### Physically-Based Modeling: Mass-Spring Systems

M. Alex O. Vasilescu









### Surface Reconstruction



### **Data Structures**

- Node *i*. *i*=1....N
  - Mass  $m_i$
  - Position:  $\mathbf{x}_{i}(t) = [x_{i}(t), y_{i}(t), z_{i}(t)]^{t}$
  - $\mathbf{v}_i(t) = d\mathbf{x}_i(t)/dt$ - Velocity:
  - Accelaration  $\mathbf{a}_i(t) = d^2 \mathbf{x}_i(t)/dt^2$
  - Net nodal force:  $\mathbf{f}_{i}^{net}(t)$
- Spring
  - Connects node *j* to node *i*
  - Natural length  $l_{ii}$
  - Stiffness k<sub>ii</sub>

### typedef struct node{ float mass; vector position; vector velocity; vector force; } node;

typedef struct spring{ node \*n1; node \*n2; double rest length; double spring\_constant; } spring;

### System Dynamics:

Lagrange equations of motion:

$$m_i \ddot{\mathbf{x}}_i + \gamma_i \dot{\mathbf{x}}_i - \mathbf{g}_i - \mathbf{f}_i = \mathbf{0},$$

*γ*<sub>i</sub> is damping coefficient

- f<sub>i</sub> is the external force at node i
- g, total internal force on the node i due to neighboring nodes

### System Dynamics:

Lagrange equations of motion:

$$\mathbf{F}_{i,\text{total}} = -\gamma_i \dot{\mathbf{x}}_i + \mathbf{g}_i + \mathbf{f}_i$$

$$\mathbf{F}_{i,\text{total}} = m_i \tilde{\mathbf{X}}_i$$

- $\gamma_i$  is damping coefficient
- f<sub>i</sub> is the external force at node i
- g<sub>i</sub> total internal force on the node *i* due to neighboring nodes

 $m_i \ddot{\mathbf{x}}_i + \gamma_i \dot{\mathbf{x}}_i - \mathbf{g}_i - \mathbf{f}_i = \mathbf{0},$ 



- for (i=0; i<num\_nodes; i++){ vscale((1 / nds[i].mass), nds[i].force, accel); vscale(dt, accel, delta\_vel);
- vplus(delta\_vel, nds[i].velocity, nds[i].velocity); vscale(dt, nds[i].velocity, delta\_pos);
- $\mathbf{x}_i(t + \Delta t) = \mathbf{x}_i(t) + \Delta t \mathbf{v}_i(t + \Delta t) \implies \text{vinc(delta_pos, nds[i].position);}$

### Force Computation (Cont.)

Damping Forces:  $\gamma \dot{\mathbf{x}}_{i}(t)$ 

void damping\_forces (int num\_nodes, node \*nds) { vector force; int i; for (i=0; i<num\_nodes; i++){ vscale(DAMPING, nds[i].velocity, force);

vdec(force, nds[i].force);}}

External force (very simple) : springs attached to the data from each mesh node

 $\boldsymbol{f}_{i}(t) = \boldsymbol{k}_{data} \| \operatorname{data}(\mathbf{x}_{i}(t)) - \mathbf{x}_{i}(t) \|$ 

- project the position of node i  $\mathbf{x}_i(t)$ into the data, and extract the value at that location  $data(\mathbf{x}_i(t))$
- calculate the force using a data spring  $\implies$  nd->force.z += IMAGE\_CONST\*(val p.z); constant  $k_{data}$

void external forces(int num nodes, node \*nds) { int i; node \*nd; vector p; float val; for (i=0: i<num nodes: i++) { nd = &ndslil

p = nd->position;

val=image\_interp(p.x, p.y, IMAGE\_DATA);

### **Mass-Spring Forces**

Spring Forces:

 $\mathbf{a}_i(t) =$ 

mi

 $\mathbf{v}_{i}(t + \Delta t) = \mathbf{v}_{i}(t) + \Delta t \, \mathbf{a}_{i}(t)$ 

 $-\mathbf{g}_i(t)$  total force on the node *i* due to springs connecting it to neighboring nodes  $j \in N_i$ 



### **Force Computation**

#### Spring Forces: • $g_i$ total force on the node i due to springs connecting it to neighboring • $g_i(t) = \sum_{j \in N_i} sij$ • the force spring ij exerts on node i $s_{ij} = k_i j e_{ij} \left| \frac{r_{ij}}{|r_{ij}|} \right|$ • where • $r_{ij} = x_j - x_i$ is separation of nodes • $r_{ij} = |r_{ij}| - l_{ij}$ is actual length of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}$ is deformation of spring $e_{ij} = |r_{ij}| - l_{ij}| - l_{ij}$





### Adaptive Meshes

• Use Lagrangian equations of motion to simulate mesh dynamics.

۰E

$$m_i \ddot{\mathbf{x}}_i + \gamma_i \dot{\mathbf{x}}_i - \mathbf{g}_i - \mathbf{f}_i = 0$$

• The internal spring forces have spring "constants" that vary according to a data dependent function, such as the magnitude of the gradient.

