

6.838: Shape Analysis

Justin Solomon

MIT, Spring 2017



<administrative>

Course Instructor

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<http://gdp.csail.mit.edu>

TA

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Office: 1-225

Office hours: Thursdays, 3pm-5pm

On the Web

`gdp.csail.mit.edu/
6838_spring_2017.html`
+ 

Register ASAP!

Deliverables

1. Four homeworks (40%)

Written + coding

2. One project (50%)

Instructions already online

3. Biweekly nanoquizzes (10%)

Designed to be easy!

Prerequisites

- **Coding**

Python or Matlab preferred

- **Math**

Fluency in linear algebra and multivariable calculus

- **Not required (won't hurt):**

Graphics, differential geometry, numerics

Experiment

A screenshot of a Jupyter Notebook interface. The title bar shows 'hw1' and the URL 'localhost:8888/notebooks/hw1.ipynb'. The main content area displays the following text:

6.838, Shape Analysis: Homework 1

As an experiment, we will be doing our homework in Jupyter notebooks. These support \LaTeX and Python, allowing us to share mathematical formulas and code easily. Please get started early to make sure that you are comfortable with this new tool.

The course staff will be extremely generous helping students figure out these problems if needed.

All homeworks will be graded out of 100 points.

- Supports \LaTeX
- Supports Python
- Plot.ly for visualization

Try early!

Problem 1: Variational calculus (30 points)

Note: This problem may be tricky to think about for computer science students who are not used to these sorts of calculations. Leave yourself plenty of time, and get help from the instructional staff!

a (10 points). Suppose $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is a smooth function with take an arbitrary vector $y \in \mathbb{R}^n$. Justify the relationship

$$\frac{d}{dt} f(x^* + t y) \Big|_{t=0} = 0$$

<http://gryd1.csail.mit.edu/>

New Course

- Schedule is **too ambitious!**
- Contact Justin with suggestions, must-cover topics, questions, etc.
- Experiment: Video
(unreliable!)

Philosophy

I want you to take this course!

Assignments intended to be interesting
(may be unintentionally easy/hard!)

Will be generous with support/grading

Quick Survey

Degree
Undergraduate
M.Eng.
M.Sc./PhD

Quick Survey

Background

EECS

Math

Engineering
Elsewhere

</administrative>

Theme

1. *Geometric data analysis:* The analysis of geometric data
 Modifier Noun

2. *Geometric data analysis:* Data analysis using geometric techniques
 Modifier Noun

Applied Geometry

- I. Theoretical toolbox
- II. Computational toolbox
- III. Application areas

Mostly a picture book!

Applied Geometry

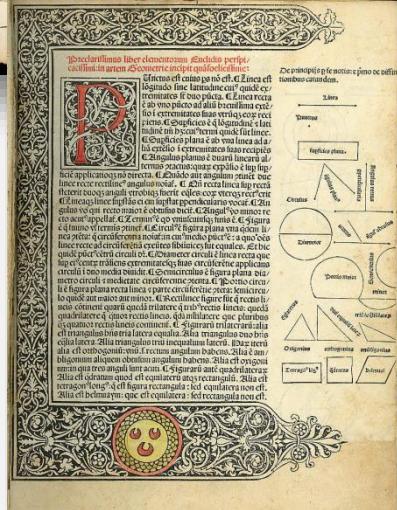
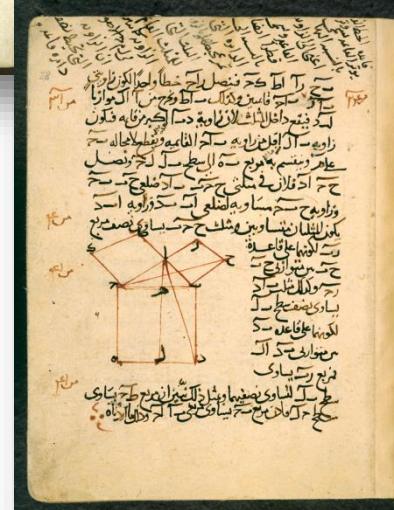
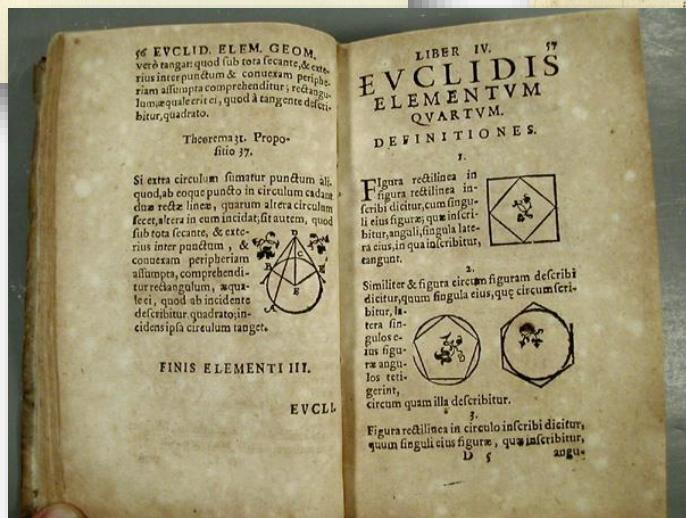
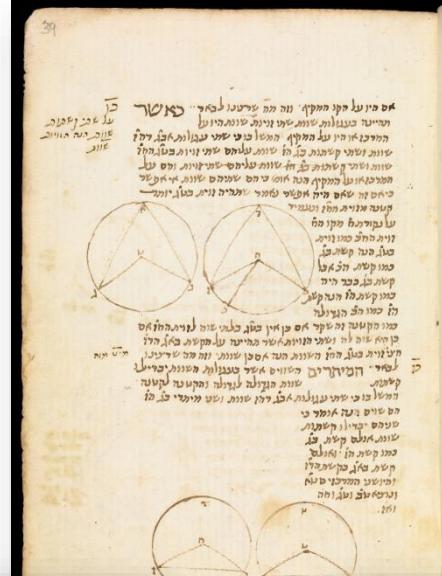
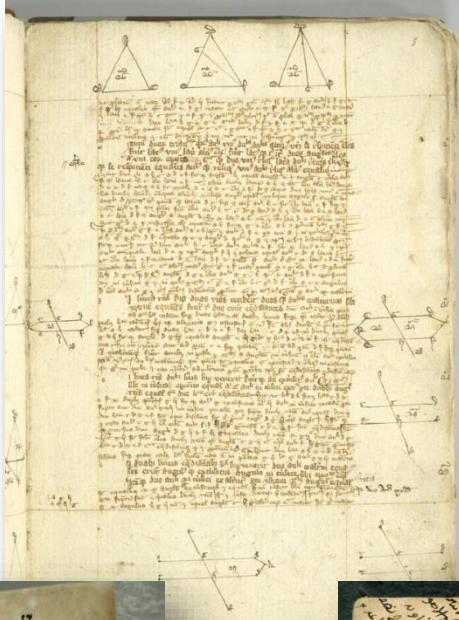
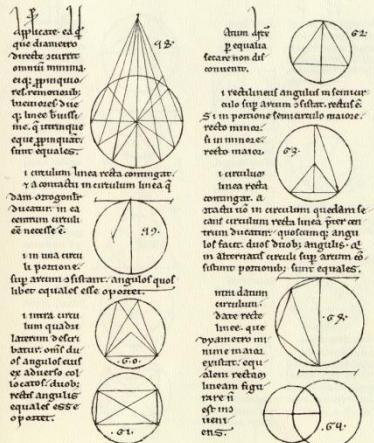
I. Theoretical toolbox

II. Computational toolbox

III. Application areas

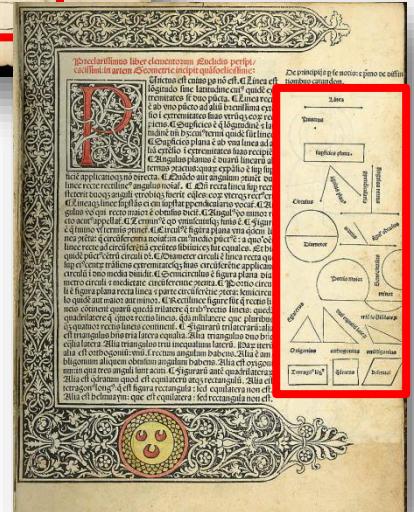
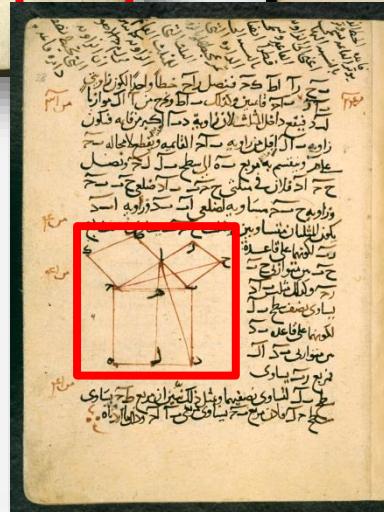
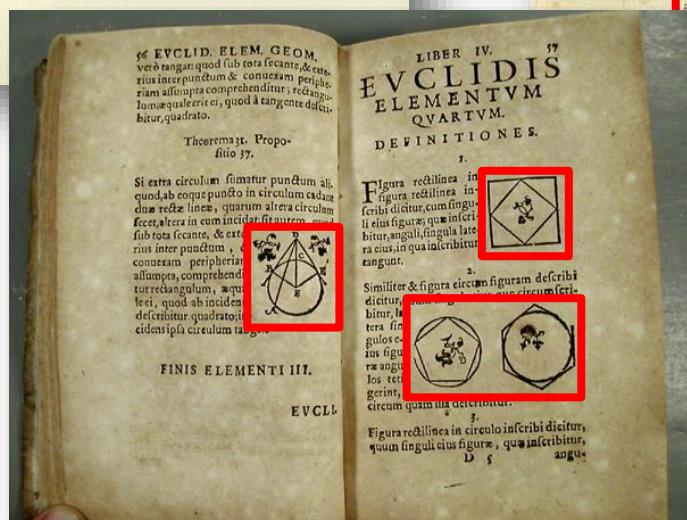
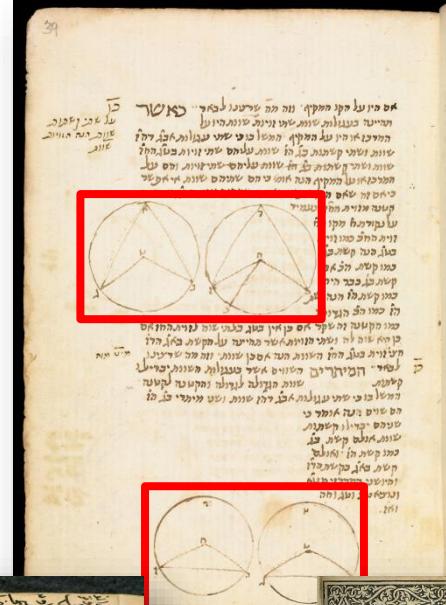
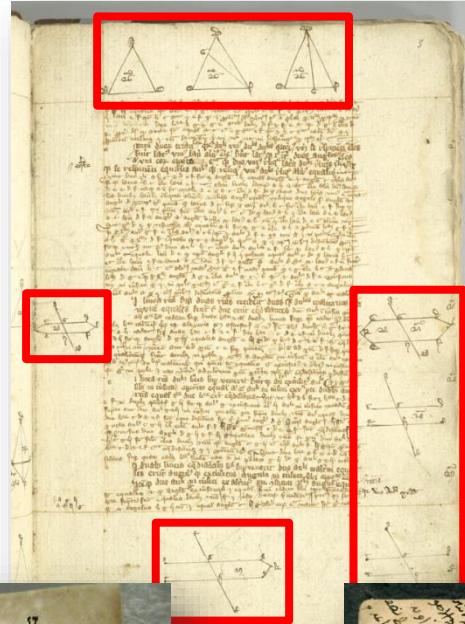
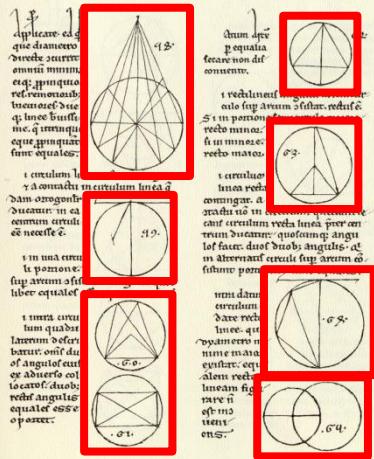
Euclidean Geometry

- 10 -



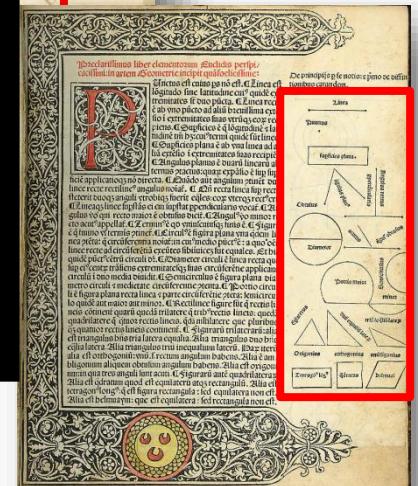
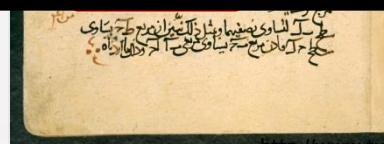
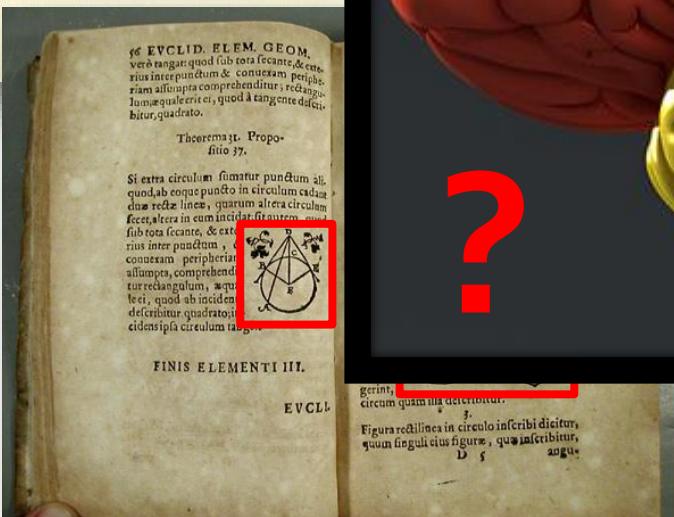
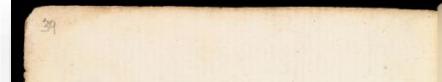
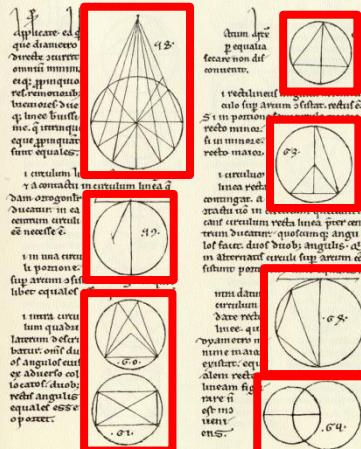
Euclidean Geometry

- 10 -

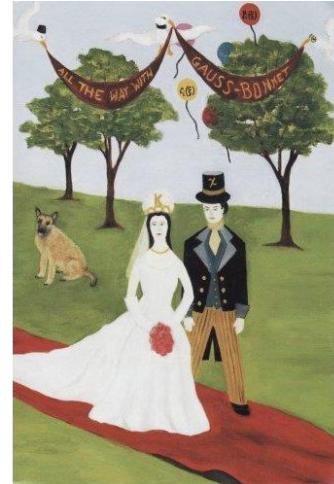
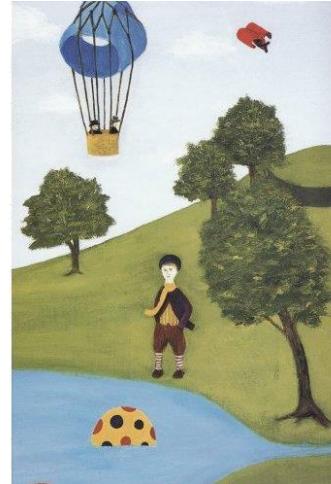
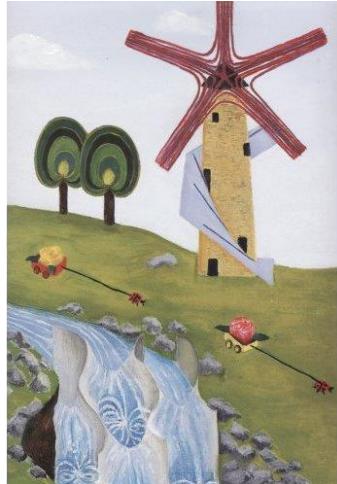
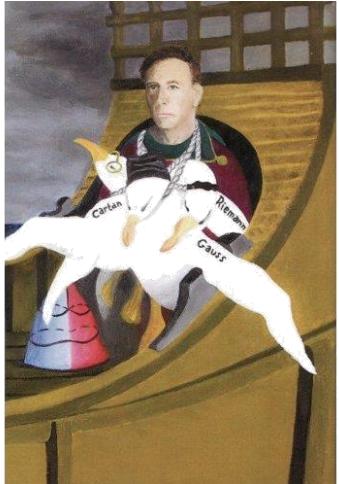
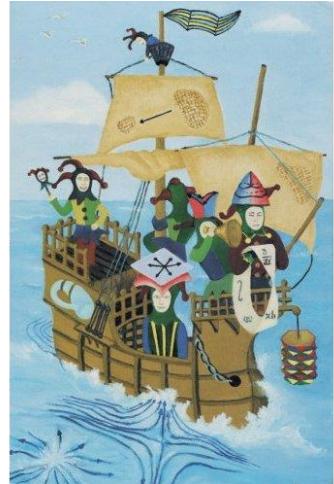


