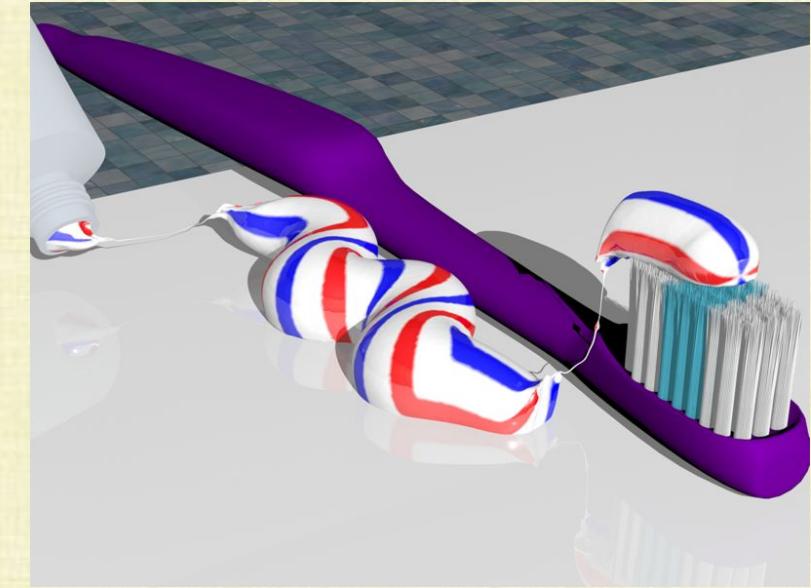


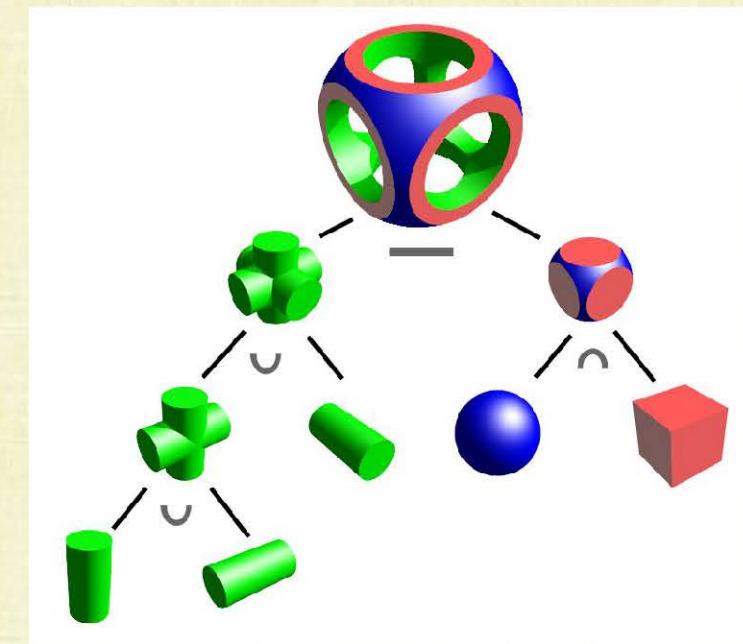
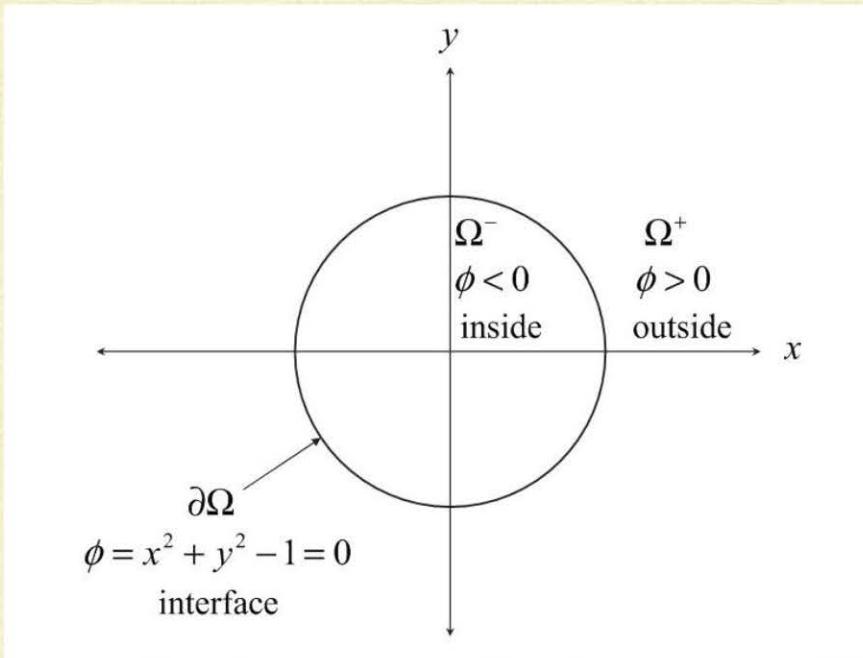
# More Geometric Modeling



# Implicit Surfaces

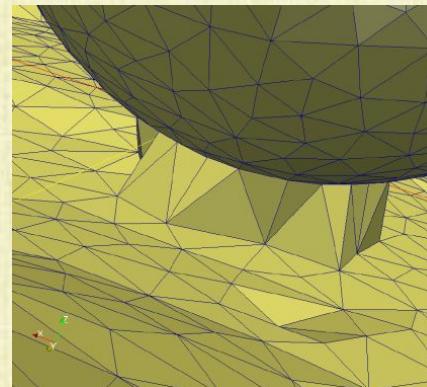
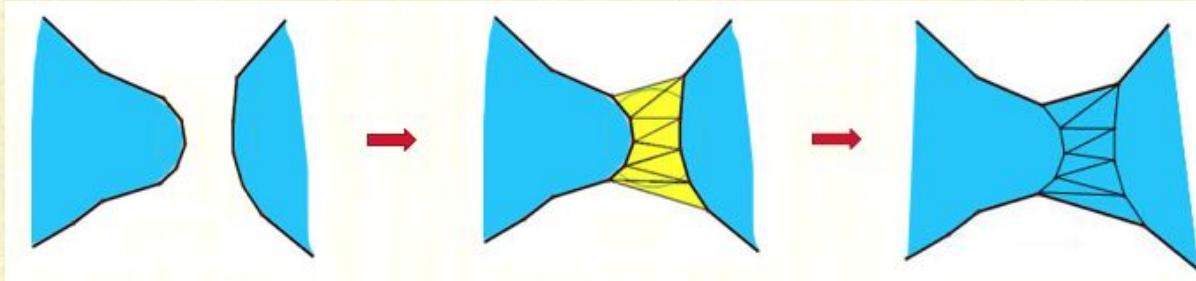
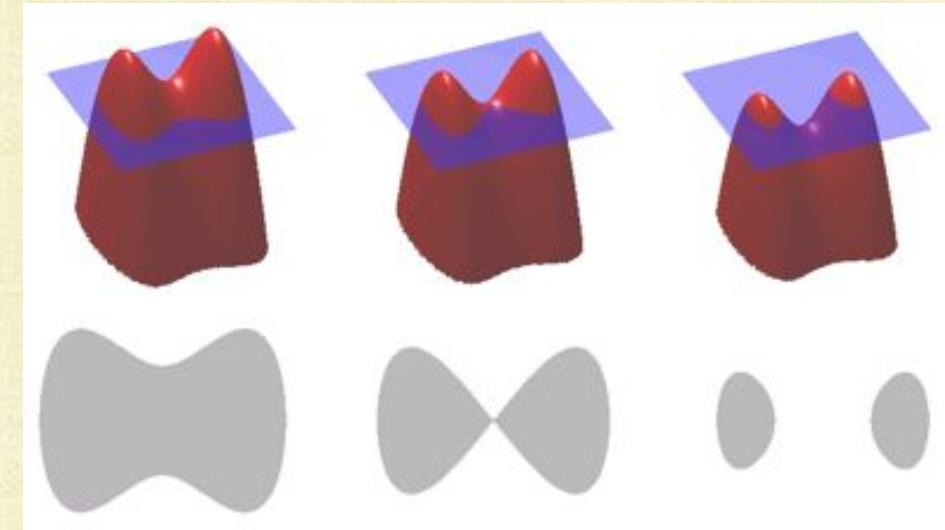
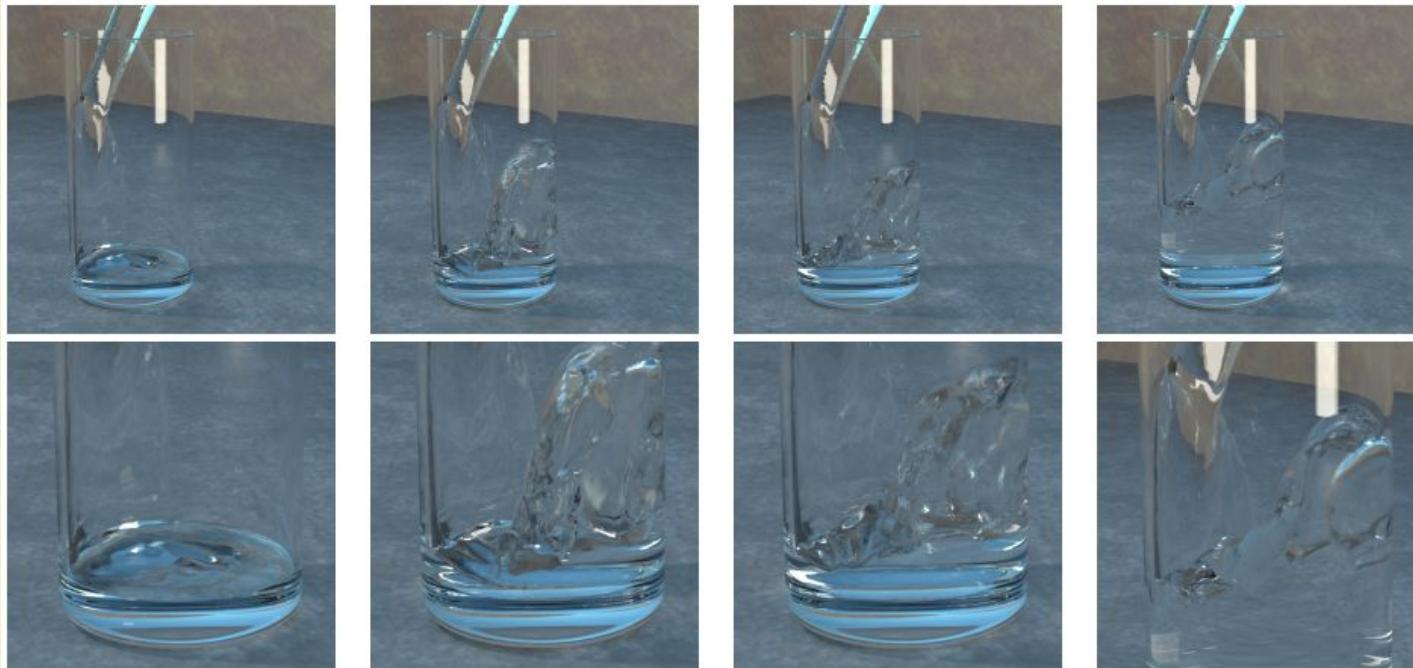
# Implicit Surfaces

