

Progressive Geometry Compression

Andrei Khodakovsky Peter Schröder Wim Sweldens



Motivation

- Large (up to billions of vertices), finely detailed, arbitrary topology surfaces
- Difficult manageability of such large amounts of information
 - Computation, Storage, Transmission, Display Strain

Progressive Compression is needed

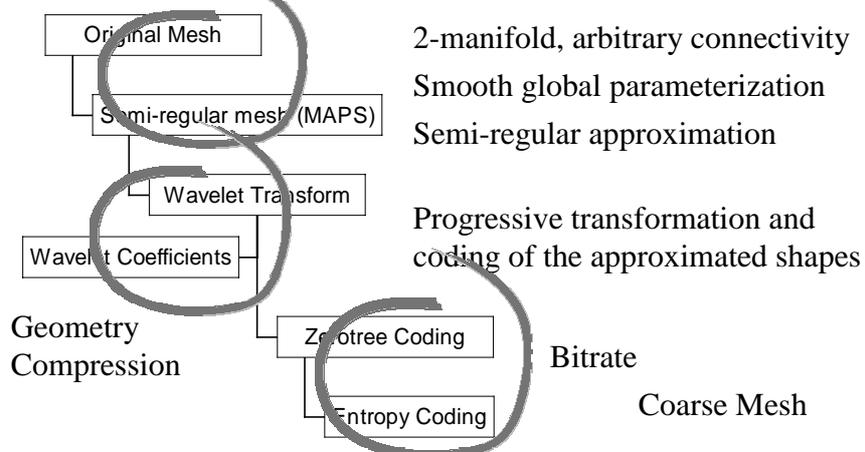
What Compression Is All About

- Accuracy/ bit per vertex
- Definition of a Geometry Error
 - Measure of the geometric distance between 3D objects
- New Problems with Respect to Image Compression
 - No direct correspondence between original and compressed surface

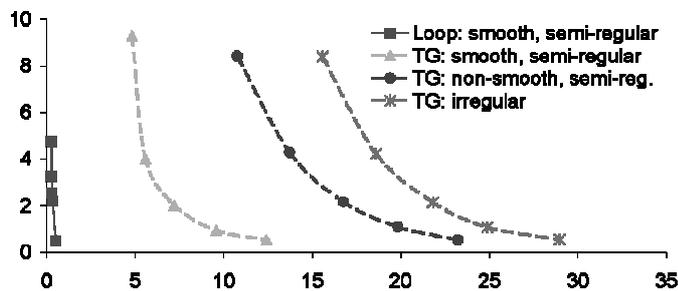
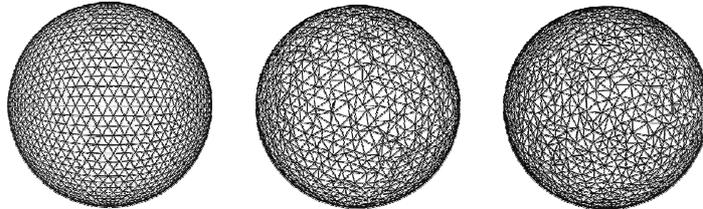
Algorithm

Detailed Mesh

Parameters & Connectivity



Geometry-Parameter-Connectivity

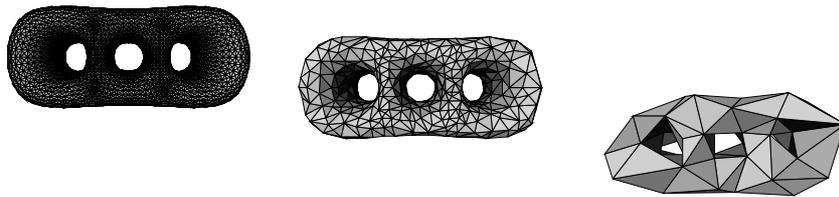


Geometry Error

- For highly detailed, densely sampled meshes, the sample location (vertices + connectivity) do not contribute in improving the geometric distance, and thus the error
- Parameter information is thus contained within the surface, and any rate-distortion improving coding should better take advantage from the geometric information only, normal to the surface

MAPS algorithm

- “Multiresolution Adaptive Parametrization of Surfaces” based on edge collapsing