Euclidean Geometry

- 10 -

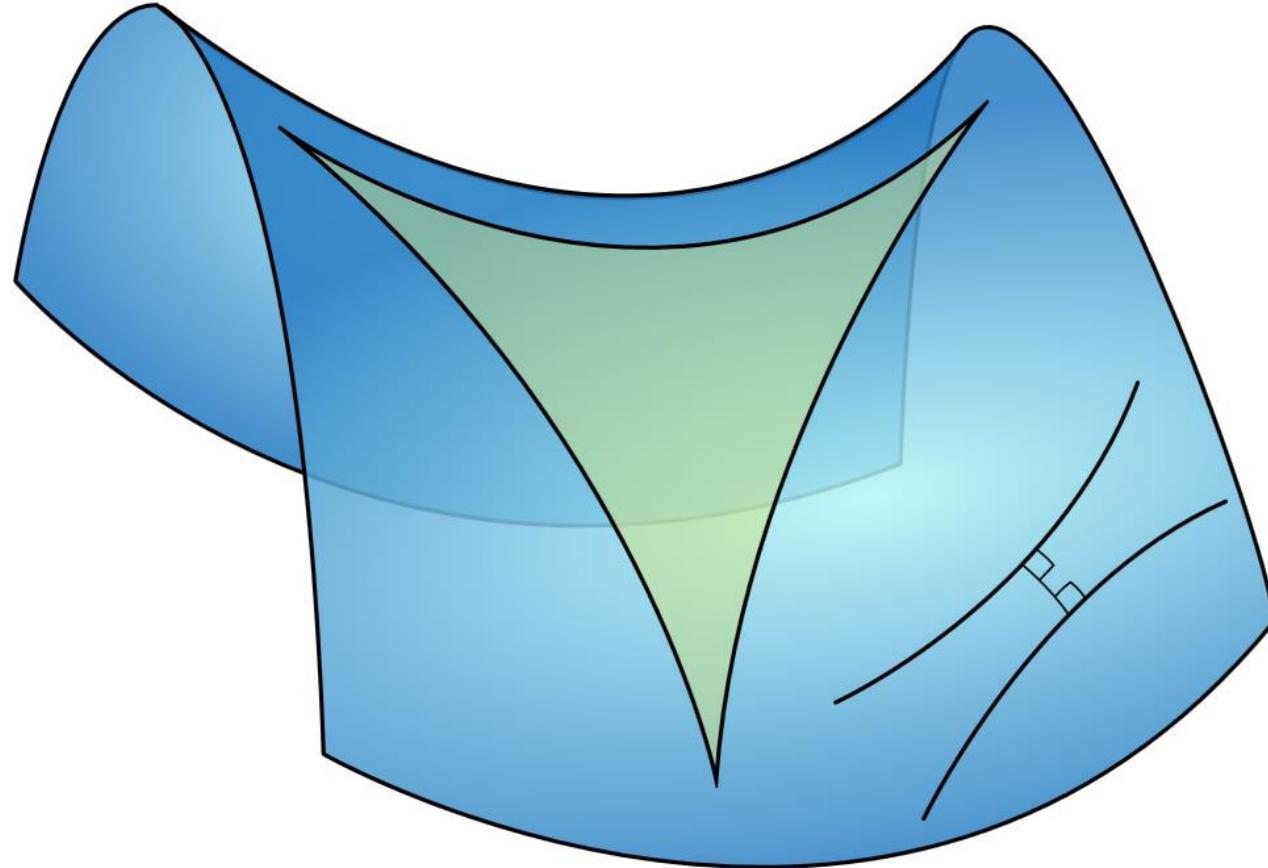


Differential Geometry



Spivak: *A Comprehensive Introduction to Differential Geometry*

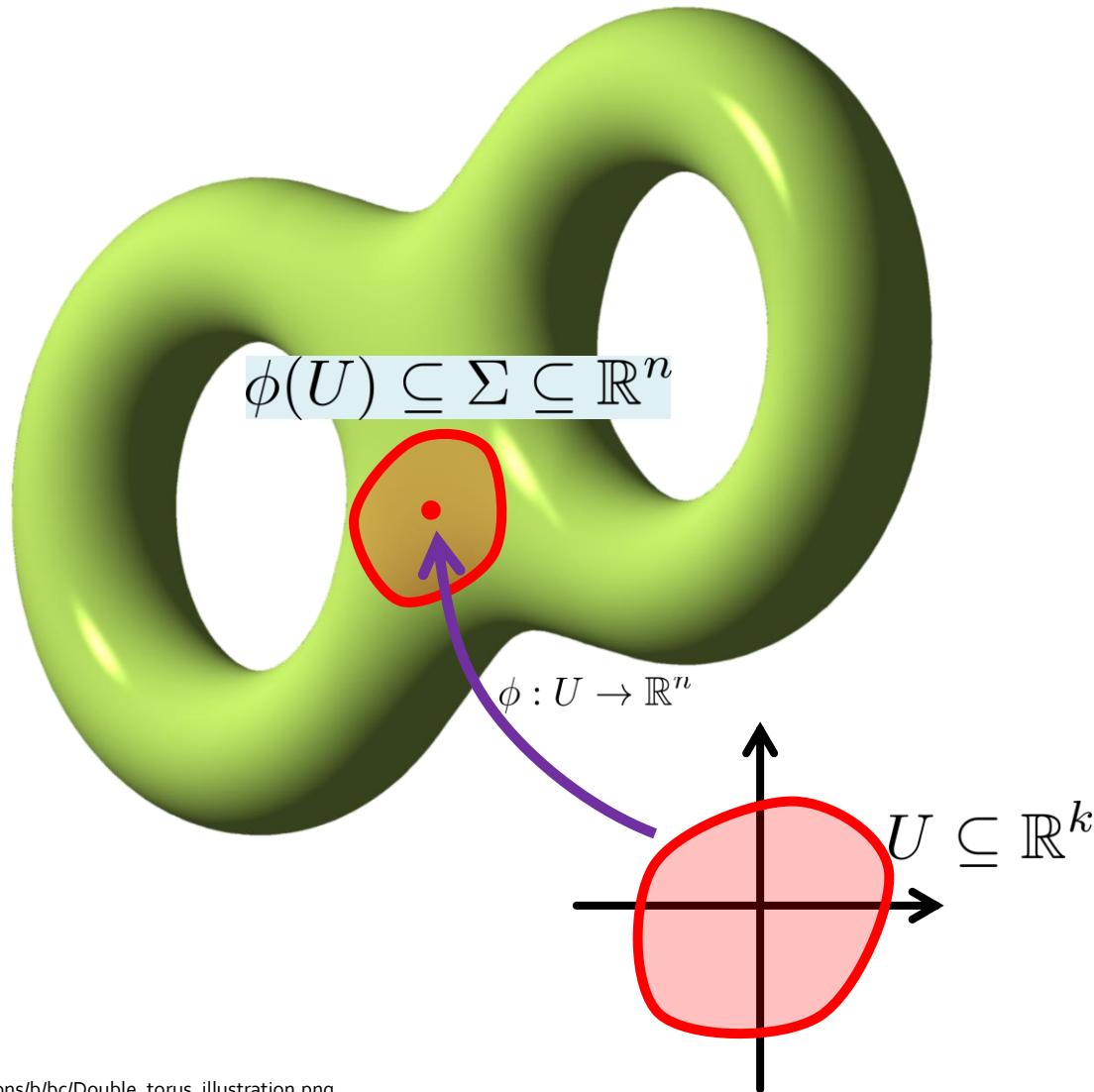
Differential Geometry



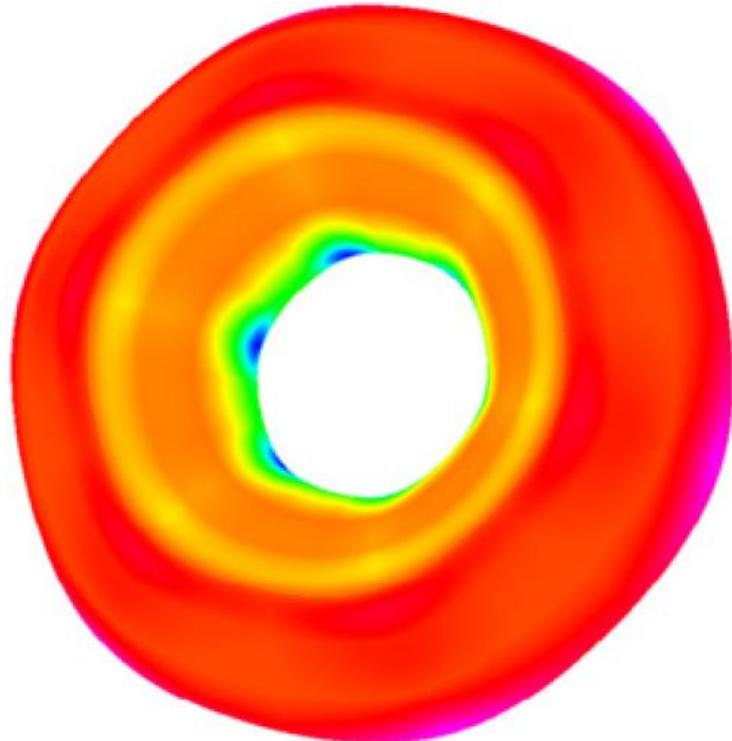
http://en.wikipedia.org/wiki/Differential_geometry

Study of smooth manifolds

Manifold



Differential Geometry Toolbox



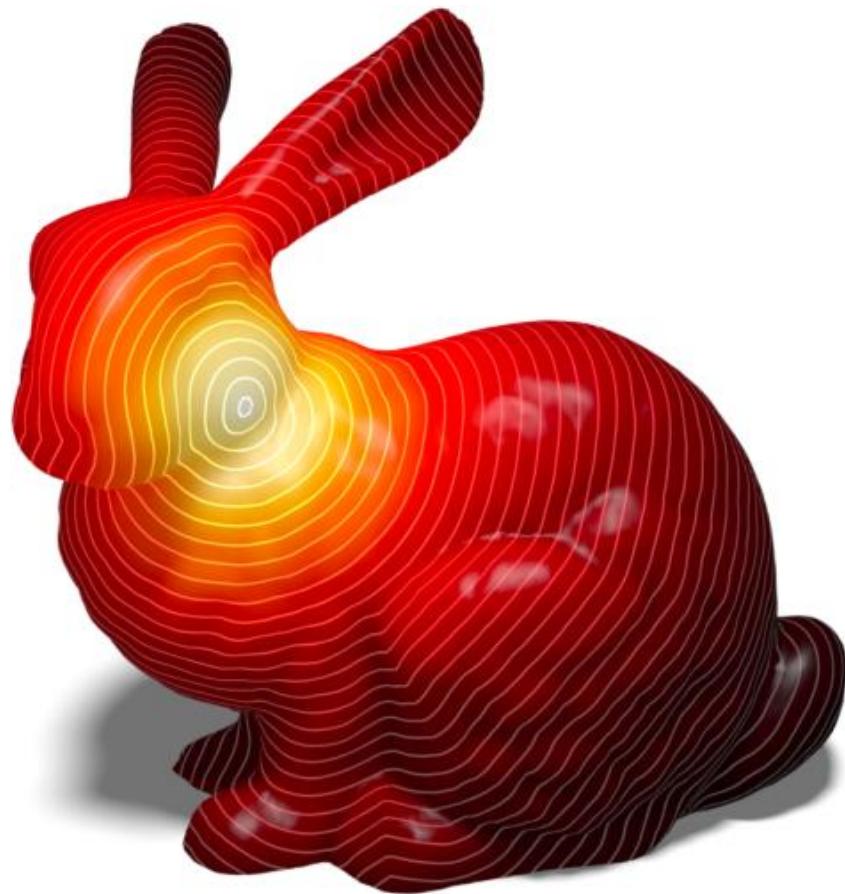
$$K := \kappa_1 \kappa_2 = \det \mathbb{II}$$

$$H := \frac{1}{2}(\kappa_1 + \kappa_2) = \frac{1}{2}\text{tr } \mathbb{II}$$

<http://www.sciencedirect.com/science/article/pii/S0010448510001983>

Curvature and shape properties

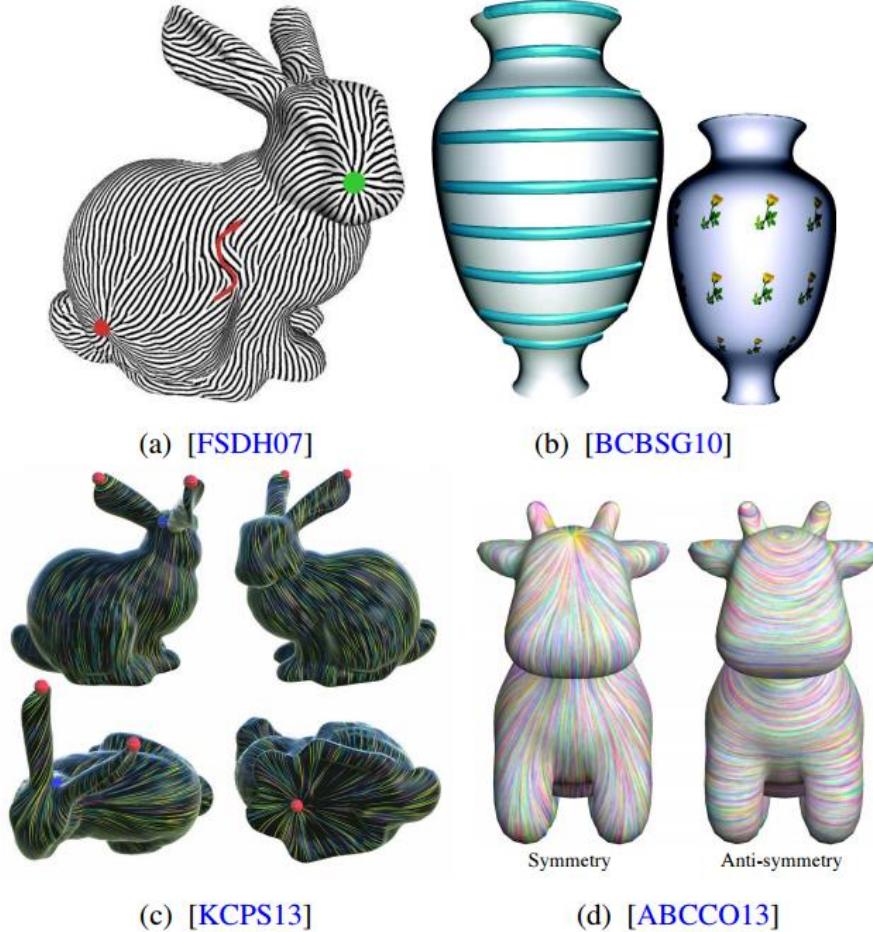
Differential Geometry Toolbox



Crane, Weischedel, Wardetzky.
Geodesics in heat. TOG 2013.

Distances

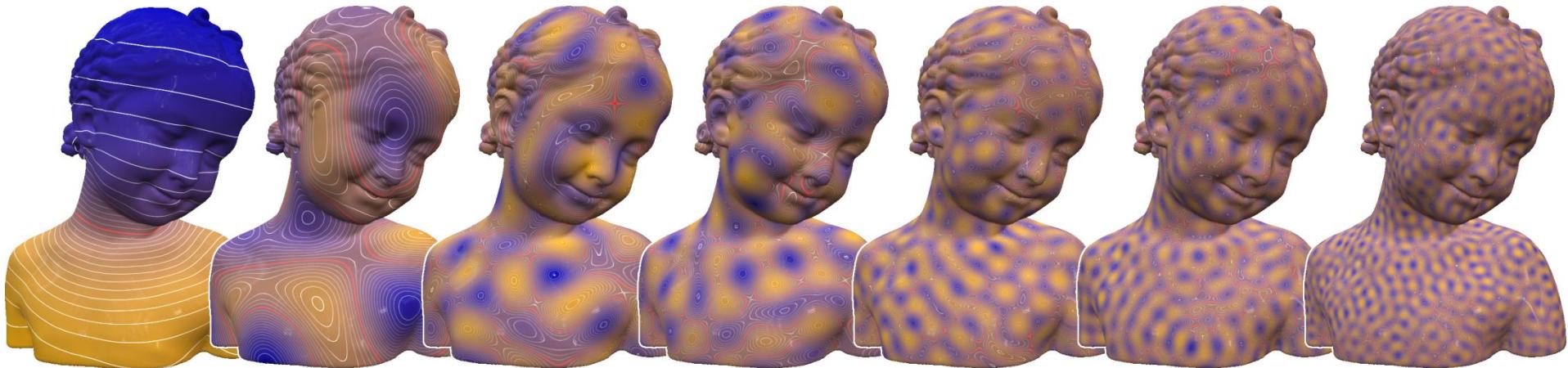
Differential Geometry Toolbox



Vaxman et al.
Directional field synthesis, design, and processing.
EG STAR 2016.