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



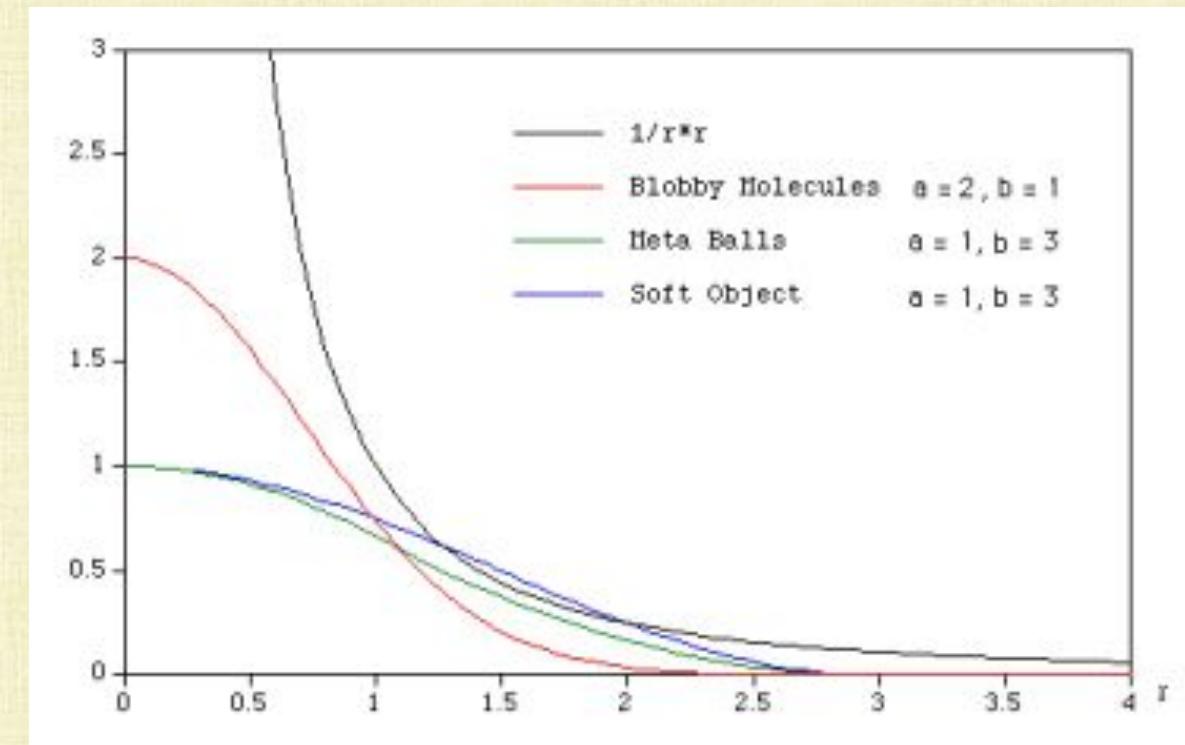
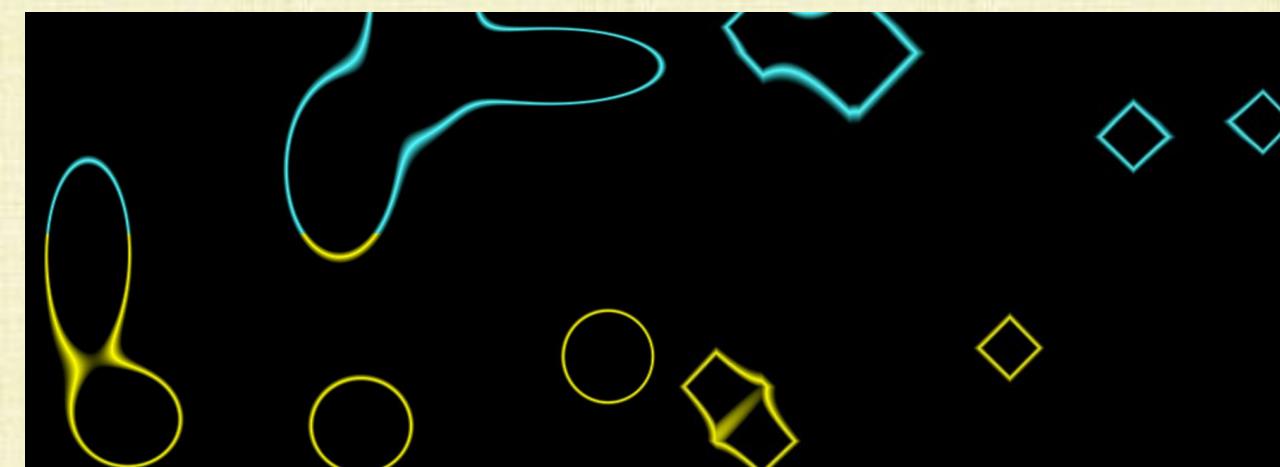
# Topological Change

- Greatly superior to triangle meshes for topological change!

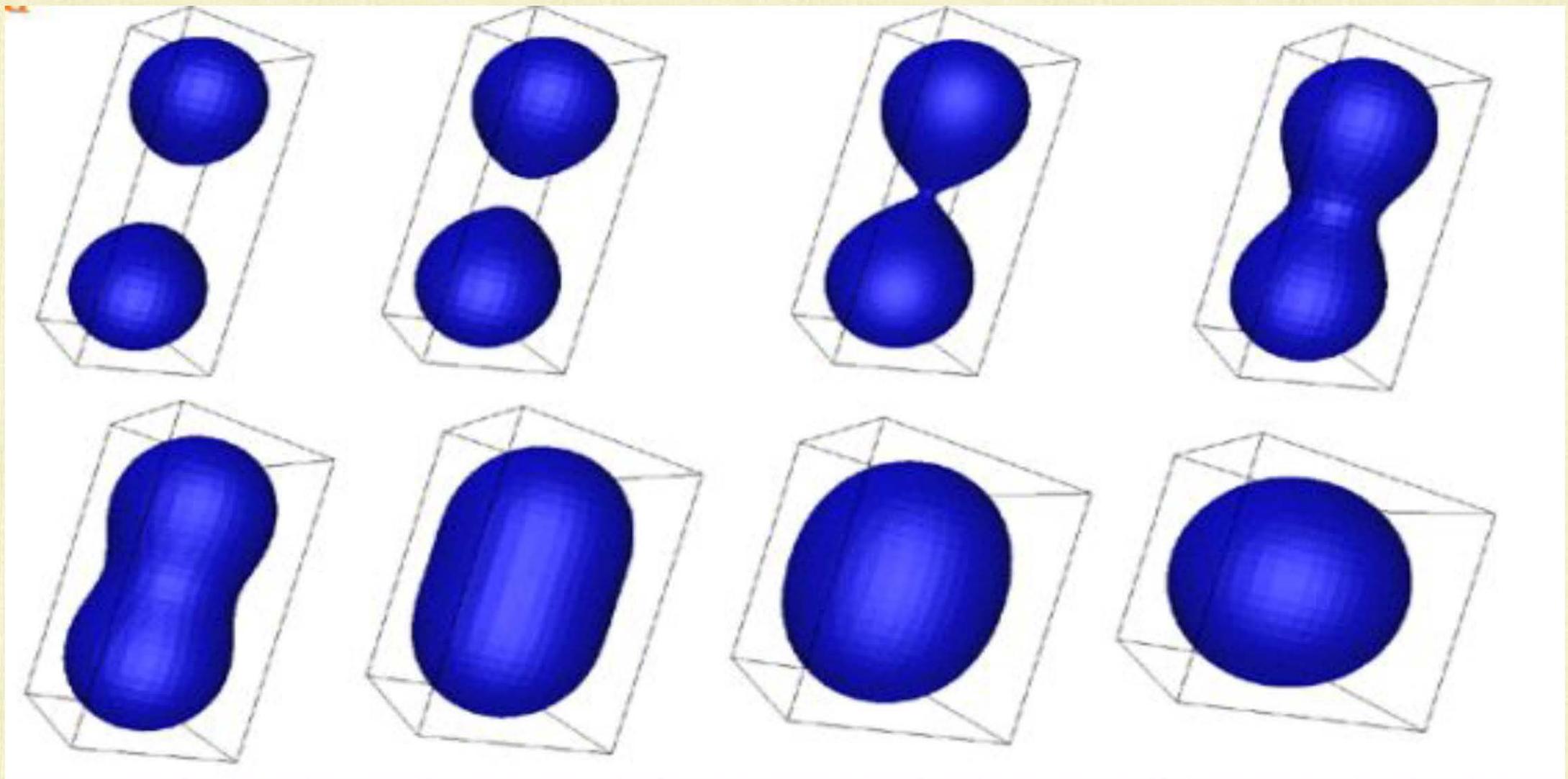


# Blobbies

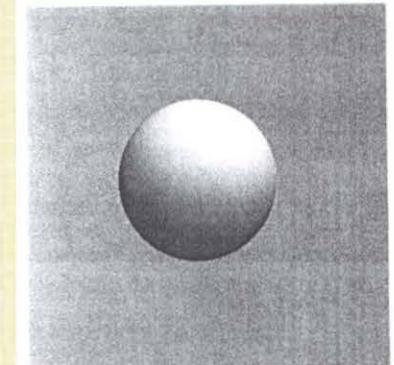
- Each blob is defined as a density function around a particle
- Blob kernels can be: 2D ellipses, 2D diamonds, 3D spheres, etc.
- For each pixel, the aggregate density is summed from all overlapping blobs
- Also known as:
  - Metaballs (in Japan), Soft objects (in Canada and New Zealand)
  - Slightly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



