Game Physics Notes 02

CSCI 321

WWU

April 13, 2016

Forces

Newton's second law of motion: F = ma

$$a = F/m$$

 $v' = a$
 $p' = v$

Corpus omne perseverare in statu suo quiescendi vel movendi uniformiter in directum, nisi quatenus a viribus impressis cogitur statum illum mutare.

Or, in English:

Every body perseveres in its state of being at rest or of moving uniformly straight forward, except insofar as it is compelled to change its state by force impressed.

Forces and Motion

$$F = ma$$

$$a = F/m$$

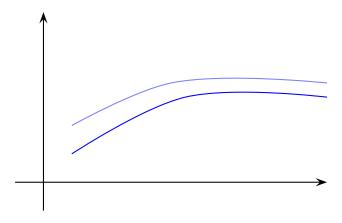
$$v' = a$$

$$p' = v$$

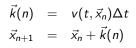
- What we really want to know is: "How do things move?"
- If we know the forces and masses, we know the acceleration.
- If we can integrate the acceleration we can get the velocity.
- If we can integrate the velocity we can get the position.
- The problem is integration—generally unsolvable.
- So we use approximate integration.

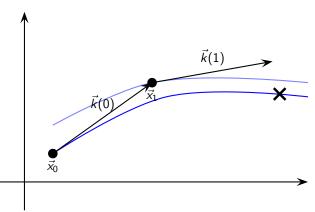
Euler Integration

Exact integration would move the point along the blue lines.



Euler Integration





Euler Integration

$$a = F/m$$

$$v' = a$$

$$p' = v$$

```
def update(F, m, dt):
    a = F / m
    v += a * dt
    p += v * dt
```

Sample Calculations

Run spring.py

Sample Calculations with smaller timestep

			Euler:				_
			t	X	V	а	
dt	=	0.5	0.0	20.0	0.0	-10.0	
m	=	10	0.5	20.0	-5.0	-10.0	
k	=	5	1.0	17.5	-10.0	-8.8	
ſ		-kx	1.5	12.5	-14.4	-6.3	
=			2.0	5.3	-17.5	-2.7	
а	=	f/m = -kx/m = -x/2	2.5	-3.4	-18.8	1.7	
x'	=	V	3.0	-12.9	-18.0	6.4	
\mathbf{v}'	=	а	3.5	-21.8	-14.8	10.9	
	_	x + y' = y + y	4.0	-29.2	-9.3	14.6	
		$x_t + x_t' = x_t + v_t$	4.5	-33.9	-2.0	16.9	
v_{t+1}	=	$v_t + v_t' = v_t + a_t$	5.0	-34.9	6.5	17.4	

• Run spring.py

Online discussions of Midpoint and Runge Kutta

Readings:

- http://www.pixar.com/companyinfo/research/pbm2001/,
 Differential equation basics, and Particle dynamics
- http://www.nrbook.com/c/, 16.0, 16.1

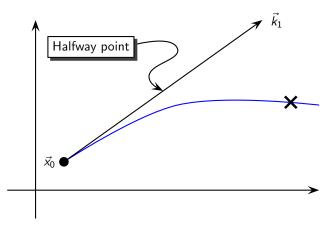
$$\vec{k}_1 = d(\vec{x}_n)\Delta t$$

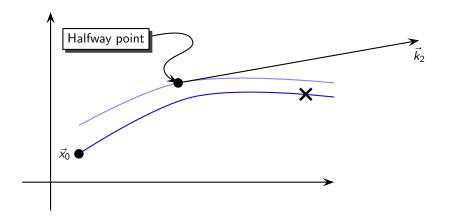
$$\vec{k}_2 = d(\vec{x}_n + \frac{1}{2}\vec{k}_1)\Delta t$$

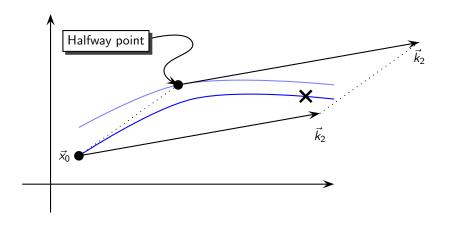
$$\vec{x}_{n+1} = \vec{x}_n + \vec{k}_2$$

