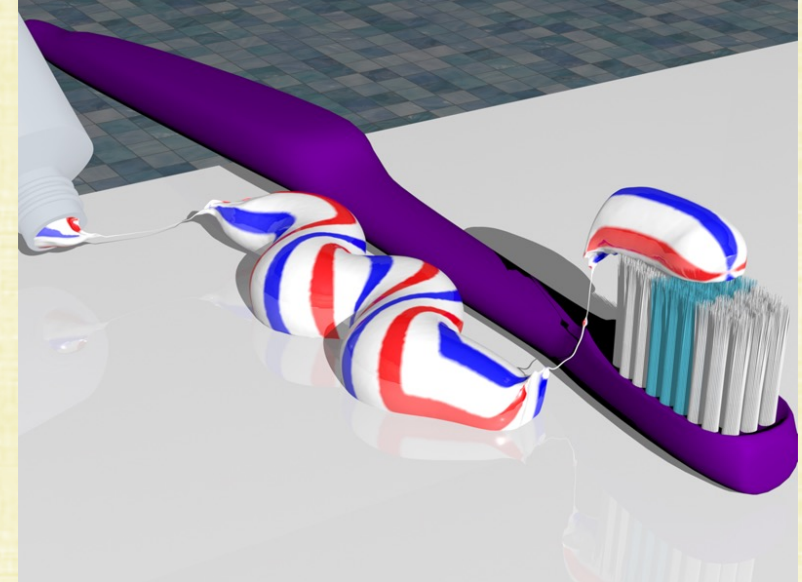
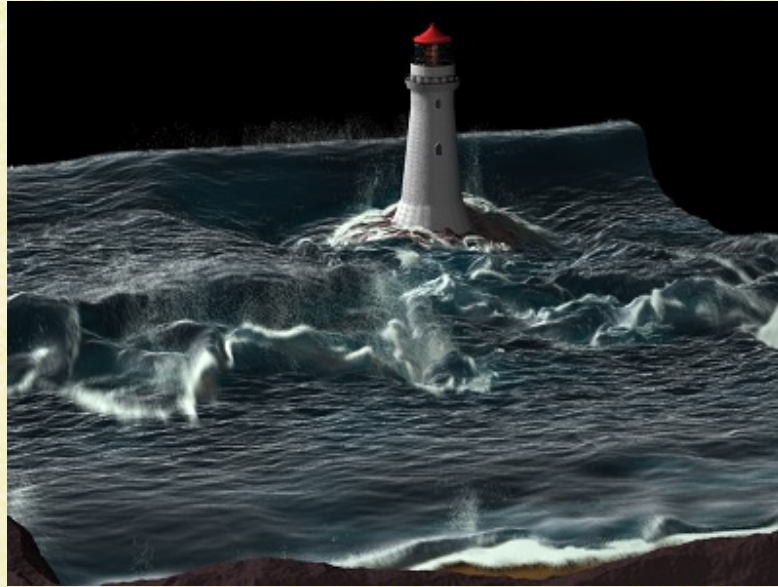
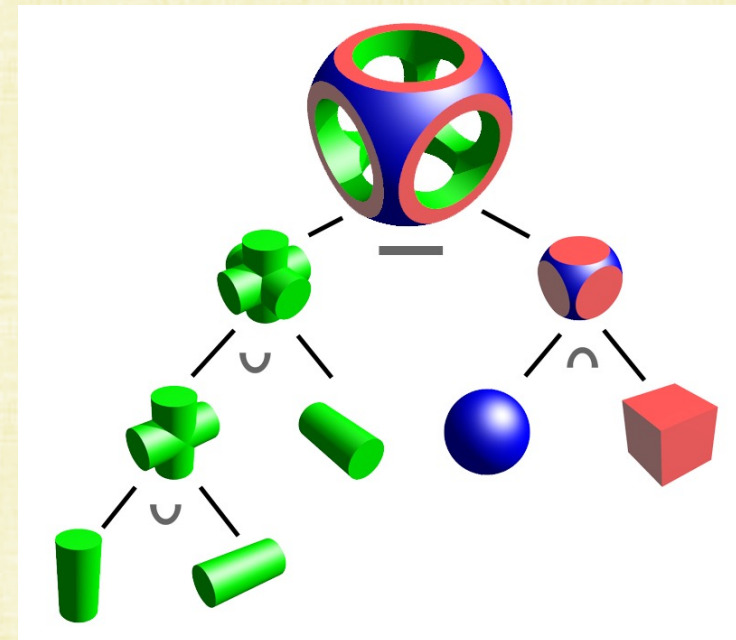
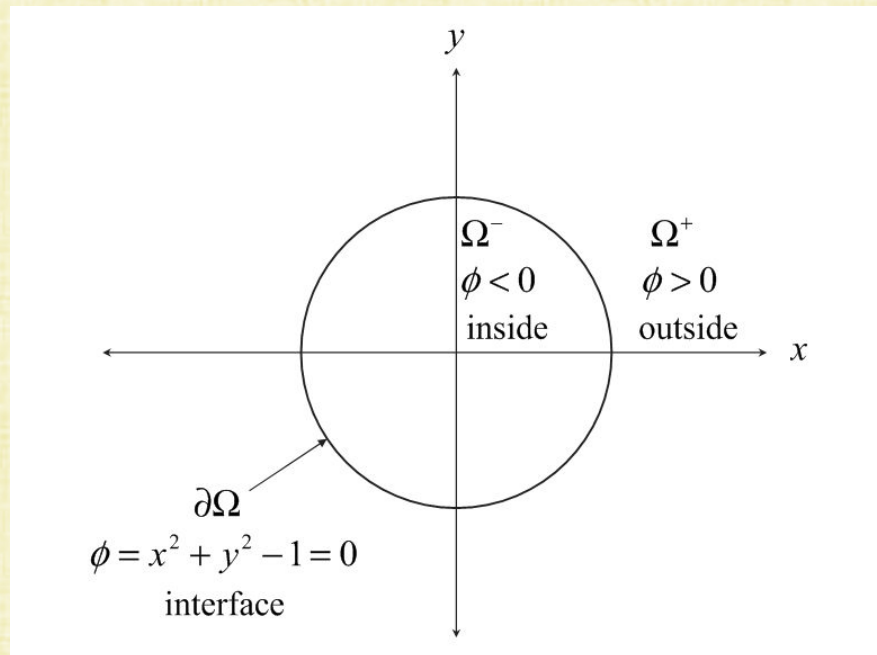


More Geometric Modeling



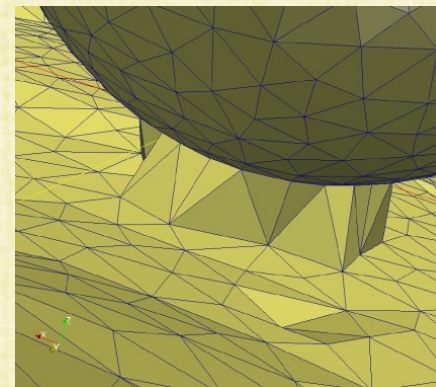
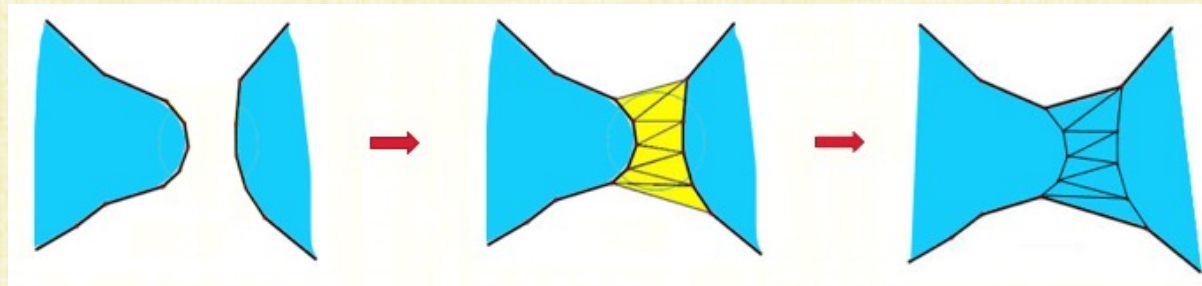
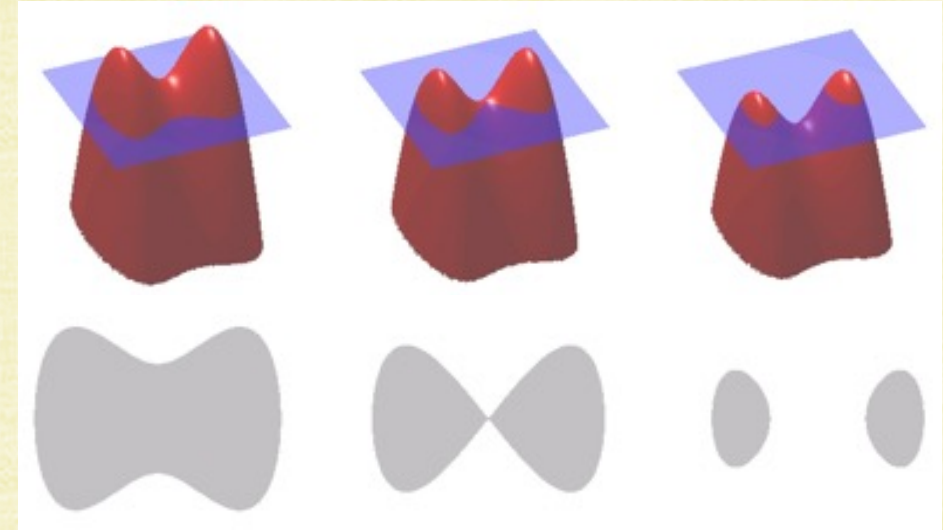
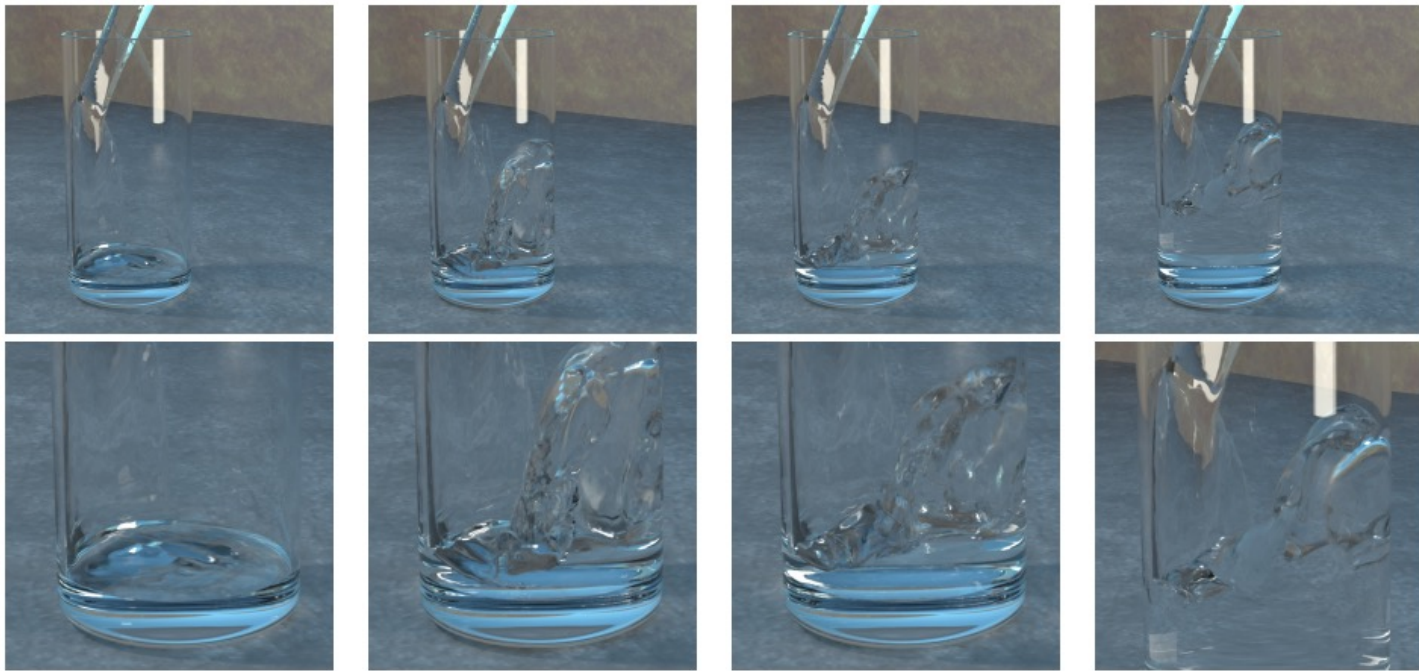
Implicit Surfaces

- Define a function $\phi(x)$ over $\forall x \in R^3$
- Interior Ω^- is defined by $\phi(x) < 0$, and exterior Ω^+ defined by $\phi(x) > 0$
- The surface $\partial\Omega$ is defined by $\phi(x) = 0$
 - We have already seen planes/spheres (and lines/rays/circles in 2D) defined by implicit surfaces
 - Easy to ray trace implicitly defined geometry
- Easy to check whether a point x is inside/outside: just evaluate $\phi(x)$
- Constructive Solid Geometry (CSG) operations: union, difference, intersection, etc.



