

AN ADAPTED LAPLACIAN OPERATOR FOR HYBRID QUAD/TRIANGLE MESHES

A thesis submitted in partial fulfillment of the requirements for the degree of:
Master in Systems Engineering and Computer Science

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INTRODUCTION

SPECTRAL MESH PROCESSING

Taubin in 1995 suggested that the **Discrete Laplace Operator** allows to do *Spectral Processing* on **Polygonal Meshes** in an analogous way to signal processing with the Fourier transform.

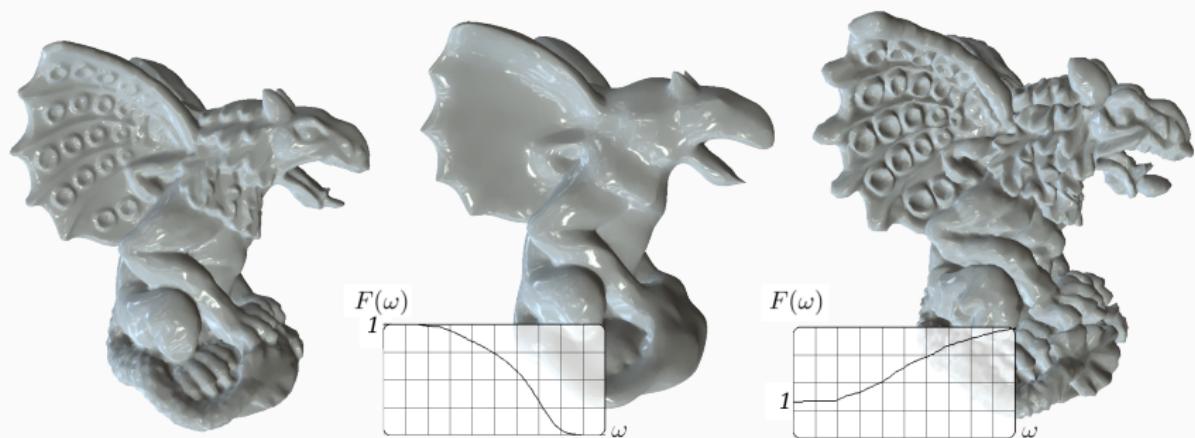


Figure: Original Mesh, Low-pass and enhancement filters [Vallet08].

LAPLACIAN OF THE FOURIER TRANSFORM

if \mathbf{u} is a eigenfunction and λ is a corresponding eigenvalue of a linear differential operator \mathbf{A} then

$$\mathbf{A}\mathbf{u} = \lambda\mathbf{u}, \quad \mathbf{u} \neq 0$$

The basis $\mathbf{e}_w = e^{2\pi iwx} = \cos(2\pi wx) - i \sin(2\pi wx)$ sine and cosine functions of the Fourier transform are eigenfunctions of the Laplacian with eigenvalue λ_w .

$$\Delta(\mathbf{e}_w) = \frac{\partial^2}{\partial x^2} \mathbf{e}_w = -(2\pi w)^2 \mathbf{e}_w = \lambda_w \mathbf{e}_w$$

$$\Delta(\mathbf{e}_w) = \lambda_w \mathbf{e}_w$$

LAPLACE-BELTRAMI OPERATOR

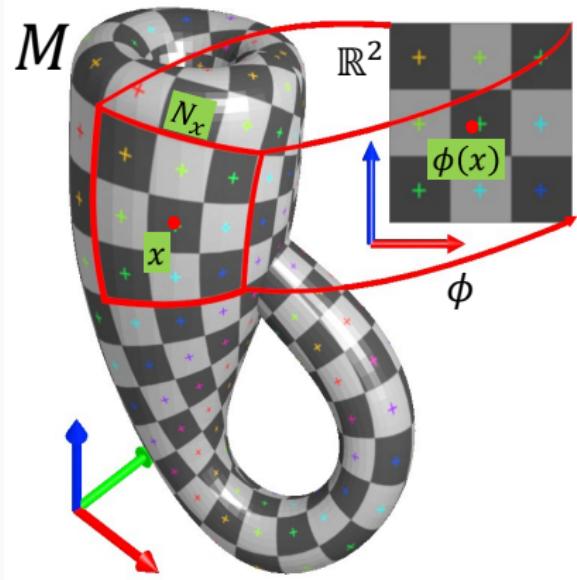
The Laplace-Beltrami operator Δ_M is a differential operator given by the divergence of a gradient field on a **surface**.

$$\begin{aligned}\frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} + \frac{\partial^2 f}{\partial z^2} &= 0 \\ \nabla_M^2 f &= \Delta_M = 0 \\ \Delta_M &= \operatorname{div}(\nabla_M f)\end{aligned}$$

where M is compact and smooth **Surface** (2D-Manifold)

The Laplace-Beltrami is a generalization of the Laplace operator for functions over surfaces and allows to perform an spectral surface analysis

SURFACE



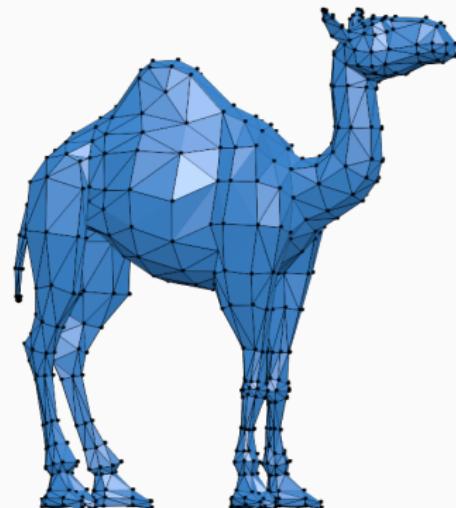
A Surface M is a 2D topological manifold
For which every point $x \in M \subseteq \mathbb{R}^3$
has a neighbourhood N_x homeomorphic to euclidean space \mathbb{R}^2 .

A homeomorphism $\phi : N_x \rightarrow \mathbb{R}^2$

SURFACE DISCRETIZATION WITH POLYGONAL MESHES

In computer graphics a **Surface** is often discretized into a **Polygonal Mesh**

Polygonal Mesh is a set of points that are connected by **triangles** and **quads**.



We need discrete versions of Laplacian operator to work with polygonal meshes.

DISCRETE LAPLACIAN OPERATOR AS A MATRIX EQUATION

The discrete version of Laplace operator for polygonal meshes.

Given a polygonal mesh $\mathbf{M} = [v_1, \dots, v_n]^T$ and $v_i \in \mathbb{R}^3$ we define the Discrete Laplacian operator matrix L with size $n \times n$

$$L(i,j) = \begin{cases} w_{ij} & \text{if } j \in N(v_i) \\ \sum w_{ik} & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}$$

w_{ij} are the weights between the vertex v_i and vertex v_j . The weights are defined depending of the polygonal structure and application.

$N(v_i)$ is the 1-ring neighborhood with shared face to vertex v_i .

DESIRED PROPERTIES FOR LAPLACIAN MATRIX

This properties ensure the construction of eigenstructure of Laplacian matrix and then those eigenvectors of L matrix are a orthogonal basis of \mathbb{R}^n

- Symmetry.
- Square.
- Locality.
- Positive Weights.
- Positive Semi-Definiteness.
- Convergence.

EIGENSTRUCTURE OF LAPLACIAN MATRIX

$$L\mathbf{e}_i = \lambda_i \mathbf{e}_i, \quad \mathbf{e}_i \neq 0$$

- The eigenvector \mathbf{e}_i of L is a *natural vibration* of the mesh [Taubin95].
- The frequency of the wave \mathbf{e}_i is the eigenvalue λ_i of L that is the *natural frequency* [Taubin95].

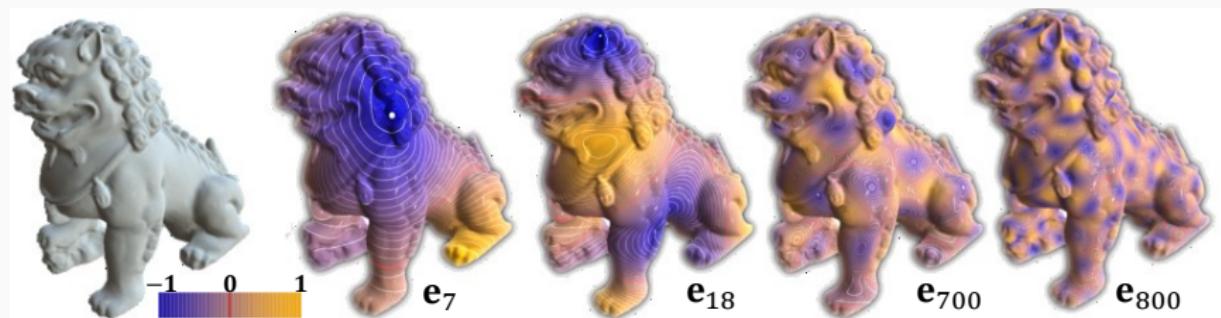


Figure: Color values are the amplitude of a wave \mathbf{e}_i projected on the mesh [Vallet08].