Flows and vector fields

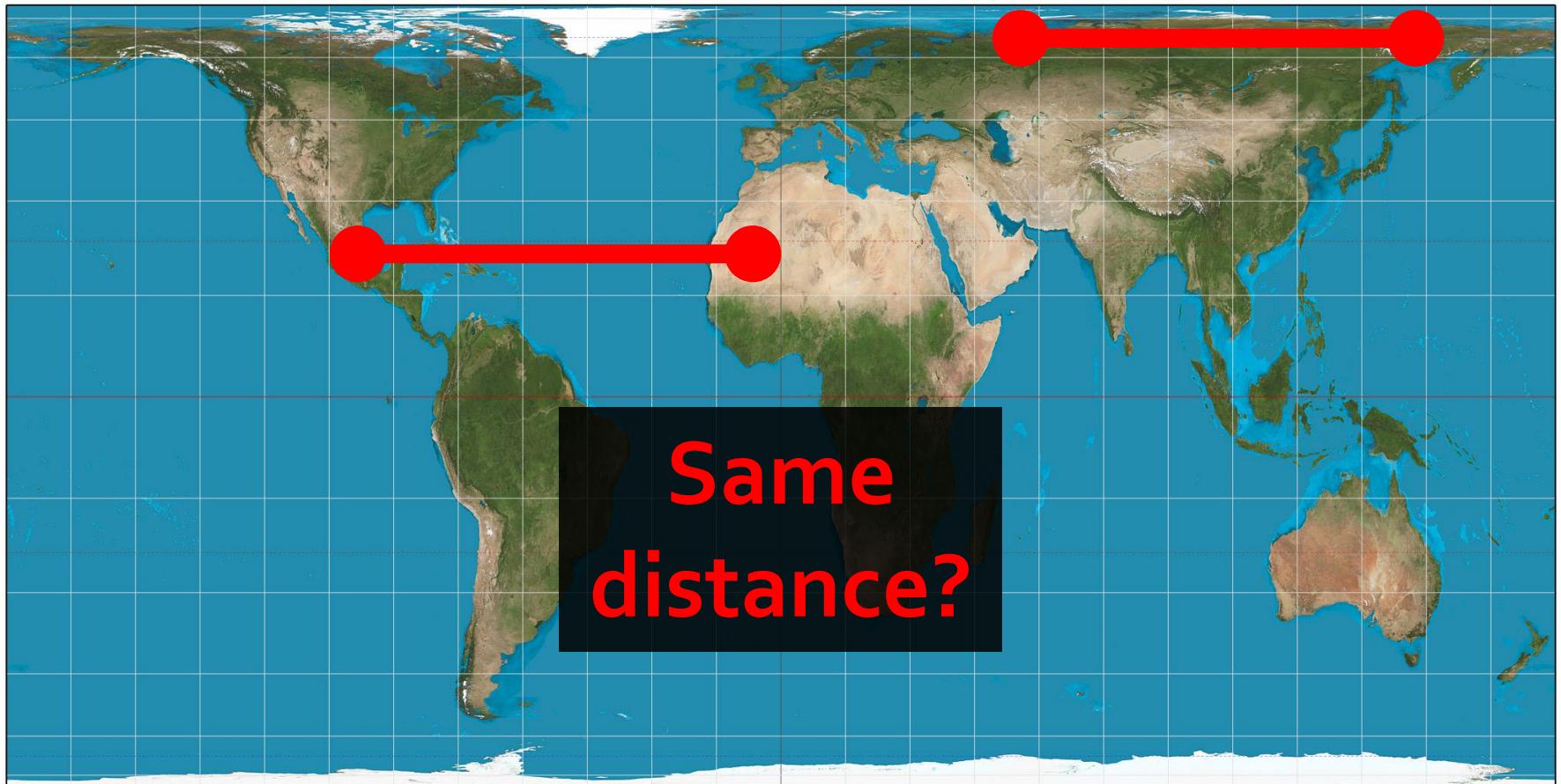
Differential Geometry Toolbox



Vallet and Lévy.
Spectral Geometry Processing with Manifold Harmonics. EG 2008

Differential operators

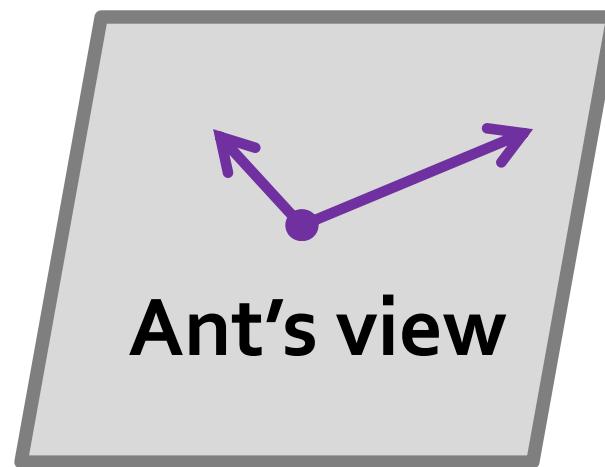
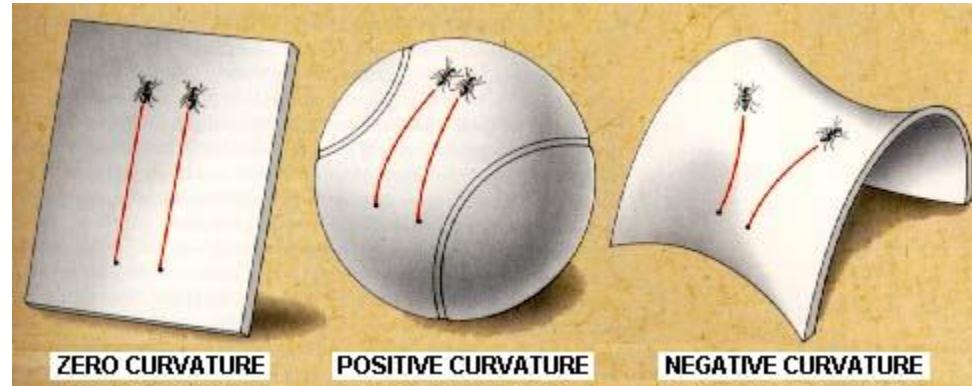
Riemannian Viewpoint



http://upload.wikimedia.org/wikipedia/commons/2/2c/Hobo%E2%80%93Dyer_projection_SW.jpg

Only need angles and distances

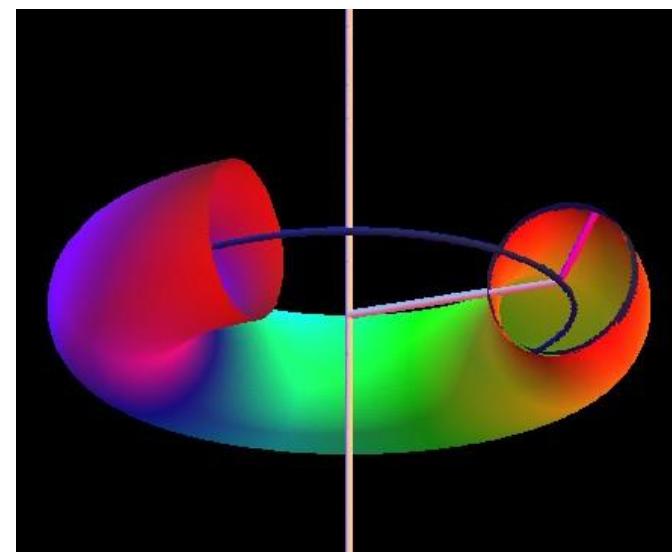
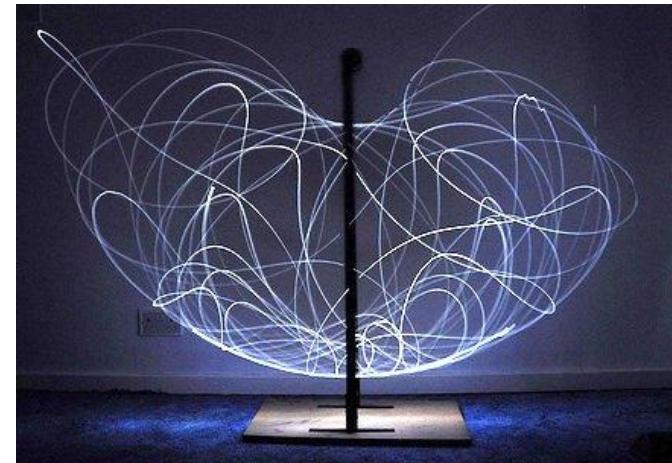
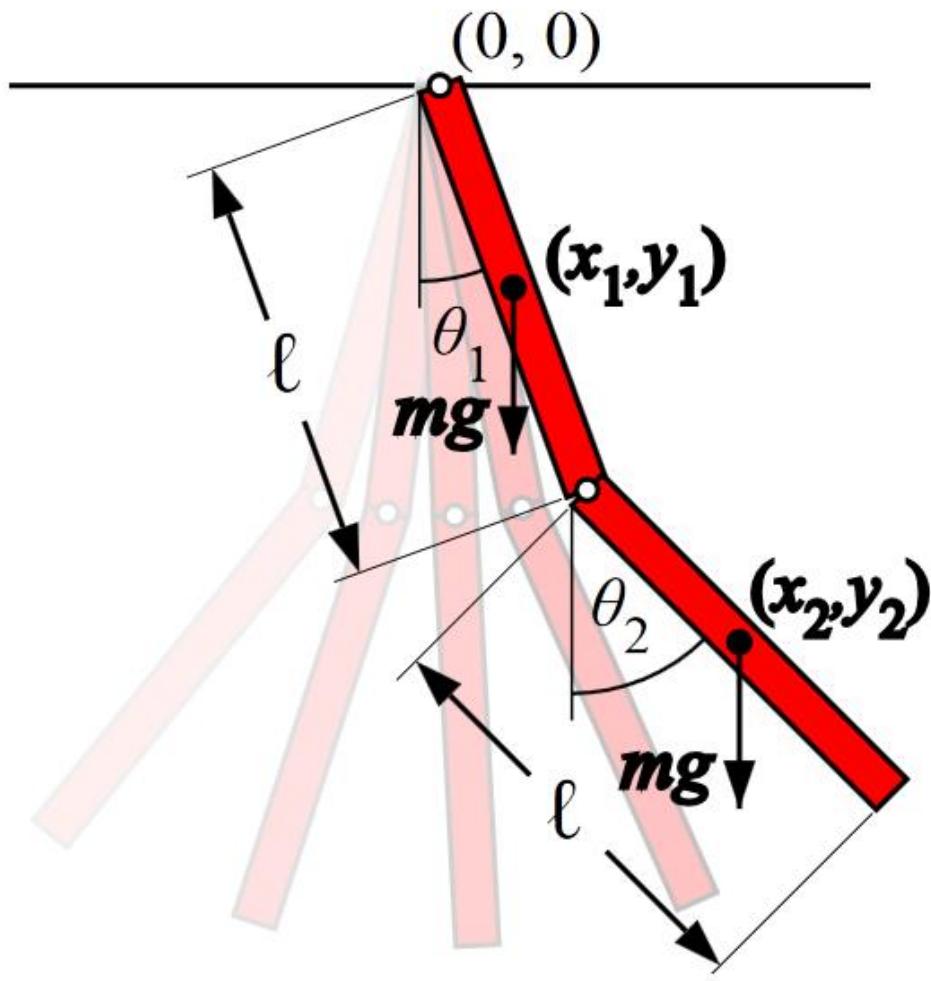
Riemannian Viewpoint