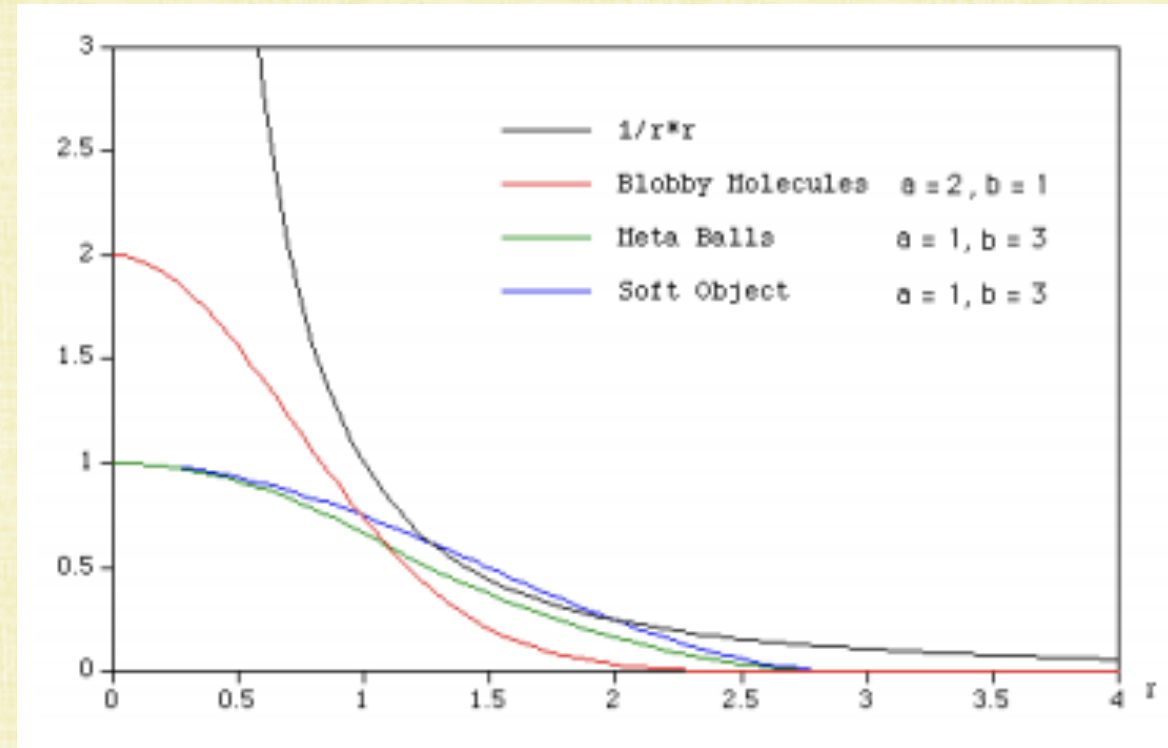
Topological Change

- Greatly superior to triangle meshes for topological change!

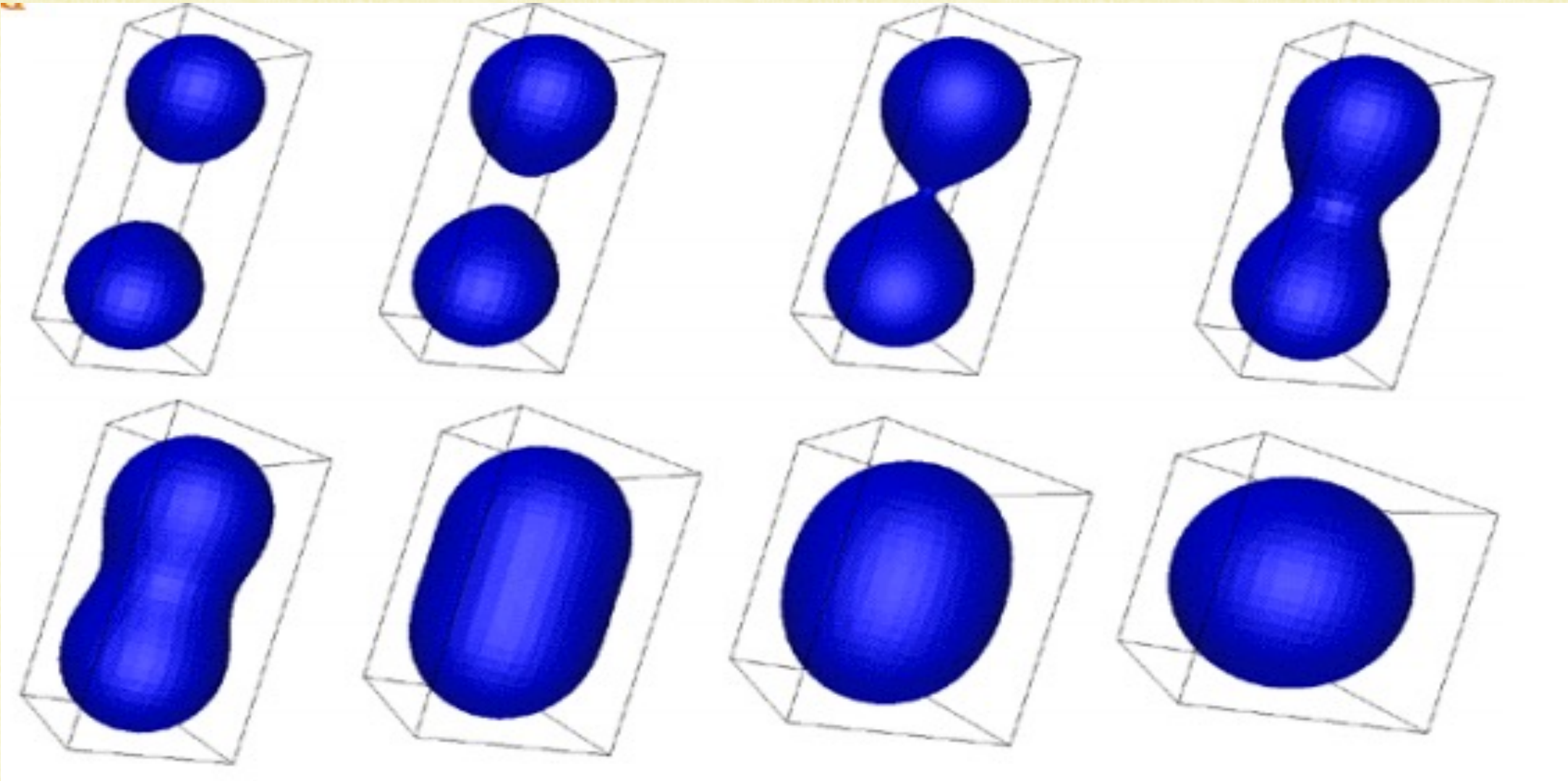


Blobbies

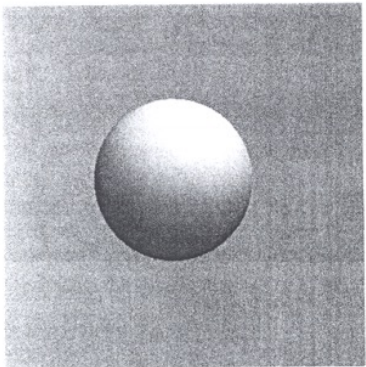
- Each blob is defined as a density function around a particle
- Blob kernels can be: 2D ellipses, 2D diamonds, 3D spheres, etc.
- For each pixel, the aggregate density is summed from all overlapping blobs
- Also known as:
 - Metaballs (in Japan), Soft objects (in Canada and New Zealand)
 - Slightly different density kernel functions