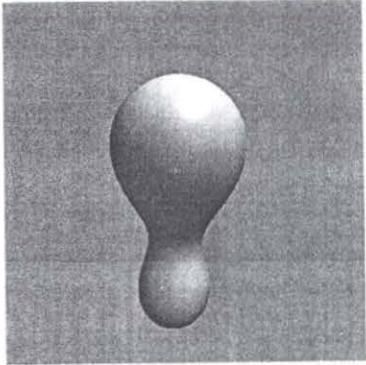
# Topological Change



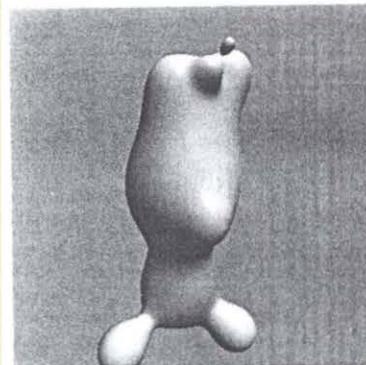
# Blobby Modeling



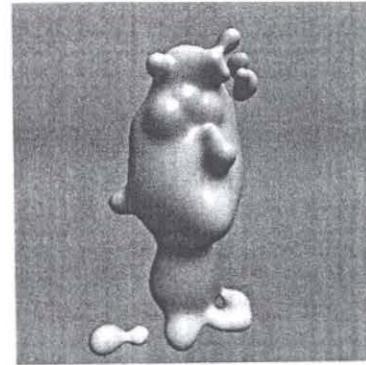
(a)  $N = 1$



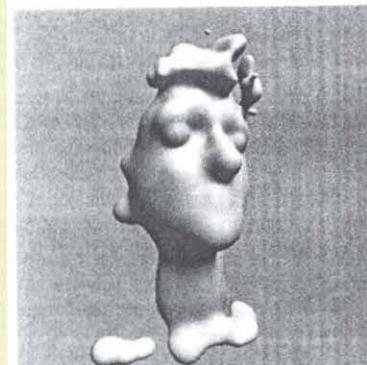
(b)  $N = 2$



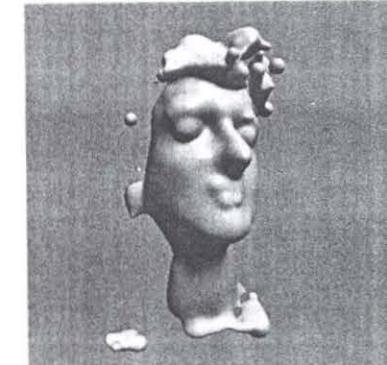
(c)  $N = 10$



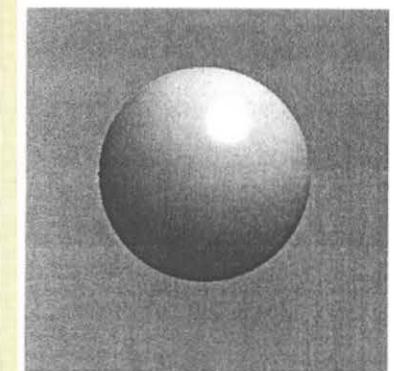
(d)  $N = 35$



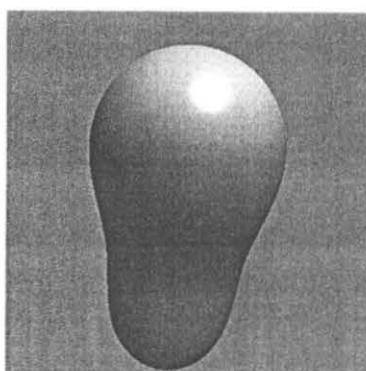
(e)  $N = 70$



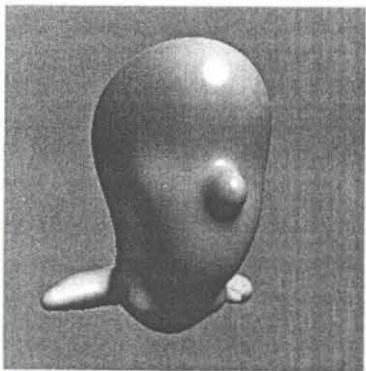
(f)  $N = 243$



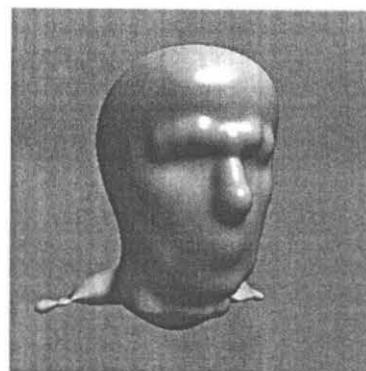
(a)  $N = 1$



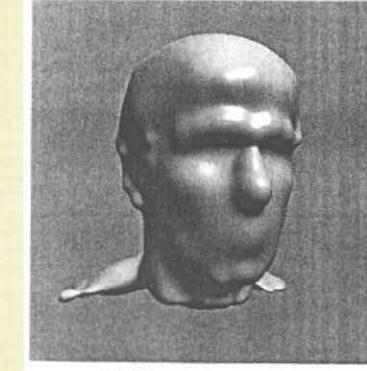
(b)  $N = 2$



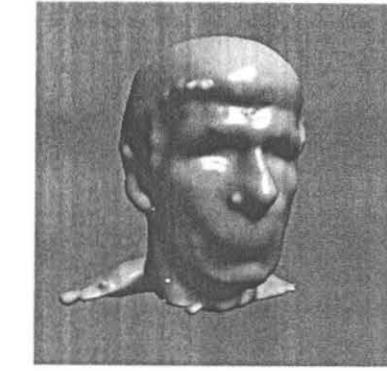
(c)  $N = 20$



(d)  $N = 60$



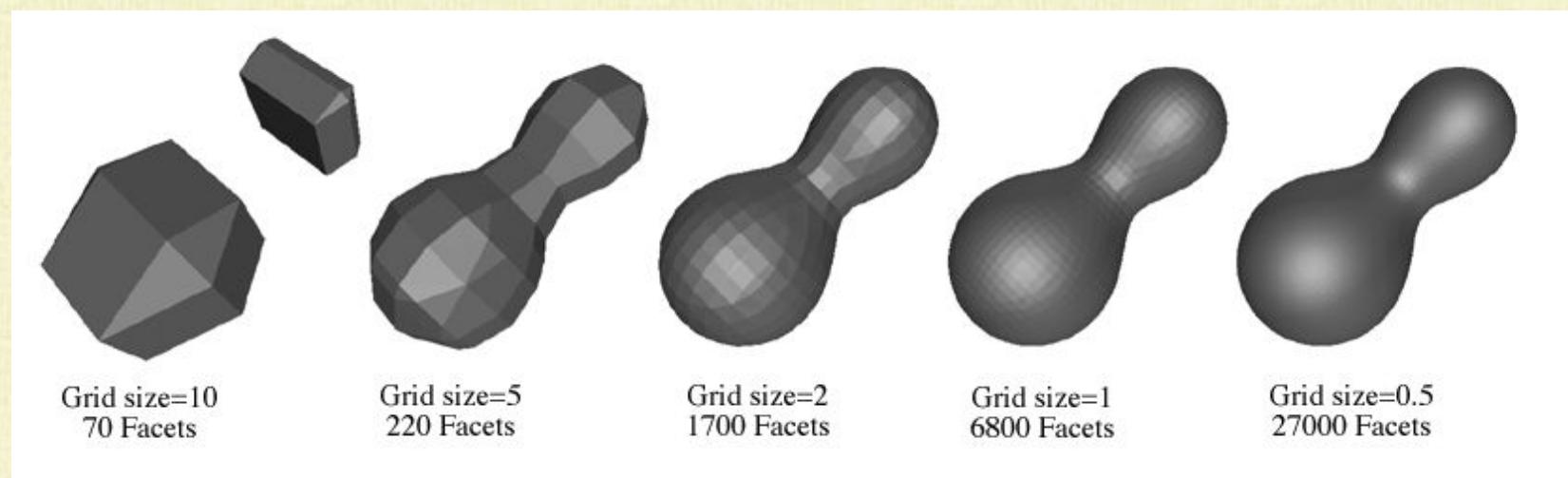
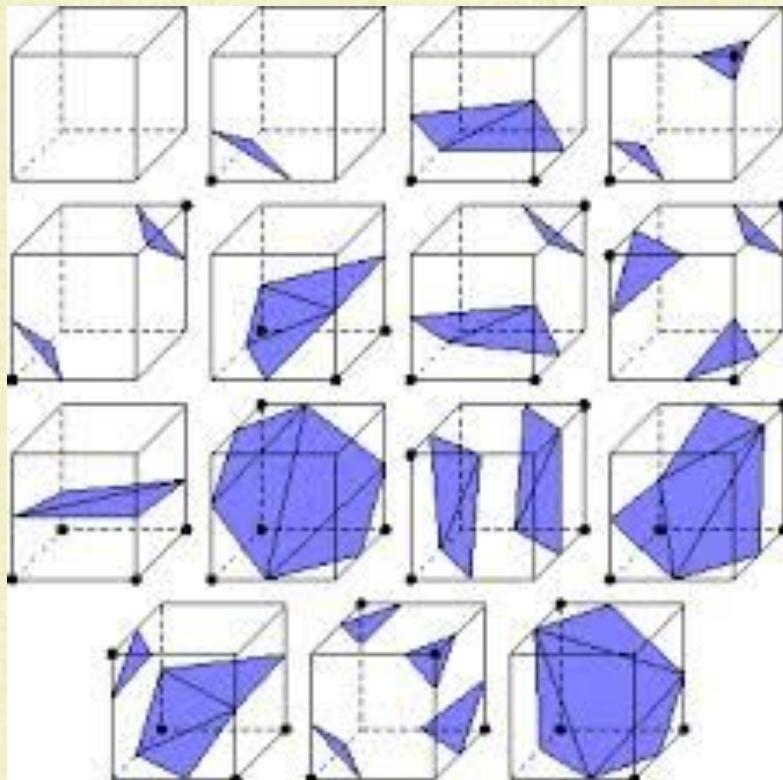
(e)  $N = 120$



(f)  $N = 451$

# Marching Cubes (or Tetrahedra)

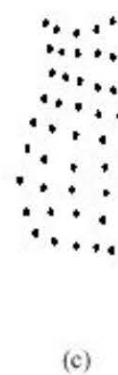
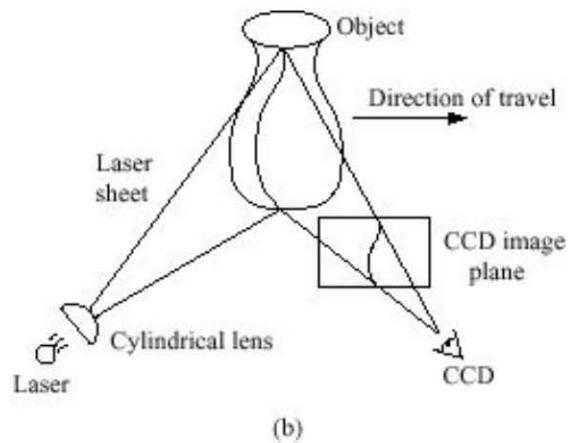
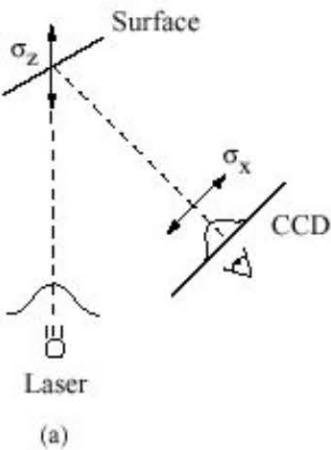
- Turns an implicit surface into triangles
  - Define the implicit surface on a 3D grid
  - For each grid cell, use the topology of the volume to construct surface triangles



# Computer Vision

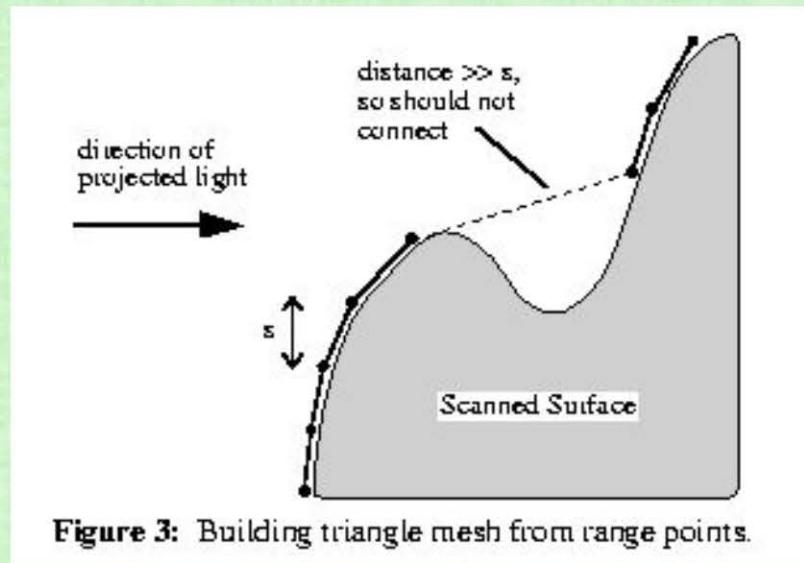
# Range Scanners

- 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

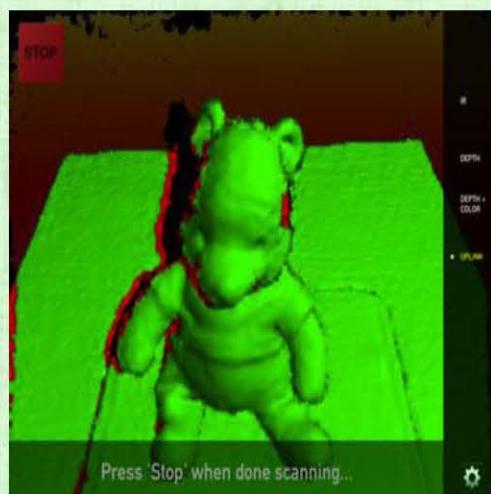
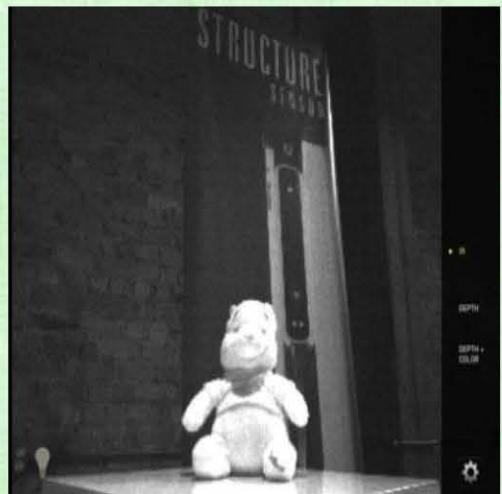


# Range Scanners

- 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



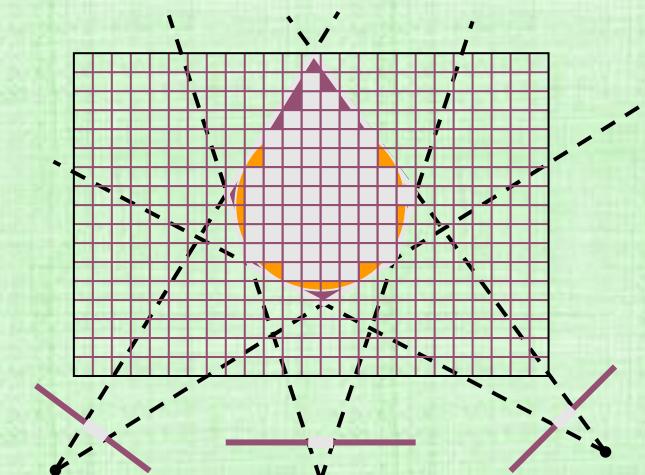
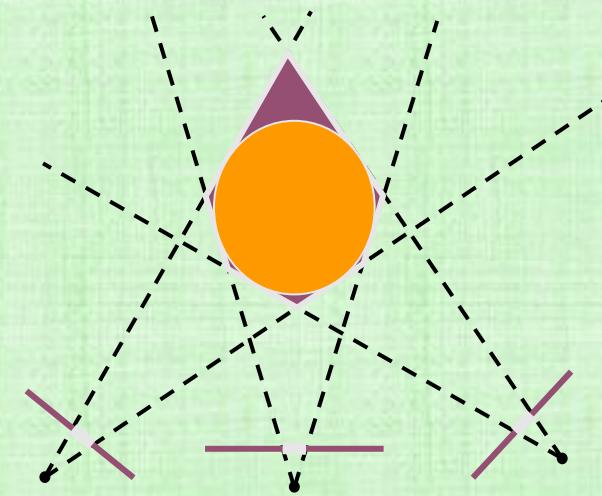
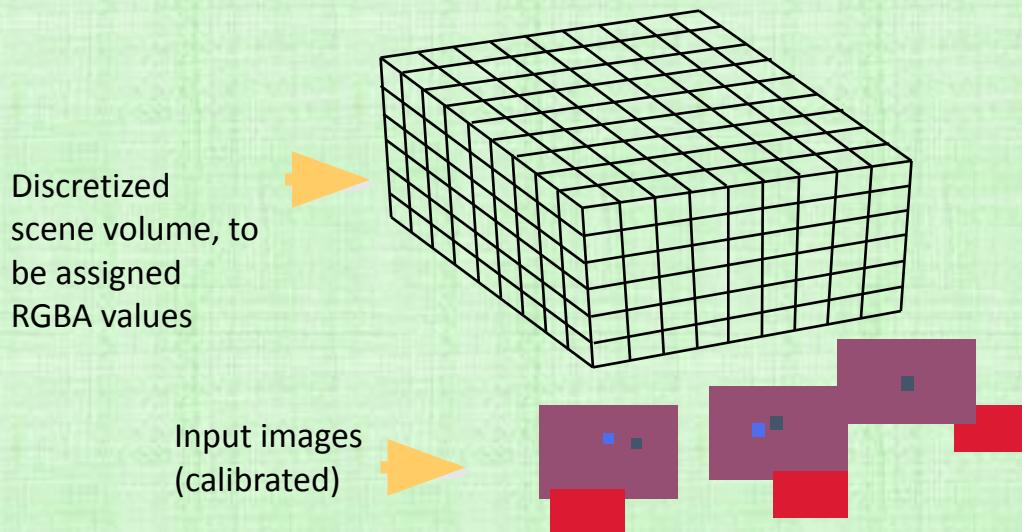
 AUTODESK® 123D®