<http://www.phy.syr.edu/courses/modules/LIGHTCONE/pics/curved.jpg>

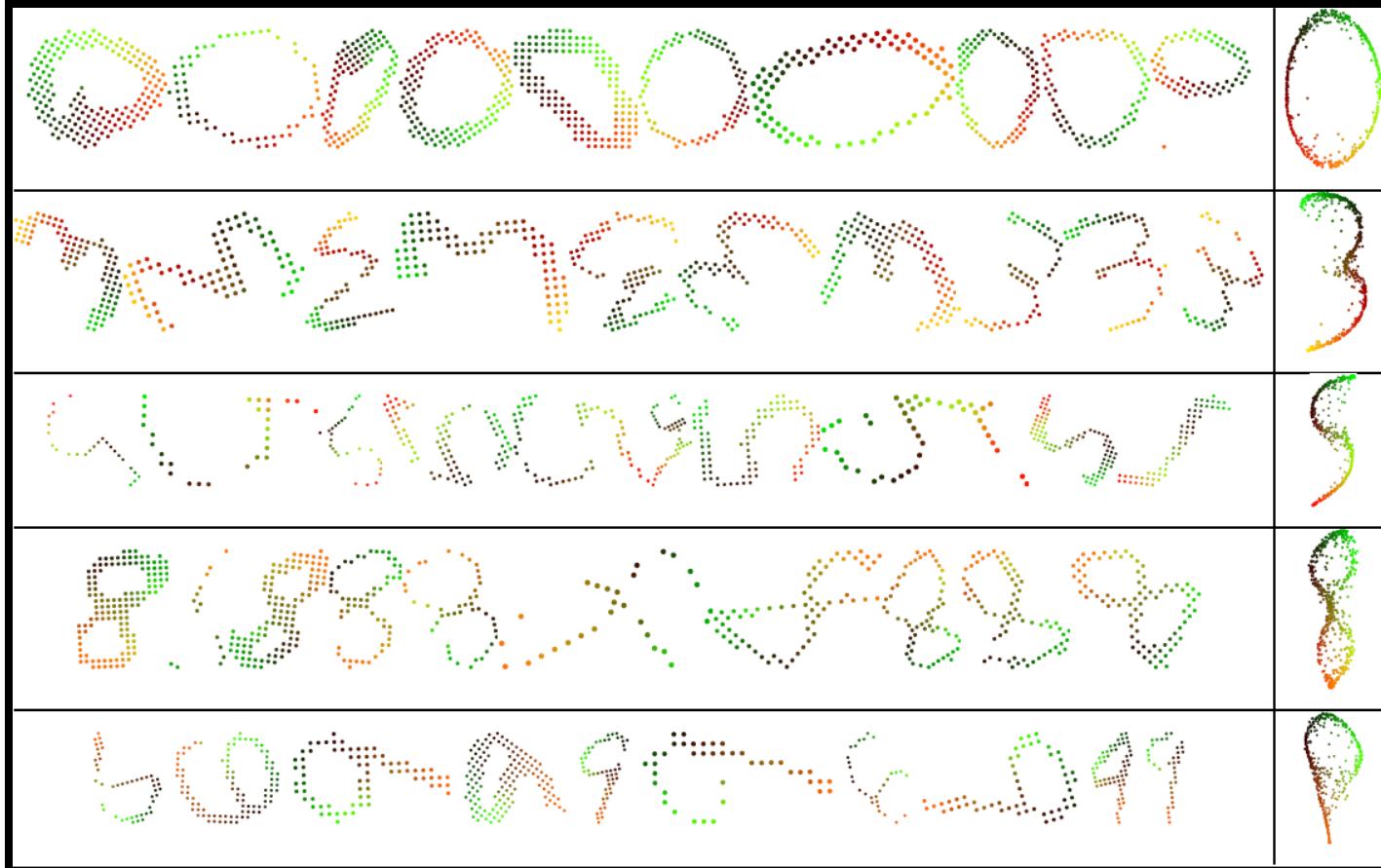
Only need angles and distances

Geometric Mechanics



Metric Geometry

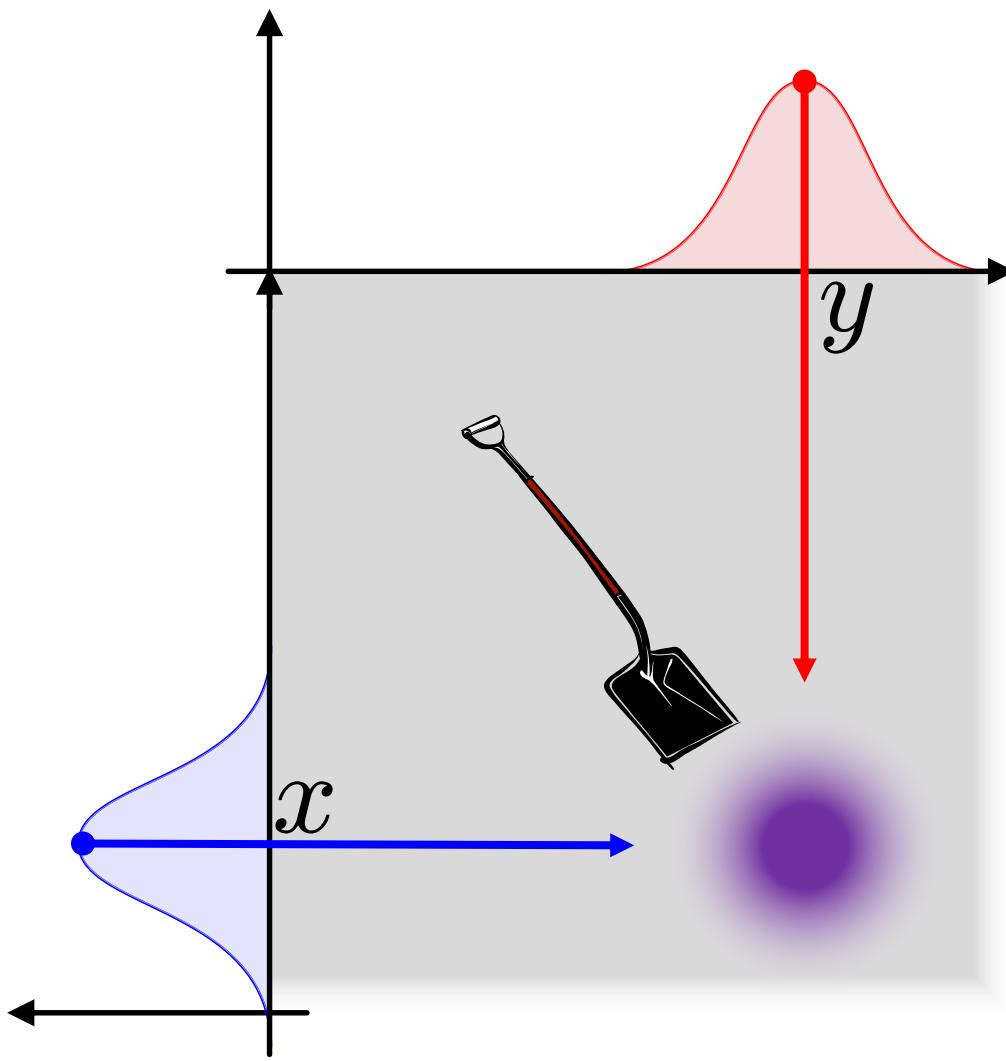
Input data



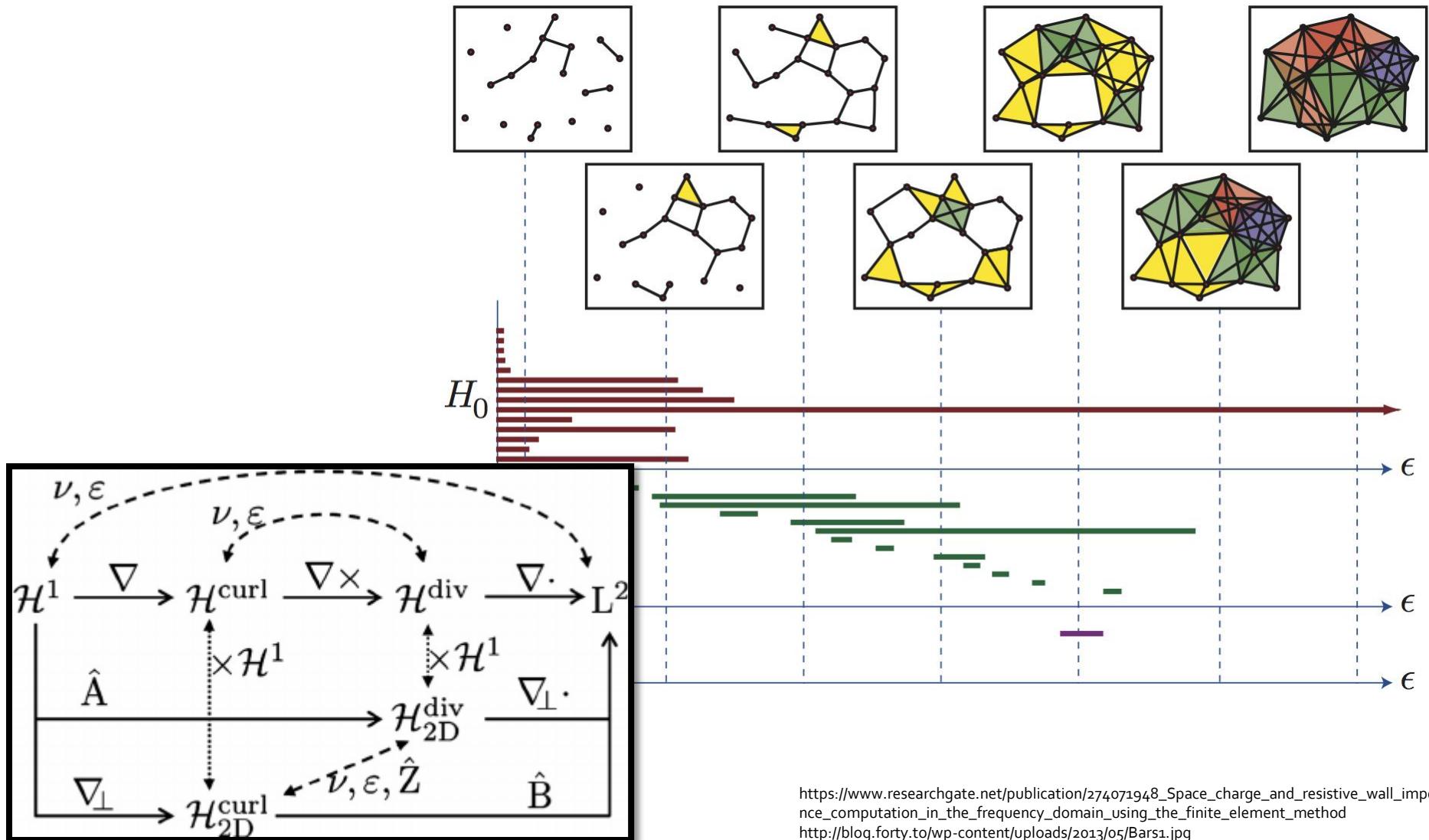
Barycenter (MDS)

Peyré, Cuturi, and Solomon.
Gromov-Wasserstein Averaging of Kernel and Distance Matrices.
ICML 2016.

Optimal Transport



{Differential/Morse/Persistent/...} Topology



Plan for Today

- I. Theoretical toolbox
- II. Computational toolbox**
- III. Application areas

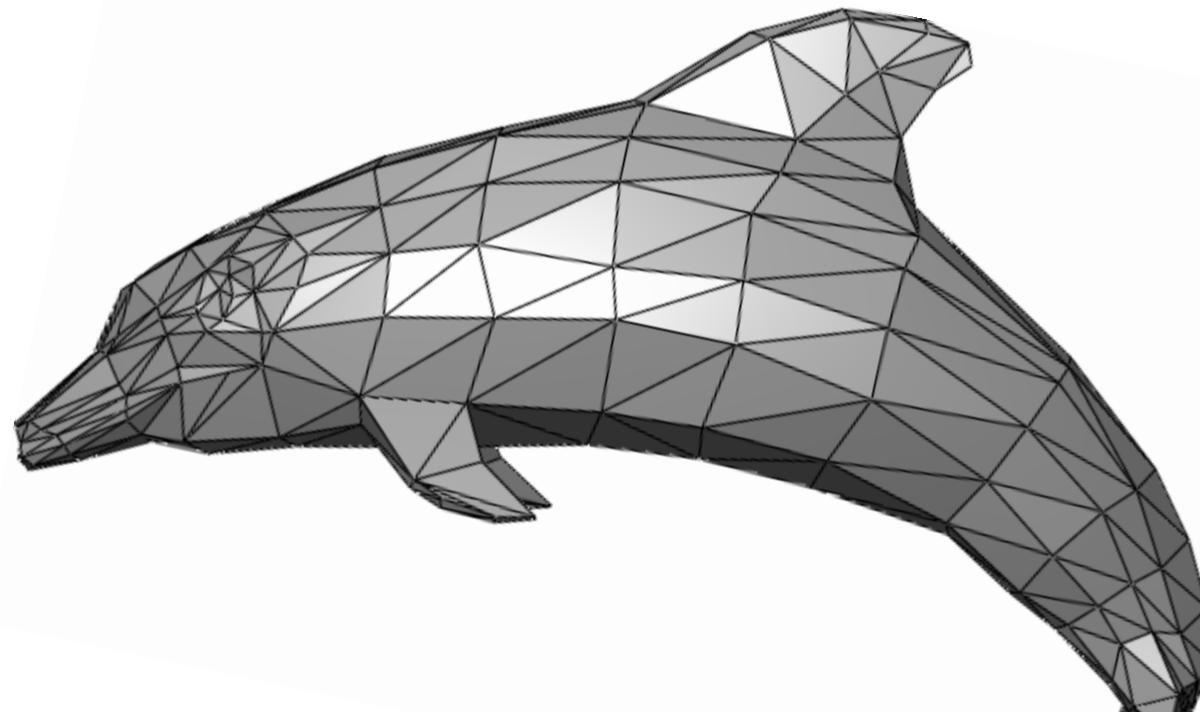
Many Notions of Shape

- Triangle mesh
- Triangle soup
 - Graph
- Point cloud
- Pairwise distance matrix

Nearly anything with a notion of proximity/distance/curvature/...

Typical issue:
Euclidean? Riemannian?

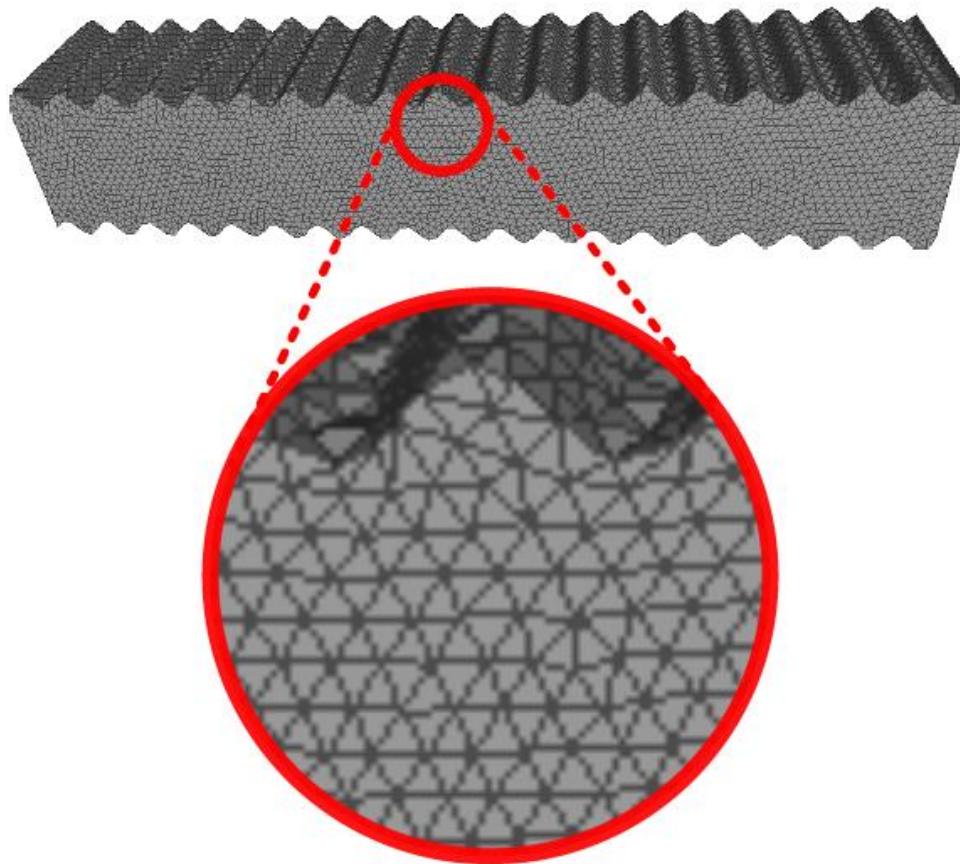
- Collection of **flat triangles**
- Approximates a **smooth surface**





Can a triangle mesh
have **curvature**?

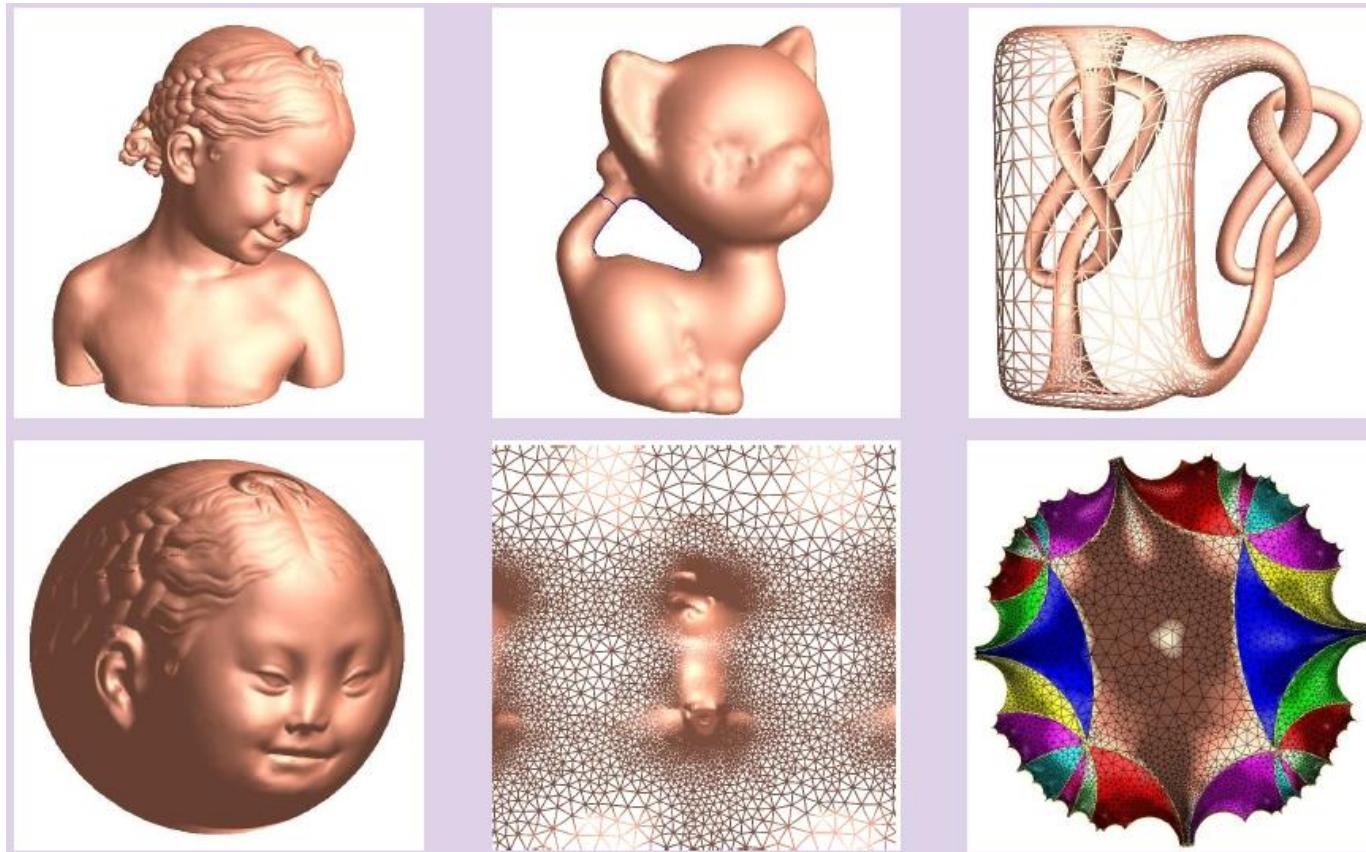
Jack of All Trades



Combine smooth and discrete

Example:

Discrete Differential Geometry



Modern Approach

Discrete

vs.

Discretized

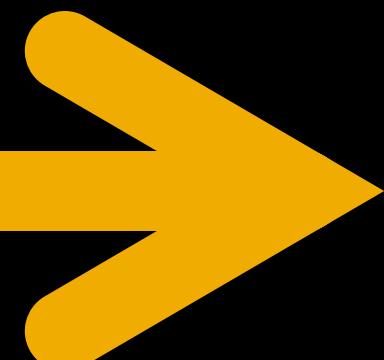
Discrete Differential Geometry

Discrete theory *paralleling*
differential geometry.