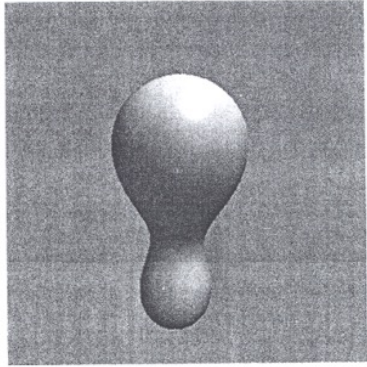
Topological Change



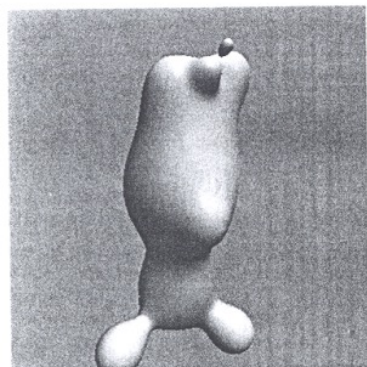
Bloppy Modeling



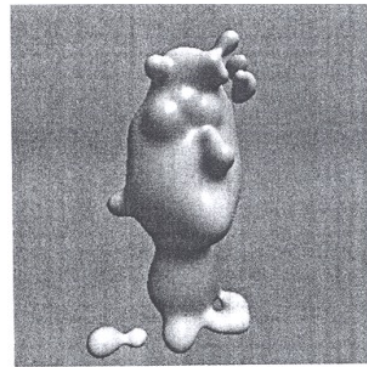
(a) $N = 1$



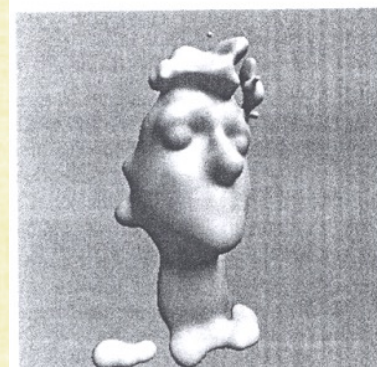
(b) $N = 2$



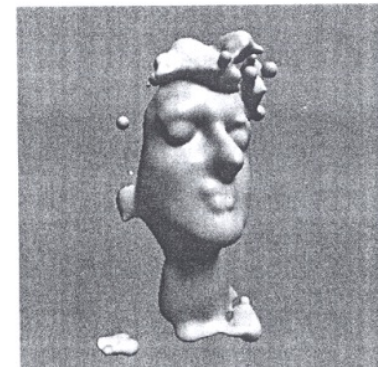
(c) $N = 10$



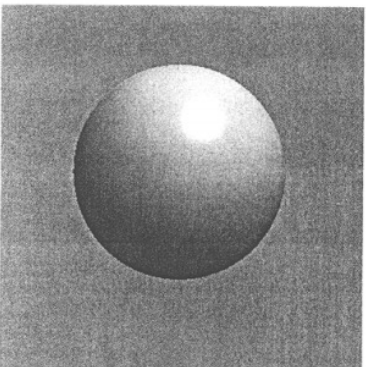
(d) $N = 35$



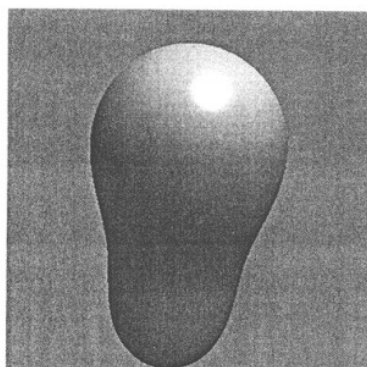
(e) $N = 70$



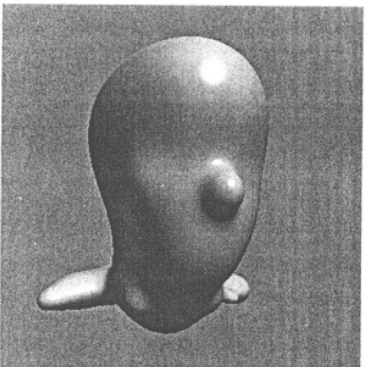
(f) $N = 243$



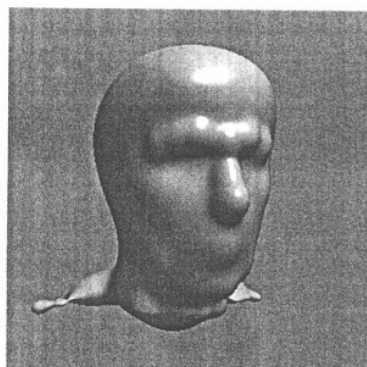
(a) $N = 1$



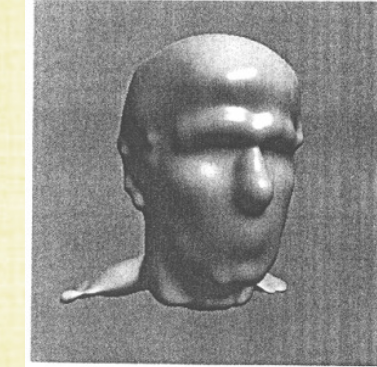
(b) $N = 2$



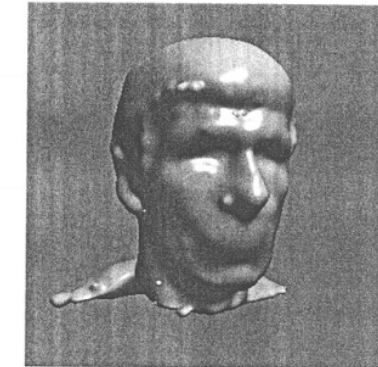
(c) $N = 20$



(d) $N = 60$



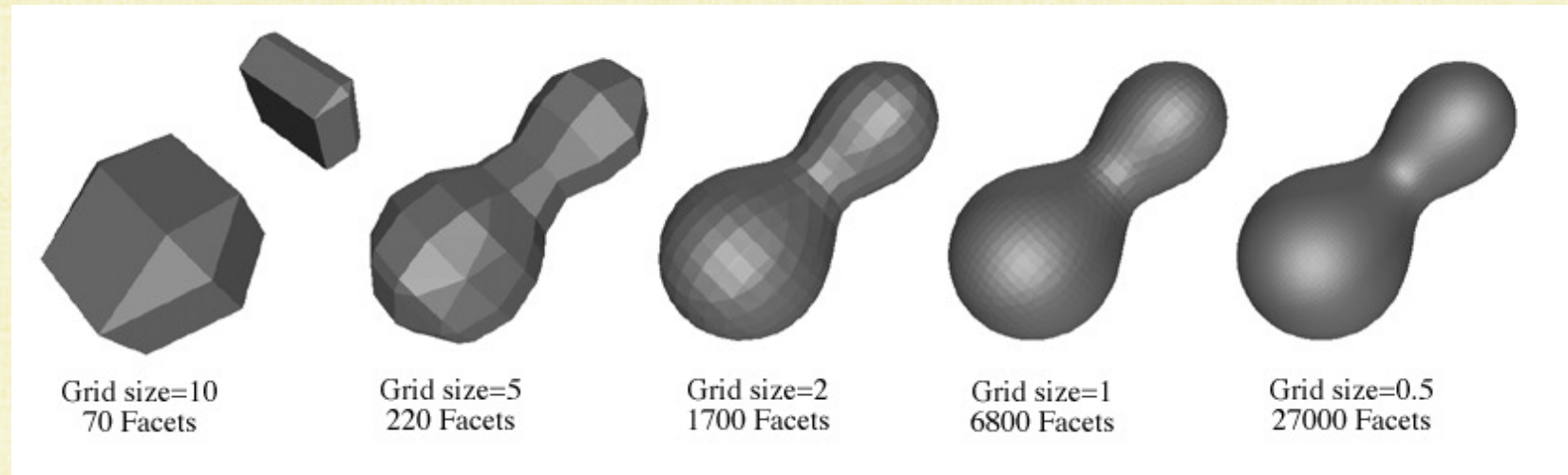
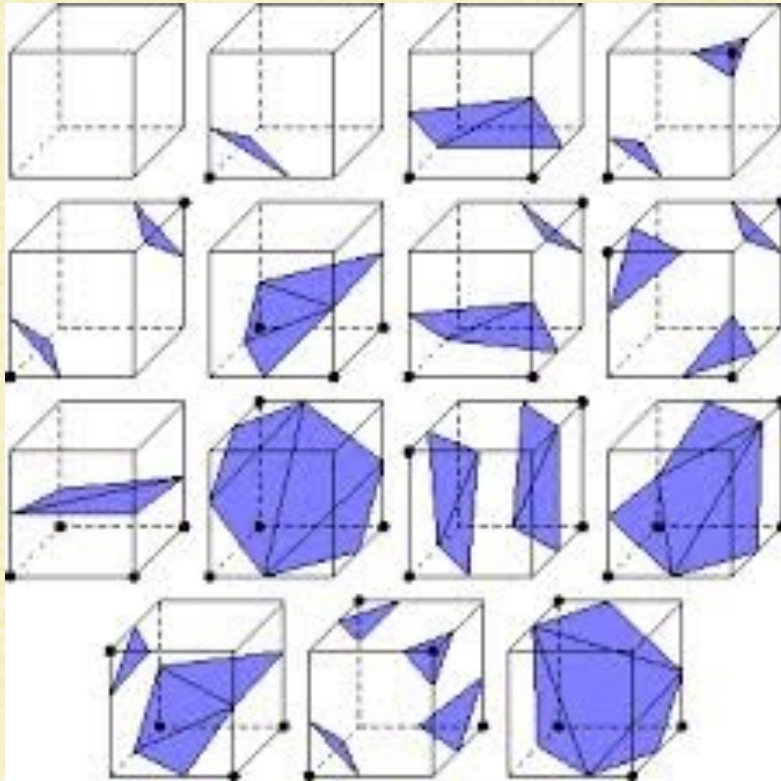
(e) $N = 120$



(f) $N = 451$

Marching Cubes (or Marching Tetrahedra)

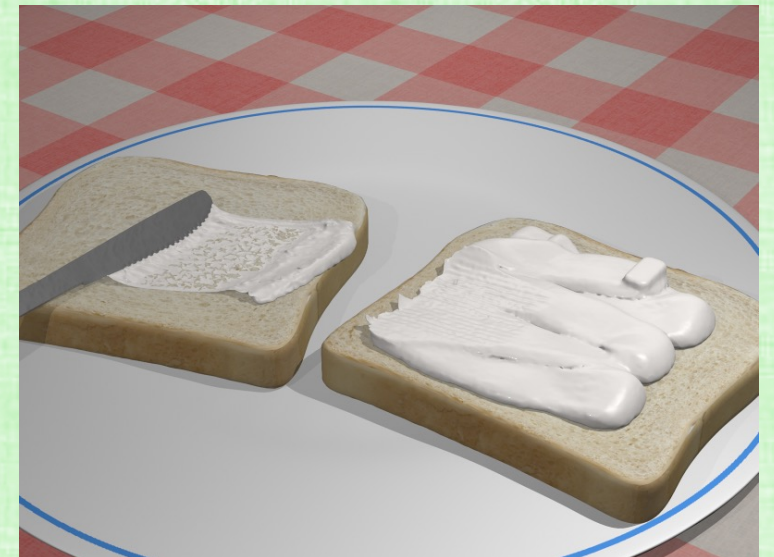
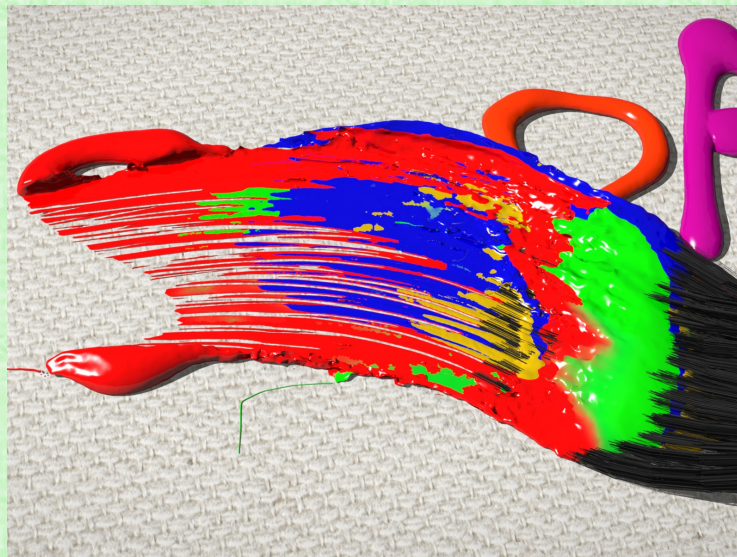
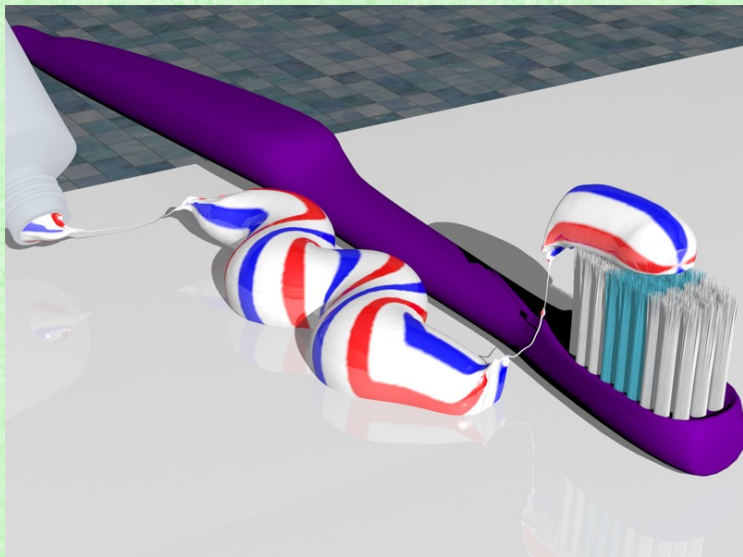
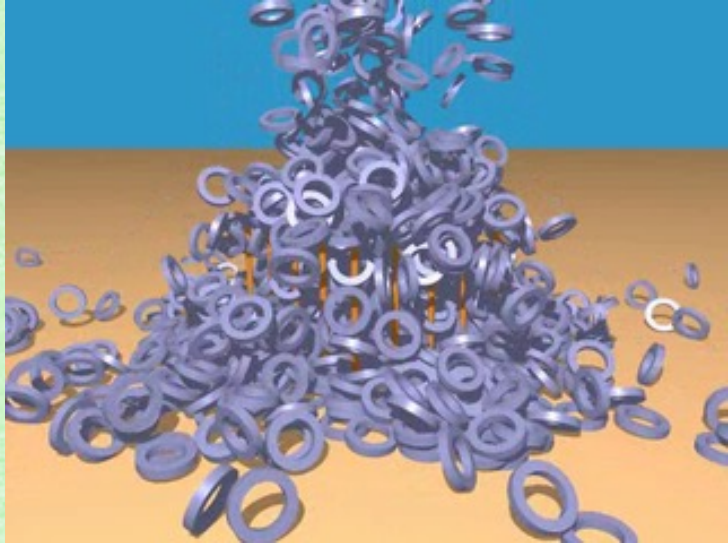
- Turns an implicit surface into triangles
 - Define the implicit surface on a 3D grid
 - For each grid cell, use the topology of the volume to construct surface triangles



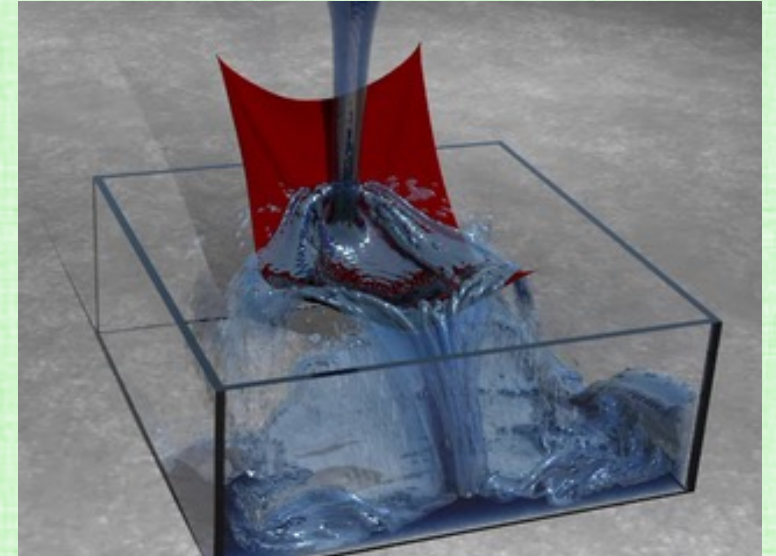
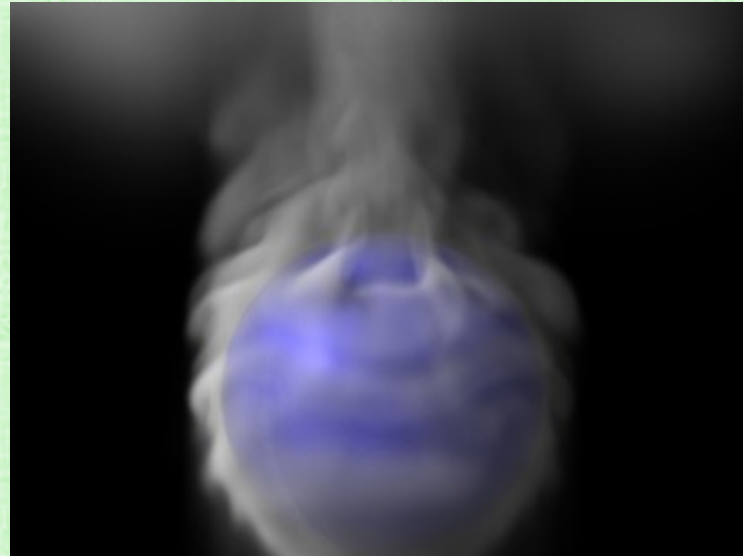
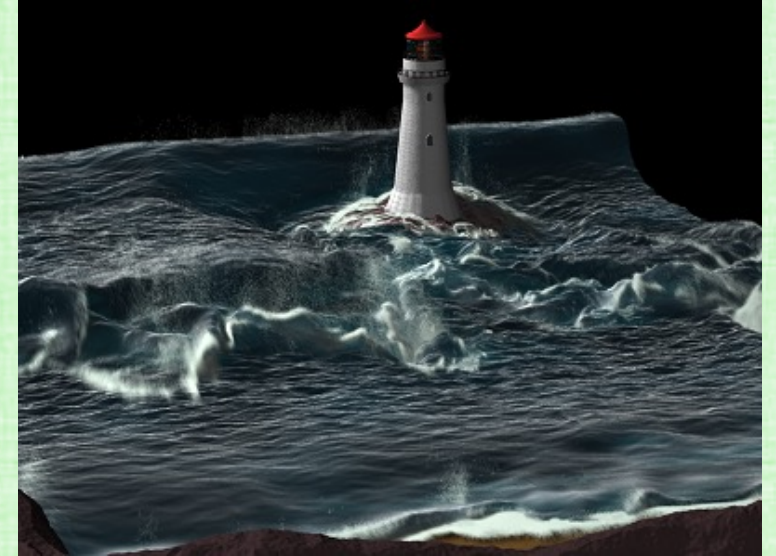
Newton's Second Law (for Physics Simulations)