MESH RECONSTRUCTION

The mesh \mathbf{M} can be reconstructed from your spectral decomposition in the analogous way that Fourier transform.

$$\mathbf{M} = \sum_{i=1}^n \langle \mathbf{e}_i, \mathbf{M} \rangle \mathbf{e}_i$$

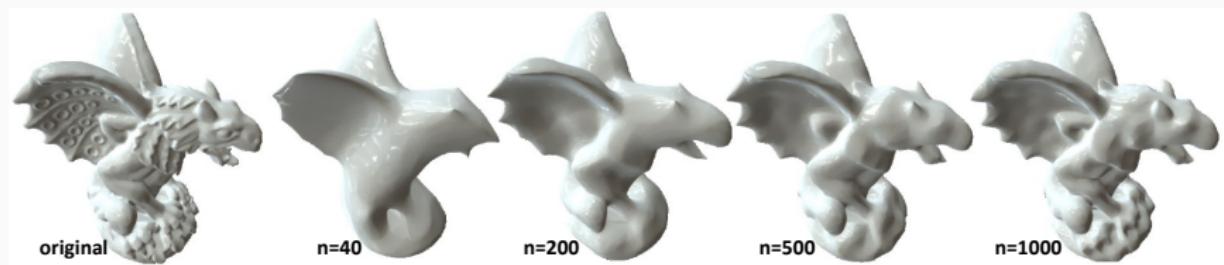


Figure: Mesh reconstruction with first n eigenvectors of the Discrete Laplacian Operator [Levy10].

WEIGHTS FOR LAPLACIAN OPERATOR

Desired property:

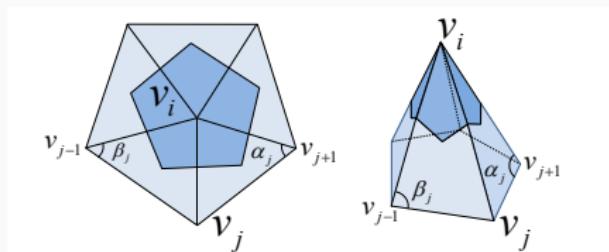
Non-negative weights ω_{ij} for $i \neq j$ ensure a positive semi-definite matrix.

- Umbrella Operator $w_{ij} = \begin{cases} 1 & \text{if } j \in N(v_i) \\ 0 & \text{else} \end{cases}$
- Fujiwara's Operator $w_{ij} = \begin{cases} \frac{1}{\|v_j - v_i\|} & \text{if } j \in N(v_i) \\ 0 & \text{else} \end{cases}$

1999 DESBRUN'S OPERATOR FOR TRIANGLE MESHES

This operator only work with meshes composed only by **triangles**.

Desbrun's operator is the discretized version of the Laplace-Beltrami operator.



$$w_{ij} = \frac{1}{4A_i} (\cot \alpha_j + \cot \beta_j)$$

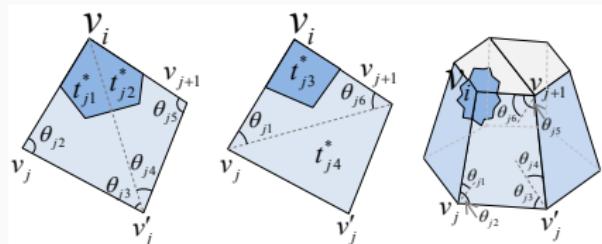
where A_i is area of the 1-ring neighborhood.

α and β are the opposite angles to edge between vertex v_i and vertex v_j .

2011 XION'S MLBO OPERATOR FOR QUAD MESHES

This operator only work with meshes composed only by **quads**.

Xion's operator is the mean of Desbrun's operator for all possible triangulations for quad meshes.



$$w_{ij} = \frac{1}{4A_i} (\cot \theta_{(j-1)1} + \cot \theta_{(j-1)4} + \cot \theta_{j3} + \cot \theta_{j6})$$

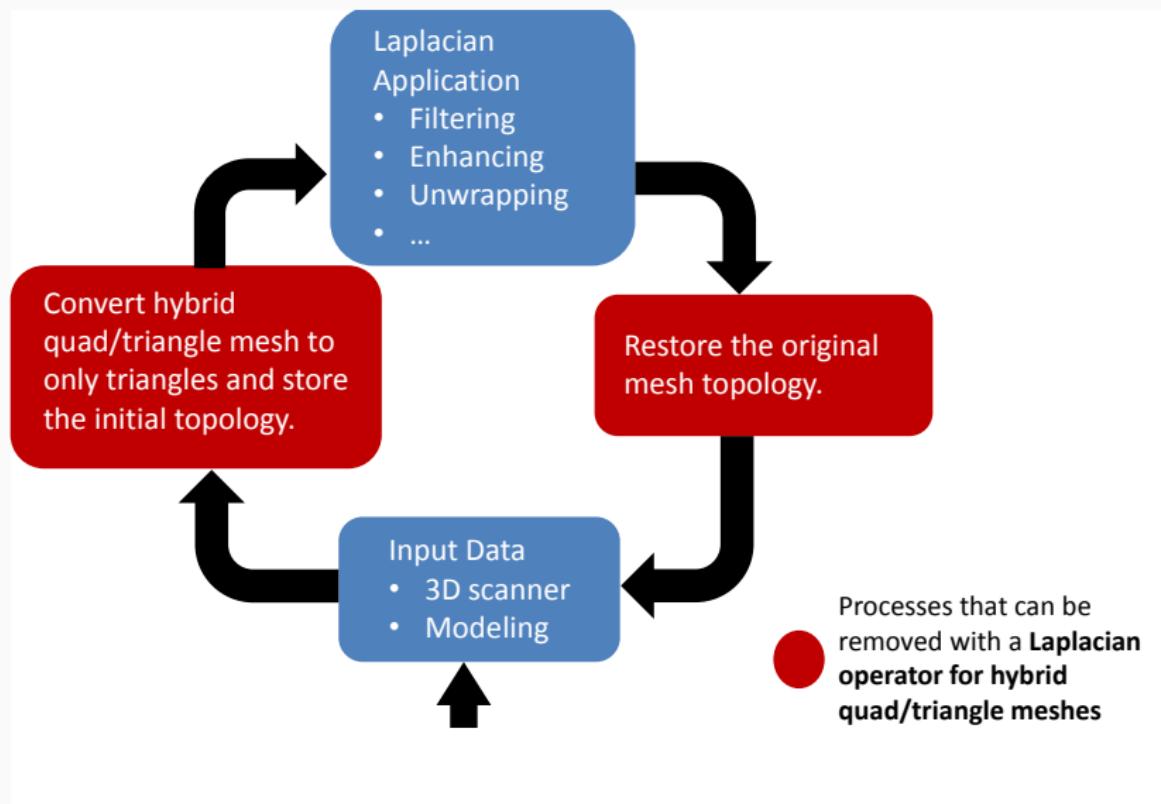
$$w_{ij'} = \frac{1}{2A_i} (\cot \theta_{j2} + \cot \theta_{j5})$$

$$A_i = \frac{1}{2} \sum_{j \in i^*} (A(t_{j2}) + A(t_{j4}) + A(t_{j3}))$$

where w_{ij} are the weights of neighbors that share an edge with v_i and $w_{ij'}$ are the weights of neighbors that share face with v_i .

LAPLACIAN APPLICATIONS ON 3D MODELING PROCESS

General 3D modeling process on polygonal meshes.



EDGE LOOPS FOR FACES

Hybrid quad/triangle meshes are necessary by artists to 3D modeling

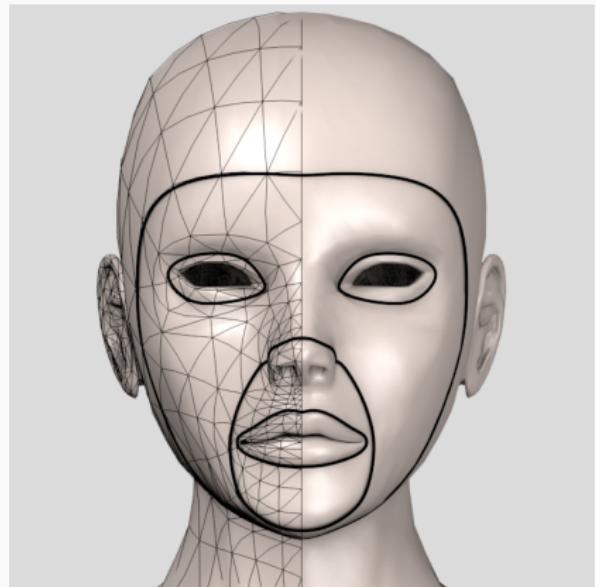


Figure: Triangle Mesh

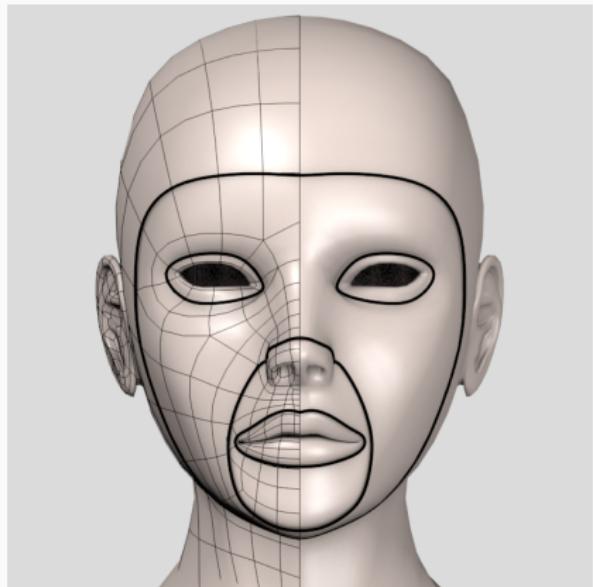
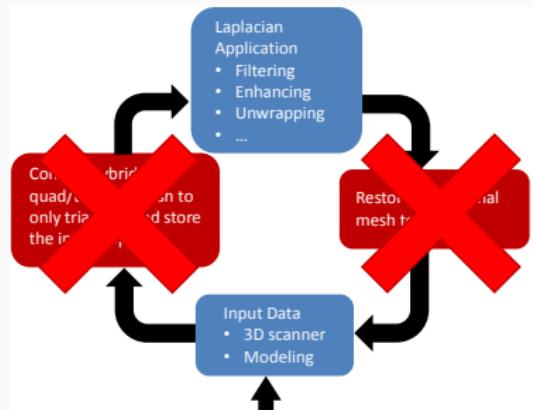


Figure: Hybrid Quad/Triangle Mesh

PROPOSED METHOD

OUR PROPOSAL

We propose an extension of the discrete Laplace-Beltrami operator to work with hybrid quad/triangle meshes and eliminates the need of triangulate the mesh and also allows the preservation of the original topology.