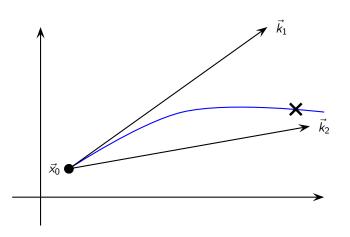
- Euler method has errors $O(\Delta t^2)$
- Midpoint method has errors $O(\Delta t^3)$
- Can take steps twice as big and get smaller errors:

$$0.05^2 = 0.0025$$
$$0.10^3 = 0.001$$

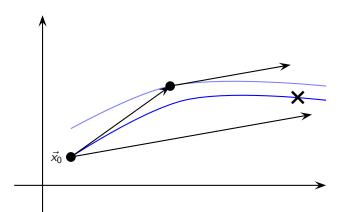








One midpoint method step of size Δt is more accurate than two Euler steps of size $\Delta t/2$.



Sample Calculations

		Midpoint:								
			t	X	V	а				
dt	=	1	0.0	20.0	<u>0.0</u>	10.0				
m	=	10	0.5	20.0	-5.0	-10.0				
k	=	5	1.0	15.0	-10.0	-7.5	İ			
f	_	-kx	1.5	10.0	-13.8	-5.0	İ			
=			2.0	1.3	-15.0	-0.6	İ			
а	=	f/m = -kx/m = -x/2	2.5	-6.3	-15.3	3.1				
x'	=	V	3.0	-14.1	-11.9	7.0				
v'	=	a	3.5	-20.0	-8.4	10.0				
V	_	$x_t + x_t' = x_t + v_t$	4.0	-22.4	-1.9	11.2	İ			
		<u>-</u>	4.5	-23.4	3.7	11.7				
v_{t+1}	=	$v_t + v_t' = v_t + a_t$	5.0	-18.7	9.8	9.3				

First add half of the derivative.

Sample Calculations

Then add all the "half-derivative."

Note that this is far more accurate than Euler with half step size.

Fourth Order Runge-Kutta

$$\vec{k}_{1} = d(\vec{x}_{n})\Delta t$$

$$\vec{k}_{2} = d(\vec{x}_{n} + \frac{1}{2}\vec{k}_{1})\Delta t$$

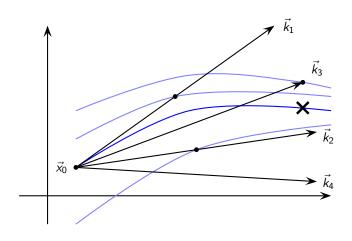
$$\vec{k}_{3} = d(\vec{x}_{n} + \frac{1}{2}\vec{k}_{2})\Delta t$$

$$\vec{k}_{4} = d(\vec{x}_{n} + \vec{k}_{3})\Delta t$$

$$\vec{x}_{n+1} = \vec{x}_{n} + \frac{\vec{k}_{1}}{6} + \frac{\vec{k}_{2}}{3} + \frac{\vec{k}_{3}}{3} + \frac{\vec{k}_{4}}{6}$$

Fourth order Runge Kutta

Tangents calculated at the dots: $\frac{\vec{k}_1}{6}+\frac{\vec{k}_2}{3}+\frac{\vec{k}_3}{3}+\frac{\vec{k}_4}{6}$



Fourth Order Runge-Kutta

- Euler method has errors $O(\Delta t^2)$
- Midpoint method has errors $O(\Delta t^3)$
- ullet Fourth order Runge Kutta has errors $O(\Delta t^5)$

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$$0.05^2 = 0.00250$$

 $0.10^3 = 0.00100$
 $0.20^5 = 0.00032$

Stepsize Matching Refresh Rate

• The simplest approach to stepsize is to use the framerate:

```
framerate = 30.0
t = 0.0
dt = 1.0/framerate
while !quitting:
   clock.tick(framerate)
   handle.input()
   integrate(state, t, dt)
   t += dt
   display()
```

- This may be OK for simple games, but if more accuracy is needed the physics should use as small a timestep as possible.
- Also, the game refresh rate may not keep up with the nominal clock rate