- Kinematics describe position $X(t)$ and velocity $V(t)$ as function of time t
 - $\frac{dX(t)}{dt} = V(t)$ or $X'(t) = V(t)$
- Dynamics describe responses to external stimuli
 - Newton's second law $F(t) = MA(t)$ is a dynamics equation
 - $V'(t) = A(t)$ implies $V'(t) = \frac{F(t)}{M}$ as well as $\frac{d^2X(t)}{dt^2} = X''(t) = \frac{F(t)}{M}$
- Combining kinematics and dynamics gives:
$$\begin{pmatrix} X'(t) \\ V'(t) \end{pmatrix} = \begin{pmatrix} V(t) \\ \frac{F(t, X(t), V(t))}{M} \end{pmatrix}$$
 - Note: forces often depend on position/velocity
- Much of the physical world can be simulated with computational mechanics (FEM) and computational fluid dynamics (CFD), using Newton's second law
 - Create degrees of freedom, specify forces, and solve the resulting ordinary differential equations (ODEs)
 - Spatially interdependent forces lead to partial differential equations (PDEs)

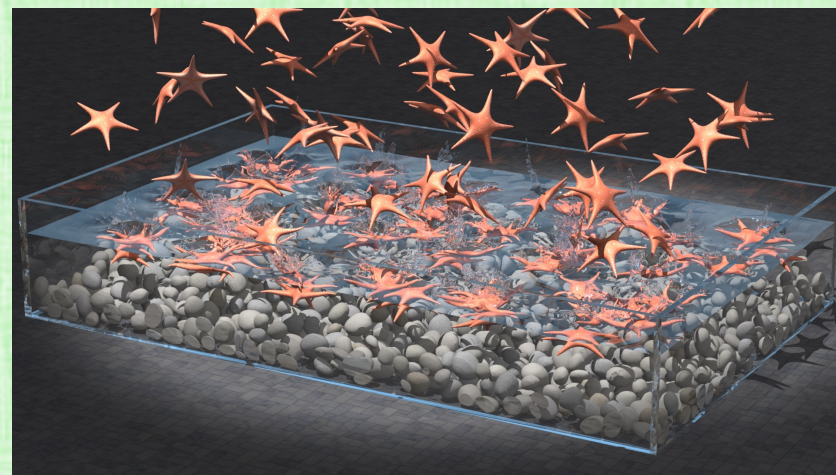
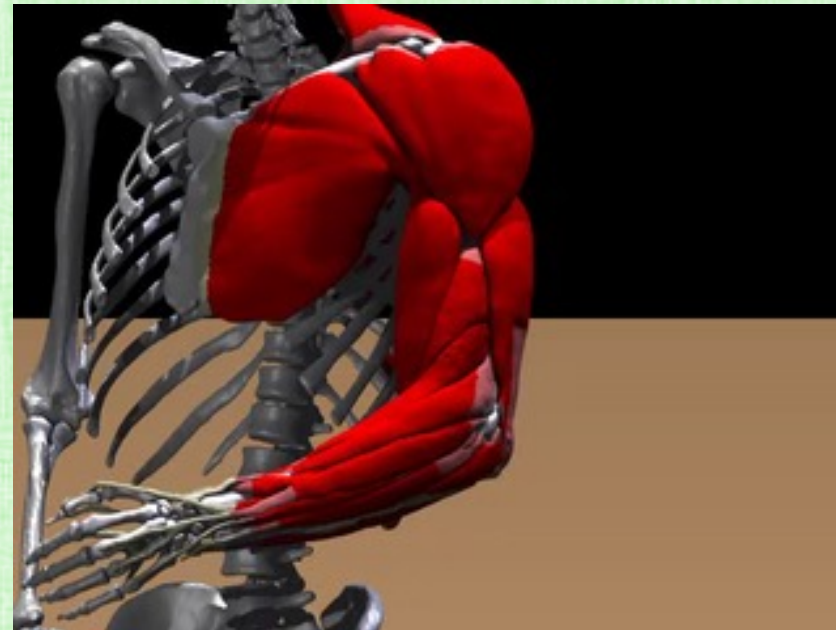
Computational Mechanics (FEM)



Computational Fluid Dynamics (CFD)



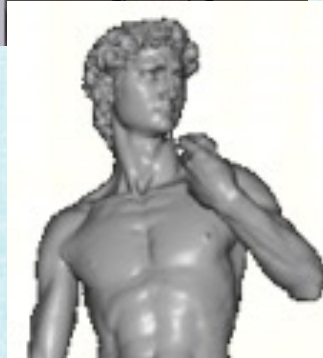
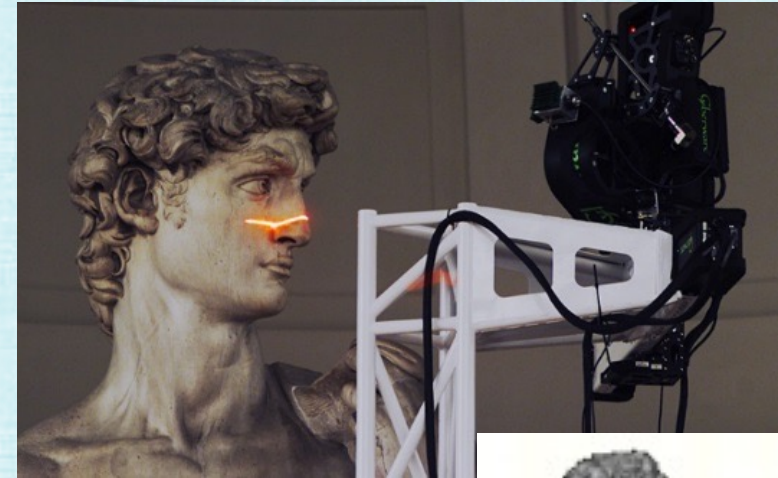
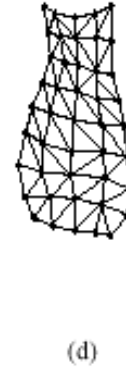
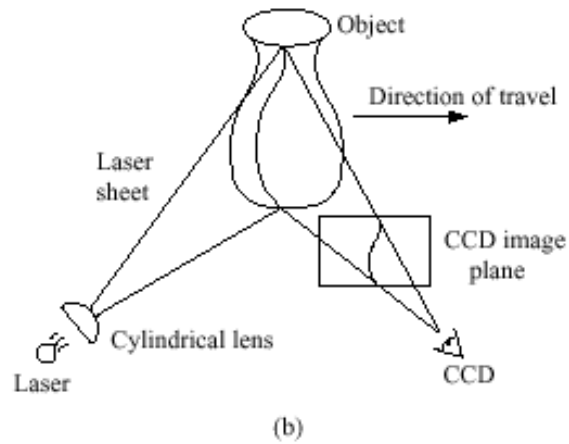
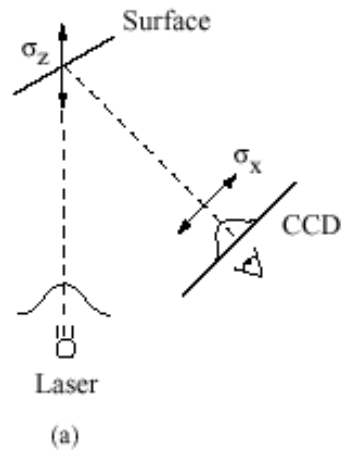
Computational Biomechanics



Computer Vision

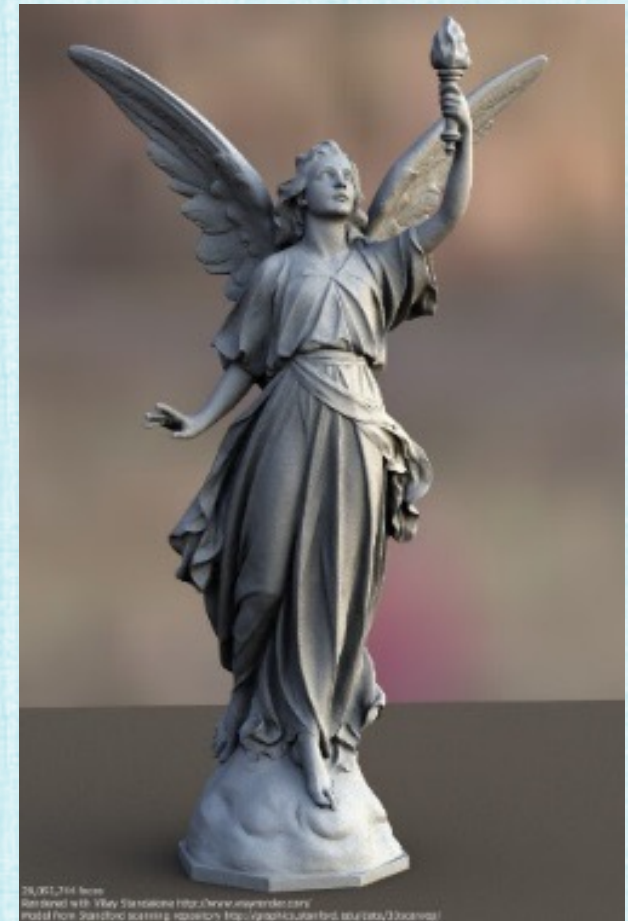
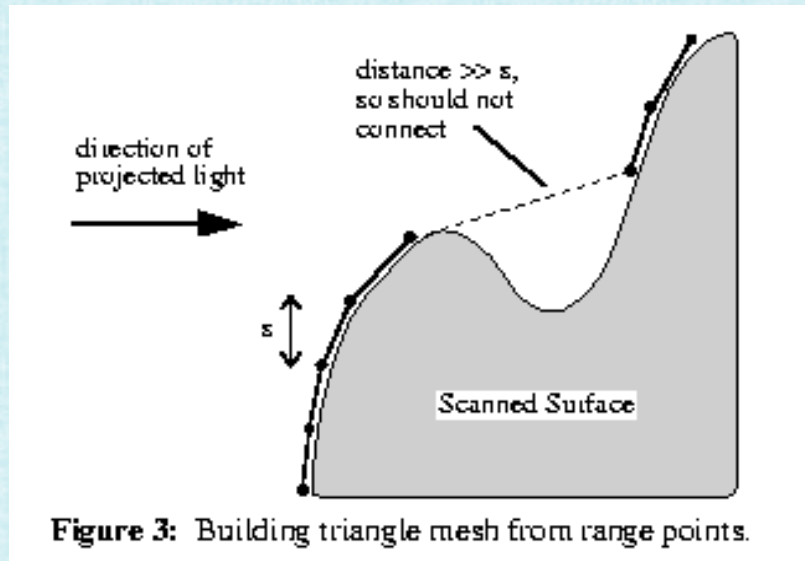
Range Scanners

- Senses 3D positions on an object's surface, and returns a range image:
 - $m \times n$ grid of distances (m points per laser sheet, n laser sheets)
- Multiple range images are aligned with transformations
 - Transformations determined via Iterative Closest Point (ICP) and related/similar methods
- Aligned range images are combined using a zippering algorithm

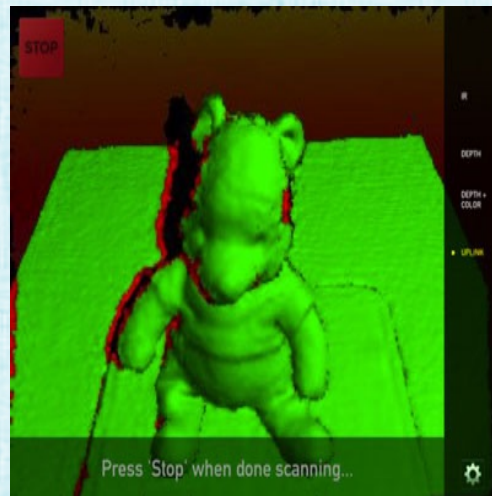
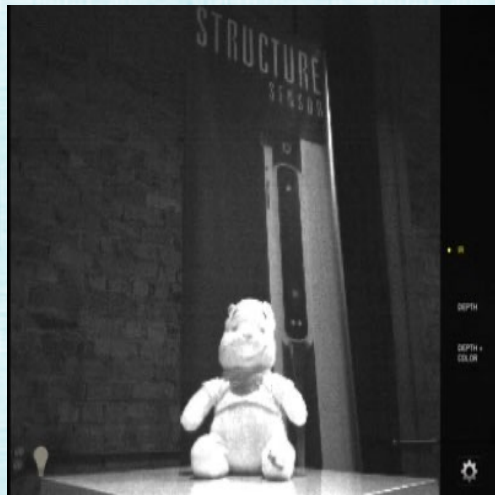
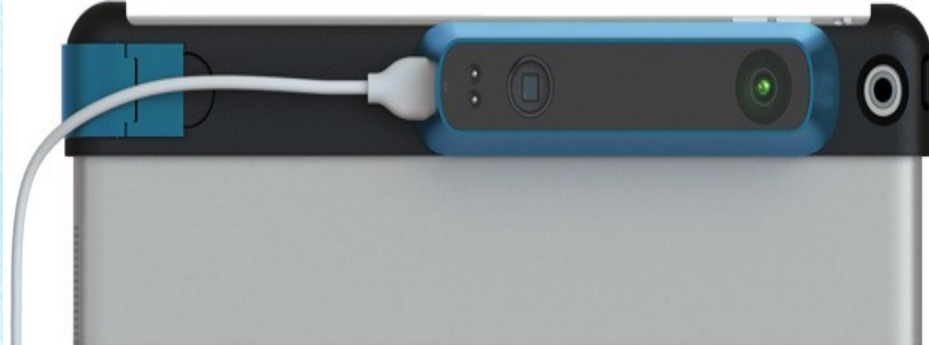