GRADIENT OF VORONOI AREA

The area change produced by the movement of v_i is called the gradient of Voronoi region [Pinkall93, Desbrun99]

$$\nabla A = \frac{1}{2} \sum_j (\cot \alpha_j + \cot \beta_j) (v_i - v_j)$$

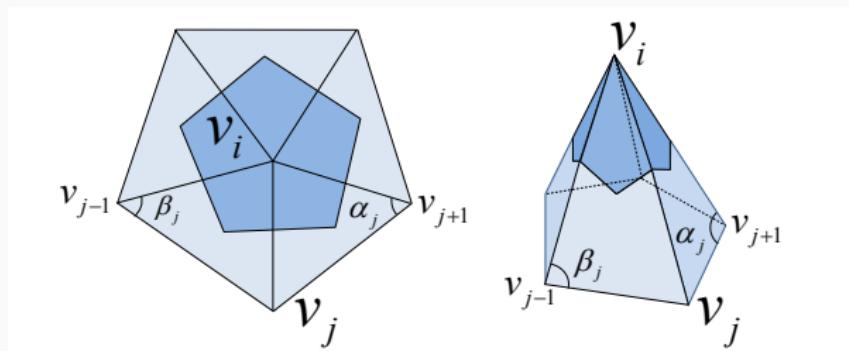


Figure: Area of the Voronoi region around v_i in dark blue. v_j belong to the first neighborhood around v_i . α_j and β_j opposite angles to edge $\overrightarrow{v_j - v_i}$.

MEAN CURVATURE OF SURFACES

In 2D, curvature κ at a given point p on a circle with radius R is defined as [Meyer01]

$$\kappa = \frac{1}{R}$$

In 3D-surface, *sectional curvature* at a given point p is the intersection of a surface with a plane parallel to a normal \mathbf{n} of p [Meyer01].

The *Mean Curvature* κ_H at a given point p is defined as

$$\kappa_H = \frac{\kappa_1 + \kappa_2}{2}$$

where κ_1 and κ_2 are the maximal and minimal sectional curvatures known as *principal curvatures* [Meyer01].

DISCRETE MEAN CURVATURE NORMAL

Gradient of voronoi region

$$\nabla A = \frac{1}{2} \sum_j (\cot \alpha_j + \cot \beta_j) (v_i - v_j)$$

If the gradient of Voronoi region is normalized by the total area of the 1-ring neighborhood around v_i , we obtained a *discrete mean curvature normal*.

$$2\kappa_H \mathbf{n} = \frac{\nabla A}{A}$$

The *Laplace Beltrami operator* Δ is used for measuring the mean curvature normal of the surface S [Pinkall93].

$$\Delta S = 2\kappa_H \mathbf{n}$$

MEAN AVERAGE AREA

Xiong's define the mean average area of voronoi region of quad around v_i as the average of areas of primal and dual triangulations.

$$Area(Q) = \frac{Area_1 + Area_2}{2} = \frac{A(t_{j1}^*) + A(t_{j2}^*) + A(t_{j3}^*)}{2}$$

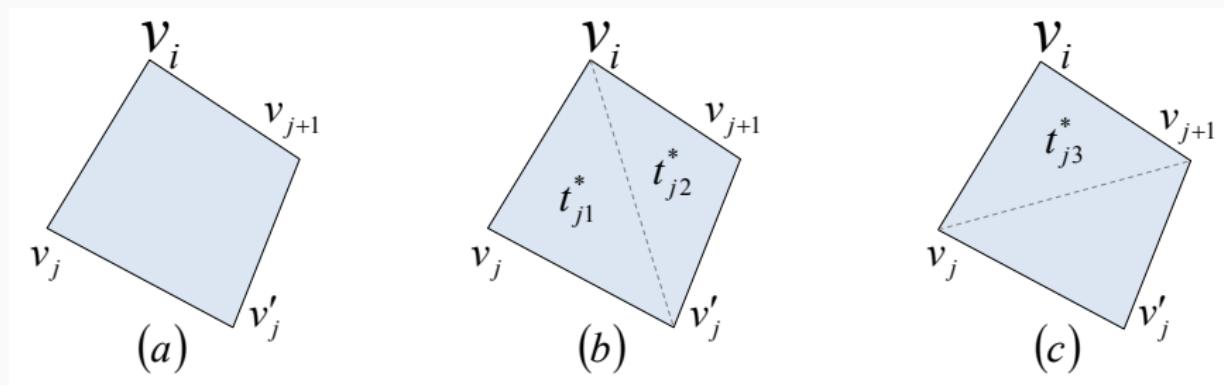


Figure: (a) Original quad. (b) Primal triangulation around v_i is $t_{j1}^* \equiv \Delta v_i v_j v'_j$, $t_{j2}^* \equiv \Delta v_i v'_j v_{j+1}$ (b) Dual triangulation around v_i is $t_{j3}^* \equiv \Delta v_i v_j v_{j+1}$.

LAPLACE BELTRAMI OPERATOR FOR HYBRID QUAD/TRIANGLE MESHES

Our proposal

Given a hybrid mesh $M = \{V, Q, T\}$, with vertices V , quads Q and triangles T , define the mean average area of all triangulations around v_i as

$$A(v_i) = \sum_{j=1}^m A(q_j) + \sum_{k=1}^r A(t_k) \quad (1)$$

where $q_j \in Q_{v_i}$ and $t_k \in T_{v_i}$

Applying Xiong's mean average area to (1)

$$A(v_i) = \frac{1}{2} \sum_{j=1}^m \left[A(t_{j1}^*) + A(t_{j2}^*) + A(t_{j3}^*) \right] + \sum_{k=1}^r A(t_k) \quad (2)$$

LAPLACE BELTRAMI OPERATOR FOR HYBRID QUAD/TRIANGLE MESHES

Our proposal

Applying the gradient operator to (2)

$$\nabla A(v_i) = \frac{1}{2} \sum_{j=1}^m [\nabla A(t_{j1}^*) + \nabla A(t_{j2}^*) + \nabla A(t_{j3}^*)] + \sum_{k=1}^r \nabla A(t_k) \quad (3)$$

Using Desbrun's equation to compute the gradient to (3), we have

$$\nabla A(t_{j1}^*) = \frac{\cot \theta_{j3}(v_j - v_i) + \cot \theta_{j2}(v'_j - v_i)}{2}$$

$$\nabla A(t_{j2}^*) = \frac{\cot \theta_{j5}(v'_j - v_i) + \cot \theta_{j4}(v_{j+1} - v_i)}{2}$$

$$\nabla A(t_{j3}^*) = \frac{\cot \theta_{j6}(v_j - v_i) + \cot \theta_{j1}(v_{j+1} - v_i)}{2}$$

$$\nabla A(t_k) = \frac{\cot \alpha_k(v_k - v_i) + \cot \beta_k(v_k - v_i)}{2}$$

LAPLACE BELTRAMI OPERATOR FOR HYBRID QUAD/TRIANGLE MESHES

Our proposal

Therefore (3) can be rewritten as

$$\nabla A(v_i) = \sum_{j=1}^n w_{ij} (v_j - v_i) \quad (4)$$

where v_j are the neighbors of v_i

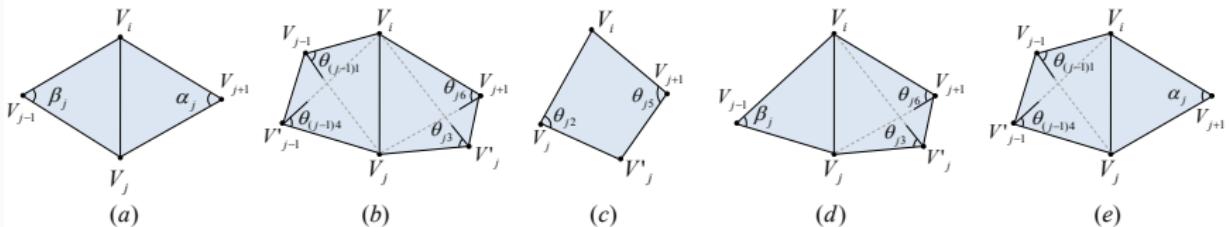
Using the relationship between Laplace-Beltrami operator Δ and Mean curvature normal in (4) we define the **Triangle Quad Laplace Beltrami Operator TQLBO** as

$$\Delta(v_i) = 2\kappa_H \mathbf{n}_i = \frac{\nabla A(v_i)}{Area_i} = \frac{1}{2Area_i} \sum_{j=1}^n w_{ij} (v_j - v_i) \quad (5)$$

where $Area_i$ is the area of the 1-ring neighborhood around v_i

WEIGHTS FOR TQLBO - OUR PROPOSAL

We define the weights of the TQLBO based on five simple cases



The 5 basic triangle-quad cases with a vertex V_i and the relationship with V_j and V'_j .

