

Notes

- ◆ Most of assignment 1 hasn't been covered in class yet, but after today you should be able to do a lot of it
- ◆ Forgot to include instructions about view_obj:
 - To navigate, hold down shift and click/drag with left, right, or middle mouse buttons (same navigation model as Maya)

Trapezoidal Rule Again

- ◆ The method:

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

- ◆ Let's work out stability:

$$\begin{aligned} x_{n+1} &= x_n + \Delta t \left(\frac{1}{2} \lambda x_n + \frac{1}{2} \lambda x_{n+1} \right) \\ \left(1 - \frac{1}{2} \lambda \Delta t \right) x_{n+1} &= \left(1 + \frac{1}{2} \lambda \Delta t \right) x_n \\ x_{n+1} &= \frac{1 + \frac{1}{2} \lambda \Delta t}{1 - \frac{1}{2} \lambda \Delta t} x_n \end{aligned}$$

Monotonicity

- ◆ Test equation with real, negative λ
 - True solution is $x(t) = x_0 e^{\lambda t}$, 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: garbage!
- ◆ So monotonicity limit stricter than stability in this case
- ◆ RK3 has the same problem
 - But the even order RK are fine for linear problems
 - TVD-RK3 designed so that it's fine when F.E. is, even for nonlinear problems!

Monotonicity and Implicit Methods

- ◆ Backward Euler is unconditionally monotone
 - No problems with oscillation, just too much damping
- ◆ Trapezoidal Rule suffers though, because of that half-step of F.E.
 - Beware: could get ugly oscillation instead of smooth damping

Summary 1

- ◆ Particle Systems: useful for lots of stuff
- ◆ Need to move particles in velocity field
- ◆ Forward Euler
 - Simple, first choice unless problem has oscillation/rotation
- ◆ Runge-Kutta if happy to obey stability limit
 - Modified Euler may be cheapest method
 - RK4 general purpose workhorse
 - TVD-RK3 for more robustness with nonlinearity (more on this later in the course!)

Summary 2

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

Second Order Motion

Second Order Motion

- ◆ If particle state is just position (and colour, size, ...) then 1st order motion
 - No inertia
 - Good for very light particles that stay suspended: 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.
- ◆ 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. $a(x,v,t)$)
- ◆ For systems (with many masses) say $dv/dt=M^{-1}F(x,v,t)$ where M is the "mass matrix" - masses on the diagonal

What's New?

- ◆ If $\mathbf{x}=(x,v)$ this is just a special form of 1st order: $d\mathbf{x}/dt=\mathbf{v}(x,t)$
- ◆ 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?

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

Linear Acceleration

- ◆ Dependence on x and v dominates:
 $a(x,v) = -Kx - Dv$
- ◆ Do the analysis as before:

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

- ◆ Eigenvalues of this matrix?

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]

Simplification

v 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

Split Into More Cases

- ◆ 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)

$$\alpha \approx \pm \sqrt{-1} \sqrt{\lambda_K}$$

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 or zero damping
 - Look at imaginary axis
- ◆ Velocity dependence (e.g. damping)
 - $a(x,v,t)=-Dv$
 - Want stability, monotone 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: great (monotone)
- ◆ Trapezoidal Rule
 - Constant acceleration: great (2nd order)
 - Position dependence: great (stable, no damping)
 - Velocity dependence: good (stable but only conditionally monotone)

Setting Up Implicit Solves

- ◆ Let's take a look at actually using Backwards Euler, for example

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

$$v_{n+1} = v_n + \Delta t M^{-1} F(x_{n+1}, v_{n+1})$$

- ◆ Eliminate position, solve for velocity:

$$v_{n+1} = v_n + \Delta t M^{-1} F(x_n + \Delta t v_{n+1}, v_{n+1})$$

- ◆ Linearize at guess v^k , solving for $v_{n+1} \approx v^k + \Delta v$

$$v^k + \Delta v = v_n + \Delta t M^{-1} \left(F(x_n + \Delta t v^k, v^k) + \Delta t \frac{\partial F}{\partial x} \Delta v + \frac{\partial F}{\partial v} \Delta v \right)$$

- ◆ Collect terms, multiply by M

$$\left(M - \Delta t \frac{\partial F}{\partial v} - \Delta t^2 \frac{\partial F}{\partial x} \right) \Delta v = M (v_n - v^k) + \Delta t F(x_n + \Delta t v^k, v^k)$$

Symmetry

- ◆ Why multiply by M?
- ◆ Physics often demands that $\frac{\partial F_{\text{position}}}{\partial x}$ and $\frac{\partial F_{\text{velocity}}}{\partial v}$ are symmetric
 - And M is symmetric, so this means matrix is symmetric, hence easier to solve
 - (physics generally says matrix is SPD - even better)
 - If the masses are not equal, the acceleration form of the equations results in an unsymmetric matrix - bad.
- ◆ Unfortunately the matrix $\frac{\partial F_{\text{velocity}}}{\partial x}$ is usually unsymmetric
 - Makes solving with it considerably less efficient
 - See Baraff & Witkin, "Large steps in cloth simulation", SIGGRAPH '98 for one solution: throw out bad part

Specialized 2nd Order 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 of the underlying physics are preserved in discretization - quite desirable visually!
- ◆ [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! (symplectic)
- ◆ 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...

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!

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?

Implicit Compromise

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

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

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

- ◆ Constant acceleration: great (2nd order)
- ◆ Position dependence: great (2nd order, no damping)
- ◆ Velocity dependence: great (unconditionally monotone)