## Surface segmentation for improved isotropic remeshing

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- Motivation and problem statement
- Centroidal Voronoi Tessellation
- Segmentation
- Stitching
- Results

# Surface remeshing

- Surface remeshing is the process of transforming one surface mesh into another
- Reasons for remeshing
  - Decimation
  - Triangle quality improvement
- $\bullet$  Many techniques sample the input surface S and then triangulate the points





# Sampling density

- Some techniques resample in parameter space [Alliez et al., 2003]
  - Global parametrization can cause distortion
  - Local parametrization: optimization is not global, stitching is required
- Direct sampling techniques require minimum sampling density



- Centroidal Voronoi Tessellation (CVT) is a method of tessellating a space
- It can be formulated as a critical point of the CVT energy function [Du et al., 1999]

$$F(X) = \sum_{i=1}^{n} \int_{\Omega_i} \rho(x) \|x - x_i\|^2 \, d\sigma \tag{1}$$

X = {x<sub>i</sub>} is the set of sample points, Ω<sub>i</sub> is the Voronoi cell of x<sub>i</sub> with respect to the other sample points, ρ is a density function
In the context of surface remeshing:

- Typical density functions are  $\rho = 1$  and  $\rho = 1/lfs^2$  [Yan et al., 2009]
- We use the Restricted Voronoi Diagram for the final mesh [Du et al., 2003]

- Restricted Voronoi Diagram: each Voronoi Cell is intersected with *S* to form Restricted Voronoi Cell (RVC)
- The dual (called Restricted Delaunay Triangulation, or RDT) is defined as usual
- Obvious problems occur when RVC is not homeomorphic to a topological disk



Two related theorems govern when RDT will be homeomorphic to  $\boldsymbol{S}$ 

Topological ball property [Edelsbrunner and Shah, 1997]

Each RVC must be a topological disk

r-sampling theorem [Amenta and Bern, 1999]

Each point p on surface S must be within  $r \cdot lfs(p)$  of a seed point, where lfs(p) is the local feature size at p

# Sampling density

- Required sample density is based on local feature size (*lfs*)
- Ifs measures curvature and thickness (see Levy short course)
- This is a disappointment: it isn't intuitive that flattish regions should still require high sampling density
- An unfortunate artifact of using euclidean distance to approximate geodesic distance



# Introducing $\kappa \text{CVT}$

- [Fuhrmann et al., 2010]:  $\rho = \sqrt{\kappa}$
- uniform:  $\rho = 1$
- Ifs:  $\rho = 1/lfs^2$
- $\kappa \text{CVT}: \rho = \sqrt{\kappa}$



• The big question: how can we use  $\kappa$  as the density function while still meeting the sampling theorem?

# Segmentation

- Segment surface S such that flattish areas have high lfs
- Let A be a triangle in  $M_i$ . Heuristic: partition S into subsurfaces  $M = \{M_i\}$  such that the ball  $\mathscr{B}(p, r_A)$  centered at any point  $p \in A$  will yield a single connected component when intersected with  $M_i$ .
  - $r_A$  is an approximation of lfs(A).



# Building compatibility table and segmentation

- $\bullet$  Find which triangles are compatible with triangle A
- Given triangle B, let  $P_{A,B}$  be the set of all points on A that are within  $r_A$  of B (shaded region of 2D triangles in image)



# Building compatibility table and segmentation

- Search out from A and use boolean set operations to build "compatibility table"
- Use region merging to find groups of compatible triangles





# Subsurface remeshing and stitching

- Remesh each subsurface  $M_i$  individually using CVT with  $\rho = \sqrt{\kappa}$  (hence the name of our method)
- $\bullet$  Stitch remeshed subsurfaces  $\{M^*_i\}$  back together using a search algorithm and cost function



## Method overview and <u>results</u>



## Results



### Remeshing the fish model with 4000 sample points.



- $\kappa$ CVT performed better than the other two CVT methods in terms of geometric error in every test but one, and showed as much as 20% improvement over the next-best method
- Topological errors were reduced to 0 in every case but one. In that case topological errors were reduced by 30% of next-best method
- Average triangle quality was similar to that of *lfs* method

- So what's the catch?
  - Min triangle quality was reduced, due to stitching
  - Improvement is specific to models with flattish areas that have low local feature size
- Future work
  - Implement feature preservation
  - Perform local CVT on stitching area

### Thank you.



### References

[Alliez et al., 2003] Alliez, P., de Verdire, E., Devillers, O., and Isenburg, M. (2003). Isotropic surface remeshing. In Share Modeling International. 2003, pages 49-58, IEEE.

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#### Theorem

*r*-sampling theorem [Amenta and Bern, 1999] If no point p on surface S is farther than  $r \cdot lfs(p)$  from a seed point  $x \in X$  where r is a constant then the Restricted Delaunay Triangulation induced by X is homeomorphic to S.

- We define  $r_p = 2 \cdot \alpha \cdot lfs(p)$  and  $r_A = \arg \min_{v_i \in V_A} r_{v_i}$ .
- All of our experiments use  $\alpha = 1.1$ .

# Flood fill segmentation





# Stitching cost function

 $t_c$  is a candidate "connector"

$$cost(t_c) = \sum_{t \in \mathcal{T}_c} area(t) \cdot Q(t)^{-\gamma}.$$
 (2)

 $\gamma$  is a user-defined parameter (we used  $\gamma = 0.5$ ) and Q(t) is the triangle quality measure

$$Q_t = \frac{6}{\sqrt{3}} \frac{r_t}{h_t} \tag{3}$$

where  $r_t$  and  $h_t$  are the inradius and longest edge length of t, respectively.



model	# seeds	method	errors	$\rm H_{mean} \times 10^{3}$	$\rm H_{RMS} \times 10^{3}$	Q <sub>min</sub>	Qave
Elk	2000	uniform	400	0.94	1.48	0.448	0.884
		/fs	15	1.31	1.76	0.347	0.858
		$\kappa \text{CVT}$	0	0.76	1.00	0.220	0.849
Elk	8000	[Fuhrmann et al., 2010]	0	0.38	0.63	0.058	0.902
		uniform	0	0.24	0.37	0.509	0.916
		/fs	0	0.36	0.49	0.451	0.893
		$\kappa CVT$	0	0.23	0.34	0.259	0.885
Fish	1000	uniform	95	0.97	0.16	0.525	0.872
		/fs	14	0.91	0.12	0.420	0.830
		$\kappa \text{CVT}$	0	0.82	0.12	0.236	0.809
Fish	4000	[Fuhrmann et al., 2010]	0	0.50	0.85	0.070	0.898
		uniform	11	0.36	0.53	0.580	0.898
		/fs	0	0.36	0.51	0.407	0.864
		$\kappa CVT$	0	0.36	0.58	0.160	0.863
Club	200	uniform	51	2.94	4.08	0.570	0.842
		/fs	31	4.25	6.36	0.362	0.770
		$\kappa \text{CVT}$	19	3.42	4.92	0.173	0.728
Club	2000	[Fuhrmann et al., 2010]	-	0.74	1.54	$\sim 0$	0.832
		uniform	0	0.39	0.70	0.555	0.893
		/fs	0	0.46	0.85	0.314	0.834
		$\kappa CVT$	0	0.34	0.62	0.082	0.855