$$w_{ij} = \begin{cases} (\cot \alpha_j + \cot \beta_j) & \text{case } a. \\ \frac{1}{2} (\cot \theta_{(j-1)1} + \cot \theta_{(j-1)4} + \cot \theta_{j3} + \cot \theta_{j6}) & \text{case } b. \\ (\cot \theta_{j2} + \cot \theta_{j5}) & \text{case } c. \\ \frac{1}{2} (\cot \theta_{j3} + \cot \theta_{j6}) + \cot \beta_j & \text{case } d. \\ \frac{1}{2} (\cot \theta_{(j-1)1} + \cot \theta_{(j-1)4}) + \cot \alpha_j & \text{case } e. \end{cases} \quad (6)$$

LAPLACE OPERATOR AS A MATRIX EQUATION

Our proposal

We define a TQLBO as a matrix equation

$$L(i,j) = \begin{cases} -\frac{1}{2A_i} w_{ij} & \text{if } j \in N(v_i) \\ \frac{1}{2A_i} \sum_{k \in N(v_i)} w_{ik} & \text{if } i = j \\ 0 & \text{otherwise} \end{cases} \quad (7)$$

Where L is a $n \times n$ matrix, n is the number of vertices, $N(v_i)$ is the 1-ring neighborhood with shared face to v_i , A_i is the ring area around v_i .

EVALUATION AND RESULTS

A standard diffusion process is used.

$$\frac{\partial V}{\partial t} = \lambda L(V)$$

To solve this equation, implicit integration is used as well as a normalized version of TQLBO matrix

$$(I - |\lambda dt| W_p L) V' = V^t$$

$$V^{t+1} = V^t + \text{sign}(\lambda) (V' - V^t)$$

where L is the TQLBO, V' are the smoothing vertices, V^t are the actual vertices positions, W_p is a diagonal matrix with vertex weights, and λdt is the inflate factor.

SCULPTING WITH ENHANCING FILTER

Inflate Brush

Real-time brushes require the Laplacian matrix is constructed with the vertices that are within the sphere radius defined by the user, reducing the matrix to be processed.

$$L(i,j) = \begin{cases} -\frac{w_{ij}}{\sum\limits_{j \in N(v_i)} w_{ij}} & \text{if } \|v_i - u\| < r \wedge \|v_j - u\| < r \\ 0 & \text{if } \|v_i - u\| < r \wedge \|v_j - u\| \geq r \\ \delta_{ij} & \text{otherwise} \end{cases}$$

Where $v_j \in N(v_i)$, u is the sphere center of radius r . The matrices should remove rows and columns of vertices that are not within the radius.

COMPUTE THE MINIMAL SURFACE WITH TQLBO RESULTS

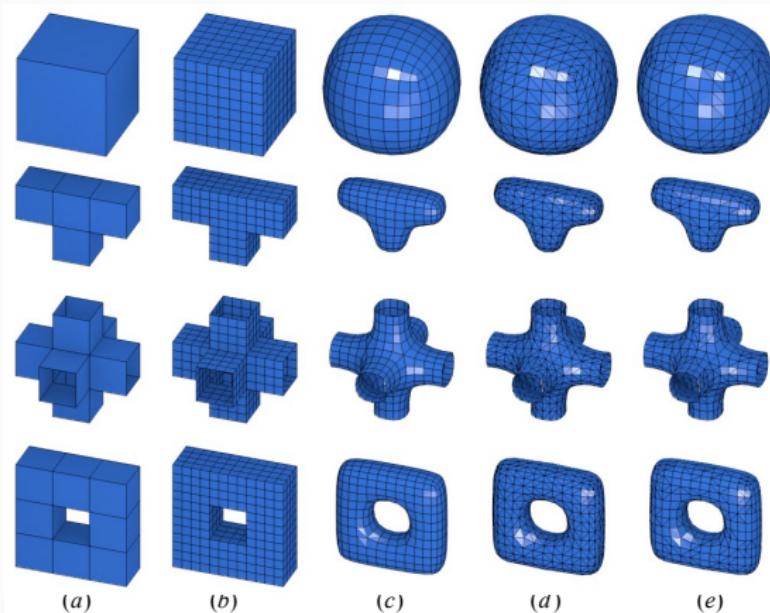


Figure: (a) Original Model. (b) Simple subdivision. (c), (d) (e) Laplacian smoothing with $\lambda = 7$ and 2 iterations: (c) for quads, (d) for triangles, (e) for triangles and quads random chosen.

SHAPE INFLATION WITH ENHANCING FILTER RESULTS

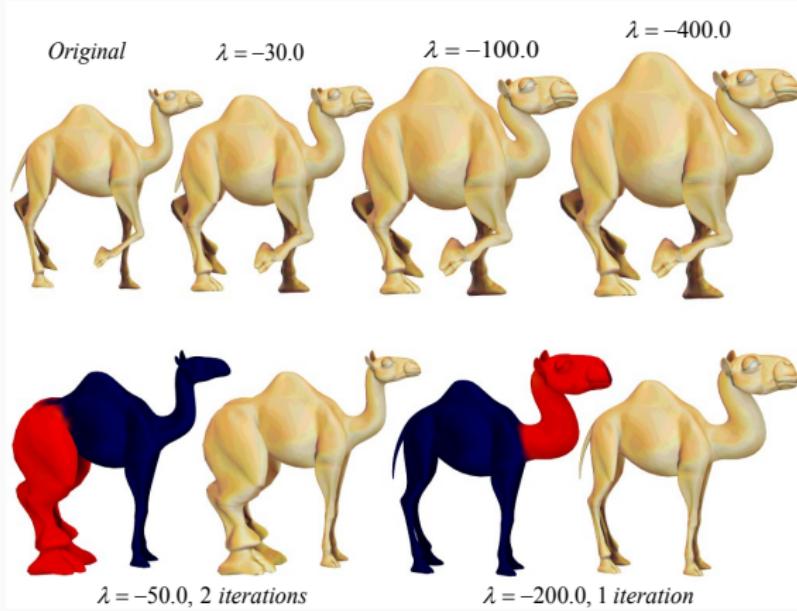


Figure: Top row: Original camel model in left. Shape inflation with $\lambda = -30.0$, $\lambda = -100.0$, $\lambda = -400.0$. Bottom row: Shape inflation with weight vertex group, $\lambda = -50.0$ and 2 iterations for the legs, $\lambda = -200.0$ and 1 iteration for the head and neck.

INFLATE BRUSH RESULTS

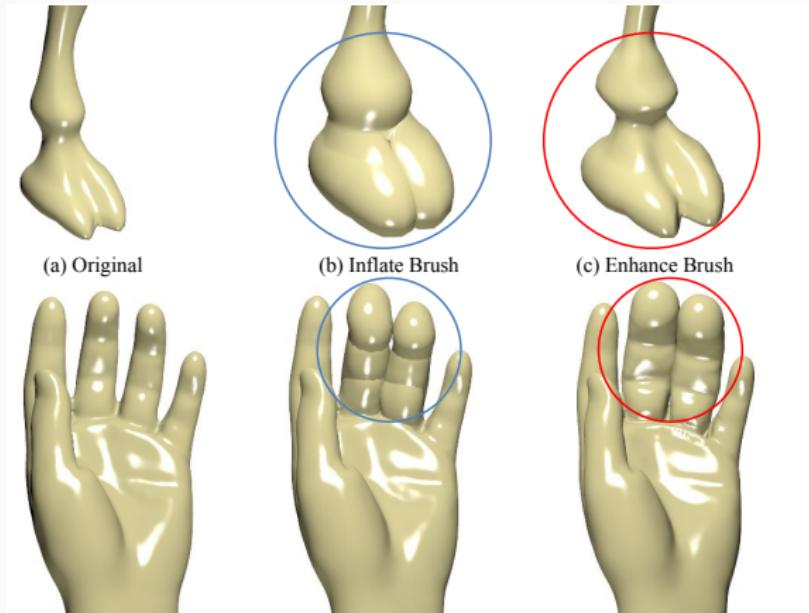


Figure: Top row: (a) Leg Camel, (b) Traditional inflate brush for leg into blue circle, (c) Shape inflation brush for leg into red circle. Bottom row: (a) Hand, (b) Traditional inflate brush for fingers into blue circle, (c) Shape inflation brush for fingers in red circle.