Range Scanners

- Each sample point in the $m \times n$ range image is a potential vertex in a triangle mesh
- Special care is required to avoid joining together vertices separated by depth discontinuities



Scanning w/Mobile Devices



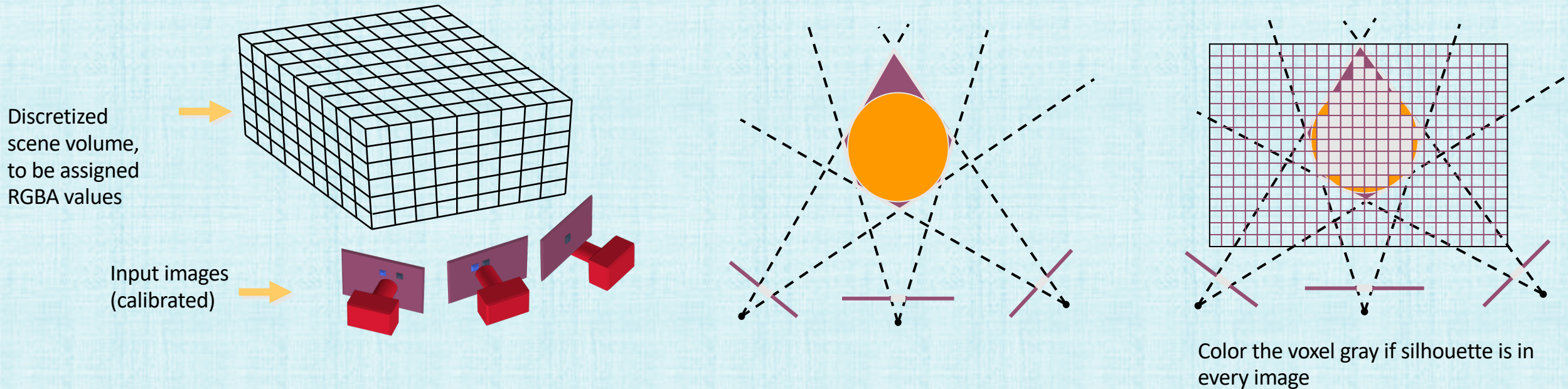
Structure Sensor for iPad



Autodesk 123D Catch

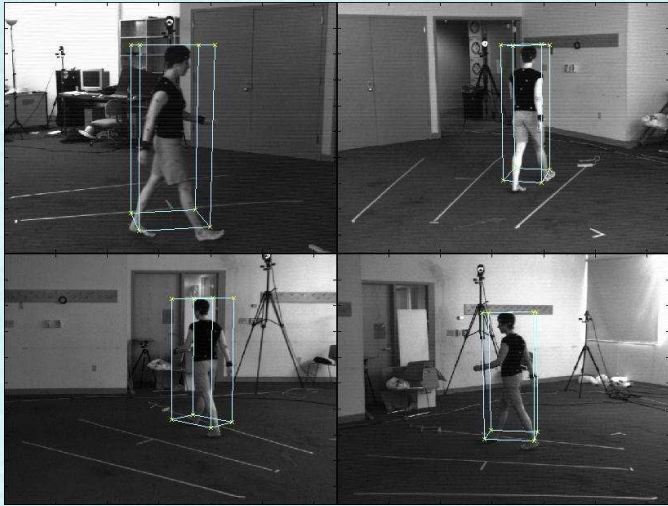
Voxel Carving

- Requires multiple images (from calibrated cameras) taken from different directions
- Construct a voxelized 3D model:
 - For each image, delete (carve away) voxels outside the object silhouette
- Colors can be projected onto the geometry

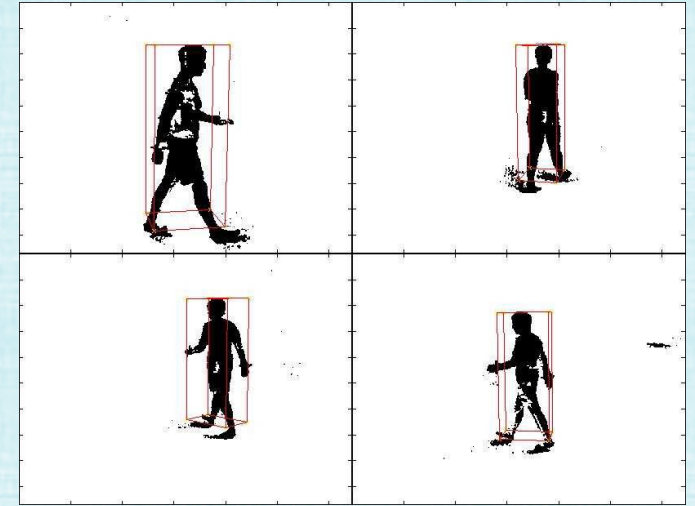


Voxel Carving

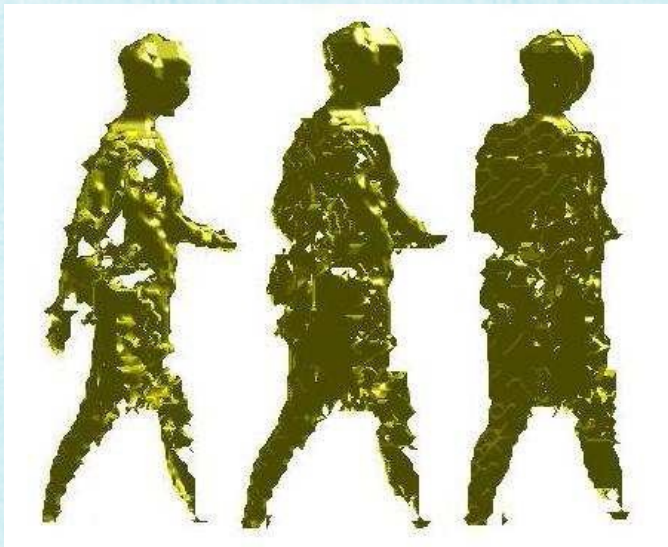
Original image



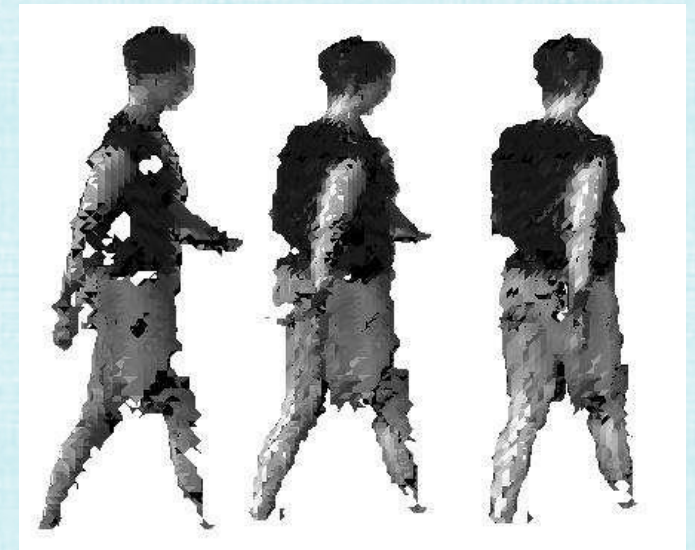
Extracted silhouettes



Carved out voxels

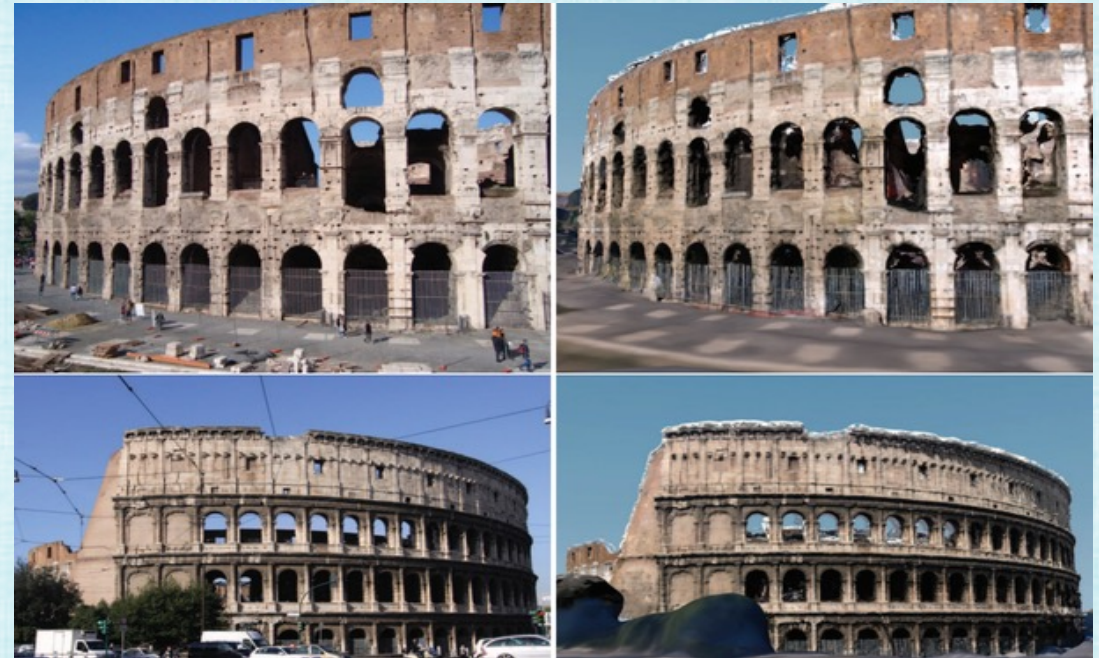
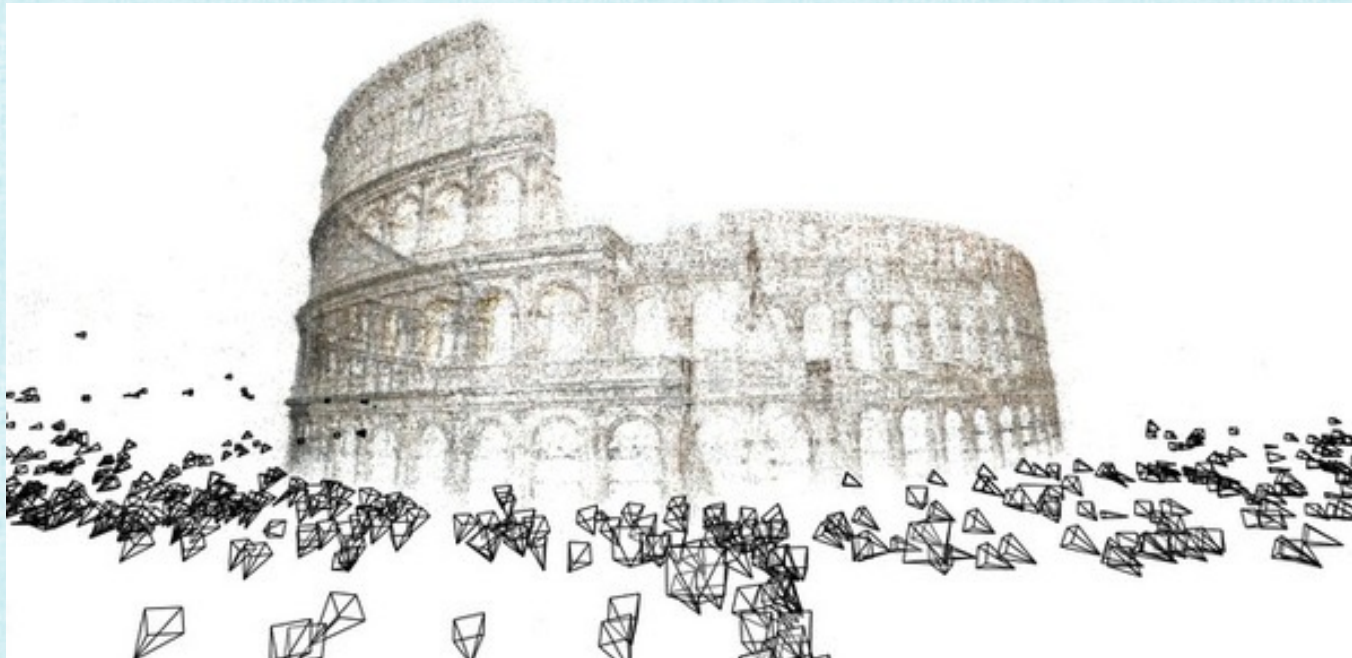