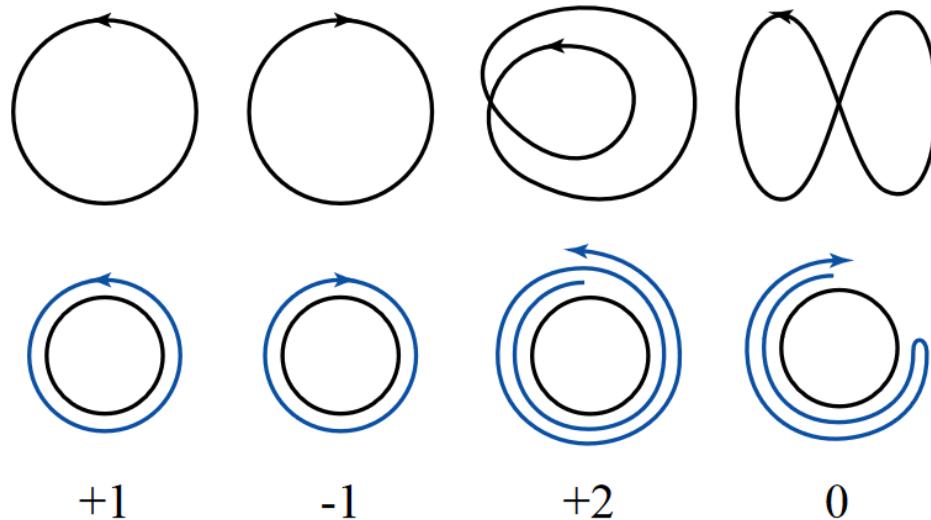
Structure preservation

[struhk-cher pre-zur-vey-shuh n]:

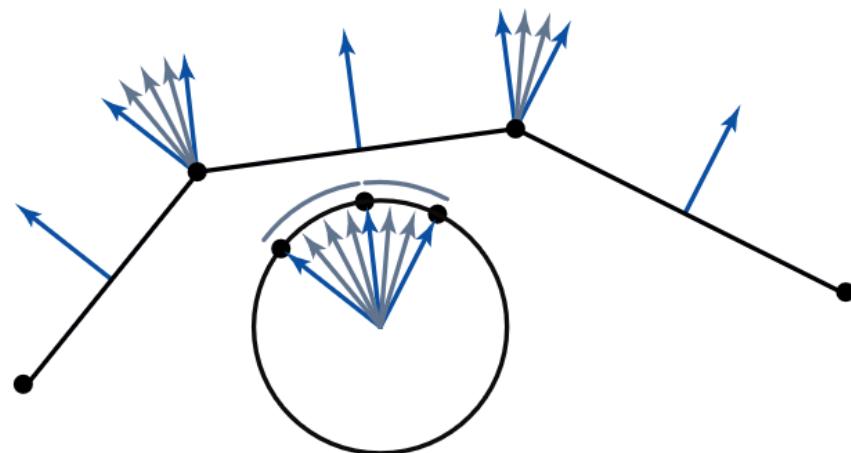
Keeping properties from the continuous abstraction exactly true in a discretization.



Example: Turning Numbers



$$\int_{\Omega} \kappa \, ds = 2\pi k$$



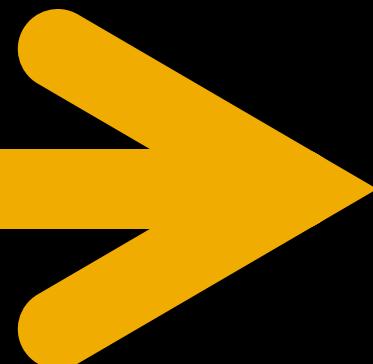
$$\sum_i \alpha_i = 2\pi k$$

Images from: Grinspun and Secord, "The Geometry of Plane Curves" (SIGGRAPH 2006)

Convergence

[kuh n-vur-juh ns]:

Increasing approximation quality as a discretization is refined.



Convergence *and* Structure

Can you have it all?



Disappointing Result

Eurographics Symposium on Geometry Processing (2007)
Alexander Belyaev, Michael Garland (Editors)

Discrete Laplace operators: No free lunch

Max Wardetzky¹

Saurabh Mathur²

Felix Kälberer¹

Eitan Grinspun² †

¹Freie Universität Berlin, Germany

²Columbia University, USA

Abstract

Discrete Laplace operators are ubiquitous in applications spanning geometric modeling to simulation. For robustness and efficiency, many applications require discrete operators that retain key structural properties inherent to the continuous setting. Building on the smooth setting, we present a set of natural properties for discrete Laplace operators for triangular surface meshes. We prove an important theoretical limitation: discrete Laplacians cannot satisfy all natural properties; retroactively, this explains the diversity of existing discrete Laplace operators. Finally, we present a family of operators that includes and extends well-known and widely-used operators.

1. Introduction

Discrete Laplace operators on triangular surface meshes span the entire spectrum of geometry processing applications, including mesh filtering, parameterization, pose transfer, segmentation, reconstruction, re-meshing, compression, simulation, and interpolation via barycentric coordinates [Tau00, Zha04, FH05, Sor05].

In applications one often requires certain structural prop-

1.1. Properties of smooth Laplacians

Consider a smooth surface S , possibly with boundary, equipped with a Riemannian metric, *i.e.*, an intrinsic notion of distance. Let the intrinsic L^2 inner product of functions u and v on S be denoted by $(u, v)_{L^2} = \int_S uv \, dA$, and let $\Delta = -\operatorname{div} \operatorname{grad}$ denote the intrinsic smooth Laplace-Beltrami operator [Ros97]. We list salient properties of this operator:

(NULL) $\Delta u = 0$ whenever u is constant

Disappointing Result

Eurographics Symposium on Geometry Processing (2007)
Alexander Belyaev, Michael Garland (Editors)

Discrete Laplace operators: No free lunch

Max Wardetzky¹

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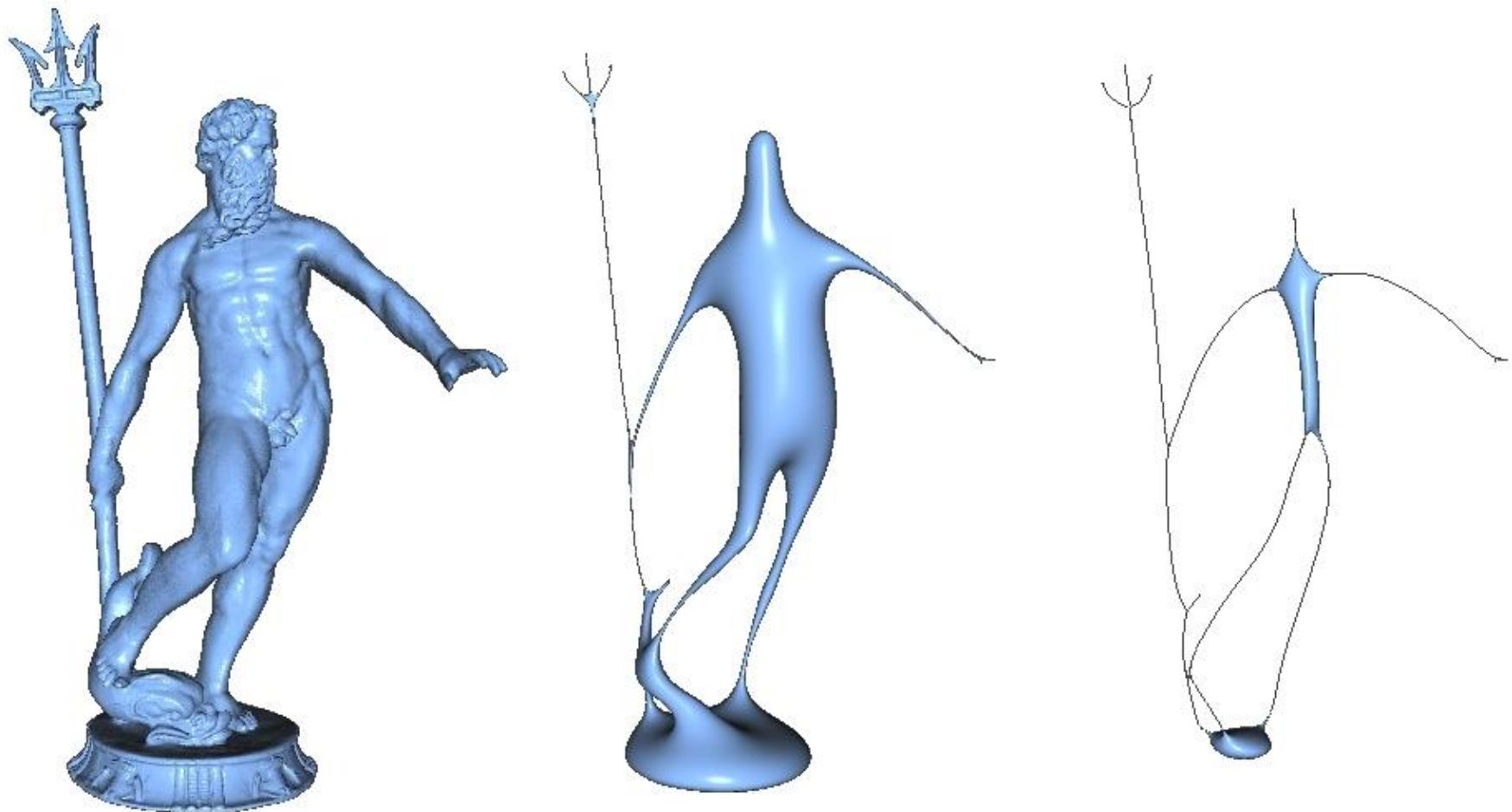
(NULL) $\Delta u = 0$ whenever u is constant

Theme

Pick and choose
which properties you need.

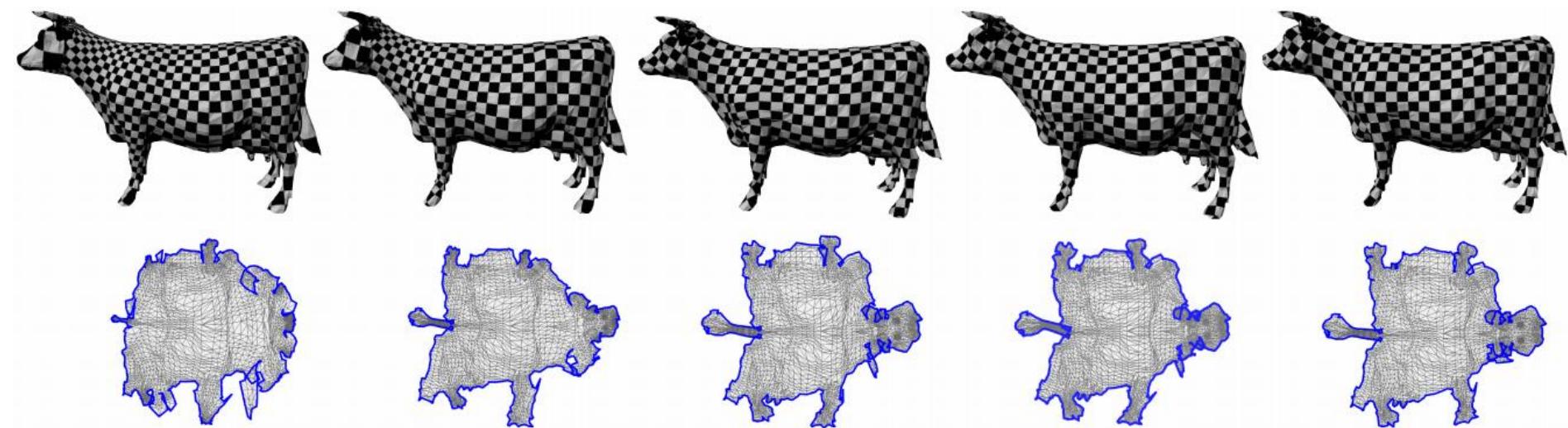
But there is a huge toolbox to draw from!

Numerical PDE



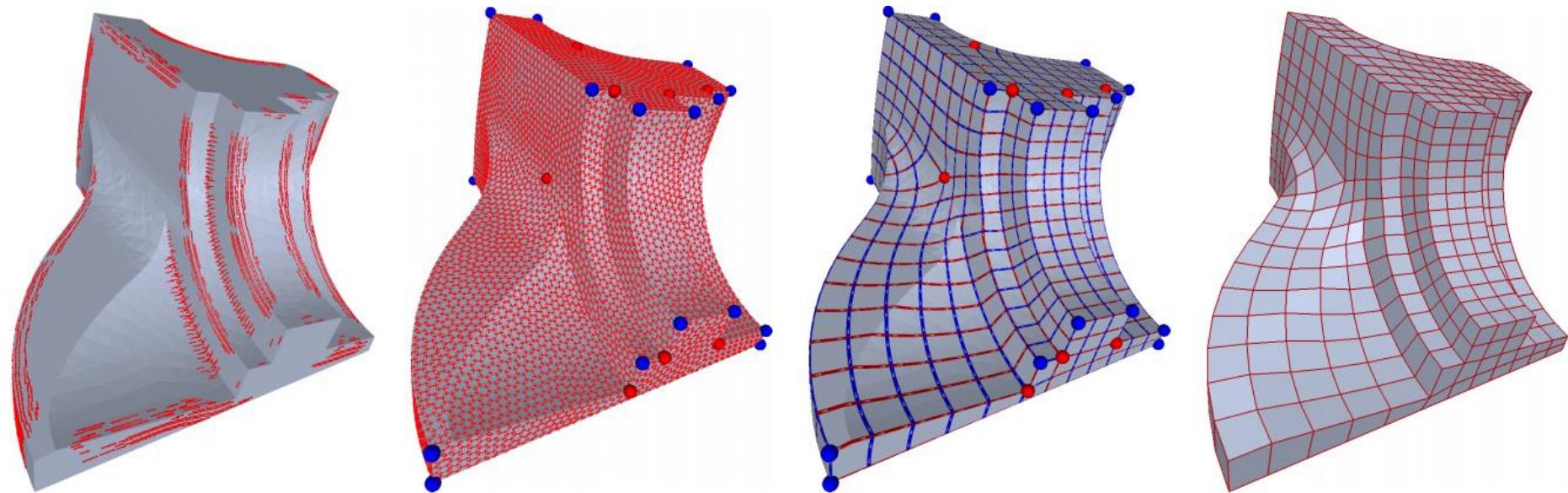
Chuang and Kazhdan.
Fast Mean-Curvature Flow via Finite-Elements Tracking.
CGF 2011.

Smooth Optimization



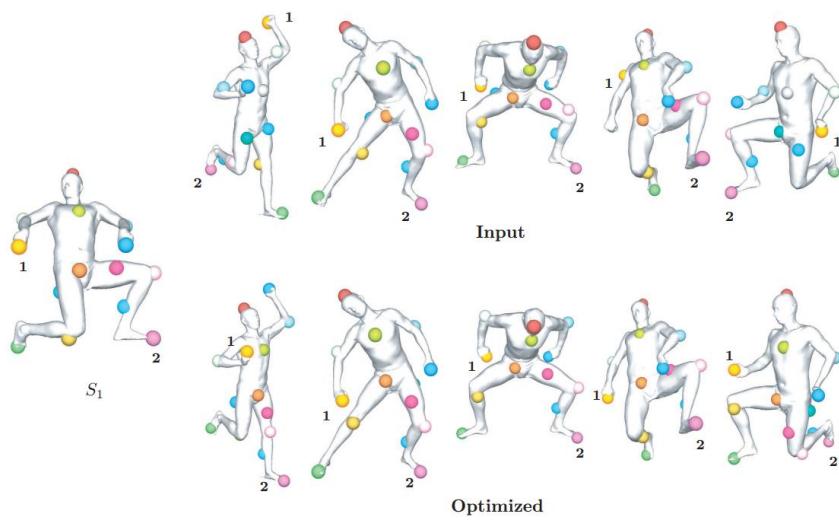
Smith and Schaefer. *Bijective parameterization with free boundaries*. SIGGRAPH 2015.

Discrete Optimization

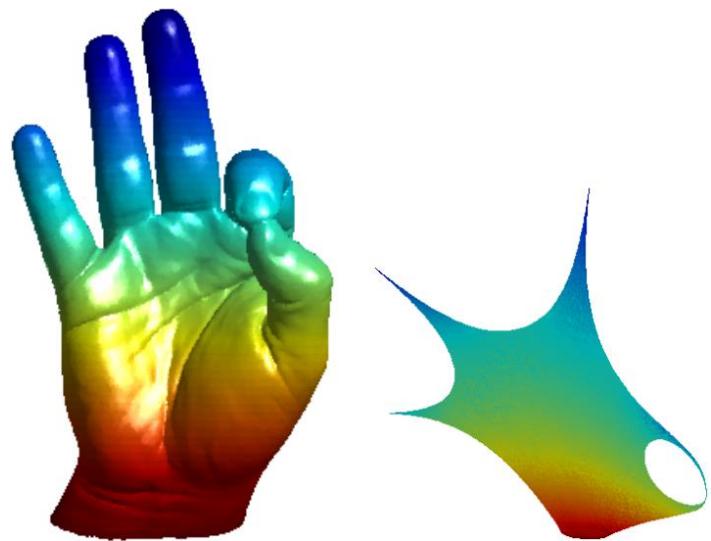


Bommes, Zimmer, Kobbelt. *Mixed-integer quadrangulation*. SIGGRAPH 2009.