INFLATE BRUSH PERFORMANCE

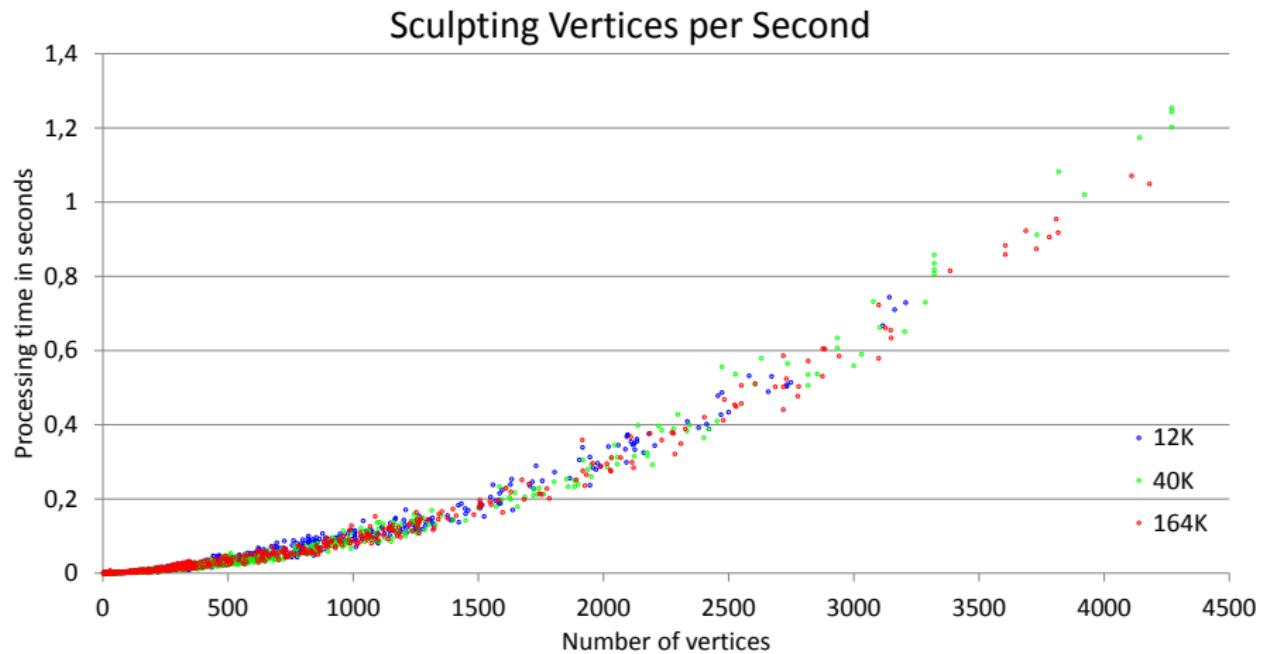


Figure: Performance of our dynamic shape inflation brush in terms of the sculpted vertices per second. Three models with 12K, 40K, 164K vertices used for sculpting in real time.

PRODUCTS

Shape Inflation With an Adapted Laplacian Operator For Hybrid Quad/Triangle Meshes

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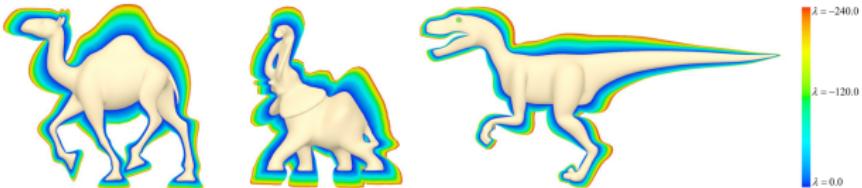


Fig. 1. A set of 48 successive shapes enhanced, from $\lambda = 0.0$ in blue to $\lambda = -240.0$ in red, with steps of -5.0 .

Abstract—This paper proposes a novel modeling method for a hybrid quad/triangle mesh that allows to set a family of possible shapes by controlling a single parameter, the global curvature. The method uses an original extension of the Laplace Beltrami operator that efficiently estimates a curvature parameter which is used to define an inflated shape after a particular operation performed in certain mesh points. Along with the method, this work presents new applications in sculpting and modeling, with subdivision of surfaces and weight vertex groups. A series of graphics examples demonstrates the quality, predictability and flexibility of the method in a real production environment with software Blender.

Keywords-laplacian smooth; curvature; sculpting; subdivision surface

[12]. Nevertheless these methods are difficult to deal with since they require a large number of parameters and a very tedious customization. Instead, the presented method requires a single parameter that controls the global curvature, which is used to maintain realistic shapes, creating a family of different versions of the same object and therefore preserving the detail of the original model and a realistic appearance.

Interest in meshes composed of triangles and quads has lately increased because of the flexibility of modeling tools such as Blender 3D [13]. Nowadays, many artists use a manual connection of a couple of vertices to perform animation processes and interpolation [14]. It is then of paramount

2ND PRODUCT - AWARDED INTERNSHIP

Mesh smoothing based on curvature flow operator in a diffusion equation

Sponsor: Google Inc - Google Summer of Code 2012 program

Project: Mesh smoothing based on curvature flow operator in a diffusion equation

Synopsis: This project proposes a new and robust mesh smoothing tool that remove the noise of the surfaces of models captured with 3d scanners, zcameras among others.

Blender software is an open source 3D application for modeling, rendering, composing, video editing and game creation.

LAPLACIAN SMOOTH TOOL FOR BLENDER

Mesh smoothing based on curvature flow operator in a diffusion equation

We define a Laplacian matrix for mesh smoothing with support for hybrid quad/triangle meshes with holes as

$$L(i, j) = \begin{cases} -\frac{1}{2A_i} W_{ij} & \text{if } j \in N(v_i) \wedge v_i \notin \text{Boundary} \\ \frac{1}{2A_i} \sum_{j \in N(v_i)} W_{ij} & \text{if } i = j \wedge v_i \notin \text{Boundary} \\ -\frac{1}{\|v_i - v_j\|} & \text{if } j \in N(v_i) \wedge \{v_i, v_j\} \in \text{Boundary} \\ \frac{2}{E_i} \sum_{j \in N(v_i)} \frac{1}{\|v_i - v_j\|} & \text{if } i = j \wedge \{v_i, v_j\} \in \text{Boundary} \\ 0 & \text{otherwise} \end{cases}$$

w_{ij} is the TQLBO defined in equation (6)

$$E_i = \sum_{j \in N(v_i)} e_{ij} .$$

USER INTERFACE

Mesh smoothing based on curvature flow operator in a diffusion equation

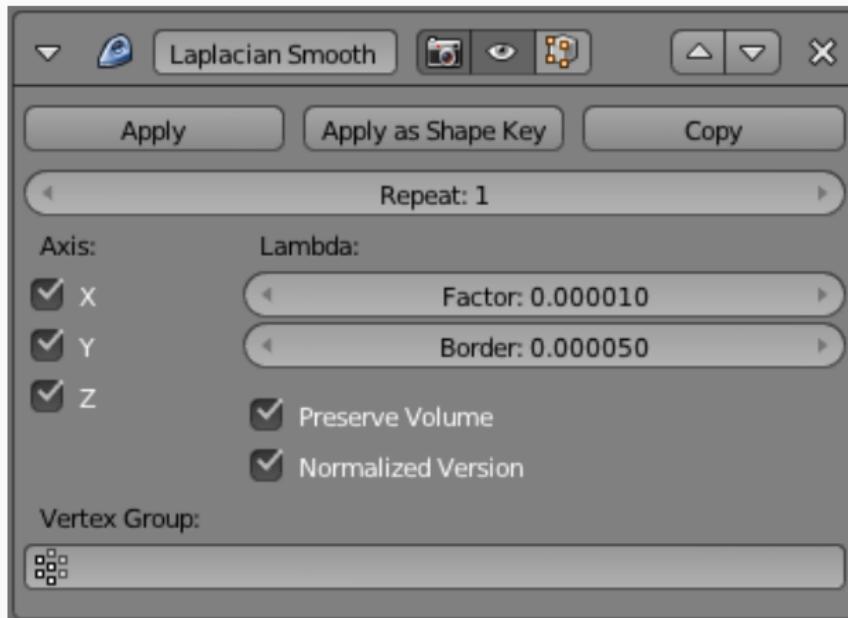


Figure: Panel inside blender user interface of the Laplacian Smooth modifier tool.

RESULTS

Mesh smoothing based on curvature flow operator in a diffusion equation

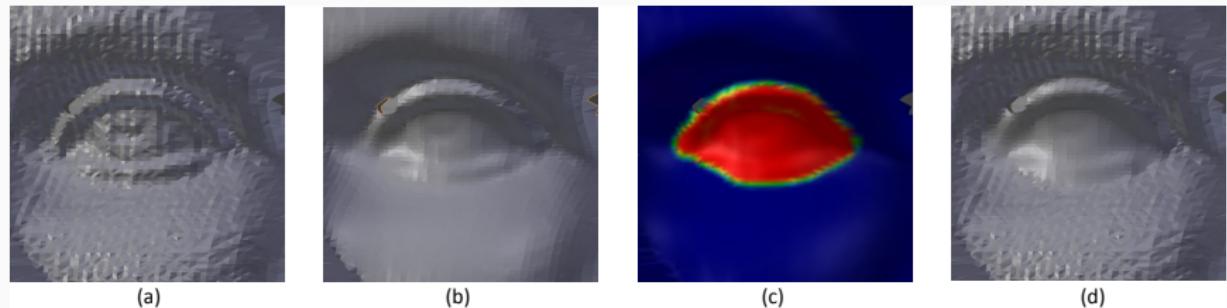


Figure: Use of weights per vertex to constrain the effect of mesh smoothing. (a) Original Model. (b) Smoothing with $\lambda = 1.5$ (c) red vertices $weight = 1.0$, blue vertices $weight = 0.0$. (d) Smoothing with $\lambda = 2.5$. The red vertices were the only vertices smoothed.