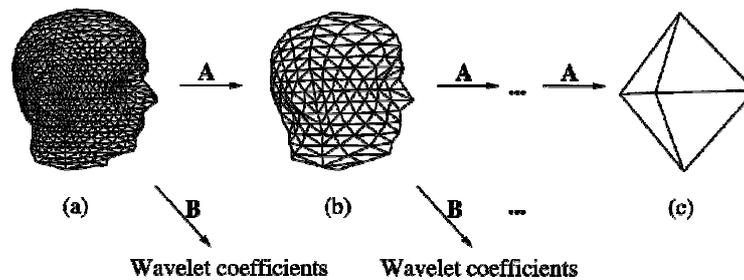
Semi-regular approximation is achieved on applying triangle quadrisection to the coarse mesh, and taking advantage of the mapping operated by the algorithm between original and coarse mesh.

Wavelet Transform-1

modeling of a complex object of arbitrary topology
at multiple levels of detail

- Replacement of level-N mesh with coarser level-(N-1) mesh + wavelet coefficients
- Generation of “nested” meshes
- Subdivision Rules and Filter-Banks

Wavelet Transform-2



Semi-regular mesh is hierarchically subdivided into a coarser mesh and some details information

Reconstruction is achieved from the coarse mesh by hierarchical addition of detail information

Wavelet Transform-3

- Filter Bank Algorithm

- Design of Analysis and Synthesis Filters.

- Decomposition of a mesh like $\mathbf{S}^{j+1}(\mathbf{x}) = \sum_i v_i^j \phi_i^j(\mathbf{x}) + \sum_j w_j^j \psi_j^j(\mathbf{x})$

- Our Paper:

- Synthesis:

$$\mathbf{p}^{j+1} = \begin{bmatrix} \mathbf{P} & \mathbf{Q} \end{bmatrix} \begin{bmatrix} \mathbf{p}^j \\ \mathbf{d}^j \end{bmatrix}$$

- Analysis: solution $[\mathbf{p}_j, \mathbf{d}_j]$ of system for a given \mathbf{p}_{j+1}

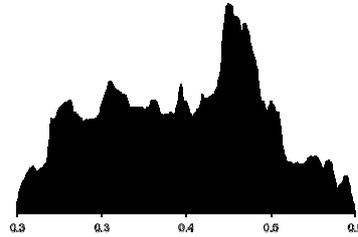
- P, low-pass reconstruction filter

- Q, high-pass reconstruction filter

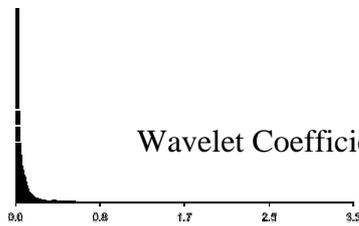
- small support

Wavelet Transform-4

Decorrelating effect of the
WV Transform



Vertex position magnitudes
for "Venus"

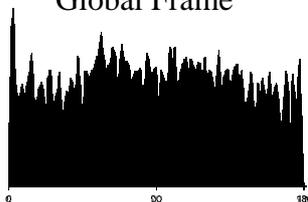


Wavelet Coefficients Magnitudes

Wavelet Coefficients

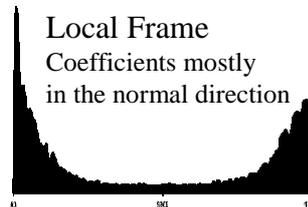
- Vector valued
- x,y,z components pretty correlated, but
- decorrelation in a local frame

Global Frame



Local Frame

Coefficients mostly
in the normal direction



- Normal component relevant for geometric information
- Code each component independently

Hierarchical Trees

- Exploit the relationship of wavelet coefficients across bands, and their exponential decay
- Areas with significant information are similar in shape and location
- Non significance in a low frequency band for a particular level of accuracy means with high probability non significance of the children nodes.
- Localizing a zerotree avoids transmitting a large amount of insignificant details with respect to a desired level of accuracy
- Progressive compression, embedded code

