嗯哼,动画。

Mass Spring System 质点弹簧系统

Idealized spring

a b
$$m{f}_{a o b}=k_S(m{b}-m{a})$$
 $m{f}_{b o a}=-m{f}_{a o b}$

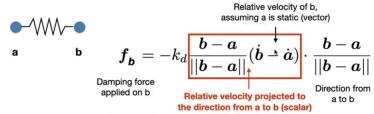
Force pulls points together

Strength proportional to displacement (Hooke's Law)

 k_s is a spring coefficient: stiffness

改变 k_s, 就是对不同材质的模拟,一种材质内部可能有不同的 k_s

Damp only the internal, spring-driven motion



- Viscous drag only on change in spring length
 - Won't slow group motion for the spring system (e.g. global translation or rotation of the group)
- Note: This is only one specific type of damping

图 1 还要考虑摩擦力,否则系统不会停止运动

Particle System 粒子系统

说了挺多哈哈,懒得记了。

逆运动学

难题

- 1. 难解
- 2. 解不唯一

Euler's Wethod

Euler's Method (a.k.a. Forward Euler, Explicit Euler)

- Simple iterative method
- Commonly used
- Very inaccurate
- Most often goes unstable

$$oldsymbol{x}^{t+\Delta t} = oldsymbol{x}^t + \Delta t \, \dot{oldsymbol{x}}^t$$

$$\dot{\boldsymbol{x}}^{t+\Delta t} = \dot{\boldsymbol{x}}^t + \Delta t \, \ddot{\boldsymbol{x}}^t$$

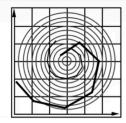
Instability of the Euler Method

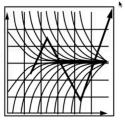
The Euler method (explicit / forward)

$$\boldsymbol{x}^{t+\Delta t} = \boldsymbol{x}^t + \Delta t \, \boldsymbol{v}(\boldsymbol{x}, t)$$

Two key problems:

- Inaccuracies increase as time step Δt increases
- Instability is a common, serious problem that can cause simulation to diverge





Witkin and Baraf

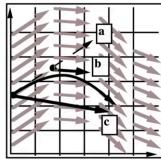
问题:除了采样率之外,误差会被无限放大才是更致命的。

Miapoint Method

Midpoint method

- Compute Euler step (a)
- Compute derivative at midpoint of Euler step (b)
- Update position using midpoint derivative (c)

$$x_{\text{mid}} = x(t) + \Delta t / 2 \cdot v(x(t), t)$$
$$x(t + \Delta t) = x(t) + \Delta t \cdot v(x_{\text{mid}}, t)$$



Witkin and Baraff

Modified Euler

Modified Euler

- Average velocity at start and end of step
- Better results

$$egin{aligned} oldsymbol{x}^{t+\Delta t} &= oldsymbol{x}^t + rac{\Delta t}{2} \ (\dot{oldsymbol{x}}^t + \dot{oldsymbol{x}}^{t+\Delta t}) \ \dot{oldsymbol{x}}^{t+\Delta t} &= \dot{oldsymbol{x}}^t + \Delta t \ \ddot{oldsymbol{x}}^t \end{aligned}$$

$$oldsymbol{x}^{t+\Delta t} = oldsymbol{x}^t + \Delta t \ \dot{oldsymbol{x}}^t + rac{(\Delta t)^2}{2} \ \ddot{oldsymbol{x}}^t$$



Adaptive Step Size

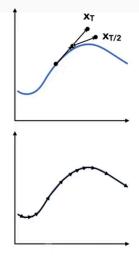
既然反正都要算这个

Adaptive step size

- Technique for choosing step size based on error estimate
- Very practical technique
- But may need very small steps!

Repeat until error is below threshold:

- Compute x_T an Euler step, size T
- Compute x_{T/2} two Euler steps, size T/2
- Compute error || x_T − x_{T/2} ||
- If (error > threshold) reduce step size and try again



How to determine / quantize "stability"?

- We use the local truncation error (every step) / total accumulated error (overall)
- Absolute values do not matter, but the orders w.r.t. step
- Implicit Euler has order 1, which means that
 - Local truncation error: O(h2) and
 - Global truncation error: O(h) (h is the step, i.e. Δt)

图 2 梦回渐进理论

Runge-Kutta

Runge-Kutta Families

A family of advanced methods for solving ODEs

- Especially good at dealing with non-linearity
- It's order-four version is the most widely used, a.k.a. RK4

Initial condition:

$$rac{dy}{dt}=f(t,y),\quad y(t_0)=y_0.$$

$$rac{dy}{dt} = f(t,y), \quad y(t_0) = y_0. \qquad \quad y_{n+1} = y_n + rac{1}{6} h \left(k_1 + 2 k_2 + 2 k_3 + k_4
ight), \ t_{n+1} = t_n + h$$

where

$$egin{aligned} k_1 &= f(t_n, y_n), \ k_2 &= f\left(t_n + rac{h}{2}, y_n + hrac{k_1}{2}
ight). \end{aligned}$$

$$k_1 = f(t_n, y_n), \qquad \qquad k_3 = f\left(t_n + rac{h}{2}, y_n + hrac{k_2}{2}
ight), \ k_2 = f\left(t_n + rac{h}{2}, y_n + hrac{k_1}{2}
ight), \qquad \qquad k_4 = f\left(t_n + h, y_n + hk_3
ight).$$

其实就是高级一点的中点法

完结撒花。

做一下作业 8

嗯,环境有问题,那就不做了。

后续的 final project, 也是我带专生涯的毕业设计, 会继续更新, 放在 https://gitee. com/isirin1131_admin/specialized-graduation-program

告一段落喽。

有什么感言呢? 我只是感觉有些无聊了,完成我的生成函数手稿后,就会开始 games001 的学 习。

之后就是 games 200 系列了? 或许会开始搞 AI 了,但那太远了,还是关注在短期目标上吧。