RESULTS

Mesh smoothing based on curvature flow operator in a diffusion equation

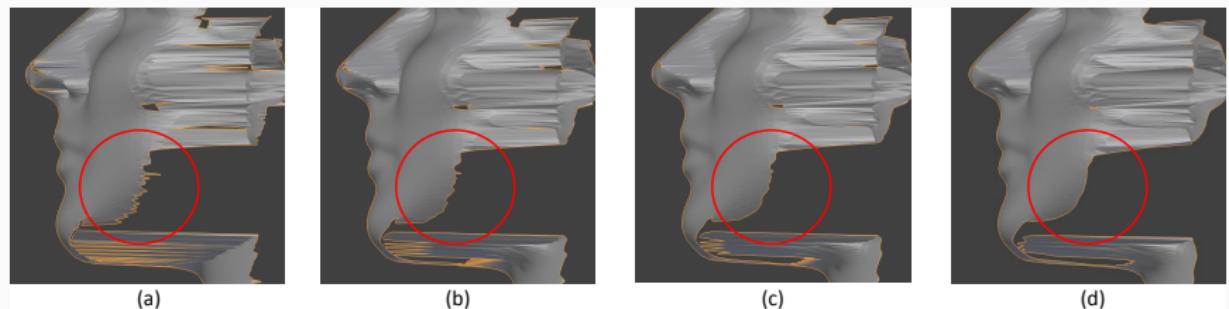


Figure: Smoothing boundary changing λ_{Border} factor. (a) Original Model. (b) Smoothing $\lambda_{Border} = 1.0$. (c) Smoothing $\lambda_{Border} = 2.5$ (d) Smoothing with $\lambda_{Border} = 10.0$.

3RD PRODUCT - AWARDED INTERNSHIP

Mesh Editing with Laplacian Deform

Sponsor: Google Inc - Google Summer of Code 2013 program

Project: Mesh Editing with Laplacian Deform

Synopsis: This project proposes a new tool that allows to pose a mesh while preserving geometric details of the surface.

Blender software is an open source 3D application for modeling, rendering, composing, video editing and game creation.

DIFFERENTIAL COORDINATES

Mesh Editing with Laplacian Deform

$$\delta_i = \sum_{j=1}^m w_{ij} (v_i - v_j) \quad (8)$$

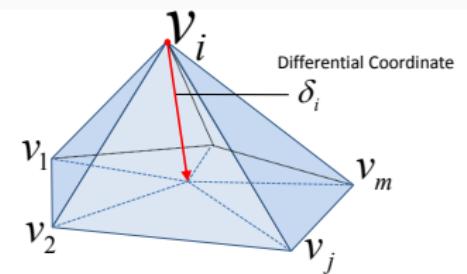


Figure: Difference between v_i and the center of mass of its neighbors v_1, \dots, v_m .

w_{ij} is the TQLBO defined in equation (6)

LAPLACIAN DEFORM

Mesh Editing with Laplacian Deform

The linear system for finding the new pose of a mesh is.

$$\begin{bmatrix} L \\ W_c \end{bmatrix} V = \begin{bmatrix} \delta \\ W_c C \end{bmatrix} \quad (9)$$

L is a matrix that used our TQLBO defined in equation 7.

W_c is a matrix that has only ones in the indices of anchor vertices.

V is the vertices of mesh.

C is a vector with coordinates of anchor vertices after several manual transformations.

δ are the differential coordinates defined in equation 8.

USER INTERFACE

Mesh Editing with Laplacian Deform

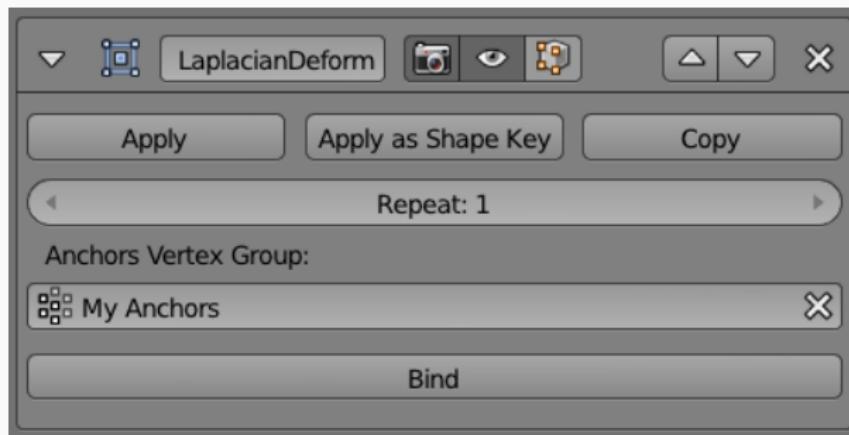


Figure: Panel inside Blender user interface of the Laplacian Deform modifier tool.

RESULTS

Mesh Editing with Laplacian Deform

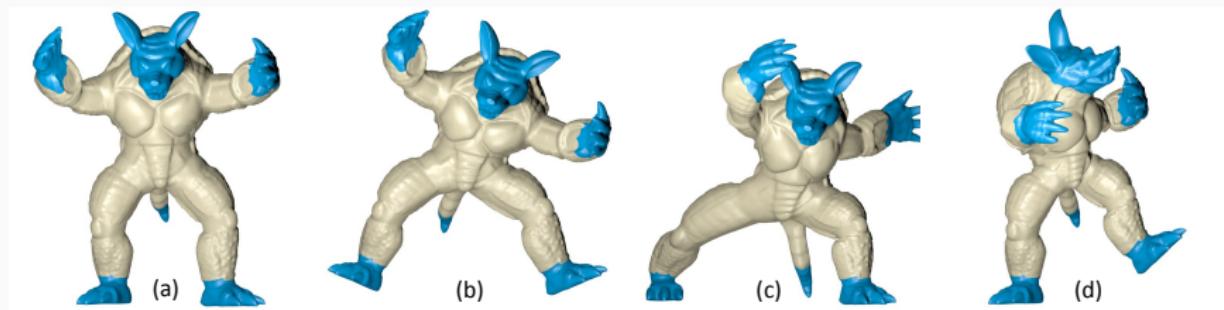


Figure: Anchor vertices in blue. (a) Original Model, (b,c,d) new poses only change the anchor-vertices, the system finds positions for vertices in yellow.

RESULTS

Mesh Editing with Laplacian Deform

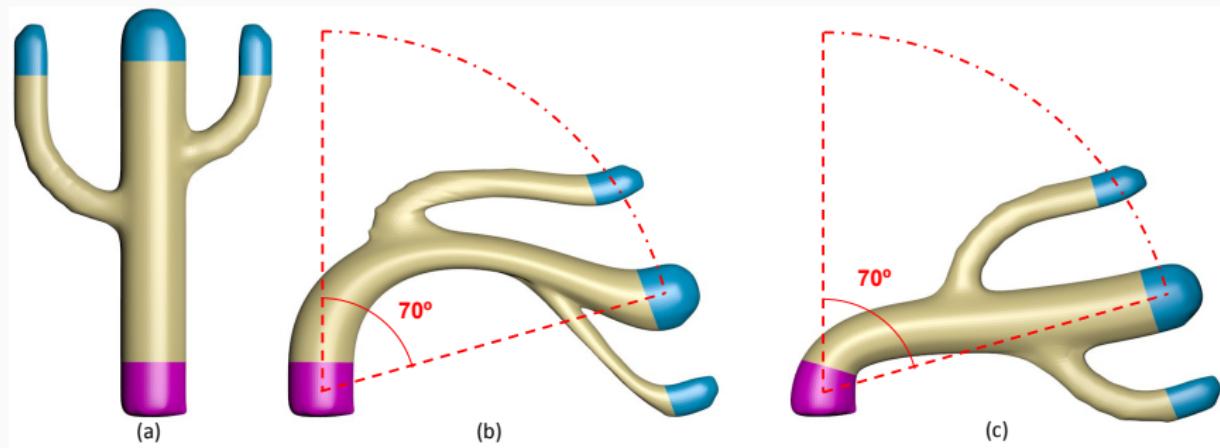


Figure: (a) Original cactus model. (b) Blue segments are rotated 70° to the right and afterwards a basic interpolation is applied to the parts in yellow (c) Blue segments are rotated 70° to the right and afterwards a Laplacian deform tool is applied to the parts in yellow.

RESULTS

Mesh Editing with Laplacian Deform

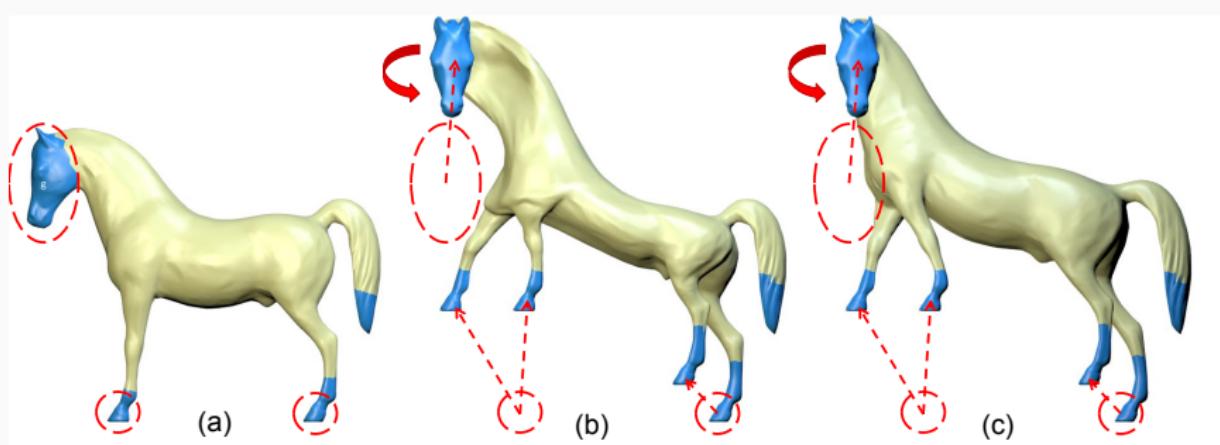


Figure: (a) Original Horse model. (b) The blue segments are translated and rotated and then basic interpolation is applied to the yellow parts (c) The blue segments are translated and rotated and then the Laplacian Deform tool is applied to the yellow parts.