Structure Sensor for iPad

Autodesk 123D Catch

# Voxel Carving

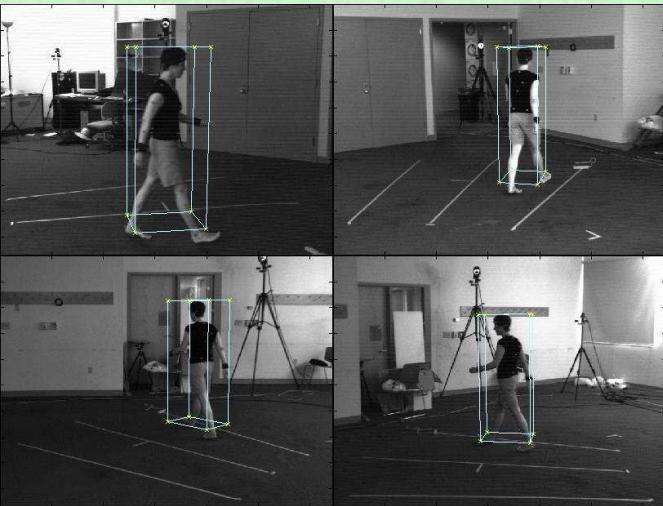
- Requires multiple images (from calibrated cameras) taken from different directions
- Construct a voxelized 3D model:
  - A silhouette is computed for each image
  - The silhouette (for each image) is back projected onto the grid
  - The voxels outside the back-projection of an image are carved away
- Colors are also back projected



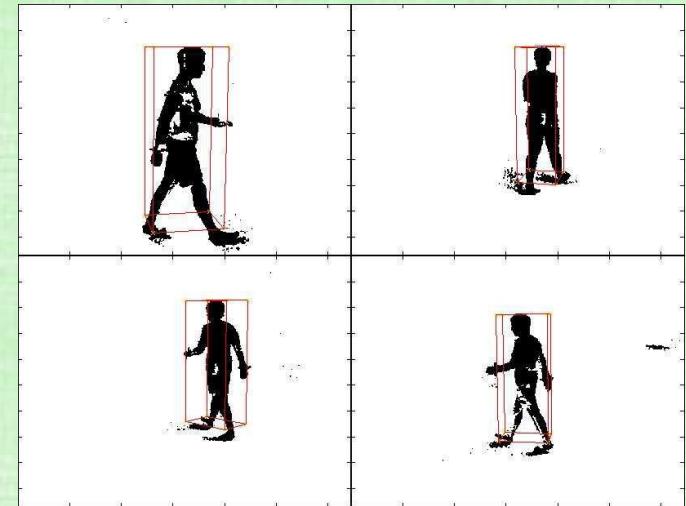
Color the voxel gray if silhouette is in every image

# Voxel Carving

Original image



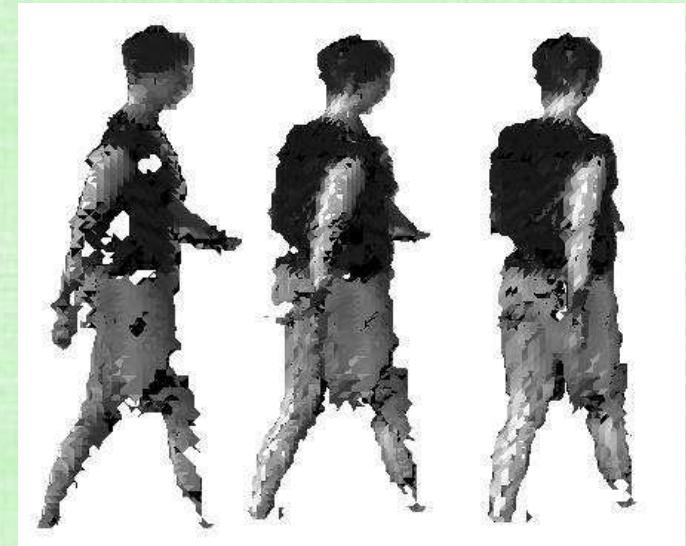
Extracted silhouettes



Carved out voxels

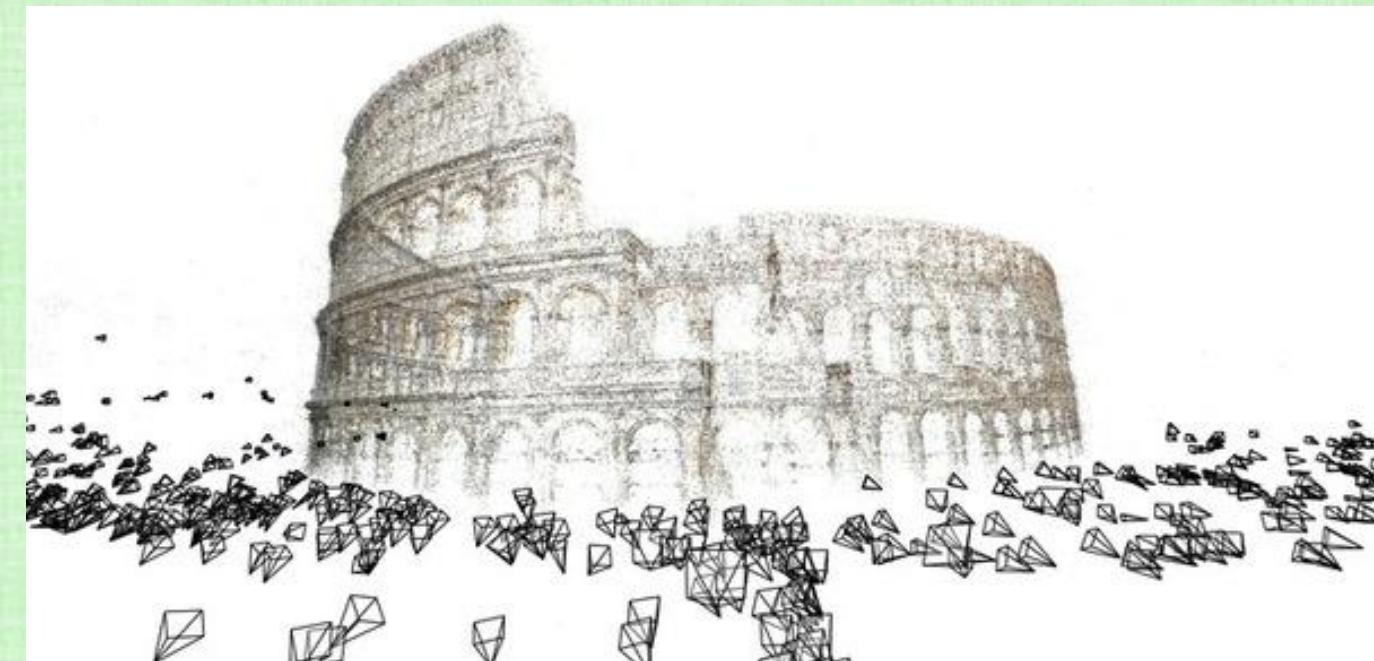


Back-projecting colors



# Reconstruction from Large Photo Collections

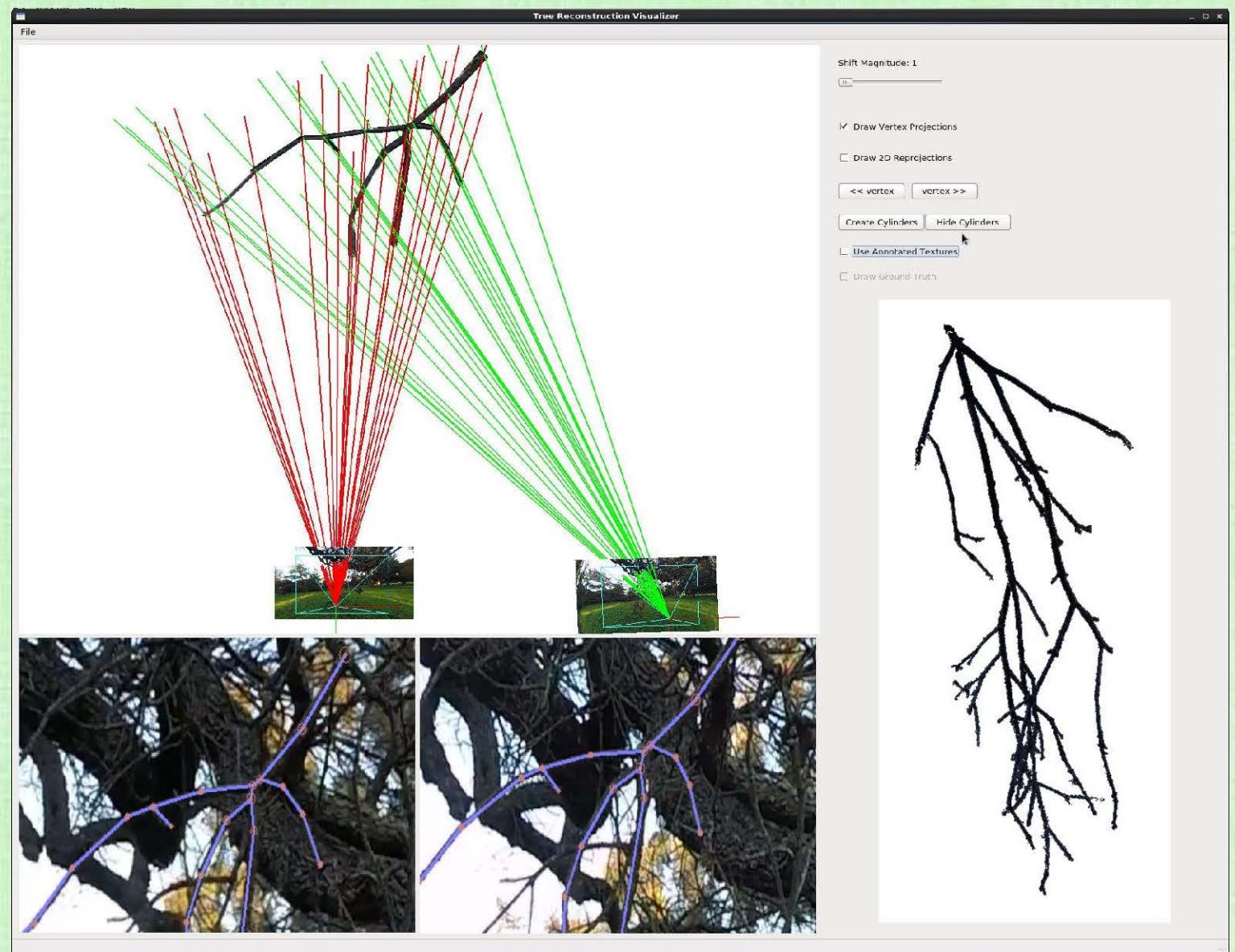
- Collect a large number of photos (e.g. from google images)
- Predict relative camera position/orientation for each image
- The position of a point that is visible in multiple images can be triangulated
- Obtain a sparse point cloud representation of the object
- Dense reconstruction algorithms can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