Zerotree Coding

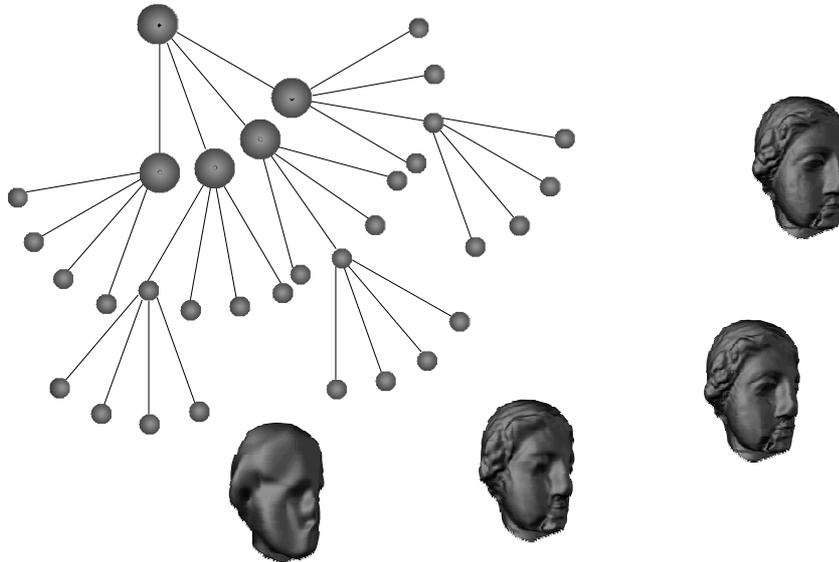
Example of the quality of the coding



SPIHT PSNR = 35.12 dB, JPEG PSNR = 31.8 dB (quality factor 15%).

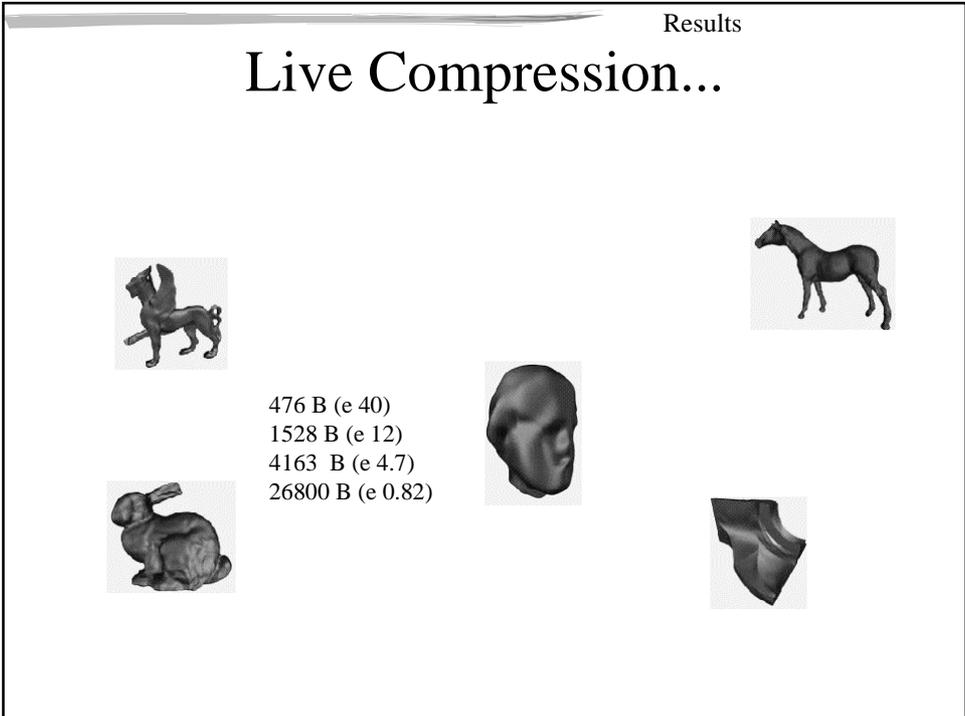
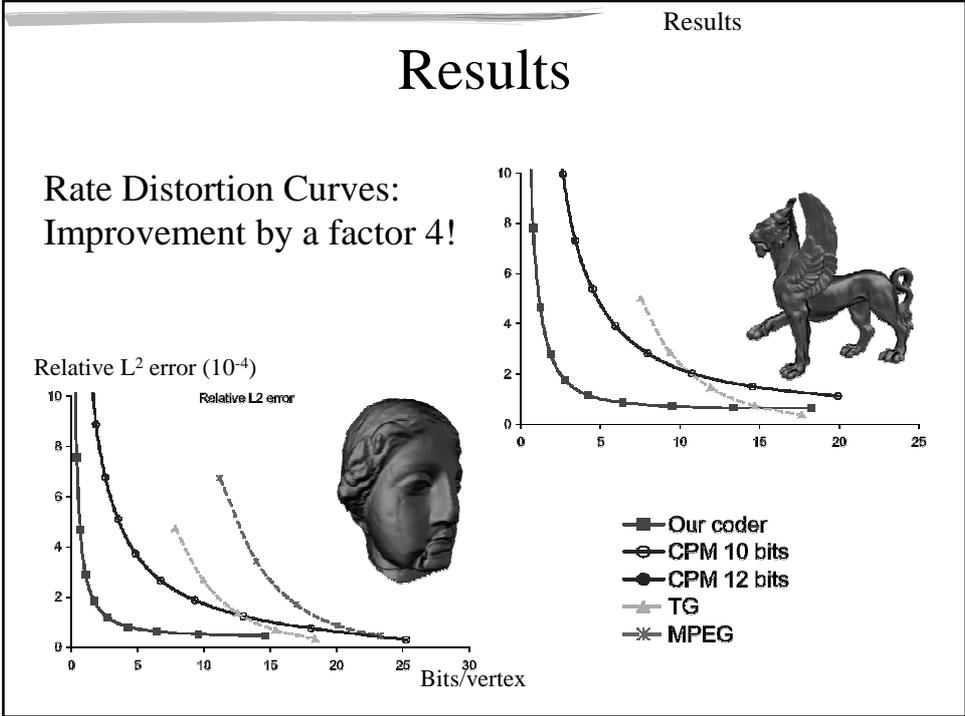
SAID AND PEARLMAN, A New, Fast, and Efficient Image Codec Based on Set Partitioning in Hierarchical Trees. *IEEE Transaction on Circuits and Systems for Video Technology* 6, 3 (1996), 243–250.