4TH PRODUCT - POSTER

6th International Seminar on Medical Image Processing and Analysis SIPAIM 2010

Análisis Experimental de la Extracción del Esqueleto por Contracción con Suavizado Laplaciano

Alexander Pizán, Fabio Martínez, Eduardo Romano

Abstract

Este artículo presenta un análisis estadístico del método de extracción del esqueleto por medios de la contracción de un volumen con suavizado Laplaciano. El trabajo realiza una evaluación experimental al problema de la extracción del esqueleto, para medir el rendimiento del método Frenet a través de comparaciones.

Diseño evaluativo: se modela la deformación armada de una persona que realizaba una caminata, a más modelos se le aplicó el esqueleto y se compararon las diferencias en dibujos instantáneos de la articulación, y distintas configuraciones del proceso de simplificación. Los resultados muestran un sólido rendimiento del método Frenet a los buenas evaluaciones de articulación, y muestra preferencia en la fase de simplificación de esqueleto.



Contracción con Suavizado Laplaciano

Este método consiste básicamente una red de polígonos por medio del suavizado laplaciano hasta tener un volumen de un figura 1.

• **Algoritmo:** Se basa en la optimización de un problema de minimización de energía, con los siguientes bloques:

$$W_1(D)^2 + \sum W_2_i B^i - V_i^2$$

- **Operador Laplaciano:** para remover las irregularidades, es decir suavizar las distorsiones de la geometría.
- **Mínima:** Fuerza de atracción que usa los vértices, para mantener información clara de la geometría.
- **Máximos:** Fuerza de contracción que hace la forma bidimensional para volúmenes.



Referencias y Agradecimientos

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Agradecimientos: The original performance data were provided courtesy of the Computer Graphics Group of the NTT CSRA, Japan Research (Cambridge, Japón).

Implementación

Se usó la implementación hecha en C# por Chung *et al.* Para los sujetos, para realizar la extracción del esqueleto de un modelo tridimensional que fue obtenido mediante la captura de imágenes de la persona en movimiento, se usó la cámara de video por Chung *et al.* De este modelo se registró una cámara durante 200 cuadros.

Para evaluar el rendimiento del método Frenet, se usó la cámara de video para describir un esqueleto (ver figura 2) y se clasificaron los sujetos así en figura 3.

En el segundo experimento se seleccionaron algunos punto para obtener la complejidad de la geometría de los esqueletos, medida para el análisis y complemento del método Frenet a transformaciones isométricas de la geometría de un cuerpo.

Resultados

El método logra de forma efectiva el esqueleto de un cuerpo, tanto transformaciones isométricas, en la tabla 2, se recuperaron en promedio 13.71 puntos de los 21 necesarios para la fase de extracción en diferentes tipos de esqueletos. La tasa de error es de 2.86 milésimas de punto, lo que resulta relativamente bueno. La linea que sirve en la figura 3 describe el numero de nodos que tienen falta para conseguir el esqueleto.



Durante la fase de simplificación (ver figura 2) el método no sufre perdida ni en puntos ni en nodos, ya que en la figura 2 se observa que los 24 nodos (linea azul), el método no pierde ninguno de los 24 nodos (linea roja). El método mantiene la estructura de los nodos de 24 nodos y permanece constante al avanzar en la linea roja de la figura 2.

Conclusiones y Trabajo Future

El método de extracción resulta en esqueletos, y tiene bajo impacto frente a cambios sutiles de la geometría, el método puede trabajar de forma automática o a lo largo de los cuadros. El método recupera de forma efectiva los nodos de los esqueletos, sin un mal resultado, demostrando la eficiencia de manejar la pose. El método provee resultados de forma sencilla y esfumada de seguimiento del esqueleto a lo largo del video.

Como trabajo futuro es posible mejorar la recuperación de información haciendo de la cohärenza espacio temporal no perderse en la técnica de extracción, para lograr la recuperación de la información temporal, se debe de usar un algoritmo de optimización que procese las simplificaciones para manejar el número óptimo de nodos con lo cual puede ser representado el esqueleto, haciendo uso de algoritmos de partición de malla.

5TH PRODUCT - POSTER

7th International Seminar on Medical Image Processing and Analysis SIPAIM 2011

Software para la Extracción del Esqueleto por Contracción y Suavizado

Alexander Pineda, Eduardo Romano

Abstract

Este artículo propone un software para el procesamiento, visualización e implementación del esqueleto de malla, más conocido como el esqueleto óseo. El software se implementó con base en un sistema de plug-ins y filtros, se implementó un plugin que contiene un filtro para la extracción del esqueleto por contracción en dirección gradiente con suavizado Laplaciano. El software producido proporciona una plataforma flexible para el diseño e implementación de plug-ins.

Métodos de Suavizado de Mallas

Los métodos para suavizar mallas reducen el ruido, o permiten iterativamente eliminar frecuencias altas presentes en el muestreo tridimensional de los modelos.

Método Laplaciano

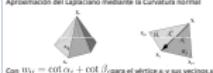
La idea básica consiste en mover un vértice en la $Eq(1)$ en la misma dirección del Laplaciano.

La ecuación 1 se implementa como la ecuación de diferencias hacia adelante es: $Eq(2) \quad X_{i+1} = (I + \lambda L)X_i$. Donde X es el conjunto de vértices, L es el Laplaciano, y $\lambda \in \mathbb{R}$ es la velocidad de difusión.

Y la aproximación discreta de la ecuación 2 es:

$$Eq(3) \quad L(x_i) = \sum_j \{x_i - x_j\} \quad x_j \in \text{Vecinos}(x_i)$$

Aproximación del Laplaciano mediante la Curvatura normal



Con $W_{ij} = \cot(\alpha_j) + \cot(\beta_j)$ para el vértice i y sus vecinos j .

Software Skeletonizer



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Implementación para la extracción del esqueleto

La Esqueletización reduce la dimensionalidad y representa un cuerpo como un estructura unidimensional.

El esqueleto puede ser obtenido suavizando la malla pero bajo dos restricciones que da prioridad al Laplaciano y λ que mantiene los vértices en su localización original.

Extracción del esqueleto: $\begin{bmatrix} W_{11}L \\ W_{22}L \end{bmatrix} X_{i+1} = \begin{bmatrix} 0 \\ W_{ii}K_i \end{bmatrix}$

Donde $L(X) = \text{Sustitución de Laplaciano con } w_i$, $w_{ii} = \cot(\alpha_i) + \cot(\beta_i)$ basado en la curvatura de fuga

Y la nueva restricción propuesta en este trabajo



Tratar de suavizar los vértices a lo largo de la linea

La distancia del punto a la linea: $d_i = \sqrt{P_i^2 + P_0^2}$, donde P_0 := el extremo (line, $P_0) = \frac{(P_0 - P_1)(x_i - P_1)}{(P_1 - P_0)}$

Cada punto en un plano satisface esta ecuación

$$P_0 = \alpha x + \beta y + \gamma z + d_0 = 0.$$

Resultados



Los vértices se pueden mover a lo largo de la linea.

El esqueleto tiene muchas ramas.

El esqueleto tiene muchas vueltas.

La solución debe ser restringida a una región particular de la linea.

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Agradecimientos: The captured performance data were provided courtesy of the Computer Graphics Group of the MIT CSAIL Vision Research (Cambridge, USA).

SKELETON EXTRACTION

Software para la Extracción del Esqueleto por Contracción y Suavizado

Au et al. [Au2008] propose the next system of equations to iteratively contract the mesh until the volume is zero, next a simplification process is done and a skeleton appears.

$$\begin{bmatrix} W_L L \\ W_H \end{bmatrix} X_{t+1} = \begin{bmatrix} 0 \\ W_H X_t \end{bmatrix} \quad (10)$$

L is a matrix that used our TQLBO defined in equation 7.

W_L is a diagonal weighting matrix for the smoothing factor.

W_H is a diagonal weighting matrix for the attraction constraint factor.

A_i^t and A_i^0 are the current area and initial area of the ring surrounding x_i .