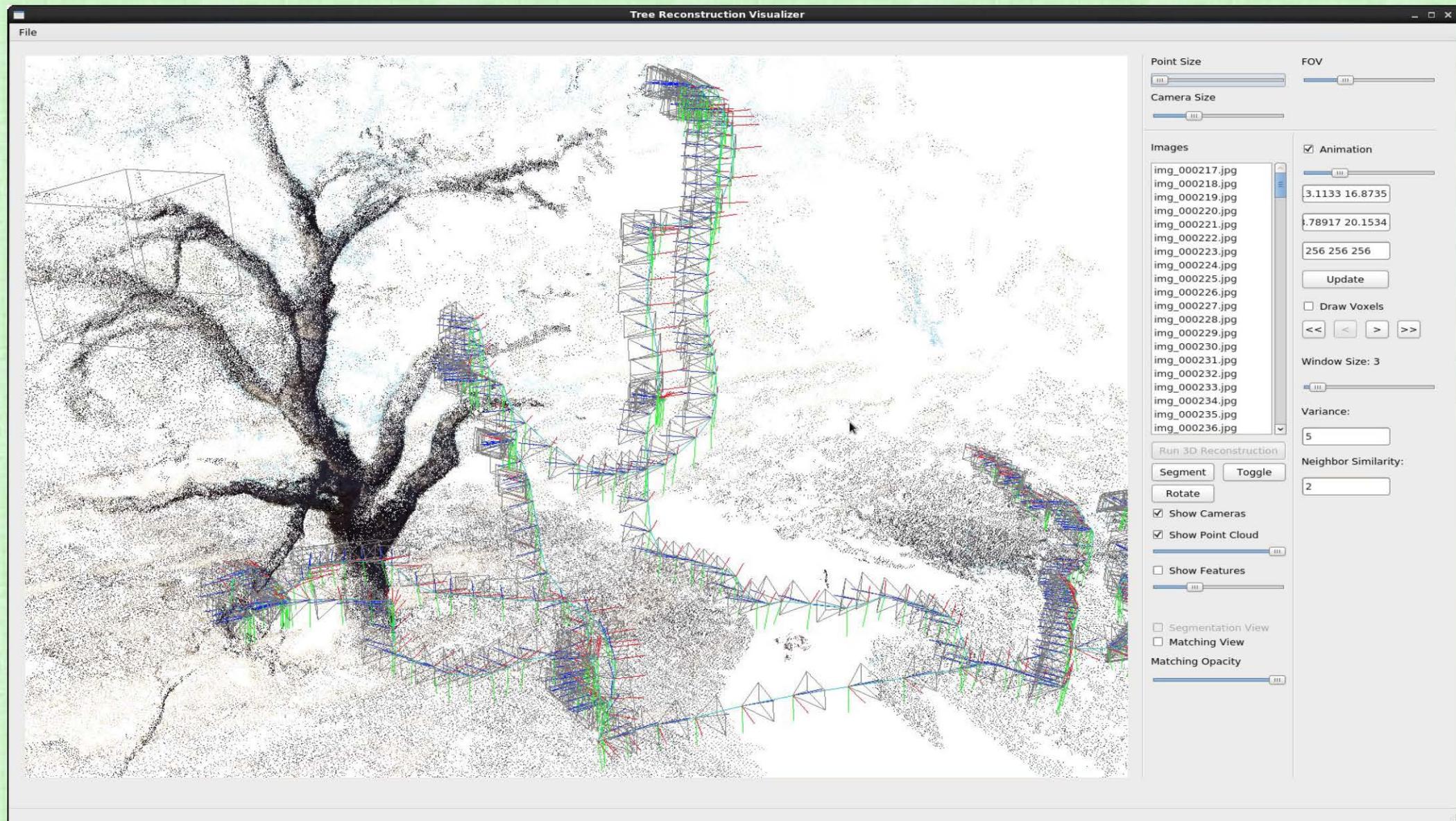
2D photos

3D geometry

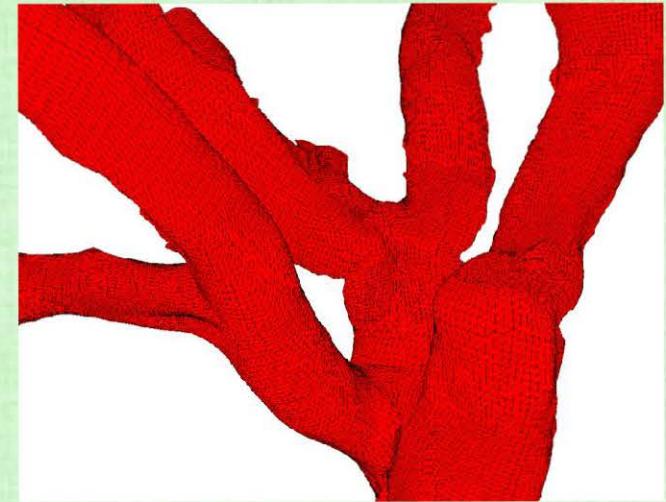
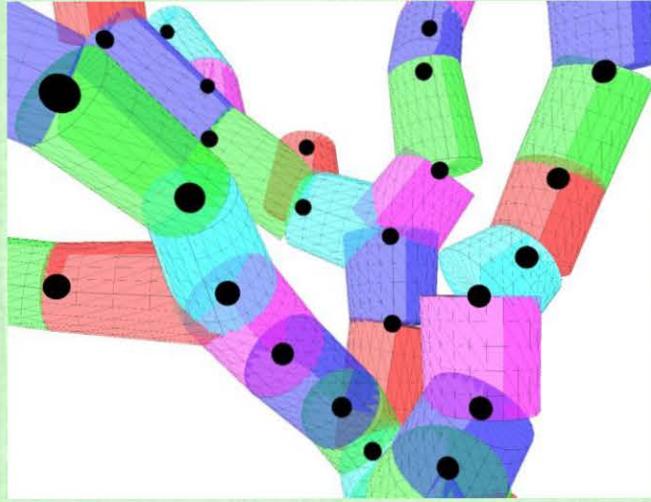
# Drones & Trees



# Drones & Trees



# Drones & Trees

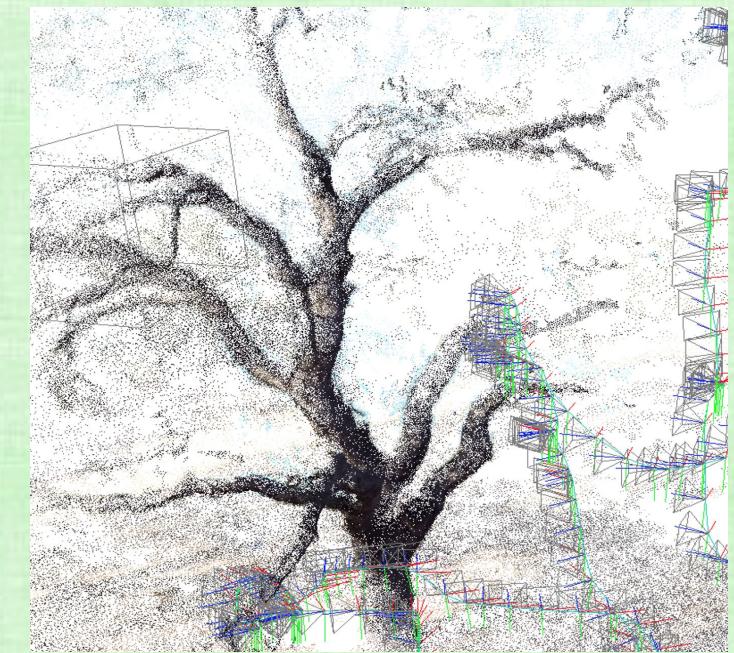
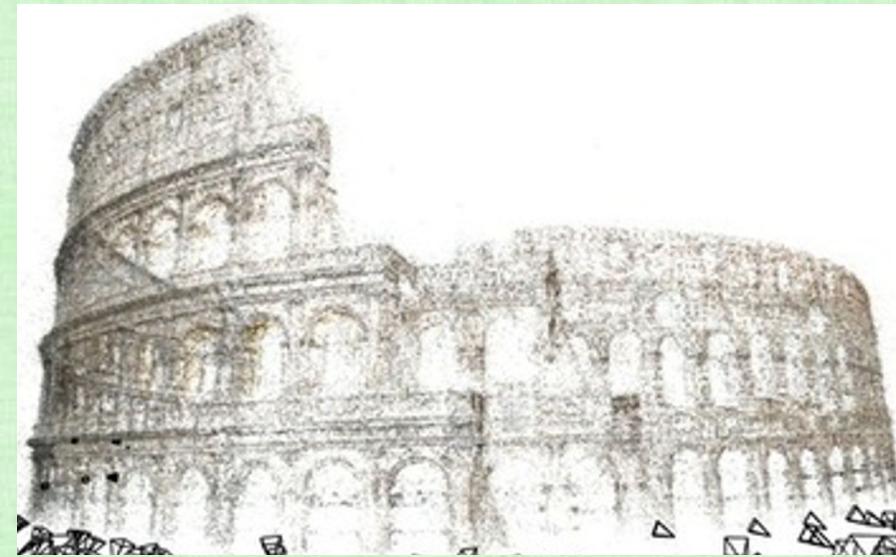
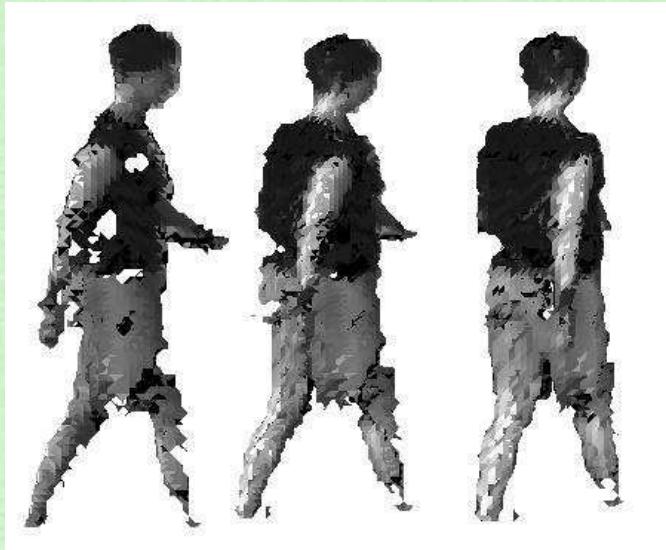


# Drones & Trees



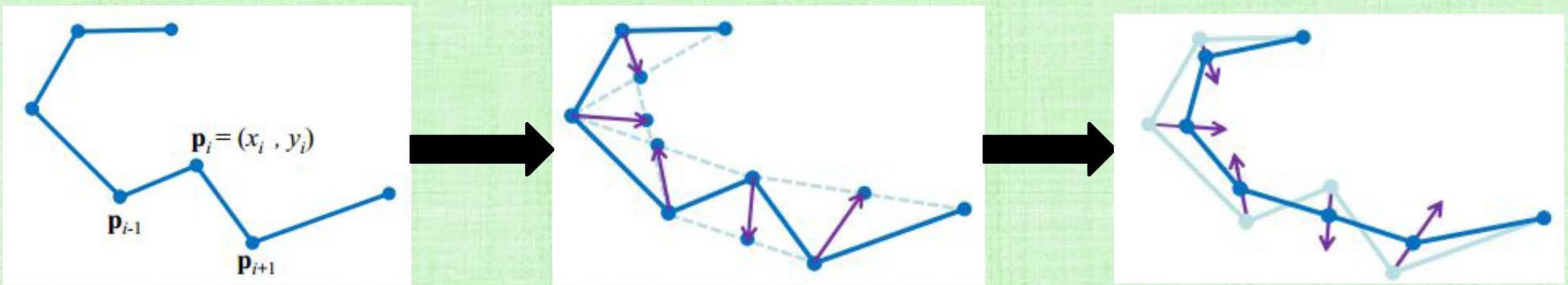
# Noise & De-Noising

- Computer Vision algorithms use real-world sensors/data
- This results in noise, which is the biggest drawback to such methods
- Denoising/smoothing algorithms become very important



# Laplacian Smoothing

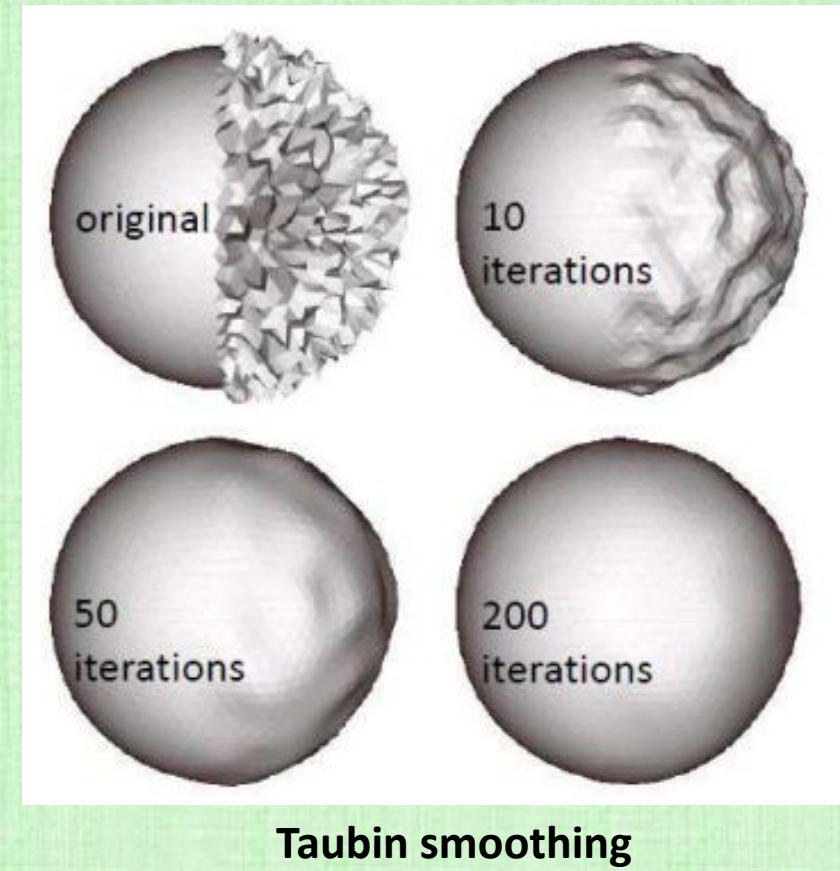
- Compute a Laplacian estimate using the one ring of vertices about a point
  - Similar to differential coordinates
- E.g., on a curve:  $L(p_i) = \frac{1}{2}((p_{i+1} - p_i) + (p_{i-1} - p_i))$
- Then, update  $p_i^{new} = p_i + \lambda L(p_i)$  where  $\lambda \in (0,1)$
- Repeat several iterations



# Taubin Smoothing

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

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



# **Physics Simulation**

# Newton's Second Law

- 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  $\frac{d^2X(t)}{dt^2} = X''(t) = \frac{F(t)}{M}$