Linear Algebra

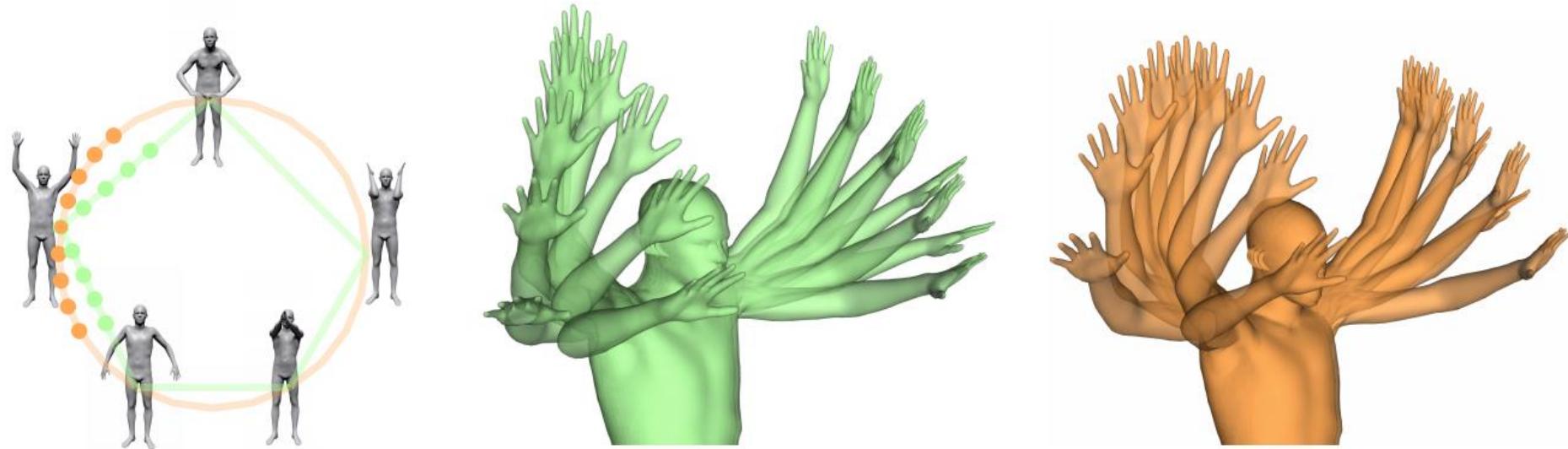


Huang, Guibas. *Consistent shape maps via semidefinite programming.* SGP 2013.



Krishnan, Fattal, Szeliski.
Efficient preconditioning of Laplacian matrices for computer graphics.
SIGGRAPH 2013.

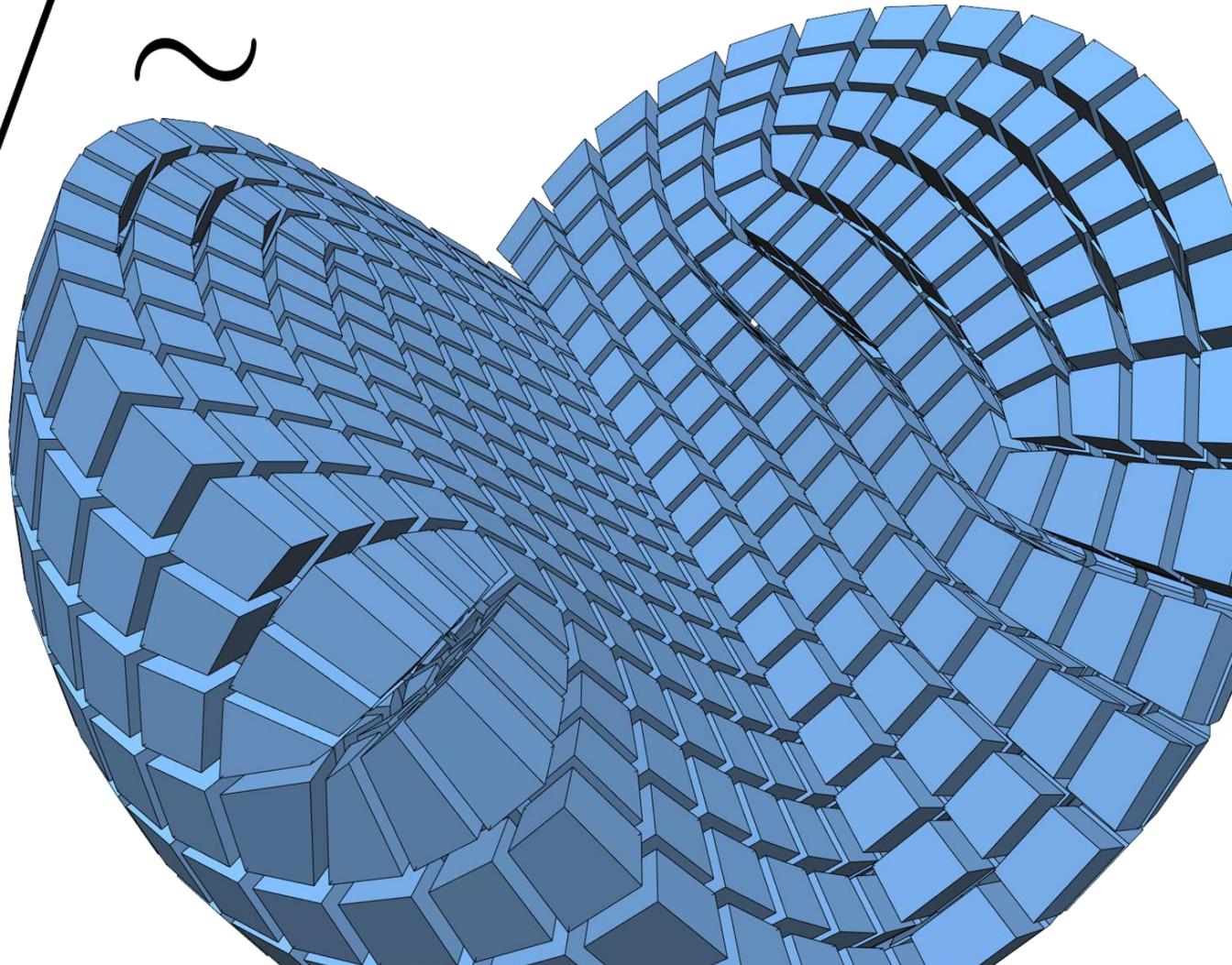
The “Geometry of Geometry”



Heeren et al. *Splines in the space of shells*. SGP 2016.

Algebra & Representation Theory

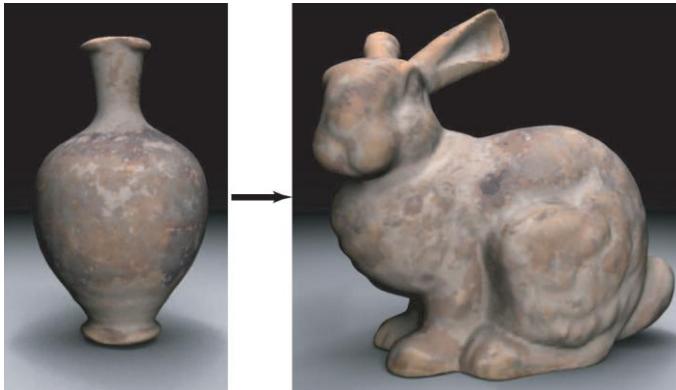
$\text{SO}(3) / \sim$



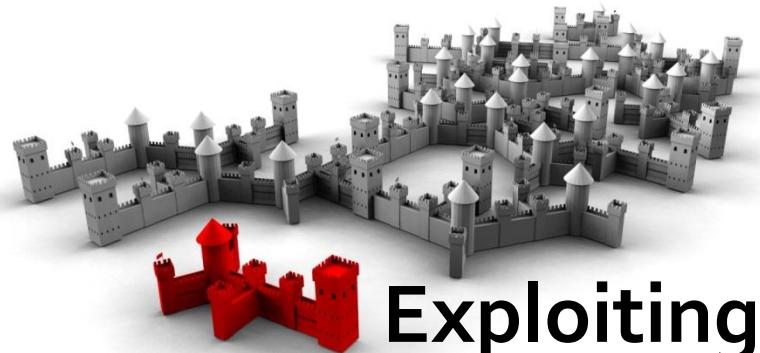
Plan for Today

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Applications

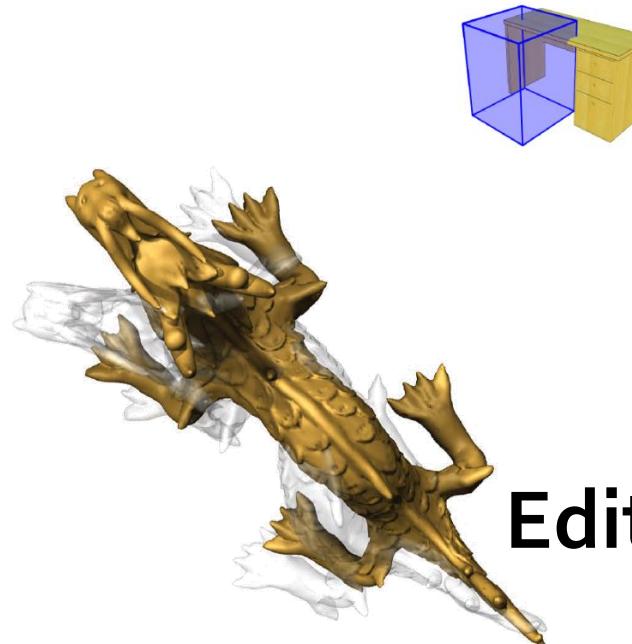


Transfer



Exploiting patterns

Retrieval



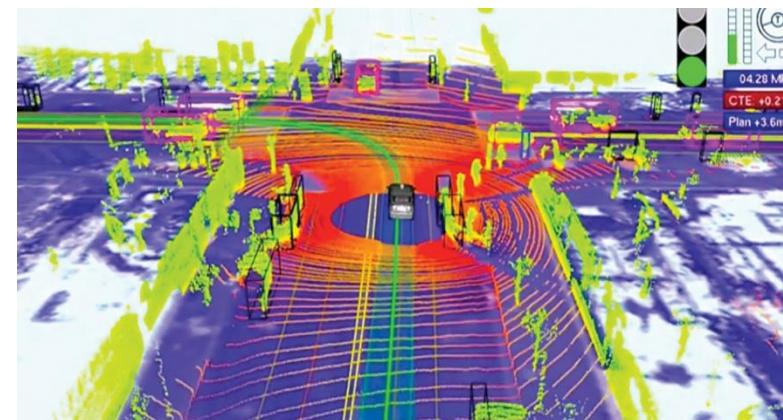
Editing

http://people.csail.mit.edu/tmertens/papers/textransfer_electronic.pdf
<http://graphics.stanford.edu/~mdfisher/Data/Context.pdf>

http://graphics.stanford.edu/~niloy/research/symmetrization/paper_docs/symmetrization_sig_07.pdf
http://www.mpi-inf.mpg.de/~mbokeloh/project_dockingSites.html

Graphics

Applications



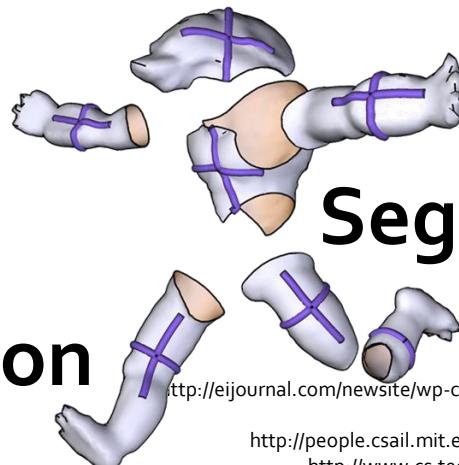
Recognition



Navigation



Reconstruction

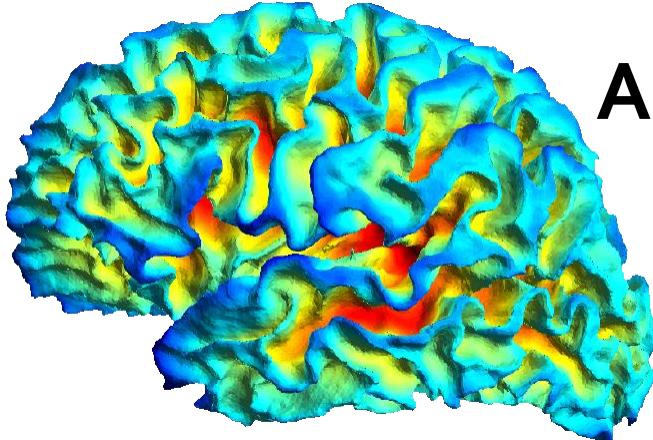


Segmentation

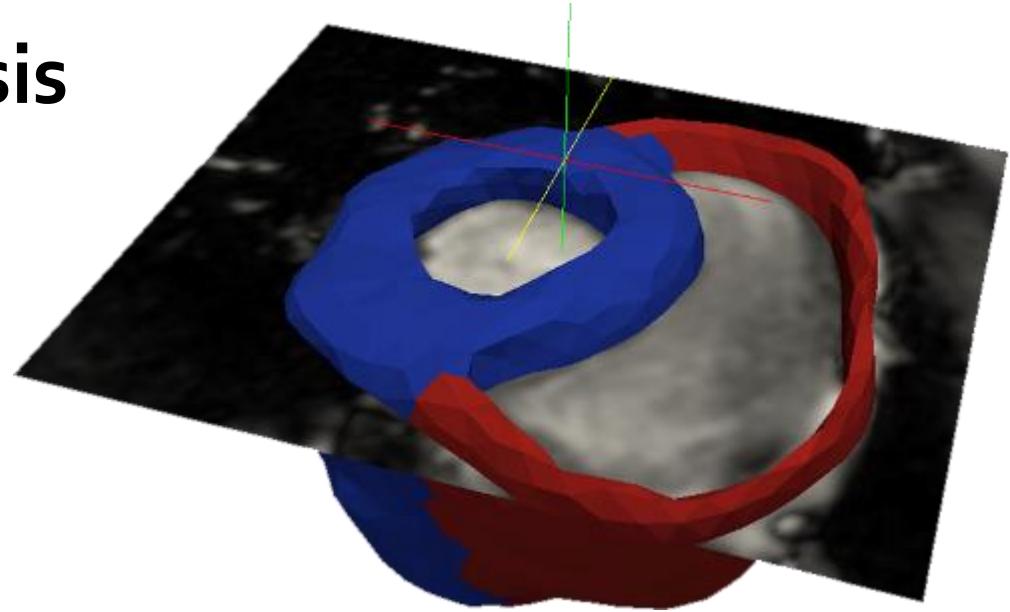
<http://eijournal.com/newsite/wp-content/uploads/2012/01/VELODYNE-IMAGE.jpg>
<http://www.stanford.edu/~jinhae/iccv09/>
http://people.csail.mit.edu/jsolomon/assets/intrinsic_part_discovery.pdf
<http://www.cs.technion.ac.il/~ron/PAPERS/BroBroKimJCVC05.pdf>