UNDESIRABLE DISPLACEMENT OF NODES FROM ROTATIONAL CENTER

Software para la Extracción del Esqueleto por Contracción y Suavizado

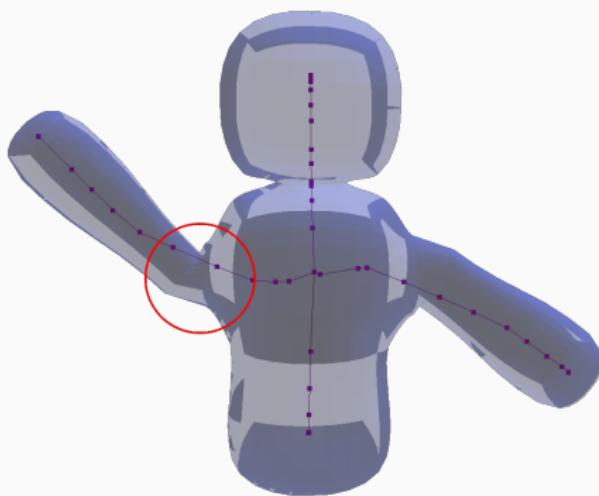


Figure: Skeleton that have a node outside of rotational center.

CONTRIBUCIÓN

Software para la Extracción del Esqueleto por Contracción y Suavizado

The basic idea is to move the vertices along a normal line, estimated at every vertex, based on the average of the normals of the faces.

This constraint eliminates the need to adjust the final skeleton.

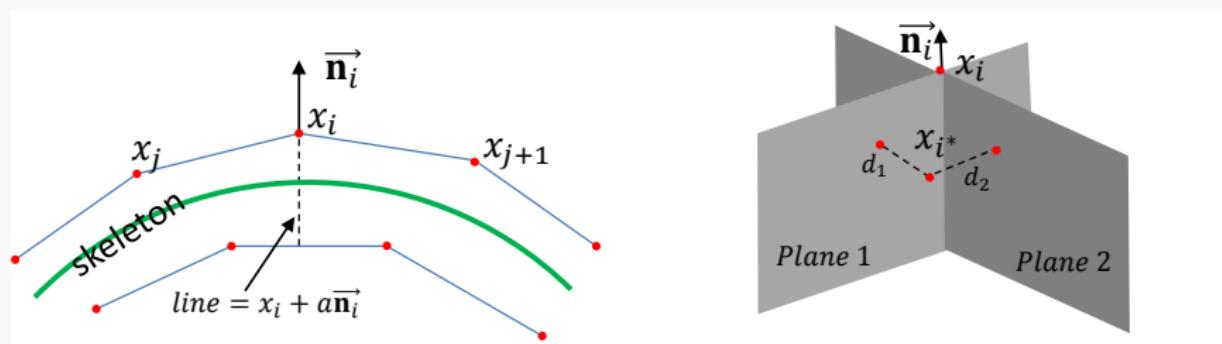


Figure: Left: The vertex x_i moves along the line constraint. Right: the distance of vertex x_i to plane 1 and plane 2 when the position in every iteration changes.

THE DISTANCE EQUATION OF POINT p TO PLANE Π

Software para la Extracción del Esqueleto por Contracción y Suavizado

The plane equation

$$ax + b_0y + c_0z + d_0 = 0$$

The distance of point $P_0 = \{x_0, y_0, z_0\}$ to a plane $\Pi = ax + b_0y + c_0z + d_0$.

$$|\Pi - P_0| = \frac{|ax_0 + by_0 + cz_0 + d|}{\sqrt{a^2 + b^2 + c^2}} \quad (11)$$

THE NEW SYSTEM OF EQUATIONS PROPOSED

Software para la Extracción del Esqueleto por Contracción y Suavizado

$$\begin{bmatrix} W_L L \\ W_H \\ \Pi_1 \\ \Pi_2 \end{bmatrix} X_{t+1} = \begin{bmatrix} 0 \\ W_H X_t \\ -D_1 \\ -D_2 \end{bmatrix} \quad (12)$$

Π_1 and Π_2 are matrix that contain a, b, c values of the plane equation for every vertex.

D_1 and D_2 are the vectors with d values of the plane equation for every vertex.

USER INTERFACE

Software para la Extracción del Esqueleto por Contracción y Suavizado

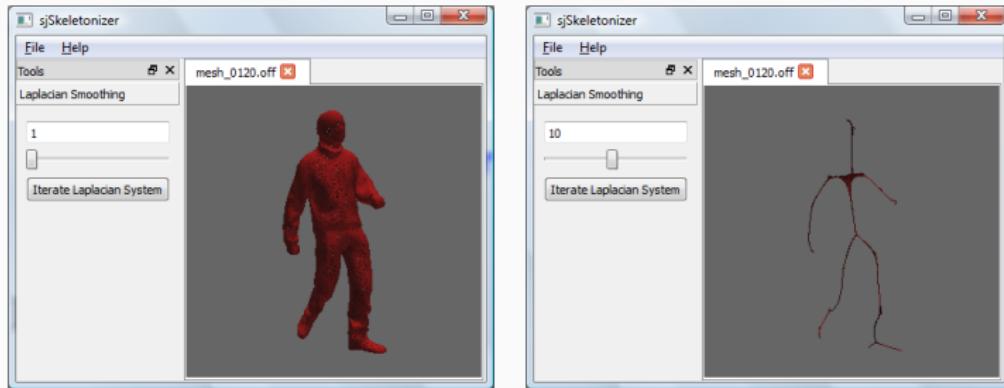


Figure: sjSkeletonizer is our prototype software application

As a result of this work, the software sjSkeletonizer permits the processing, visualization and extraction of the skeleton from polygonal hybrid meshes composed of triangles and quads.

RESULTS

Software para la Extracción del Esqueleto por Contracción y Suavizado

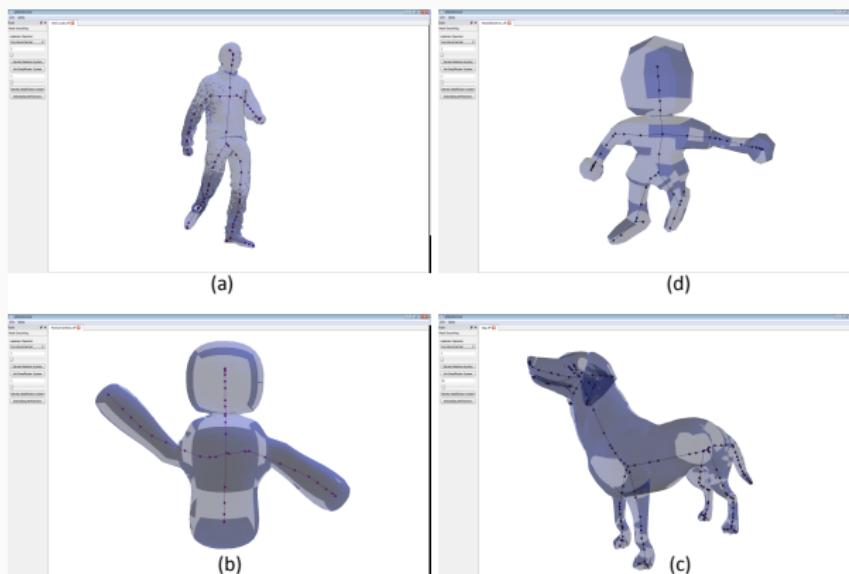


Figure: Skeleton extracted from different models. (a) Dog model (b) Character model. (c) Person model. (d) Clay model.

RESULTS

Software para la Extracción del Esqueleto por Contracción y Suavizado

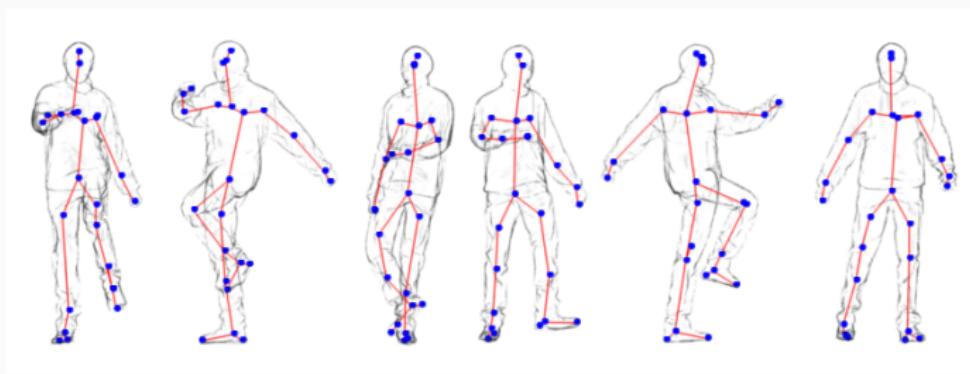


Figure: Model with different poses and skeleton obtained with our skeleton extraction software.

CONCLUSIONS

CONCLUSIONS

- This work presented a novel extension of the Laplace Beltrami operator for hybrid quad/triangle meshes, and the successful application of such principles in different types of problems in computer geometric modeling like smoothing, enhancing, sculpting, deformation, reposing and skeleton extraction.
- A new sculpting brush to make a proper inflation while preserving the geometric details in real-time sessions was proposed, implemented and tested.
- We have largely demonstrated that our method has good performance, stability and robustness of the extension proposed.
- This novel extension of the Laplace Beltrami operator was introduced in the computer modeling industry inside the Blender 3D computer graphics software.

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- Livingstone elephant model is provided courtesy of INRIA and ISTI by the AIM@SHAPE Shape Repository. Hand model is courtesy of the FarField Technology Ltd. Camel model by Valera Ivanov is licensed under a Creative Commons Attribution 3.0 Unported License. Dinosaur and Monkey models are under public domain, courtesy of Blender Foundation.

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QUESTIONS?