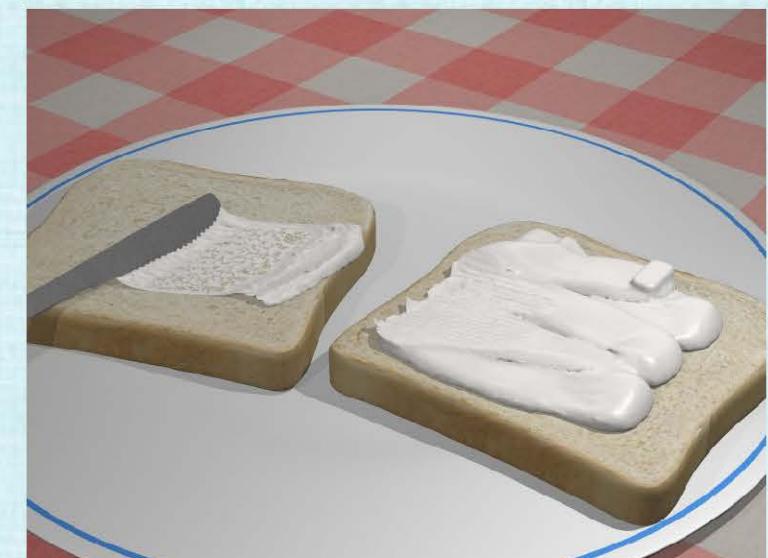
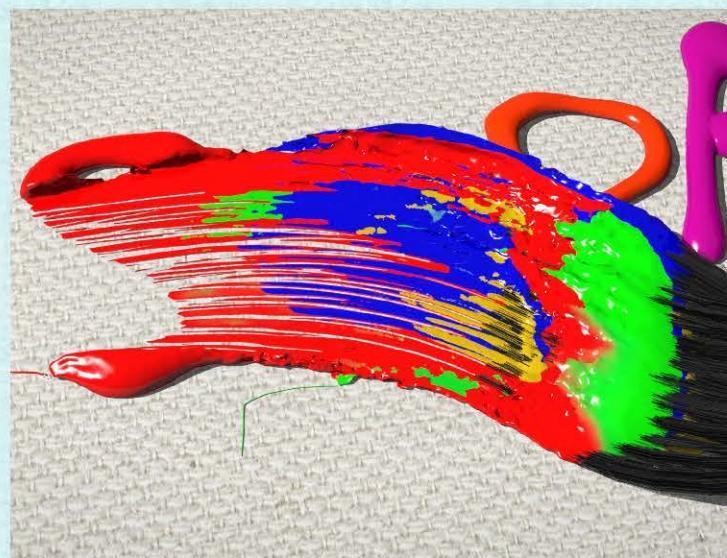
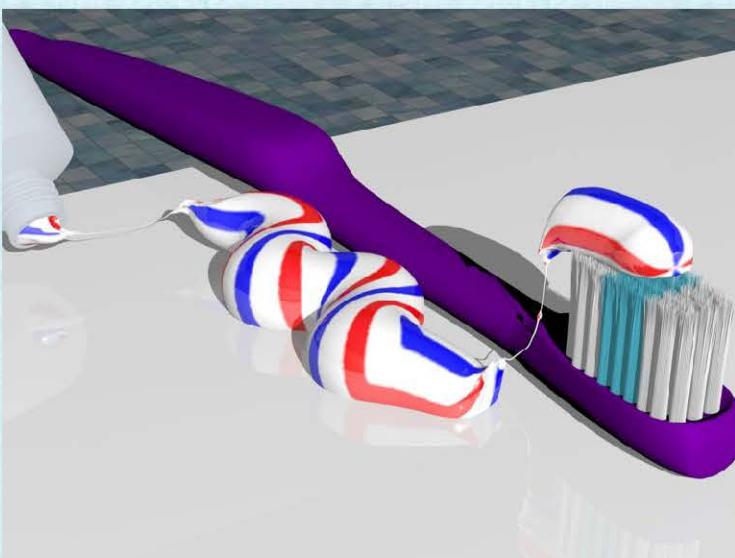
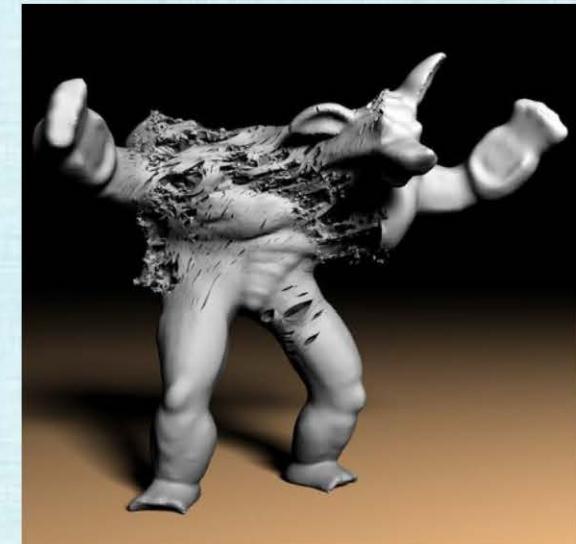
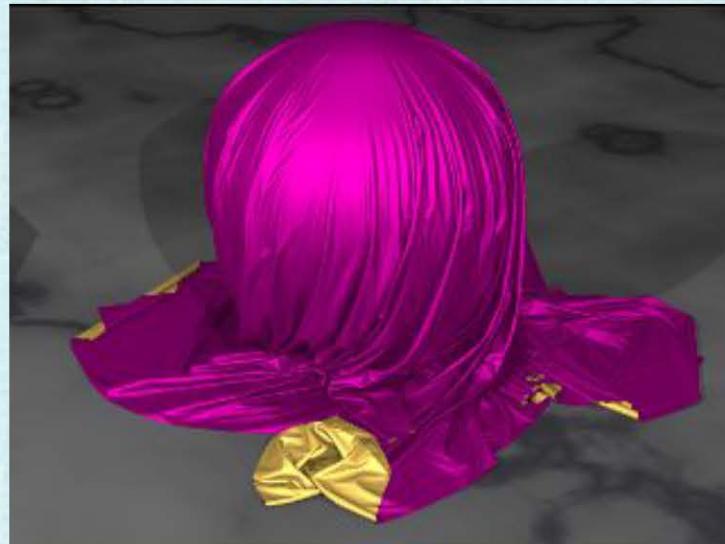
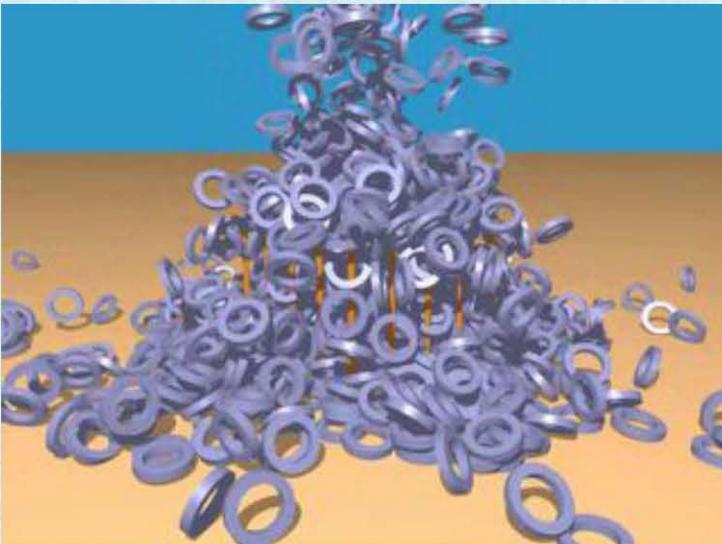
- Combining kinematics and dynamics gives: 
$$\begin{pmatrix} X'(t) \\ V'(t) \end{pmatrix} = \begin{pmatrix} V(t) \\ \frac{F(t, X(t), V(t))}{M} \end{pmatrix}$$

- Note: forces often depend on position/velocity

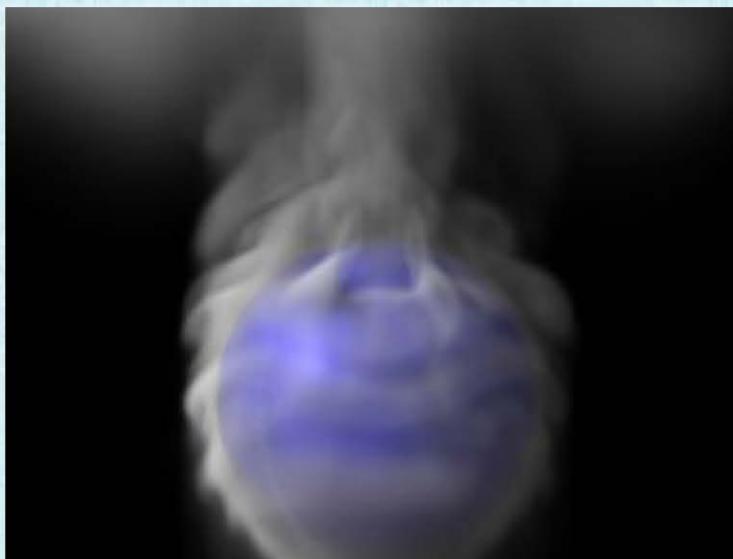
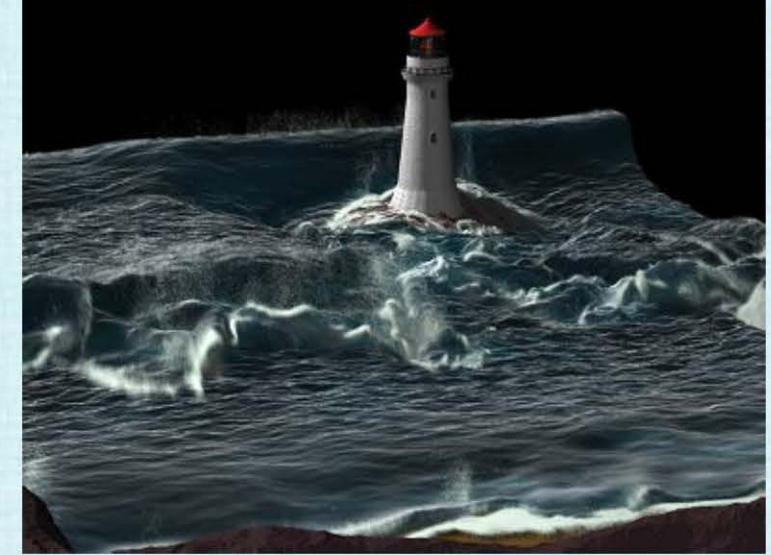
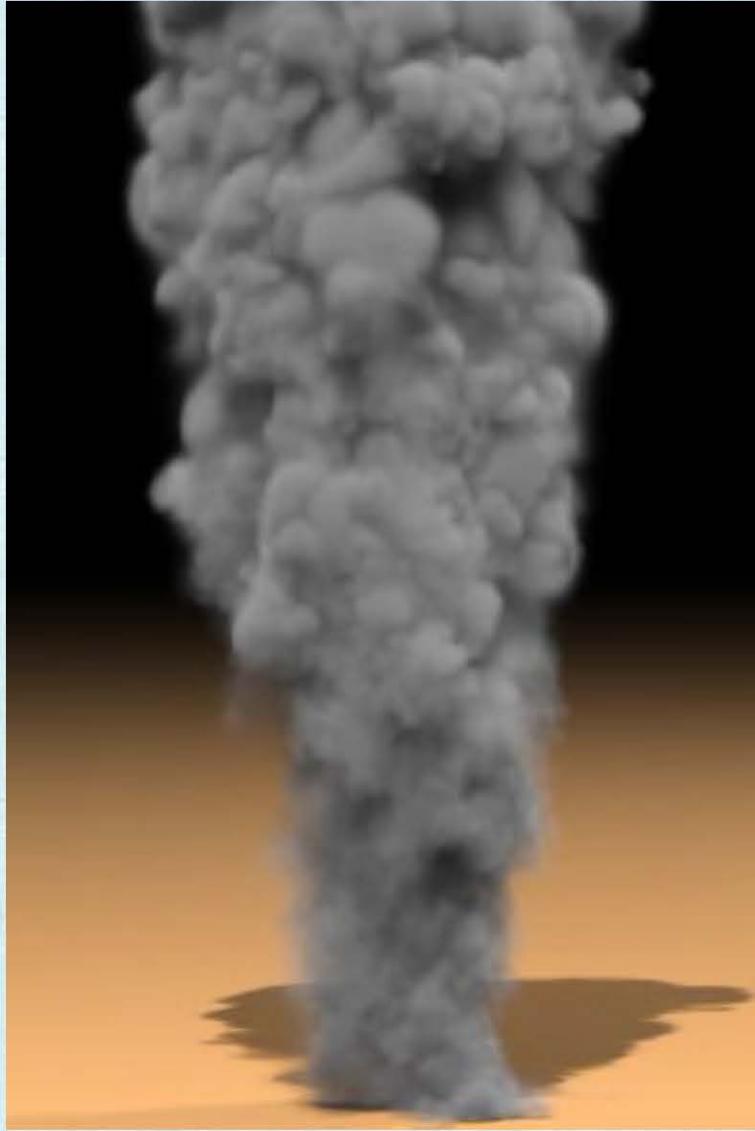
- Much of the physical world can be simulated with computational mechanics (FEM) and computational fluid dynamics (CFD), using Newton's second law

- Create degrees of freedom, specify forces, and solve the resulting ordinary differential equations (ODEs)
- When the forces have spatial interdependence, one obtains partial differential equations (PDEs)

