

Sampling and Reconstruction with Adaptive Meshes

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Outline

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Problem

- Adaptively sample the data to lower resolution
 - for compression or simplification
- Conventional approaches: uniformly sampling

Background (Numerical Grid Generation)

- A tool for numerical solution of partial differential equations on arbitrarily shaped regions.
- The numerical solution is built on those grids.
- Make best use of those finite grids to represent the physical solution with sufficient accuracy.
- The grid is influenced by both the geometric configuration and by the physical solution being done thereon.

Dynamic Mesh

- Essential component of adaptive mesh
- Constructed from physically-based nodes and springs
- Finite elements: spring and its attached nodes



Dynamic Mesh (Node)

- Node: $i, i = 1, \dots, N$
 - mass: m_i
 - position : $X_i(t) = [x_i(t), y_i(t), z_i(t)]$
 - velocity: $v_i = dX_i / dt$
 - acceleration: $a_i = d^2 X_i / dt^2$
 - force $f_i^n(t)$
- Be subject to a net nodal force: $f_i^n(t)$

Dynamic Mesh (Spring)

- Spring: ij
 - connect node j to node i
 - natural length l_{ij}
 - stiffness C_{ij}
- Force exerted on node i : $i \bullet \text{---} \bullet j$

$$S_{ij} = \frac{C_{ij} e_{ij}}{\|r_{ij}\|} r_{ij},$$

$$r_{ij} = X_j - X_i, e_{ij} = \|r_{ij}\| - l_{ij}$$

Dynamic Mesh (Motion equations)

- For each node:

$$m_i \frac{d^2 X_i}{dt^2} + \gamma_i \frac{dX_i}{dt} + g_i = f_i; i = 1, \dots, N$$

m_i : mass X_i : position γ_i : damping coefficient
 g_i : the total force exerted from all connected springs
 f_i : an external force applied at node i

Dynamic Mesh (Numerical time-integration)

- At each time step t :

$$1. f_i^{nt} = f_i^t - \gamma_i v_i^t - g_i^t$$

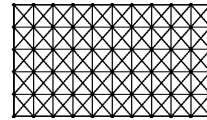
$$2. a_i^t = \frac{B_i f_i^{nt}}{m_i} \quad \boxed{m_i \frac{d^2 X_i}{dt^2} + \gamma_i \frac{dX_i}{dt} + g_i = f_i; i = 1, \dots, N}$$

$$3. v_i^{t+\Delta t} = v_i^t + \Delta t a_i^t$$

$$4. X_i^{t+\Delta t} = X_i^t + \Delta t v_i^{t+\Delta t}$$

Dynamic Mesh (Utilized in this paper)

- Quadrilateral elements are assembled into bounded surfaces (actually rectangular)



- Boundary condition: The boundary nodes can only move along the boundary in (x,y) visual domain

Adaptive Meshes

- Dynamic meshes with an adaptation function

-- stiffness of springs are changed according to the adaptation function during the process.

-- stiffness of spring increases in regions with rapid variation.

Adaptive Meshes (Adaptation functions)

- Adaptation function, $\sigma^k(\mathcal{D}) \in [0,1]$

$$\sigma^k(\mathcal{D}) = \alpha(\mathcal{D}) * H(\nabla \mathcal{D})$$

G is a normalized spatial filter; H is some function of partial derivative of the input data field.

- Nodal Observation, $\sigma^k(\mathcal{V})$

$$\sigma^k(\mathcal{V}) = \sigma^k(\mathbf{I}(\mathbf{X}))$$

Adaptive Meshes (Feedback procedure)

■ Feedback procedure

$$c_{ij}^t = (1 - \rho_{ij}^t) c_{\min} + \rho_{ij}^t c_{\max},$$

$$\rho_{ij}^t = \frac{1}{2} (O_i^t + O_j^t)$$

More interesting area \rightarrow Higher $\rho_{ij}^t \rightarrow$ larger $c_{ij}^t \rightarrow$ larger c_{ij}^t

Algorithm

$$m_i \frac{d^2 X_i}{dt^2} + \gamma_i \frac{dX_i}{dt} + g_i = f_i; i = 1, \dots, N$$

- Input: A 2D scalar-valued intensity or range image $d(k,l)$
- At each time step t :
 1. Evaluate the adaptation function and nodal observation for each node i
 2. Adjust the stiffness for each spring ij
 3. Compute the total force from springs for each node i
 4. Evaluate the current nodal forces, acceleration, the new velocity, and the new position.

Algorithm (Data forces)

- External data force at time t :

$$f_i^t = [0, 0, \alpha(d(\Pi X_i^t) - z_i^t)]$$
- The only trigger to shift nodes in z direction.
- For surface reconstruction, it deflects the mesh to fit the input data.

Results

(Image Reconstruction)

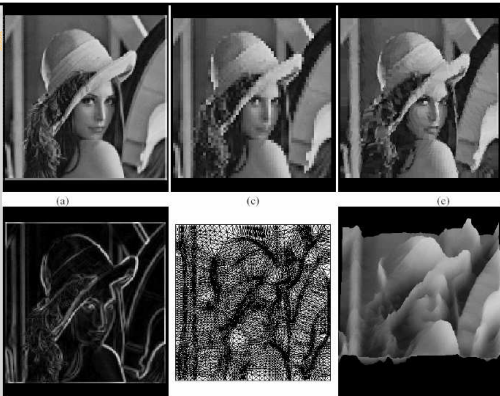
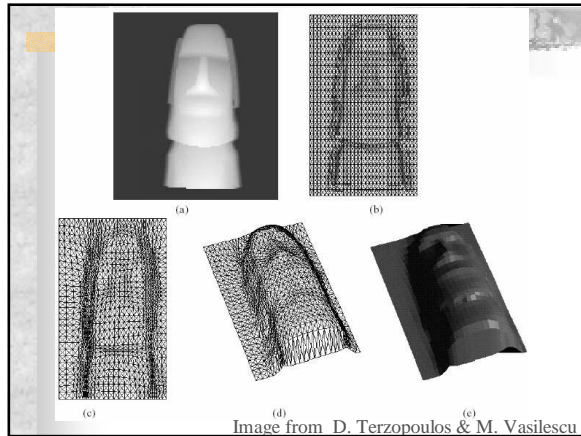


Image from D. Terzopoulos & M. Vasilescu

Results

(Surface Reconstruction)



Advantages

- Applicable to arbitrary dimensional data sets
- Nodes are optimally distributed to preserve the interesting properties of the input data.
- Interactive rates (no actual timing data)

Disadvantages

- Not suitable for arbitrary mesh in 3D
 - Assume a planar surface as initial guess
- No theoretical justification