Back-projecting colors



Reconstruction from Large Photo Collections

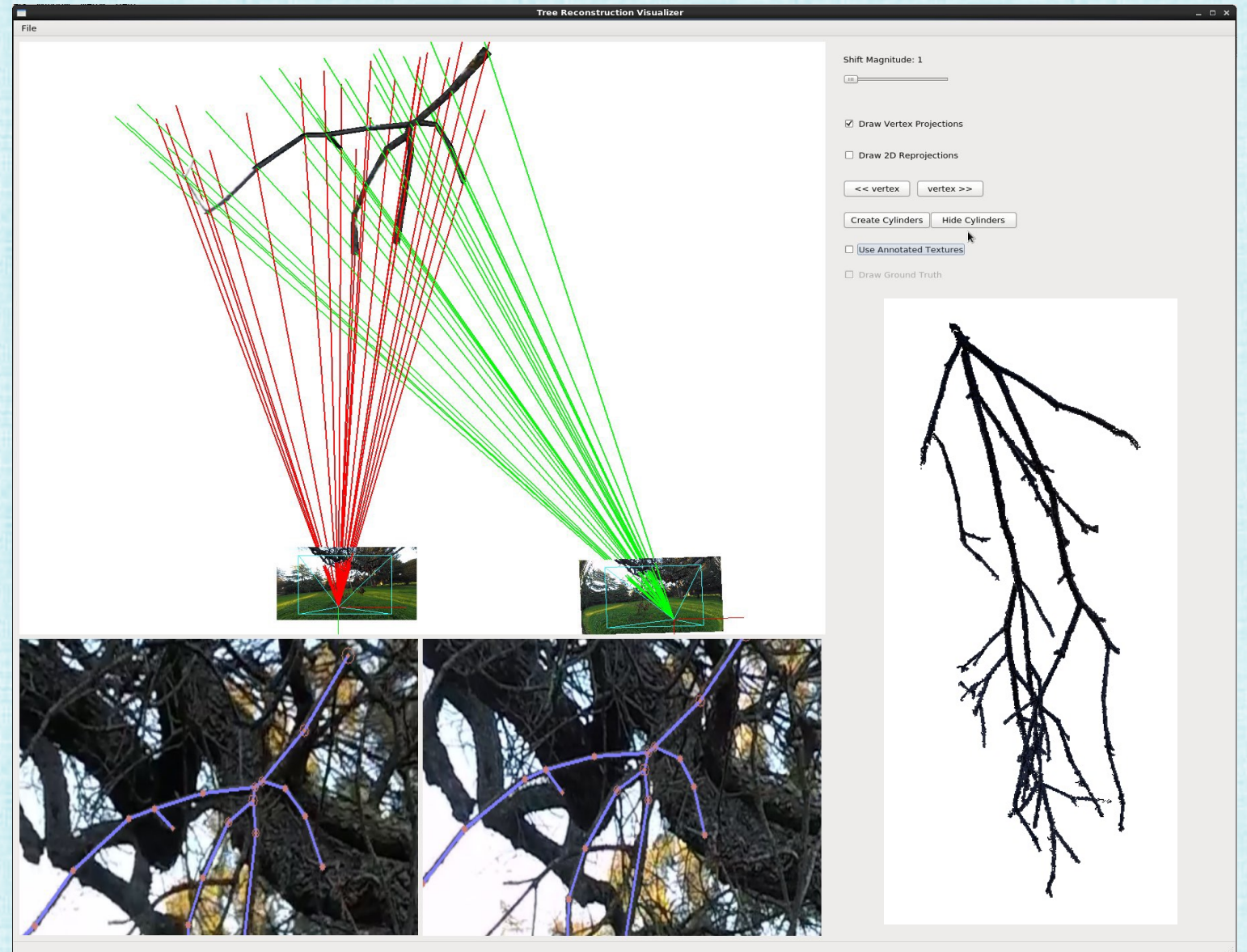
- Collect a large number of photos (e.g. from google images)
- Predict relative camera position/orientation for each image
- The position of a point that is visible in multiple images can be triangulated
- Obtain a sparse point cloud representation of the object
- Dense reconstruction algorithms can be used to improve the results



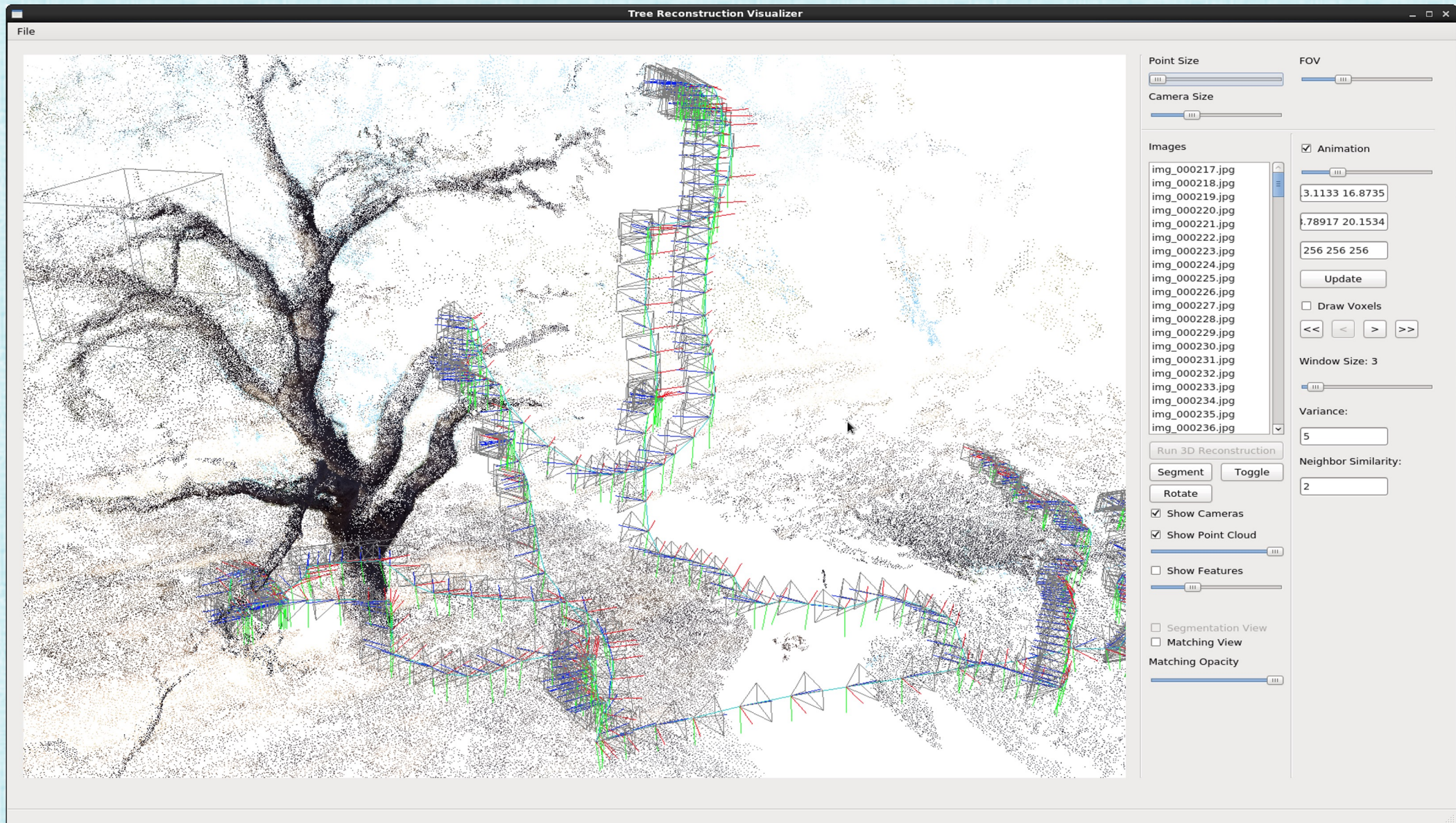
2D photos

3D geometry

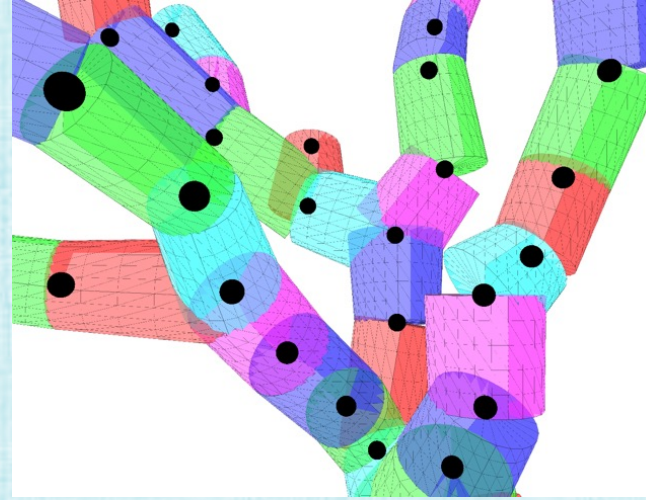
Drones & Trees



Drones & Trees



Drones & Trees

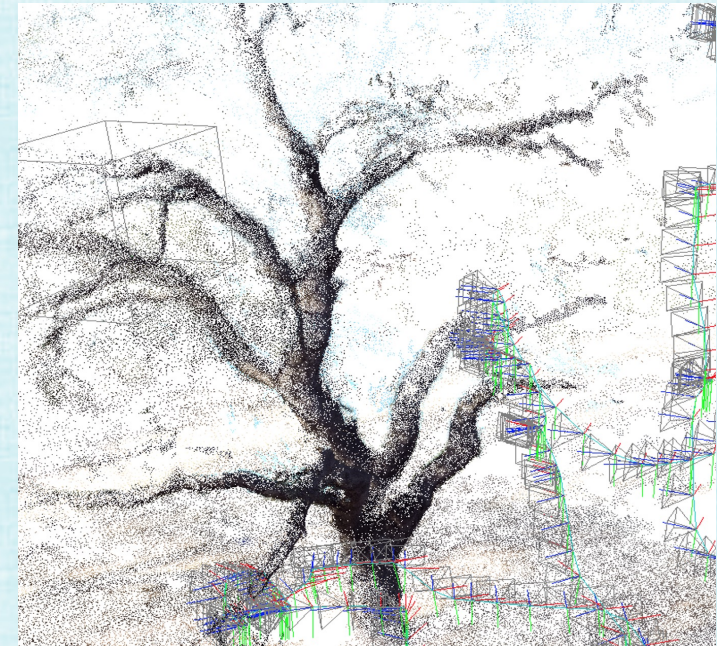
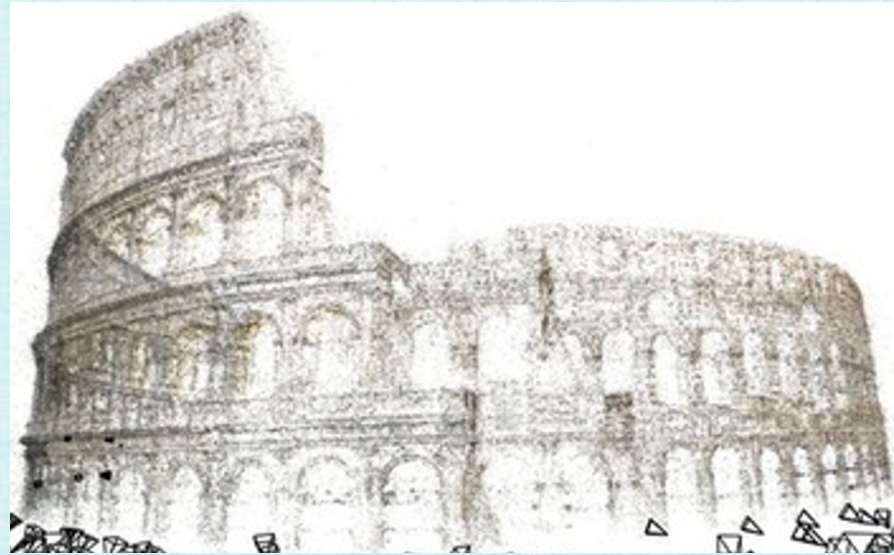


Drones & Trees



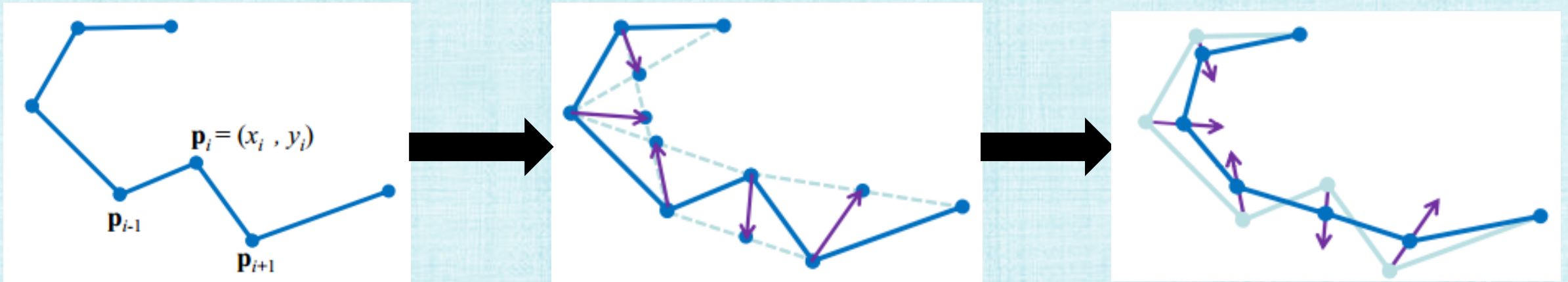
Noise & De-Noising

- Computer Vision algorithms use real-world sensors/data
- This results in noise corruption, which is the biggest drawback to such methods
- Denoising/smoothing algorithms are very important (for alleviating these issues)