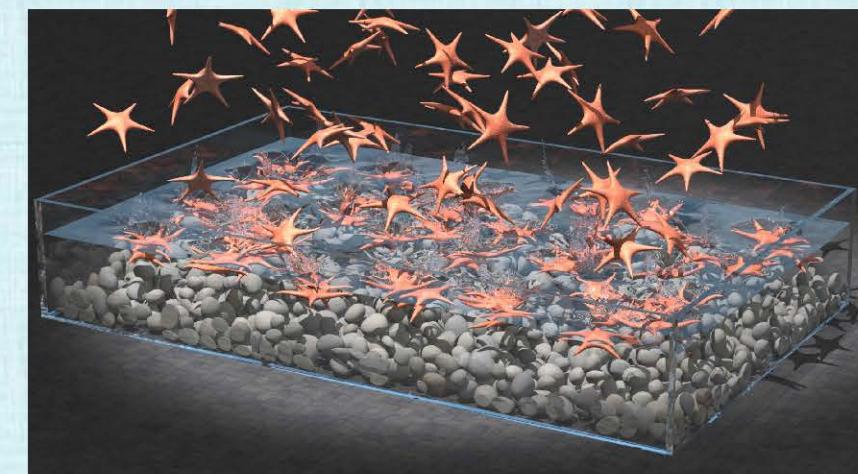
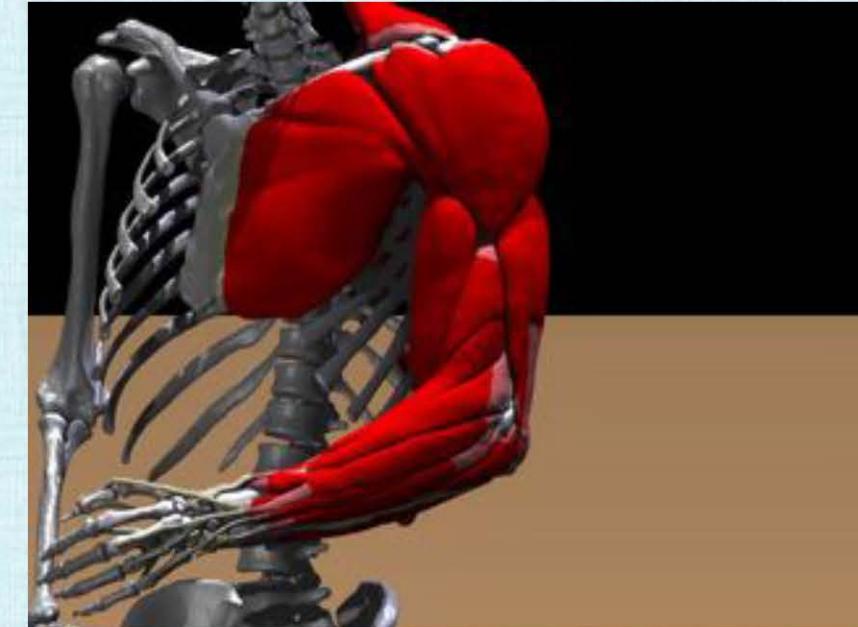
# Computational Mechanics (FEM)



# Computational Fluid Dynamics (CFD)



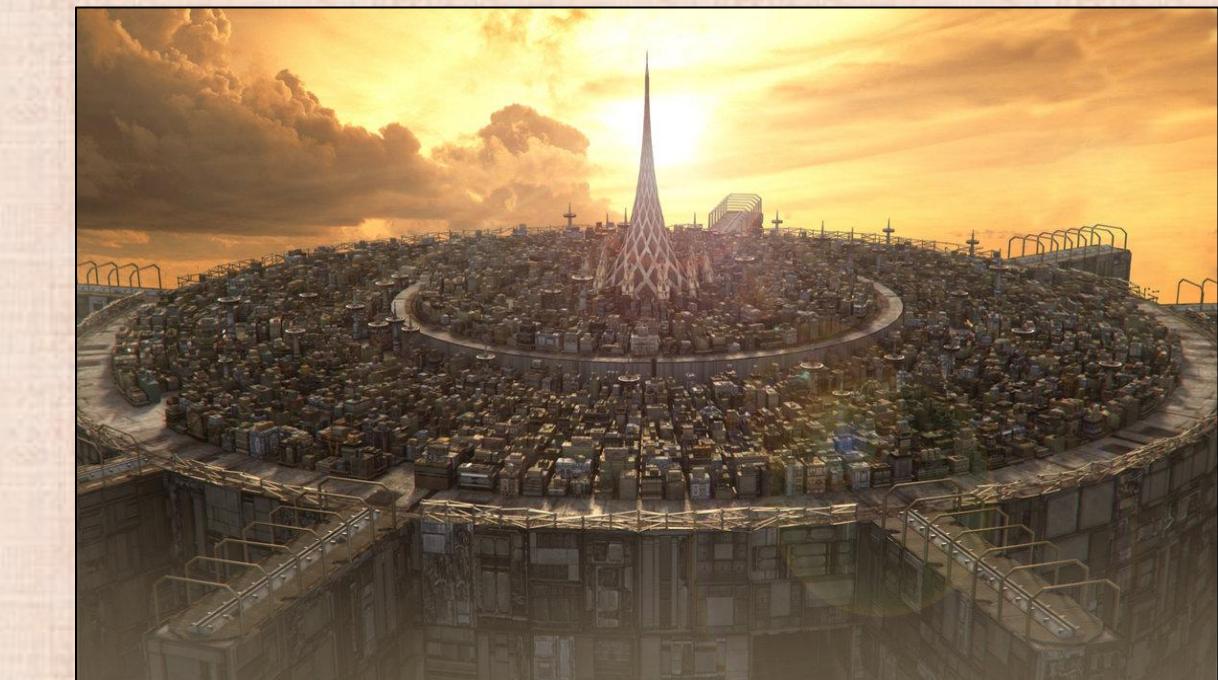
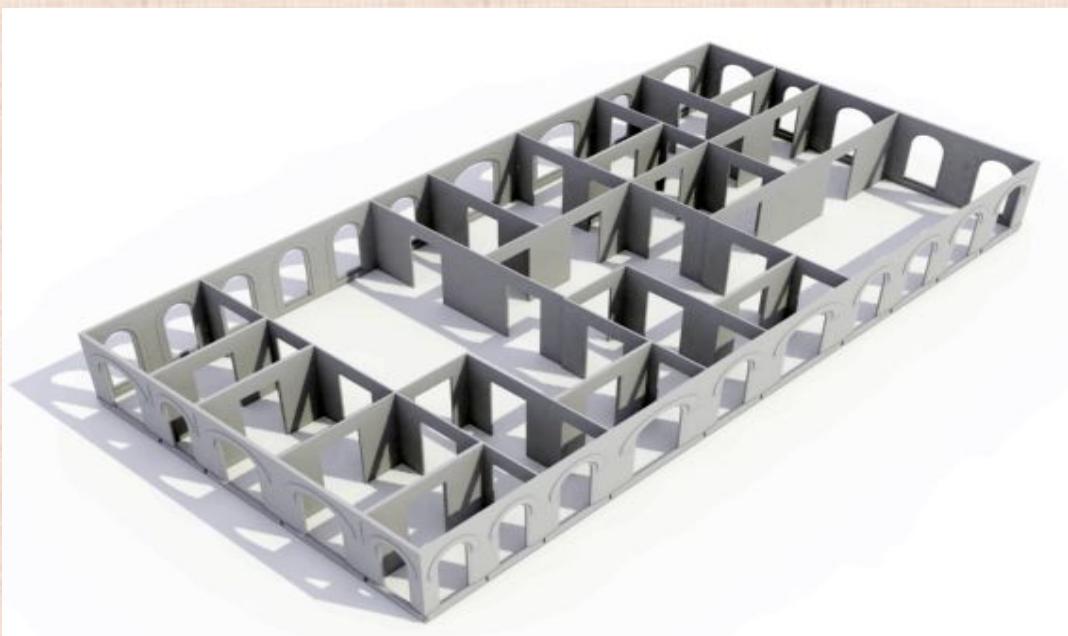
# Computational Biomechanics



# Procedural Methods

# Procedural Geometry

- 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



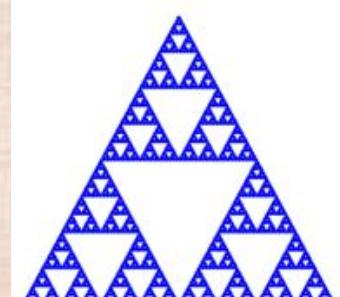
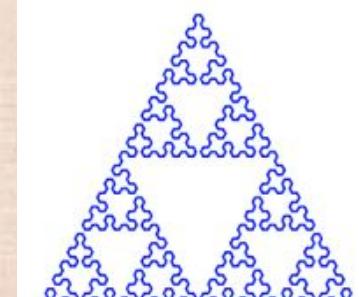
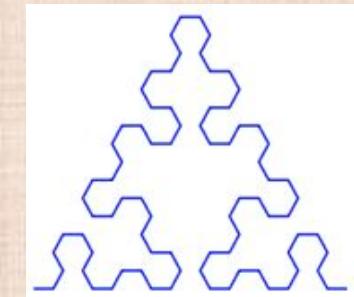
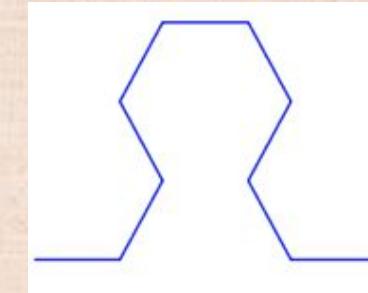
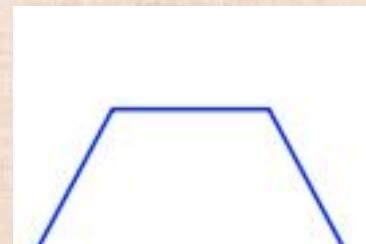
# L-Systems

- 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)
- Starts with an initial string (axiom), and apply production rules
- A translator turns symbols into geometric structures

Nonterminals: A and B both mean to “draw forward”

Terminals: +/- mean to turn right/left (respectively) by 60 degrees

Rules:  $A \rightarrow B + A + B$  and  $B \rightarrow A - B - A$



Intial 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
- Rules:  $X \rightarrow F - [[X]+X] + F [+FX] - X$ ,  $F \rightarrow FF$



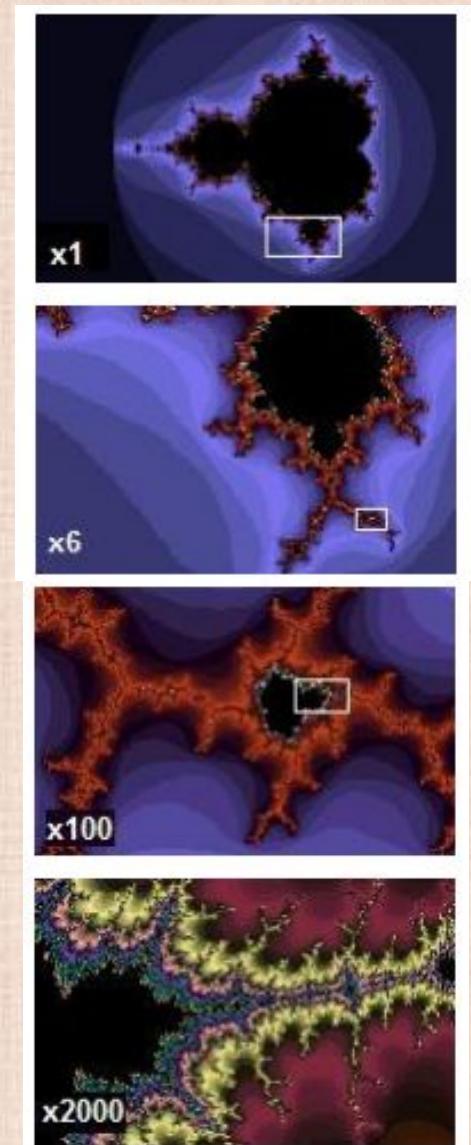
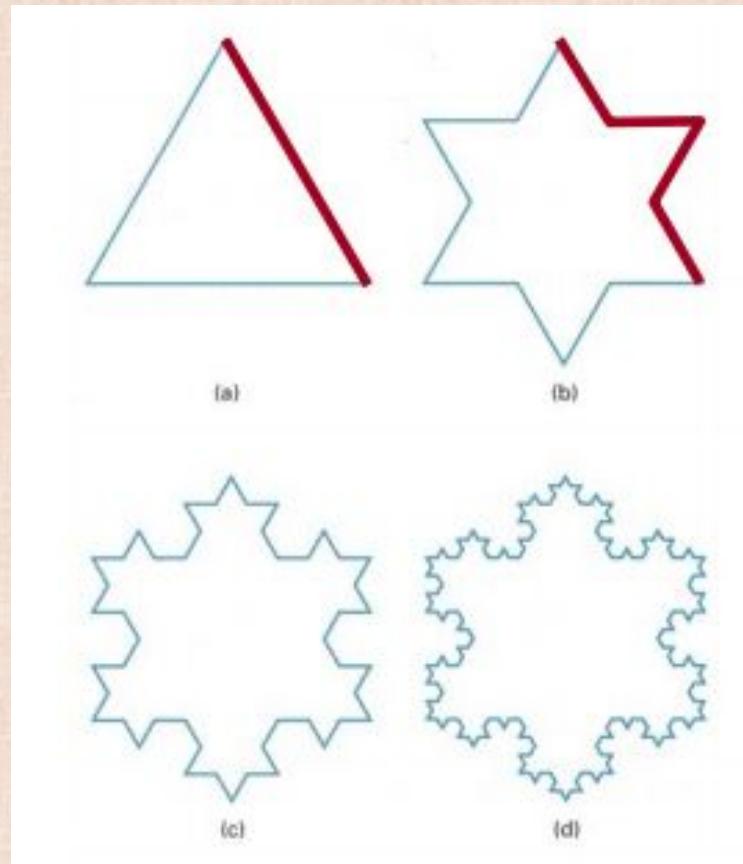
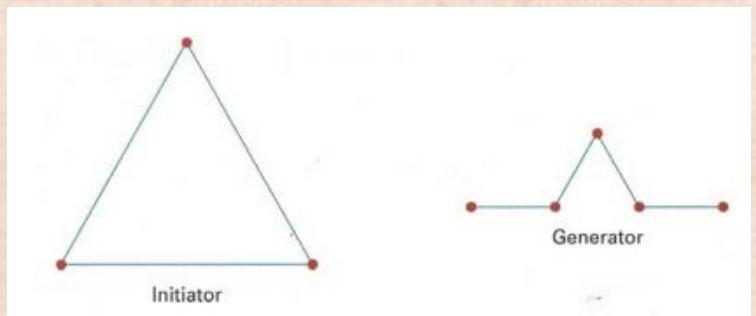
# L-Systems