	$\mathbf{g}_i(t) = \sum_{\substack{j \in N_i}} \mathbf{s}_{ij} ,$	$\mathbf{s}_{ij} = k_{ij}  e_{ij} \frac{\mathbf{r}_{ij}}{ \mathbf{r}_{ij} }$
igital Image:	$\mathbf{d}(k,l)$	
daptation function:	$\mathbf{a}_{d} = G * \left\  \nabla \mathbf{d} \right\ $	G = gaussian
bservation:	$O_i = \mathbf{a}_d(x_i, y_i)$	
pring "Constant": - var	ies according to the adaptation fu	nction
	$k = (1 \circ k) + \circ k$	a = 5(0 + 0)



### Adaptive Surface Reconstruction





## SpringLens

### **Distributed Nonlinear Magnifications**

Tobias Germer Timo Götzelmann Martin Spindler Thomas Strothotte

Department of Simulation and Graphics Otto-von-Guericke University of Magdeburg

### Mass-Spring Systems Applications

Surface Reconstruction:

Deformable objects capable of:

- Heat conduction,
- Thermoelasticity,
- Melting and fluid-like behavior in the molten state

Cloth simulation



### Heating and Melting Deformable Models





### Heat/Diffusion Equation

• Diffusion of heat in materials:

$$\stackrel{\partial}{\partial t}(\mu\sigma\theta) - \nabla \cdot (\mathbf{C}\nabla\theta) = \mathbf{d}$$

Heat is conducted from high temperature to low temperature .

The rate of heat conduction per unit area is proportional to the gradient of the temperature.

The amount of heat required to raise the temperature of a material theta degrees is proportional to the mass of the sample per unit volume and the proportionality factor sigma, the specific heat of the material.



### Heat/Diffusion Equation

• Homogeneous, isotropic material:

$$\frac{\partial}{\partial t}(\mu\sigma\theta) - c\nabla^2\theta = q$$

• Discretize the heat equation:

$$\mu\sigma \frac{\left(\theta^{t+\Delta t} - \theta^{t}\right)}{\Delta t} - c \left[ \frac{\theta^{t}_{u+\Delta u,v,w} - 2\theta^{t}_{u,v,w} + \theta^{t}_{u-\Delta u,v,w}}{\Delta t^{2}} + \frac{\theta^{t}_{u,v+\Delta u,w} - 2\theta^{t}_{u,v,w} + \theta^{t}_{u,v-\Delta v,w}}{\Delta v^{2}} + \frac{\theta^{t}_{u,v,w+\Delta w} - 2\theta^{t}_{u,v,w} + \theta^{t}_{u,v-\Delta v,w}}{\Delta w^{2}} \right] = q$$





### Liquids - Particle Models

Model long range attraction and short range repulsion forces between pairs of particles according to Lennard-Jones potentials.

Forces involving inverse powers of particle separation



### **Discrete Fluid Model**

The total force on a particle, i, due to all other particles

$$\mathbf{g}_{i}(t) = \sum_{j \neq i} \mathbf{g}_{ij}(t) \qquad \text{attraction term repulsion term}$$
$$\mathbf{g}_{ij}(t) = m_{i}m_{j}(\mathbf{x}_{i} - \mathbf{x}_{j}) \left( -\frac{\alpha}{(r_{ij} + \varsigma)^{\alpha}} + \frac{\beta}{(r_{ij})^{b}} \right)$$

a=2 and b=4

 $\alpha$  and  $\beta$  determine the strength of the attraction and repulsion forces  $r_{ij} = \left\| \mathbf{x}_{j} - \mathbf{x}_{i} \right\|$ 

 $\zeta$  minimum required separation between particles









# Implicit Numerical Solution $\mathbf{M}\ddot{\mathbf{x}} + \mathbf{G}\dot{\mathbf{x}} + \mathbf{K}\mathbf{x} = \mathbf{f}$ discretize time: $t \Rightarrow 0, \Delta t, 2\Delta t, \dots, t, t + \Delta t, \dots$ $\mathbf{x}^{t+\Delta t} = \mathbf{x}^{t} + \Delta t \, \dot{\mathbf{x}}^{t+\Delta t} \qquad \mathbf{x}^{t+\Delta t} = \mathbf{x}^{t} + \Delta t \, \dot{\mathbf{x}}^{t}$ $\dot{\mathbf{x}}^{t+\Delta t} = \dot{\mathbf{x}}^{t} + \Delta t \, \ddot{\mathbf{x}}^{t+\Delta t} \qquad \dot{\mathbf{x}}^{t+\Delta t} = \dot{\mathbf{x}}^{t} + \Delta t \, \ddot{\mathbf{x}}^{t}$ $\Rightarrow \ddot{\mathbf{x}}^{t+\Delta t} = (\dot{\mathbf{x}}^{t+\Delta t} - \dot{\mathbf{x}}^{t})/\Delta t \qquad \ddot{\mathbf{x}}^{t+\Delta t}$ $\left(\frac{\mathbf{M}}{\Delta t}\dot{\mathbf{x}}^{t+\Delta t} - \frac{\mathbf{M}}{\Delta t}\dot{\mathbf{x}}^{t}\right) + \left(\mathbf{G}^{t}\dot{\mathbf{x}}^{t+\Delta t}\right) + \left(\mathbf{K}^{t}\mathbf{x}^{t} + \Delta t\mathbf{K}^{t}\dot{\mathbf{x}}^{t+\Delta t}\right) = \mathbf{f}^{t}$





### Skyline Storage Schemes for System Matrix A

A is a  $n^2 \times n^2$  matrix that typically requires O(n<sup>3</sup>) storage.