Laplacian Smoothing

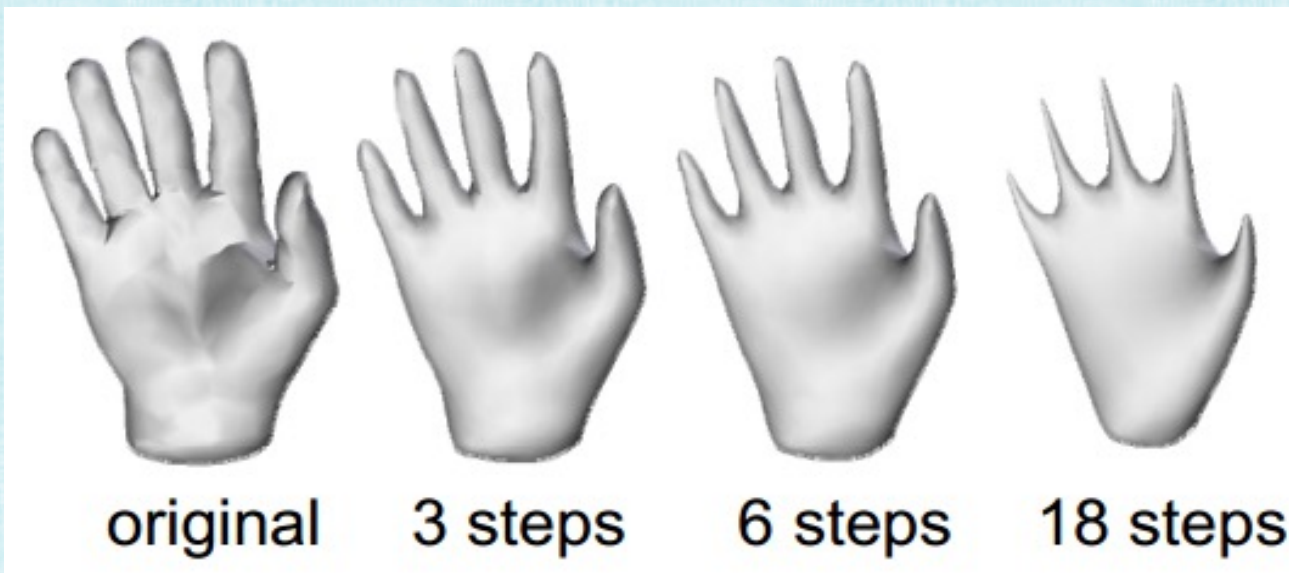
- Compute a Laplacian estimate using the one ring of vertices about a point
 - Similar to differential coordinates
- E.g., on a curve: $L(p_i) = \frac{1}{2}((p_{i+1} - p_i) + (p_{i-1} - p_i))$
- Then, update $p_i^{new} = p_i + \lambda L(p_i)$ where $\lambda \in (0,1)$
- Repeat several iterations



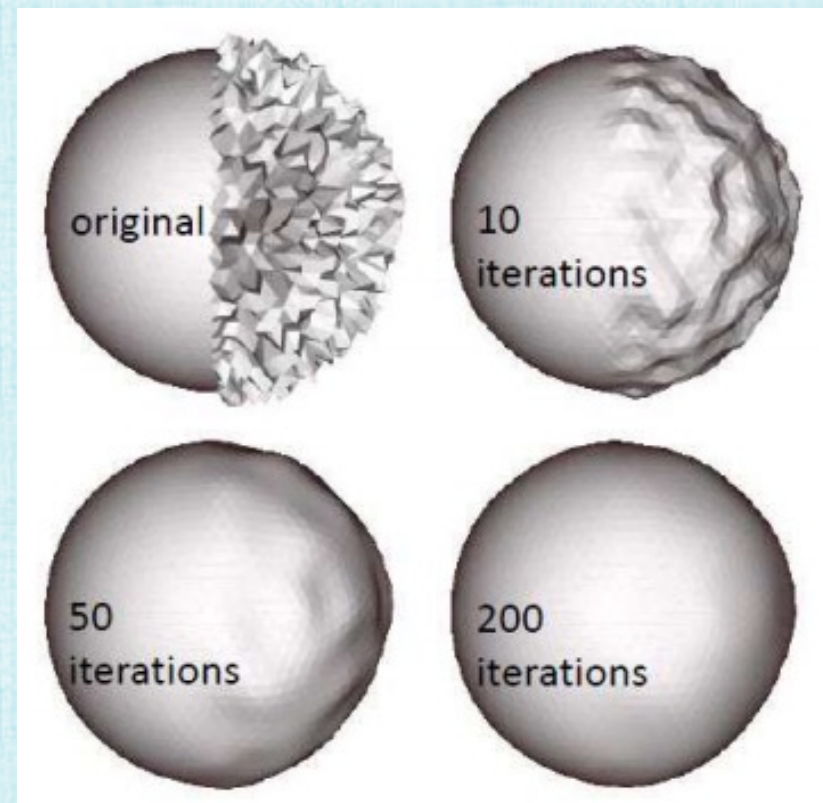
Taubin Smoothing

- Laplacian smoothing suffers from volume loss
- Taubin smoothing periodically performs an inflation step to add back volume:

$$p_i^{new} = p_i - \mu L(p_i) \quad \text{with } \mu > 0$$



Laplacian smoothing (only)



Taubin smoothing

Procedural Methods (for Geometry Construction)

- Generate geometry with an algorithm
 - Typically used for complex or tedious-to-create models
 - Perturb the algorithm to make variations of the geometry
- Start with a small set of data
- Use rules to describe high level properties of the desired geometry
- Add randomness, and use recursion

Neural Networks?
Generative AI?

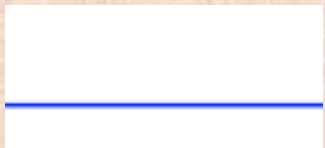


L-Systems

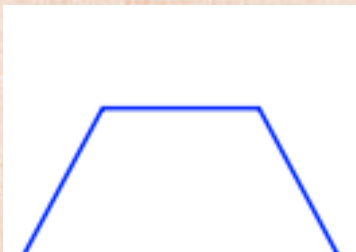
Natural Language
Processing (NLP)?

- Developed by a biologist (Lindenmayer) to study algae growth
- A recursive formal grammar:
 - An alphabet of symbols (terminal and non-terminal)
 - Production rules: non-terminal symbols recursively create new symbols (or sequences of symbols)
- Start with an initial string (axiom), and apply production rules
- A translator turns symbols into geometric structures

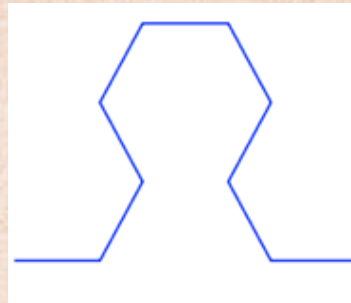
Nonterminals: A and B both mean "draw forward"
Terminals: +/- mean "turn" right/left (respectively) by 60 degrees
Production Rules: $A \rightarrow B + A + B$ and $B \rightarrow A - B - A$



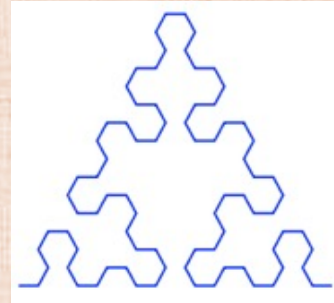
Initial Axiom: A



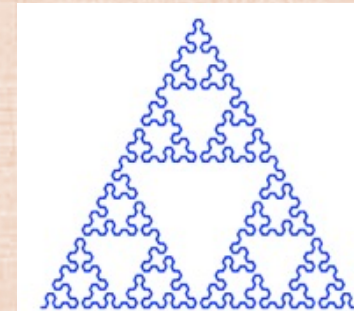
B+A+B



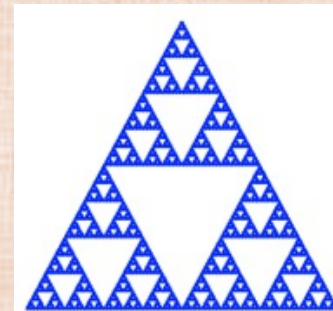
A-B-A + B+A+B + A-B-A



B+A+B-(A-B-A)-(B+A+B)
+A-B-A +B+A+B +A-B-A
+B+A+B-(A-B-A)-(B+A+B)



Etc.



Sierpinski
Triangle

L-System + Stack = Branches

- Nonterminals: X is no action, F is draw forward
- Terminals: +/- means turn right/left by 25 degrees
 - **[** means to store current state on the stack
 - **]** means to load the state from the stack
- Initial Axiom: X
- Production Rules: $X \rightarrow F - [[X]+X] + F [+FX] - X$, $F \rightarrow FF$



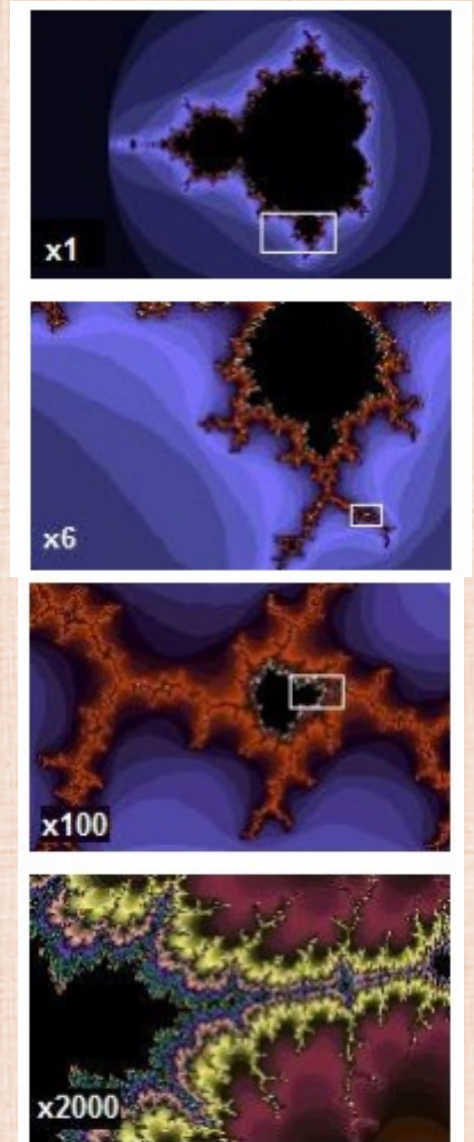
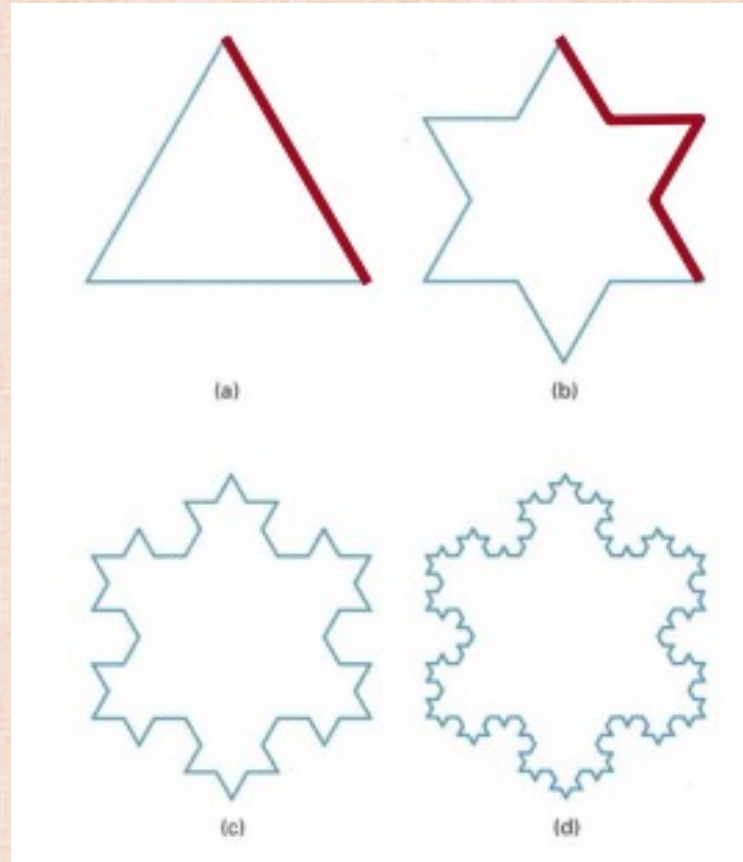
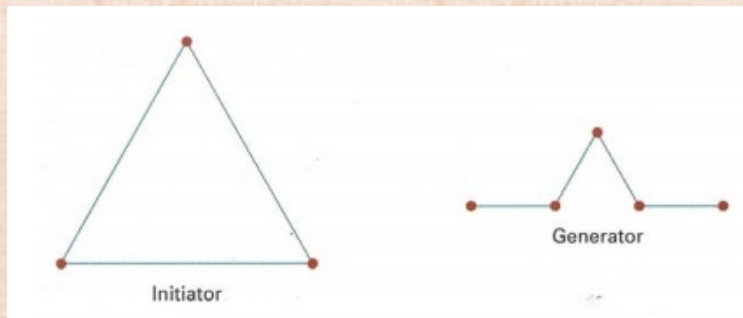
L-Systems

- Easily extended to 3D
- Model trunk/branches as cylinders
- As recursion proceeds:
 - shrink cylinder size
 - vary color (from brown to green)
- Add more variety with a stochastic L-system
 - Multiple (randomly-chosen) rules for each symbol
- Practice and experimentation is required in order to obtain good results
 - (just like for ML)