Vision

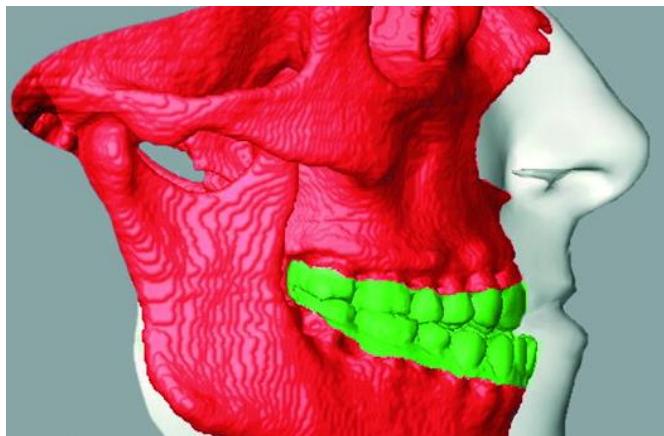
Applications



Analysis



Segmentation

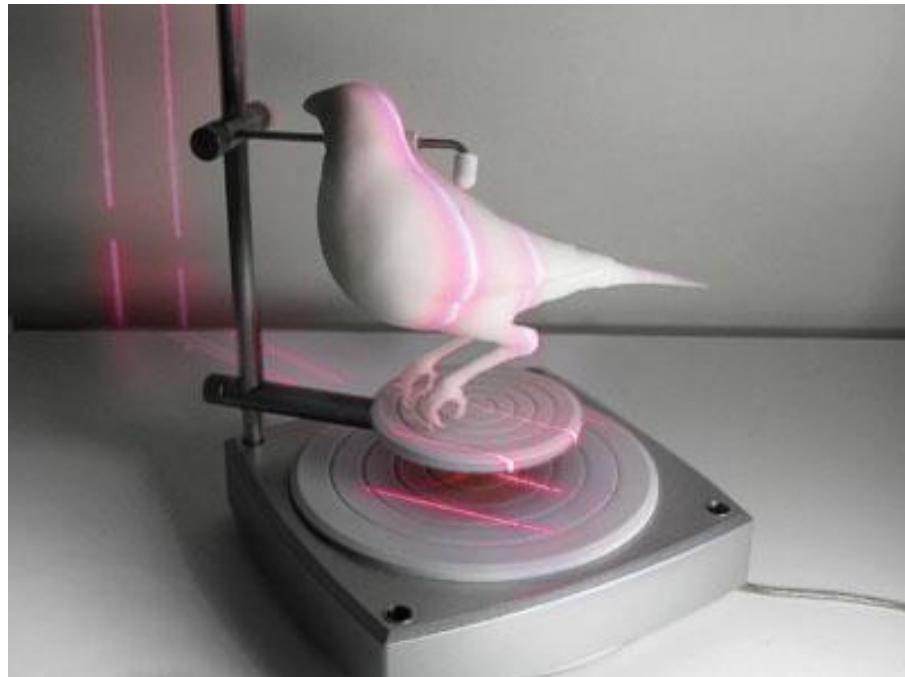


Registration

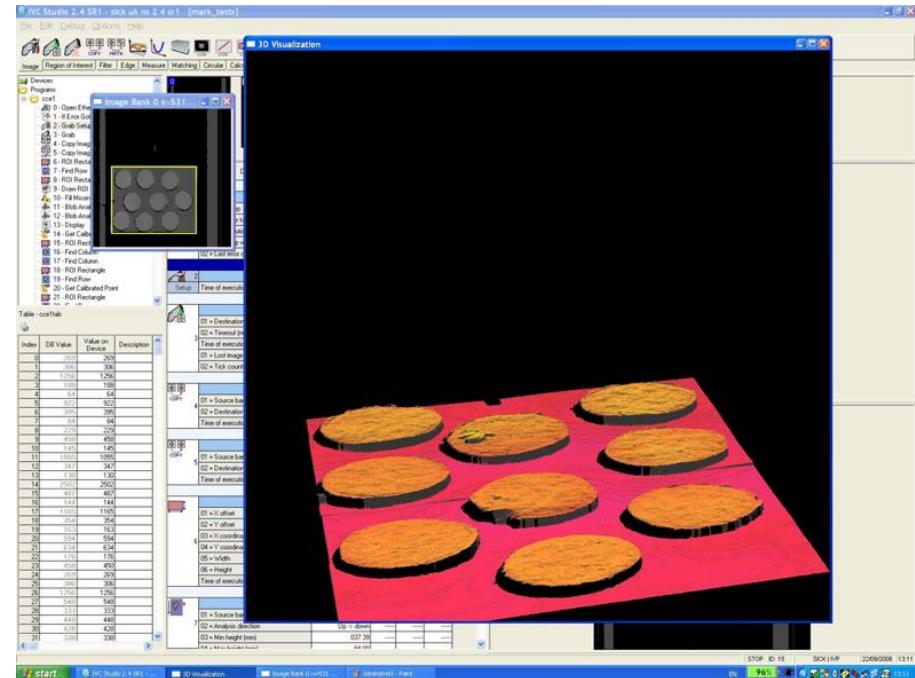
<http://dmfr.birjournals.org/content/33/4/226/F3.large.jpg>
<http://www-sop.inria.fr/asclepios/software/inriaviz4d/SphericalImTransp.png>
<http://www.creatis.insa-lyon.fr/site/sites/default/files/segm2.png>

Medical Imaging

Applications



Scanning



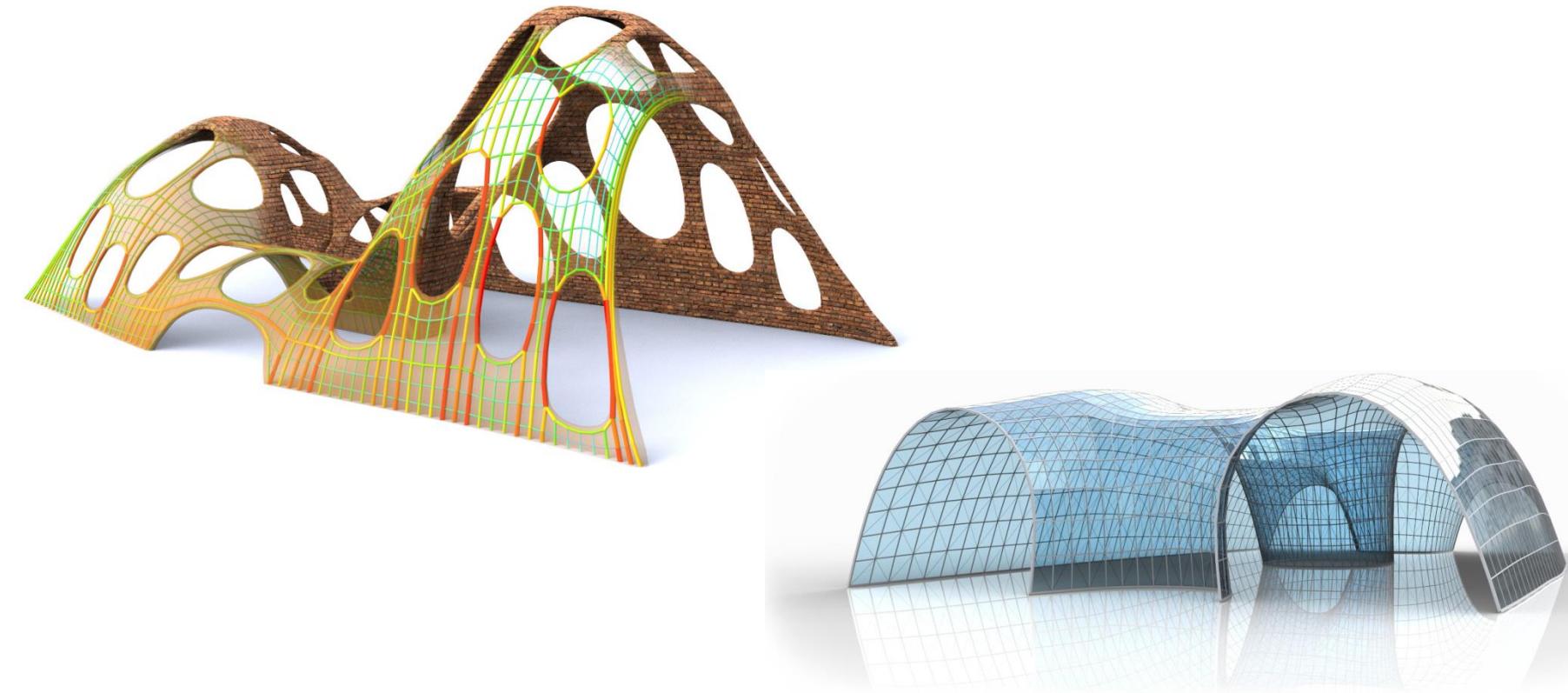
Defect detection

<http://www.conduitprojects.com/php/images/scan.jpg>

http://www.emeraldinsight.com/content_images/fig0330290204005.png

Manufacturing and Fabrication

Applications

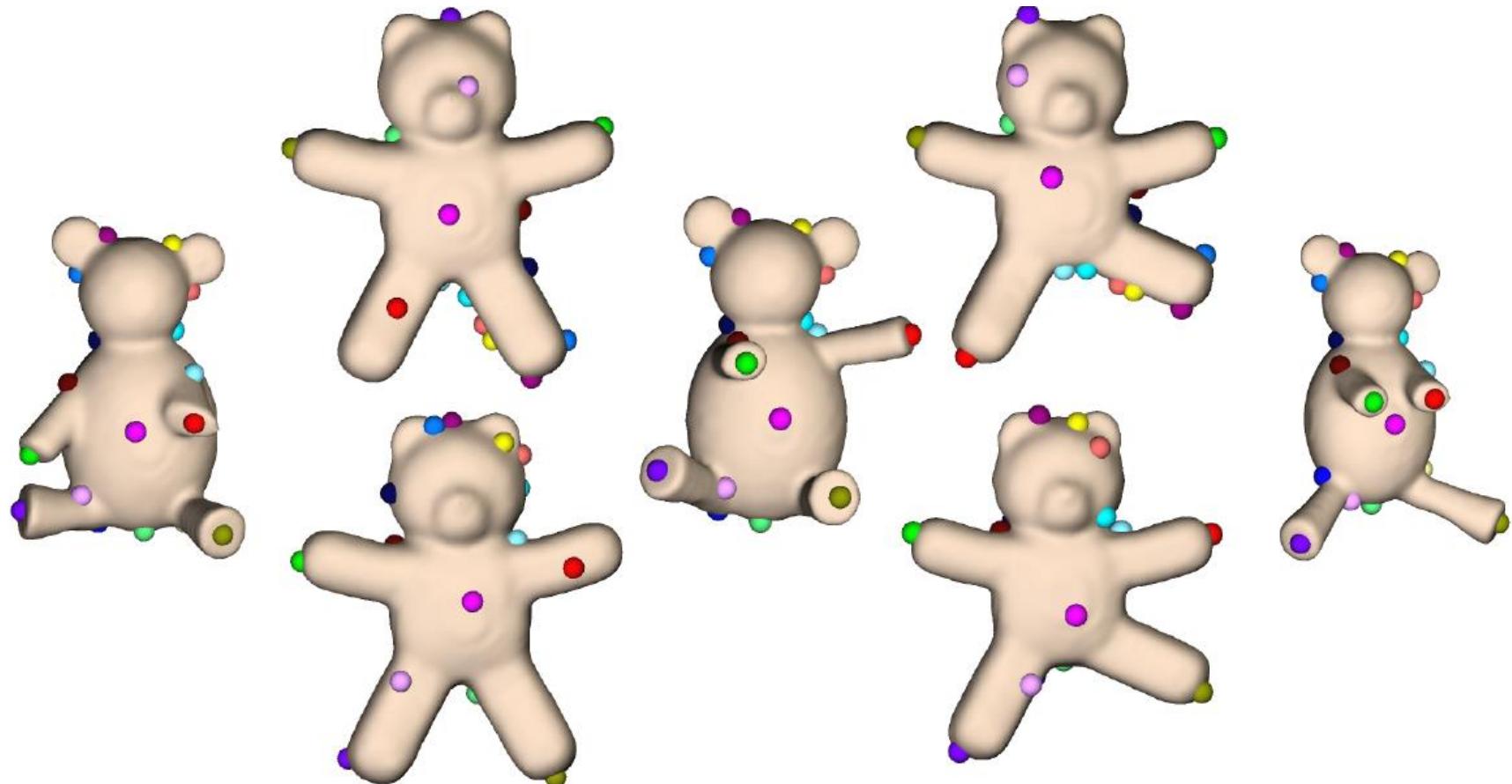


Design and analysis

http://gmsv.kaust.edu.sa/people/faculty/pottmann/pottmann_pdf/selfsupporting.pdf

Architecture

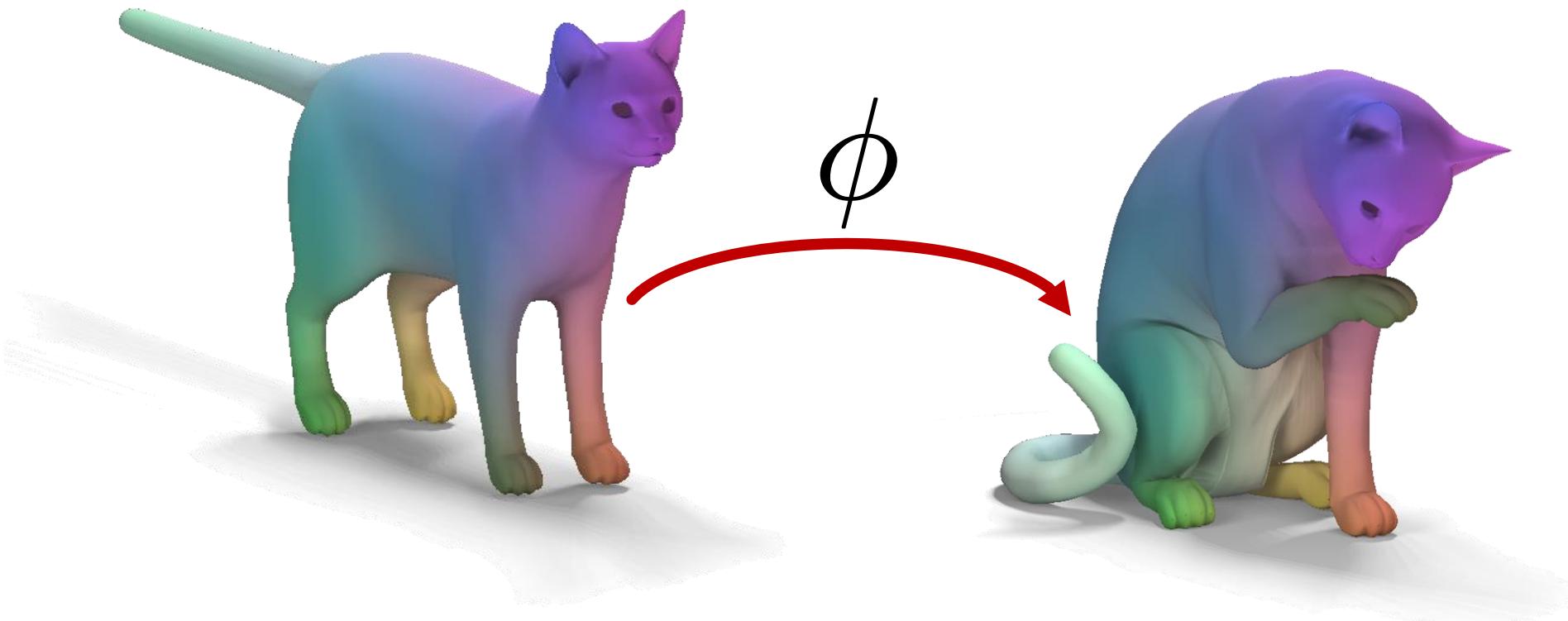
Applications



<http://graphics.stanford.edu/projects/lgl/papers/nbwyg-oaicsm-11/nbwyg-oaicsm-11.pdf>

Shape collection analysis

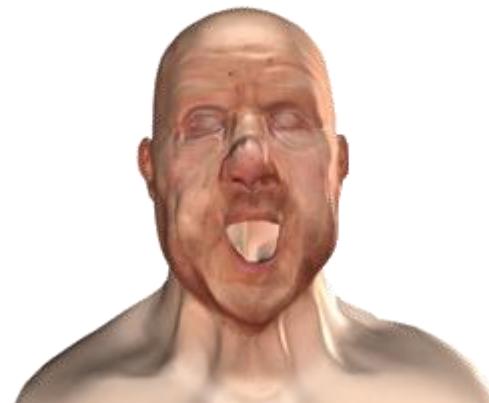
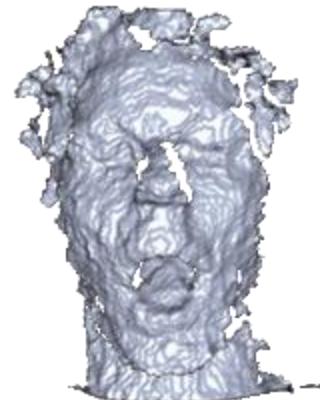
Applications