- 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 modern day ML)



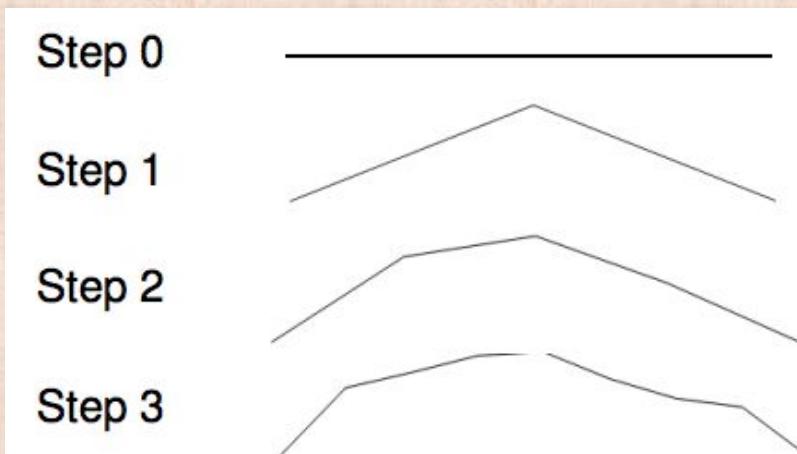
# Fractals

- Initiator: start with a shape
- Generator: replace subparts with scaled copy of the original
- Repeat



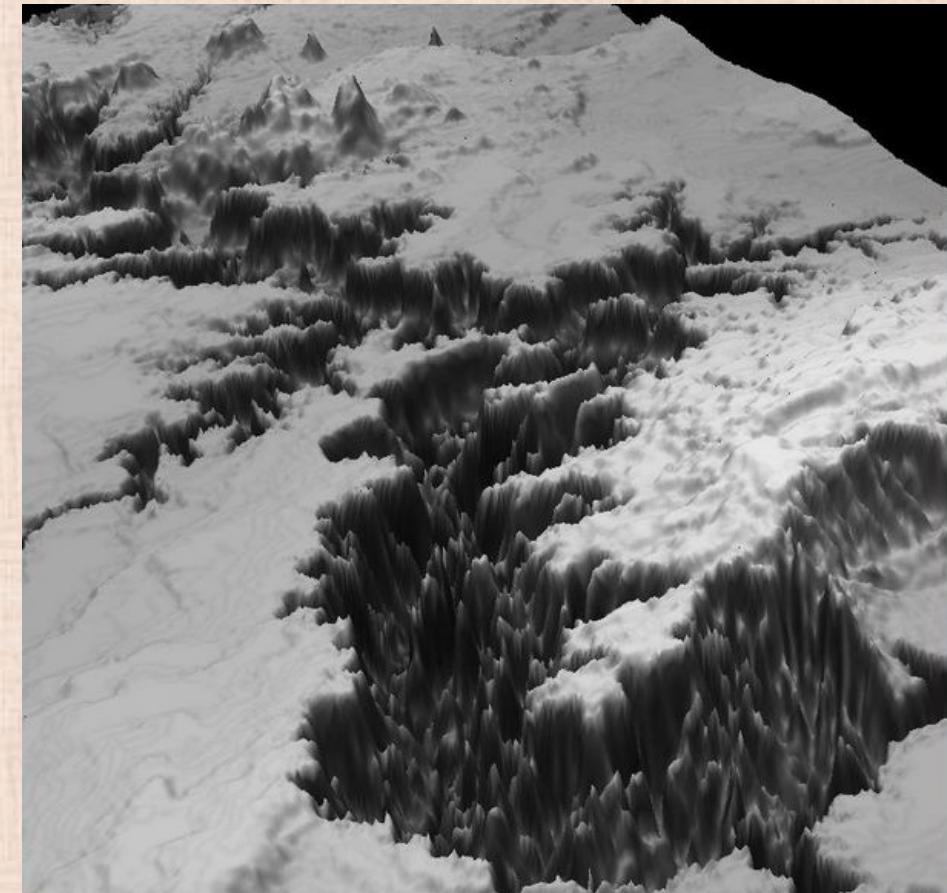
# Fractals

- Add randomness (such as random midpoint location/displacement)
  - Create an irregular 2D silhouette (for far away mountains)



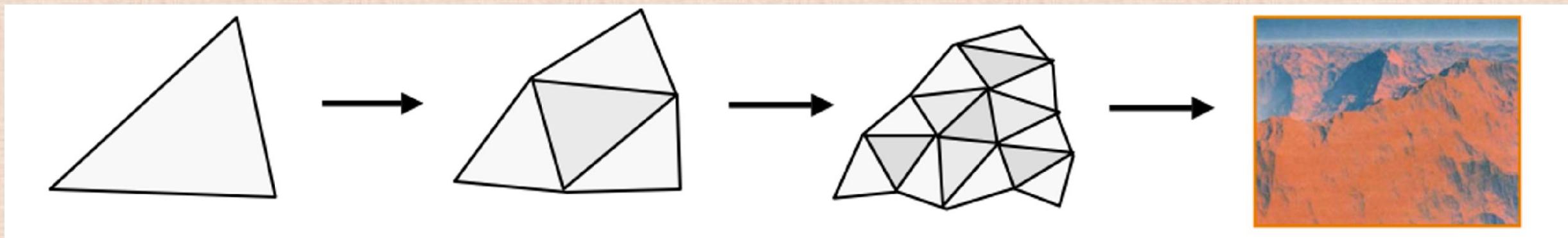
# Height Fields

- 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, vary the height based on pixel intensity



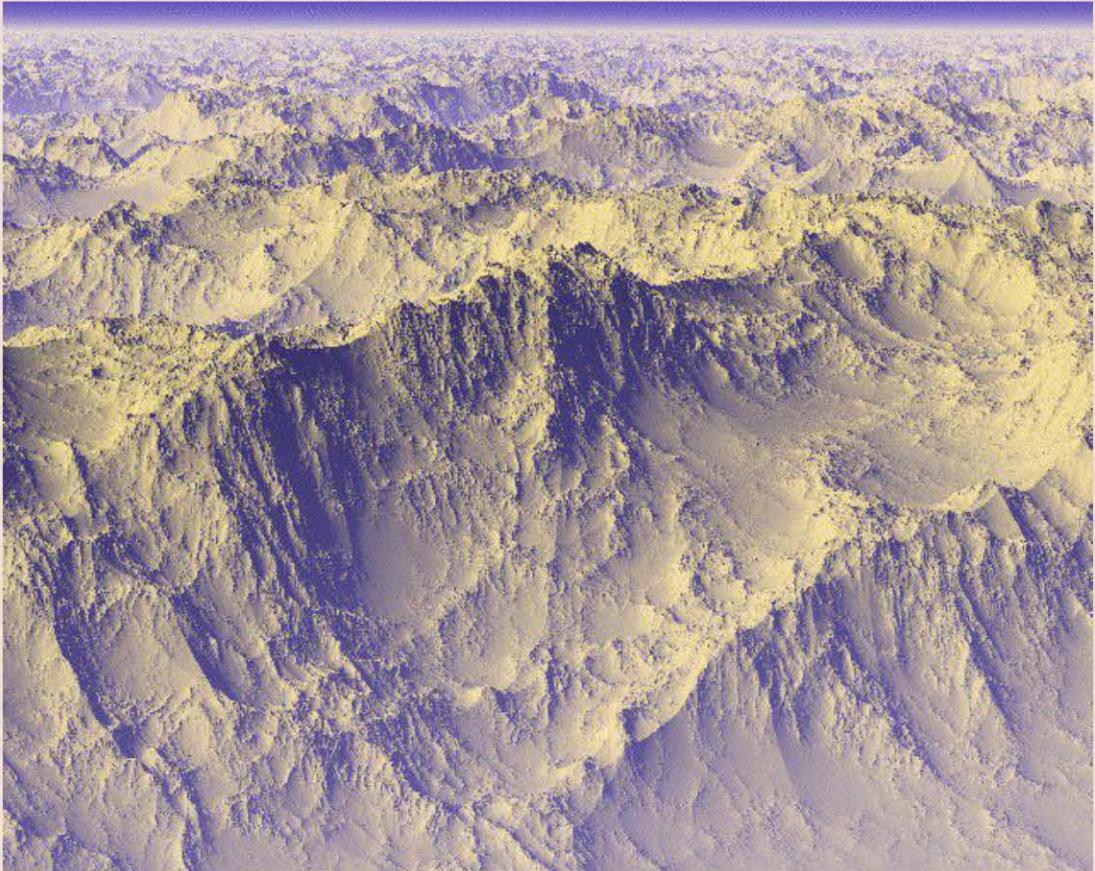
# 3D Landscapes

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

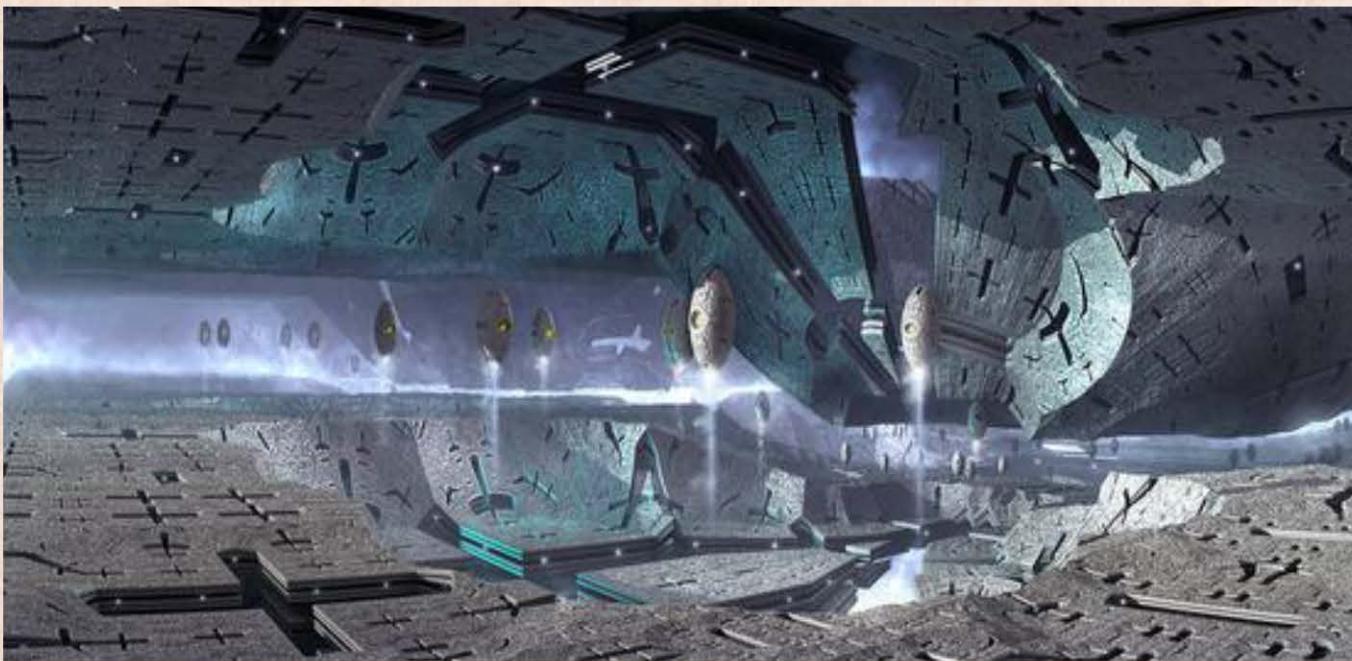
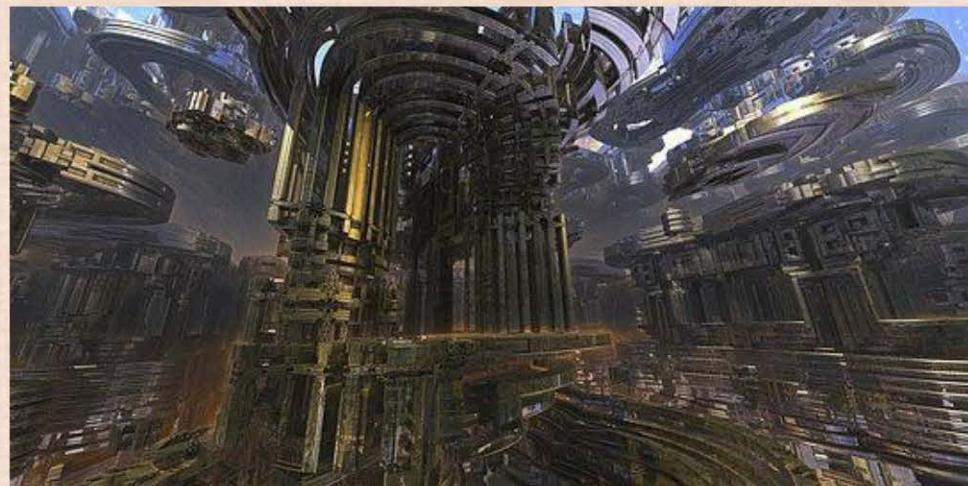


- Similar to subdivision, but with much more interesting rules for setting vertex positions

# 3D Landscapes



# Fractal Worlds



# Machine Learning

# Machine Learning

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