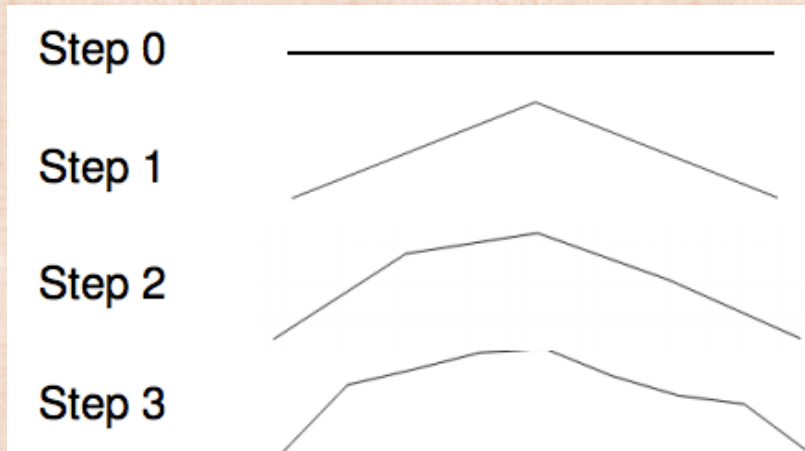
Fractals

- Initiator: start with a shape
- Generator: replace subparts with a (scaled) generator
- Repeat



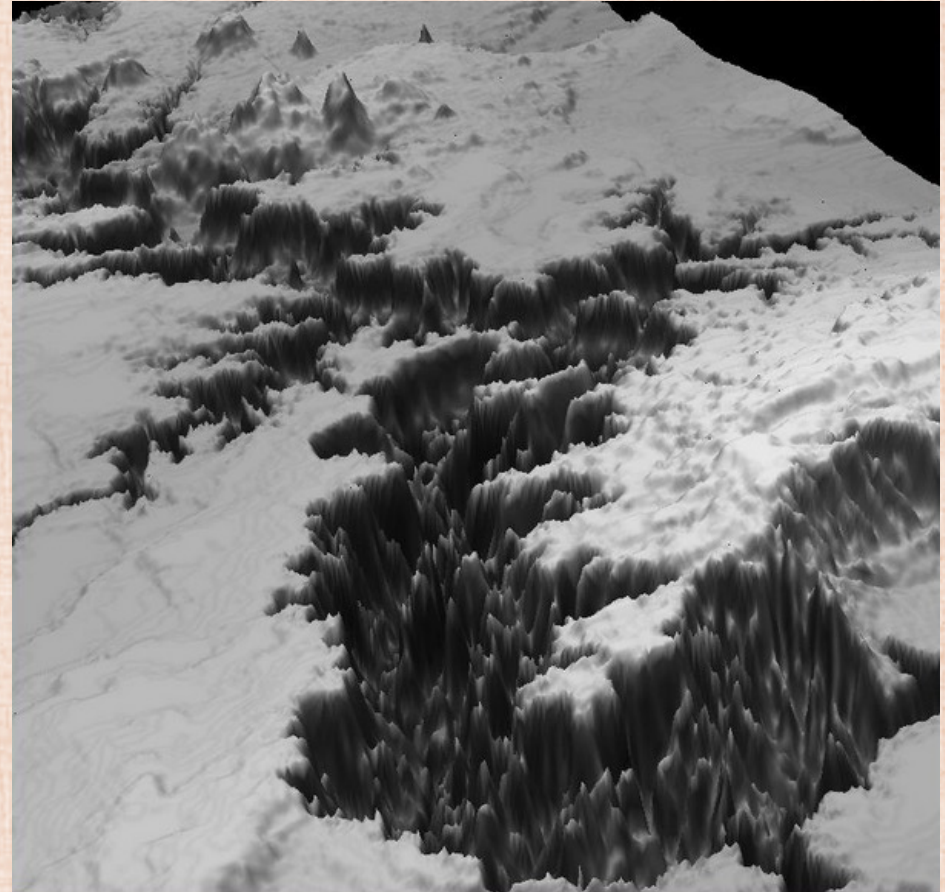
Fractals

- Add randomness to the new vertex locations
- E.g., can create an irregular 2D silhouette (for far away mountains)



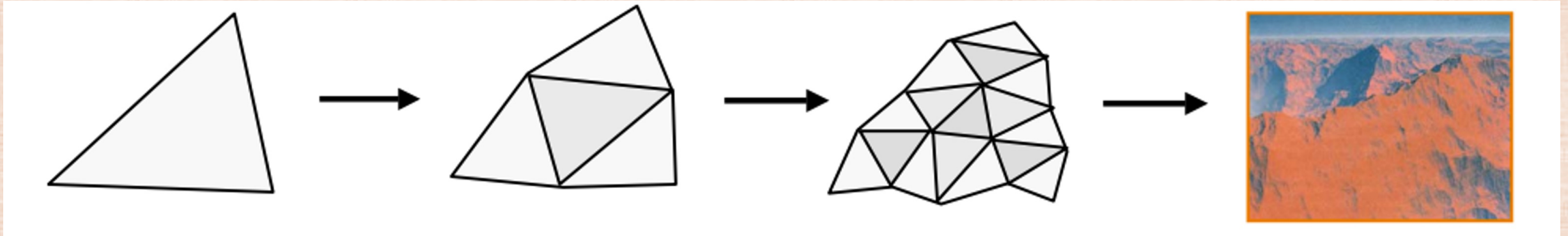
Height Fields

- Start with a 2D fractal (or any 2D grey-scale image)
- Place the image on top of a ground plane (subdivided into triangles)
- For each triangle vertex, displace its height based on the local pixel intensity



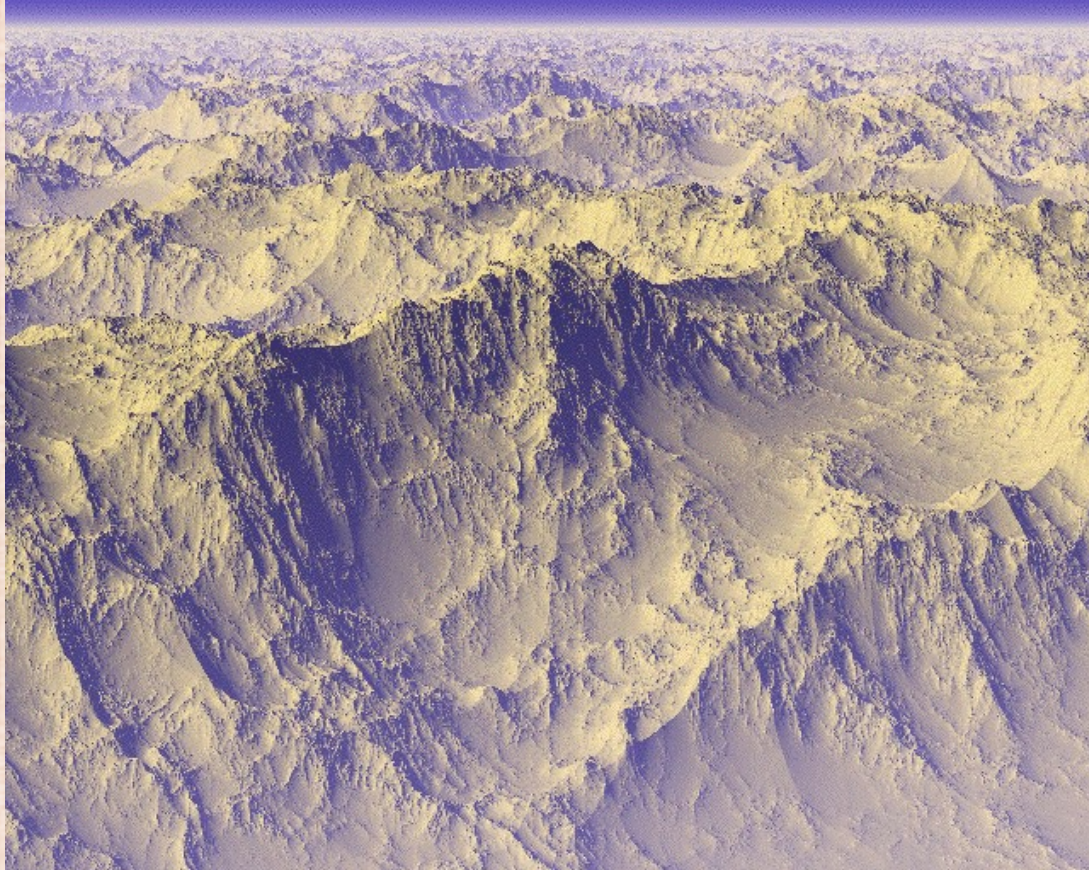
3D Landscapes

- Initiator: start with a shape
- Generator: replace random subparts with a self-similar (somewhat random) pattern

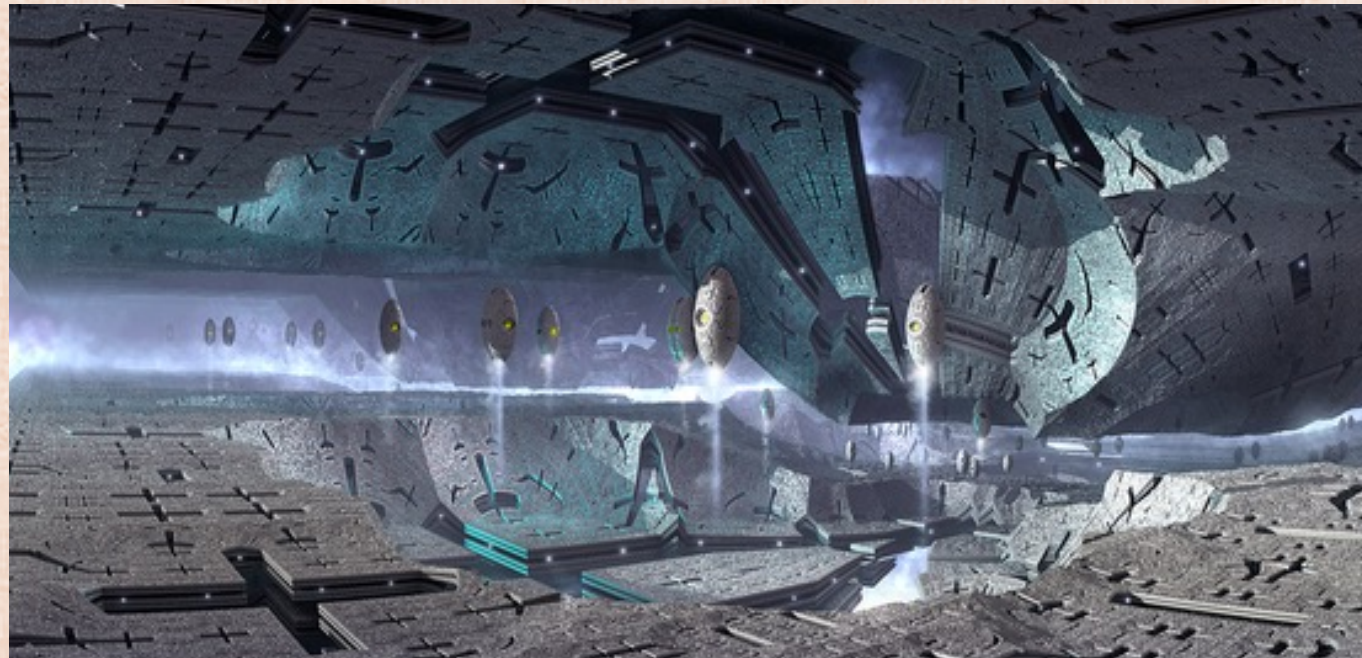
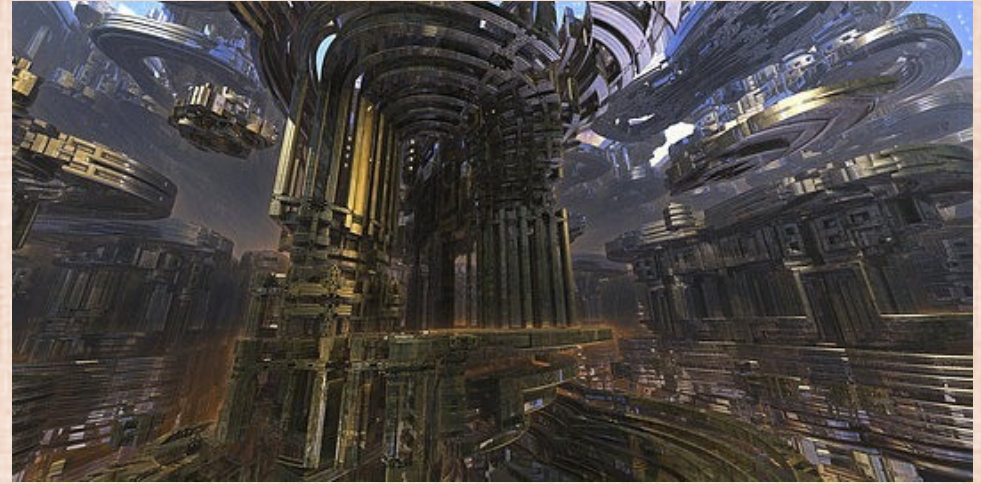


- Similar to subdivision, but with much more interesting rules for setting vertex positions

3D Landscapes



Fractal Worlds



Machine Learning

- Interactive Example-Based Terrain Authoring with Generative Adversarial Networks
 - Siggraph 2017



Generative AI

Rapidly Generate 3D Assets for Virtual Worlds with Generative AI

Jan 03, 2023

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Nvidia