<http://people.csail.mit.edu/jsolomon/assets/fmaps.pdf>

Correspondence

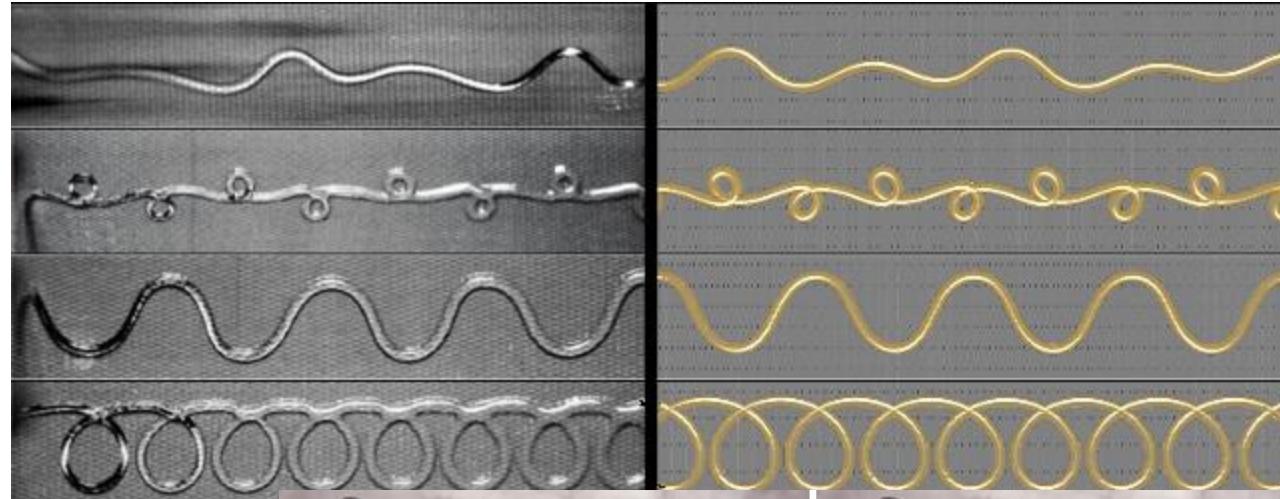
Applications



<http://www.hao-li.com/publications/papers/siggraph2011RPBFA.pdf>

Deformation transfer

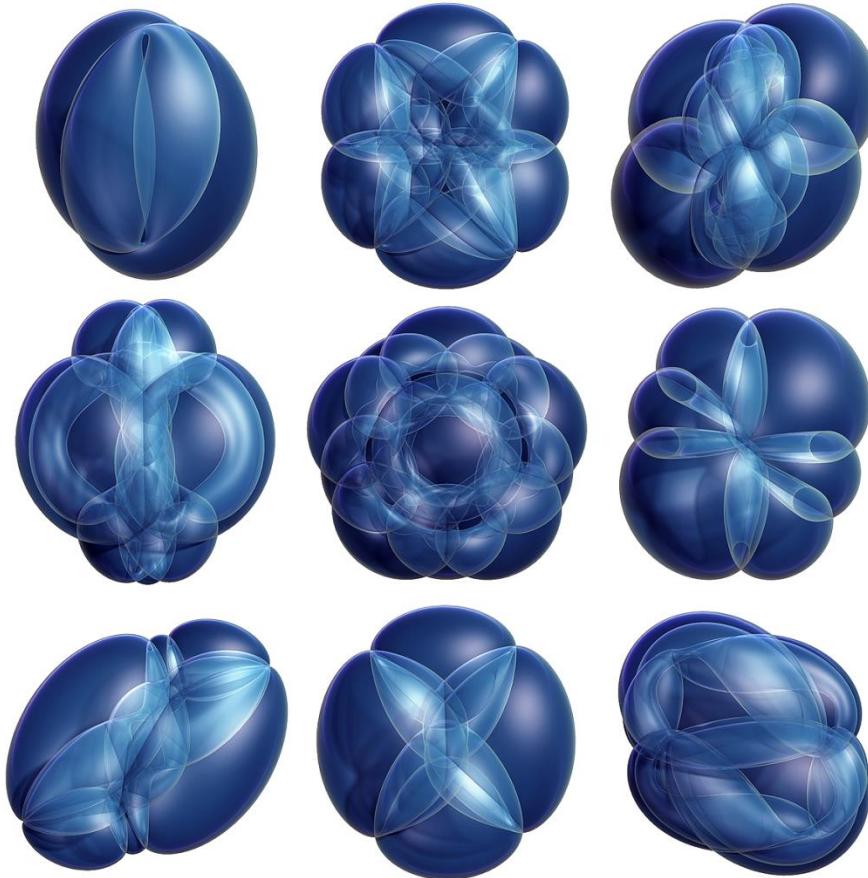
Applications



<http://www.cs.columbia.edu/cg/threads/> <http://mbergou.com/>

Simulation

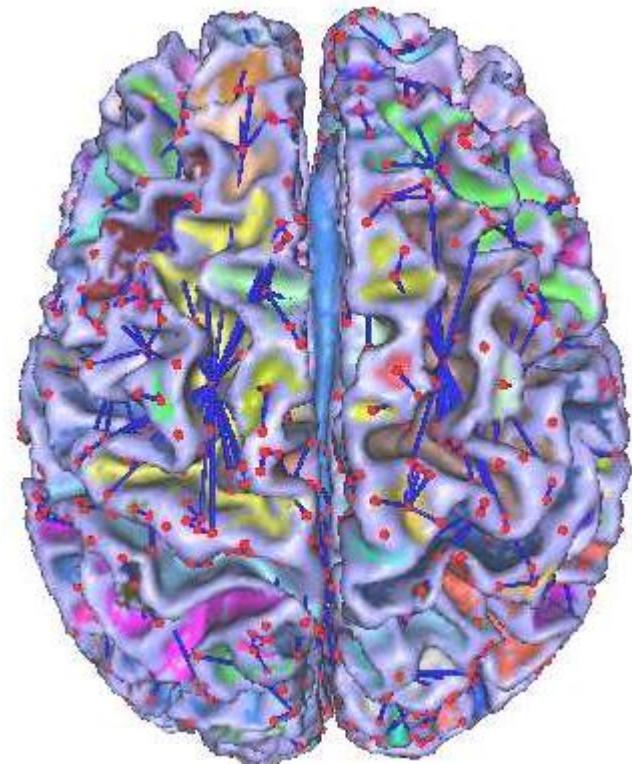
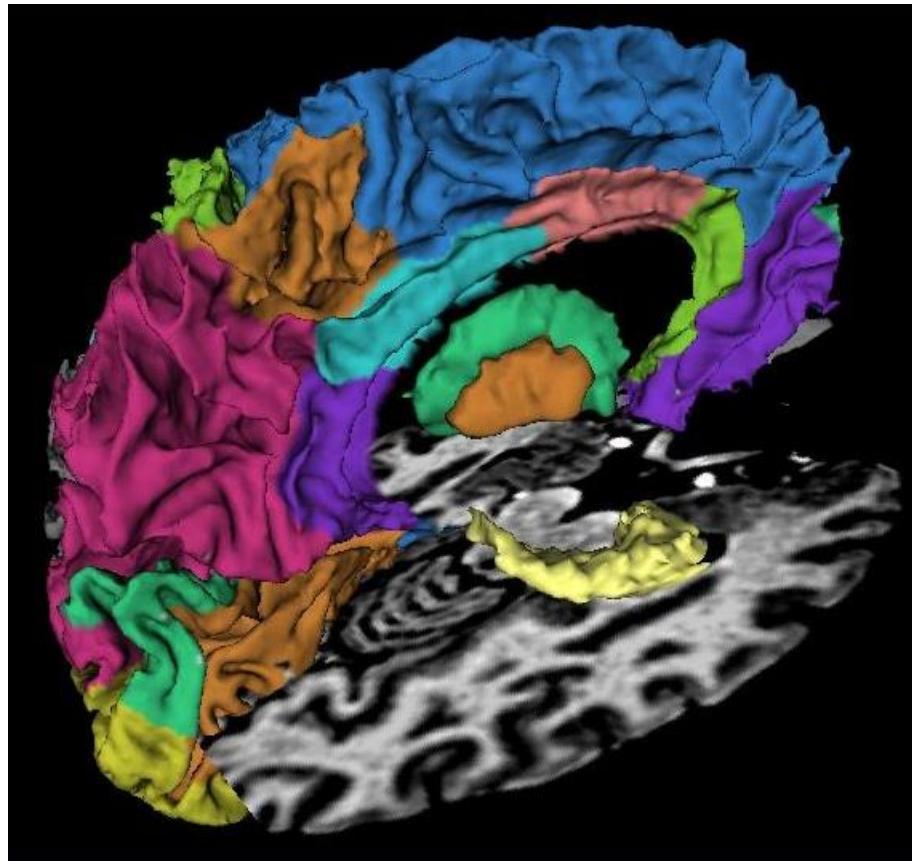
Applications



<http://multires.caltech.edu/~keenan/pdf/spinxform.pdf>

Scientific visualization

Applications



<http://www.bioinformaticslaboratory.nl/twiki/pub/EBioScience/News/freesurfer-3d.jpg>
http://hal.inria.fr/docs/00/40/21/30/IMG/vivodtzev_et_al-Dagstuhlo3.jpg

Segmentation

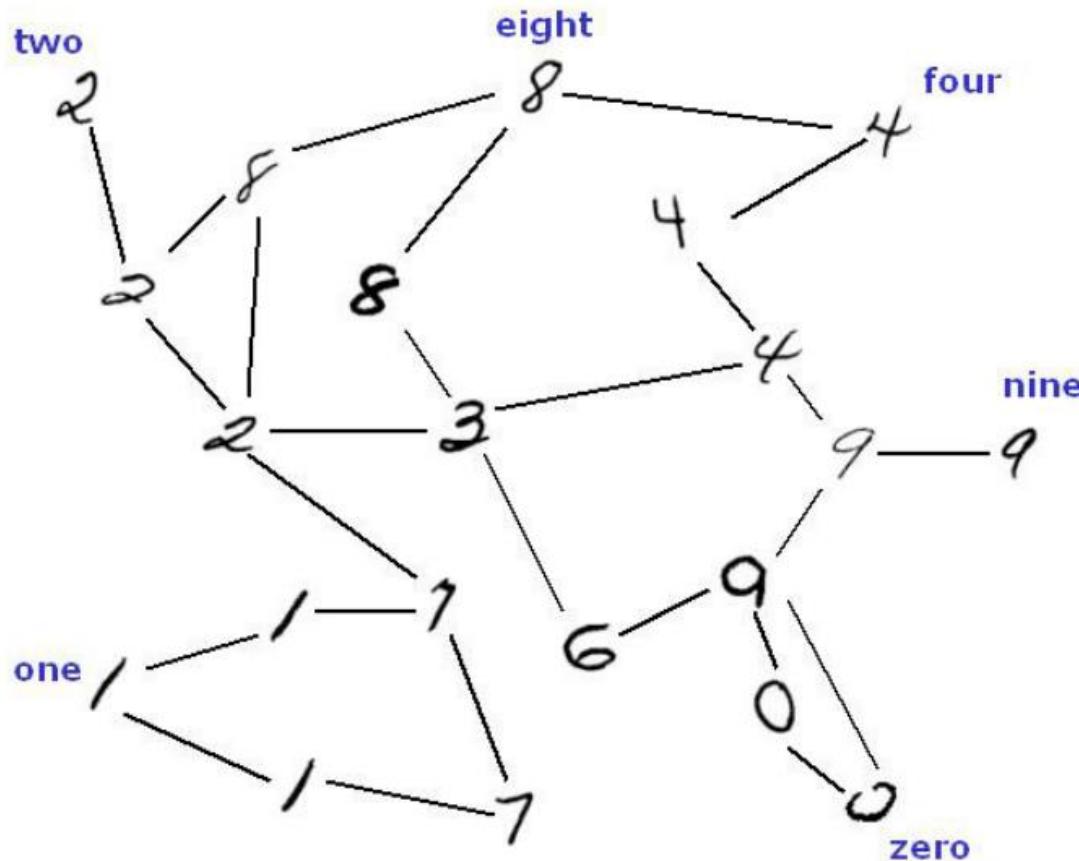
Applications



Su et al. *Estimating image depth using shape collections*. SIGGRAPH 2014.

Computer vision

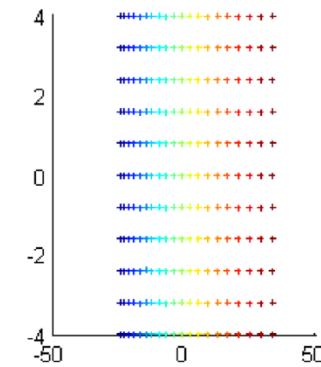
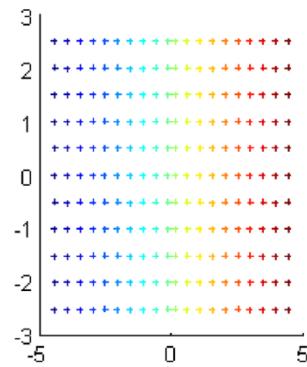
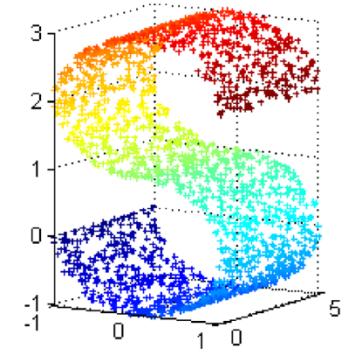
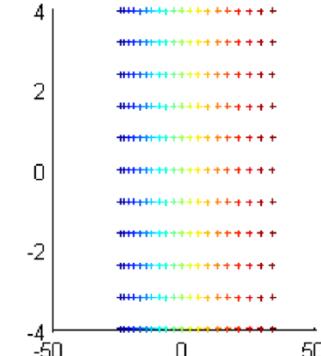
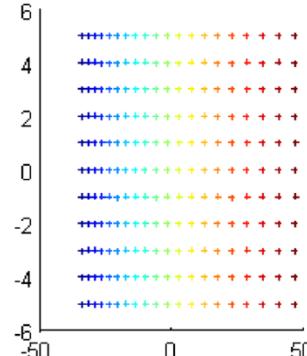
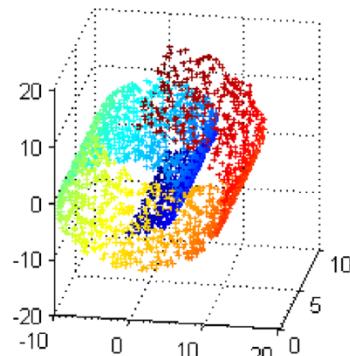
Applications



Zhu et al. *Semi-Supervised Learning Using Gaussian Fields and Harmonic Functions*. ICML 2003.

Machine learning

Applications



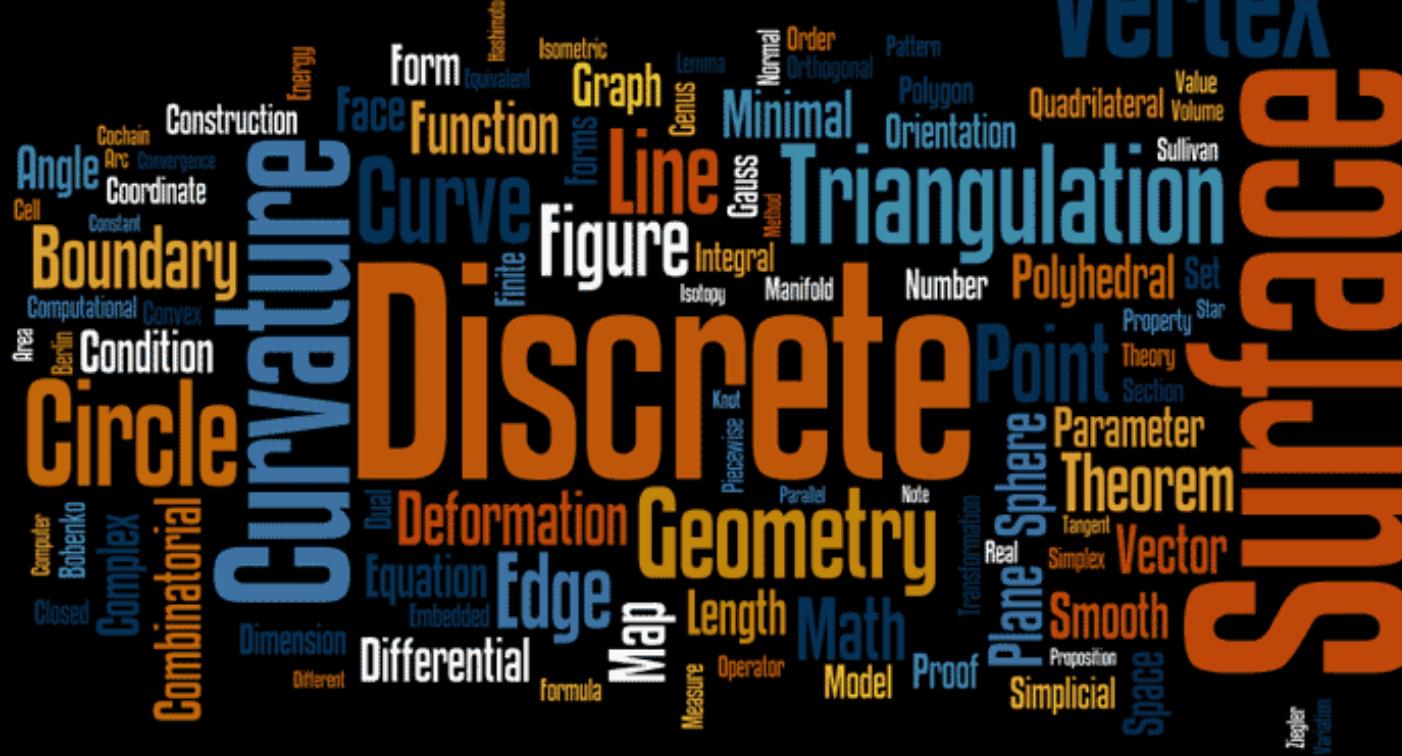
(a)

(b)

(c)

Hou et al. *Novel semisupervised high-dimensional correspondences learning method*. Opt. Eng. 2008.

Statistics



6.838: Shape Analysis

Justin Solomon

MIT, Spring 2017