Zerotree-Example



Entropy Coding

- Further improvement of the bit budget by eliminating redundancy due to non-uniform distribution
- Refinement and sign bits are found to be distributed uniformly
- Significance is a function of bitplane: “early” bitplanes will contain many insignificant coefficients, which will become significant in “later” bitplanes



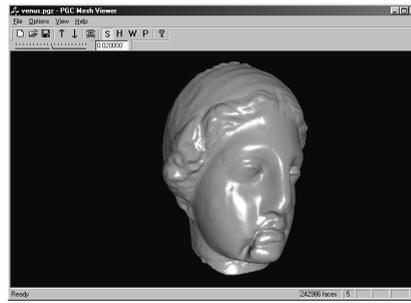
Conclusions

- Very effective compression algorithm
 - smooth appearance, low (hardware) strain
- Many details in very early stages of decompression
- Very fruitful distinction between parameters and geometry
- Still at the beginning of wavelet 3D model compression...

In Fact...

- Effectiveness of compression inherits very much from a bunch of “tested” approaches
- No sound treatment of multiresolution analysis
 - orthogonality, stability...
- “Tentative” choice of wavelets
- Elimination of tangential information (Normal Meshes...)

Demo



Thanks

- Andreas Hubeli for the supervision and for help
- You for your attention....

It's your turn...

Extra Slides...

Wavelet Transform-2

Ingredients of Multiresolution Analysis (Mallat and Meyer) :
Existence of nested linear spaces and of an inner product relative to a subdivision rule.

$$V^0 \subset V^1 \subset V^2 \subset \dots$$

$$W^j := \{f \in V^{j+1} \mid \langle f, g \rangle = 0 \quad g \in V^j\} \quad f^{j+1} = f^j + h^j$$

Nested spaces are generated by translations and dilations of a single function, $\phi(x)$

$$\phi(x) = \sum_i p_i \phi(2x - i)$$

$$V^j := \text{Span}\{\phi(2^j x - i) \mid i = -\infty, \dots, \infty\}$$

Subdivision Rules can be used to define such functions

Distance Function

- Euclidean Distance (L^2) $d(X, Y)$ between two surfaces X, Y

$$d(X, Y) = \left(\frac{1}{\text{area}(X)} \int_{x \in X} d(x, Y)^2 dx \right)^{1/2}$$

- Symmetrize by taking the max of $d(X, Y)$ and $d(Y, X)$