explicit methods, implicit methods
- Also can treat position and velocity dependence differently: mixed implicit-explicit methods

Symplectic Euler

- Like Forward Euler, but updated velocity used for position

$$v_{n+1} = v_n + \Delta t a(x_n, v_n)$$

$$x_{n+1} = x_n + \Delta t v_{n+1}$$

- Some people flip the steps (= relabel v_n)
- (Symplectic means certain qualities are preserved in discretization; useful in science, not necessarily in graphics)
- [work out test cases]

Symplectic Euler performance

- Constant acceleration: bad
 - Velocity right, position off by $O(\Delta t)$
- Position dependence: good
 - Stability limit $\Delta t < \frac{2}{\sqrt{K}}$
 - No damping!
- Velocity dependence: ok
 - Monotone limit $\Delta t < 1/D$
 - Stability limit $\Delta t < 2/D$

Tweaking Symplectic Euler

- [sketch algorithms]
- Stagger the velocity to improve x
- Start off with $v_{1/2} = v_0 + \frac{1}{2}\Delta t a(x_0, v_0)$
- Then proceed with
$$v_{n+1/2} = v_{n-1/2} + \frac{1}{2}(t_{n+1} - t_{n-1})a(x_n, v_{n-1/2})$$
$$x_{n+1} = x_n + \Delta t v_{n+1/2}$$
- Finish off with $v_N = v_{N-1/2} + \frac{1}{2}\Delta t a(x_N, v_{N-1/2})$

Staggered Symplectic Euler

- Constant acceleration: great!
 - Position is exact now
- Other cases not effected
 - Was that magic? Main part of algorithm unchanged (apart from relabeling) yet now it's more accurate!
- Only downside to staggering
 - At intermediate times, position and velocity not known together
 - May need to think a bit more about collisions and other interactions with outside algorithms...

An Implicit Compromise

- Backward Euler is nice due to unconditional monotonicity
 - Although only 1st order accurate, it has the right characteristics for damping
- Trapezoidal Rule is great for everything except damping with large time steps
 - 2nd order accurate, doesn't damp pure oscillation/rotation
- How can we combine the two?

A common explicit method

- May see this one pop up:

$$v_{n+1} = v_n + \Delta t a(x_n, v_n)$$

$$x_{n+1} = x_n + \Delta t \left(\frac{1}{2} v_n + \frac{1}{2} v_{n+1} \right) = x_n + \Delta t v_n + \frac{1}{2} \Delta t^2 a_n$$

- Constant acceleration: great
- Velocity dependence: ok
 - Conditionally stable/monotone
- Position dependence: **BAD**
 - Unconditionally unstable!

Implicit Compromise

- Use Backward Euler for velocity dependence, Trapezoidal Rule for the rest:

$$x_{n+1} = x_n + \Delta t \left(\frac{1}{2} v_n + \frac{1}{2} v_{n+1} \right)$$

$$v_{n+1} = v_n + \Delta t a \left(\frac{1}{2} x_n + \frac{1}{2} x_{n+1}, v_{n+1}, t_{n+\frac{1}{2}} \right)$$

- Constant acceleration: great (2nd order)
- Position dependence: great (2nd order, no damping)
- Velocity dependence: good (unconditionally monotone, but only 1st order accurate)

Time scales

- [work out]
- For position dependence, characteristic time interval is
$$\Delta t = O\left(\frac{1}{\sqrt{K}}\right)$$
- For velocity dependence, characteristic time interval is
$$\Delta t = O\left(\frac{1}{D}\right)$$
- Note: matches symplectic Euler stability limits

Newmark Methods

- A general class of methods
$$x_{n+1} = x_n + \Delta t v_n + \frac{1}{2} \Delta t^2 [(1 - 2\beta)a_n + 2\beta a_{n+1}]$$
$$v_{n+1} = v_n + \Delta t [(1 - \gamma)a_n + \gamma a_{n+1}]$$
- Includes Trapezoidal Rule for example ($\beta=1/4, \gamma=1/2$)
- The other major member of the family is Central Differencing ($\beta=0, \gamma=1/2$)
 - This is mixed Implicit/Explicit