Stepsize Matching Refresh Rate $\times n$

• Can also match *n* steps to each frame:

```
framerate = 30.0
t = 0.0
dt = 1.0/framerate
while !quitting:
   clock.tick(framerate)
   handle.input()
   for i in range(n):
        integrate(state, t, dt/n)
   t += dt
   display()
```

Use actual timestep

clock.tick returns milliseconds since last call.

```
framerate = 30.0
t = 0.0
while !quitting:
   dt = clock.tick(framerate) * 0.001
   handle.input()
   integrate(state, t, dt)
   t += dt
   display()
```

- Physics will be "same" regardless of computer's speed.
- But again physics update should be as fast as possible for most realism.
- We could increase the framerate, but then we'd be doing unnecessary rendering.

Use smaller time step

```
framerate = 30.0
t = 0.0
dt = 0.01
while !quitting:
   timespan = clock.tick(framerate) * 0.001
handle.input()
   while (timespan > 0):
    integrate(state, t, dt)
      timespan -= dt
   display()
```

- Problem with the fractional part of dt?
 - Can interpolate for fractional dt.
- What if the physics gets behind? Spiral of death!
 - Make sure your physics can keep up with dt.

Use separate time for display and physics

```
framerate = 30.0
rendertime, physicstime = 0.0, 0.0
dt = 0.01
while !quitting:
   rendertime += clock.tick(framerate) * 0.001
   handle.input()
   while (physicstime < rendertime):
     integrate(state, physicstime, dt)
     physicstime += dt
   display()</pre>
```

- Spiral of death still possible.
 - Note: matching stepsize to framerate $\times n$ avoids death spiral.
- Leftover fraction of dt is carried forward to next render.
- Can interpolate again for fractional dt.
- Note that dt can be longer than time for a frame and it still works.

Differential Equations

Reading:

- Strange attractors http://en.wikipedia.org/wiki/Attractor
- Run: strange??.py
- The Limits to Growth

http://www.manicore.com/fichiers/Turner_Meadows_vs_historical_data.pdf

 $\verb|http://www.theguardian.com/commentisfree/2014/sep/02/limits-to-growth-was-right-new-research-shows-were-new-research-shows$

Symplectic Euler/Semi-implicit Euler

- http://en.wikipedia.org/wiki/Semi-implicit_Euler_method
- Two forms:

$$v_{n+1} = v_n + a_n \Delta t$$

 $p_{n+1} = p_n + v_{n+1} \Delta t$

and

$$p_{n+1} = p_n + v_n \Delta t$$

$$v_{n+1} = v_n + a_{n+1} \Delta t$$

- Can use either one by itself, or alternate between them.
- Not accurate, but almost conserves energy.
- Easy to program when updates are by assignment.

Verlet Integration

Begin with symplectic Euler

$$v_{n+1} = v_n + a_n \Delta t$$

$$p_{n+1} = p_n + v_{n+1} \Delta t$$

• Substitute for v_{n+1}

$$v_{n+1} = v_n + a_n \Delta t$$

$$p_{n+1} = p_n + (v_n + a_n \Delta t) \Delta t$$

$$= p_n + v_n \Delta t + a_n \Delta t^2$$

• Use old positions to approximate $v_n \Delta t \approx p_n - p_{n-1}$

$$p_{n+1} = p_n + v_n \Delta t + a_n \Delta t^2$$

$$= p_n + (p_n - p_{n-1}) + a_n \Delta t^2$$

$$= 2p_n - p_{n-1} + a_n \Delta t^2$$

• This is velocityless Verlet. There are other versions.

Verlet Integration

• A Verlet based approach for 2D game physics (www.gamedev.net)

http://www.gamedev.net/page/resources/_/technical/math-and-physics/a-verlet-based-approach-for-2d-

A nice web demo:

 $\verb|http://gamedev.tutsplus.com/tutorials/implementation/simulate-fabric-and-ragdolls-with-simple-version-defined and the substitution of the subs$

- Can be used as the basis of a collision response system.
- Run VerletPhysicsDemo.py

True elastic collisions

- http://en.wikipedia.org/wiki/Elastic_collision
- Run BouncingBalls.py