Mixed Implicit/Explicit

- For some problems, that square root can mean velocity limit **much** stricter
- Or, we know we want to properly resolve the position-based oscillations, but don't care about damping
- Go explicit on position, implicit on velocity
 - Also cuts the number of equations to solve in half
 - Often, $a(x,v)$ is linear in v , though nonlinear in x ; this way we avoid Newton iteration

Central Differencing

- Rewrite it with intermediate velocity:
$$v_{n+1/2} = v_n + \frac{1}{2} \Delta t a(x_n, v_n)$$
$$x_{n+1} = x_n + \Delta t v_{n+1/2}$$
$$v_{n+1} = v_{n+1/2} + \frac{1}{2} \Delta t a(x_{n+1}, v_{n+1})$$
- Looks like a hybrid of:
 - Midpoint (for position), and
 - Trapezoidal Rule (for velocity - split into Forward and Backward Euler half steps)

Central: Performance

- Constant acceleration: great
 - 2nd order accurate
- Position dependence: good
 - Conditionally stable, no damping
- Velocity dependence: good
 - Stable, but only conditionally monotone
- Can we change the Trapezoidal Rule to Backward Euler and get unconditional monotonicity?

Time Integration Summary

- Depends a lot on the problem
 - What's important: gravity, position, velocity?
- Explicit methods from last class are bad
- Symplectic Euler is a great fully explicit method (particularly with staggering)
 - Switch to implicit velocity step for more stability
- Implicit Compromise method
 - Fully stable, nice behaviour
- Central Differencing and Trapezoidal Rule
 - More accurate velocity, but may have monotonicity issues for strong damping...

Staggered Implicit/Explicit

- Like the staggered Symplectic Euler, but use B.E. in velocity instead of F.E.:

$$v_{n+1/2} = v_{n-1/2} + \frac{1}{2}(t_{n+1} - t_{n-1})a(x_n, v_{n+1/2})$$
$$x_{n+1} = x_n + \Delta t v_{n+1/2}$$

- Constant acceleration: great
- Position dependence: good (conditionally stable, no damping)
- Velocity dependence: good (unconditionally monotone, but 1st order)

Example Forces

- Gravity: $F_{\text{gravity}} = mg$ ($a=g$)
- If you want to do orbits

$$F_{\text{gravity}} = -GmM_0 \frac{x - x_0}{|x - x_0|^3}$$

- Note x_0 could be a fixed point (e.g. the Sun) or another particle
 - But make sure to add the opposite and equal force to the other particle if so!

Spring Forces

- Springs: $F_{\text{spring}} = -K(x - x_0)$
 - x_0 is the attachment point of the spring
 - Could be a fixed point in the scene
 - ...or somewhere on a character's body
 - ...or the mouse cursor
 - ...or another particle (but please add equal and opposite force!)

Nonzero Rest Length Spring

- Need to measure the “strain”:
the fraction the spring has stretched
from its rest length L

$$F_{\text{spring}} = -K \left(\frac{|x - x_0|}{L} - 1 \right) \frac{x - x_0}{|x - x_0|}$$

Spring Damping

- Simple damping: $F_{\text{damp}} = -D(v - v_0)$
 - But this damps rotation too!
- Better spring damping:
 $F_{\text{damp}} = -D(v - v_0) \cdot u \ u$
 - Here u is $(x - x_0) / |x - x_0|$, the spring direction
- [work out 1d case]
- Critical damping $D = 2\sqrt{mK}$

Drag Forces

- Air drag: $F_{\text{drag}} = -Dv$
 - If there's a wind blowing with velocity v_w then
 $F_{\text{drag}} = -D(v - v_w)$
- D should be proportional to cross-section
exposed to wind
 - Think sheets of paper, leaves...
- Depends in a difficult way on shape too
- How do you come up with a good wind
velocity?

Wind

- Later in the course: actually directly simulate the wind
- For now: fake it
 - Random “turbulence”
 - Superposition of basic flow elements
 - Constant wind, vortices, ...
 - Key ingredient is incompressibility

Incompressibility

- Air is basically incompressible
 - Acoustic waves are so small as to be ignored usually
 - Large shock waves only around supersonic objects
- The volume of air going into a region of space equals the volume leaving it
- [derive divergence condition]