



中国科学技术大学
University of Science and Technology of China



GAMES 102在线课程

几何建模与处理基础

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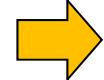


GAMES 102在线课程：几何建模与处理基础

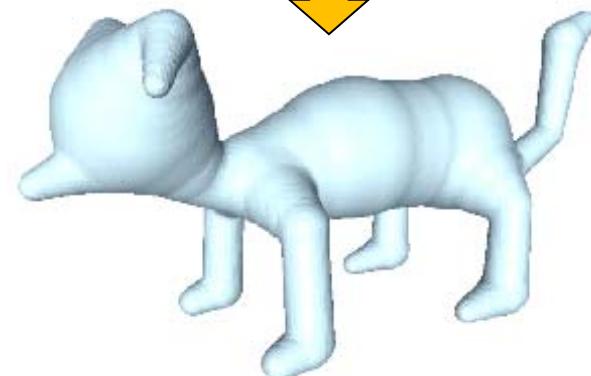
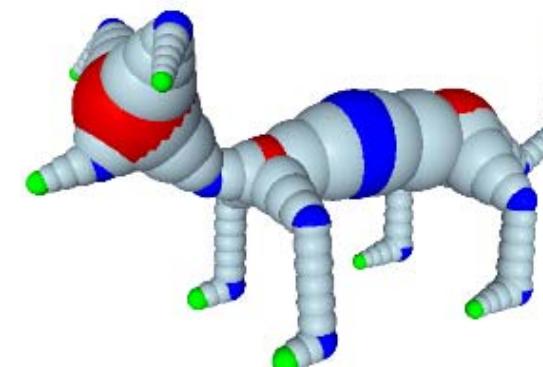
纹理合成

回顾：曲面重建或建模

曲面重建



曲面建模



Textured Surfaces



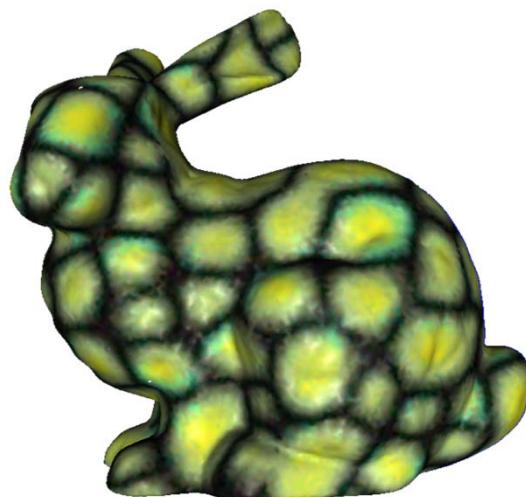
Real object



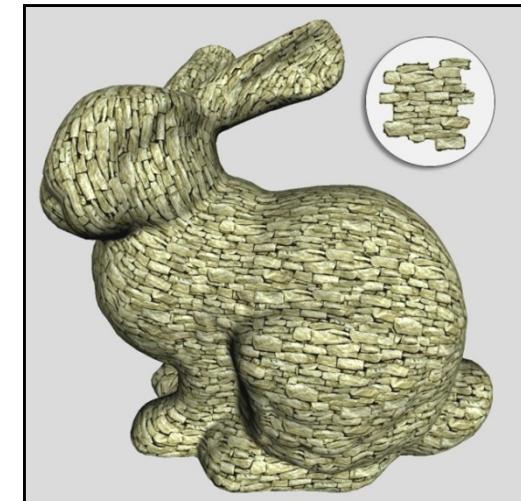
Model without texture



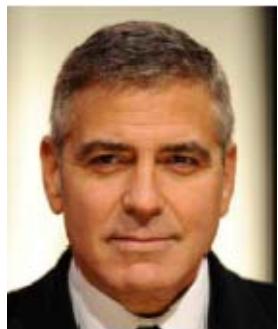
Textured model



Textured models



Texturing 3D Models

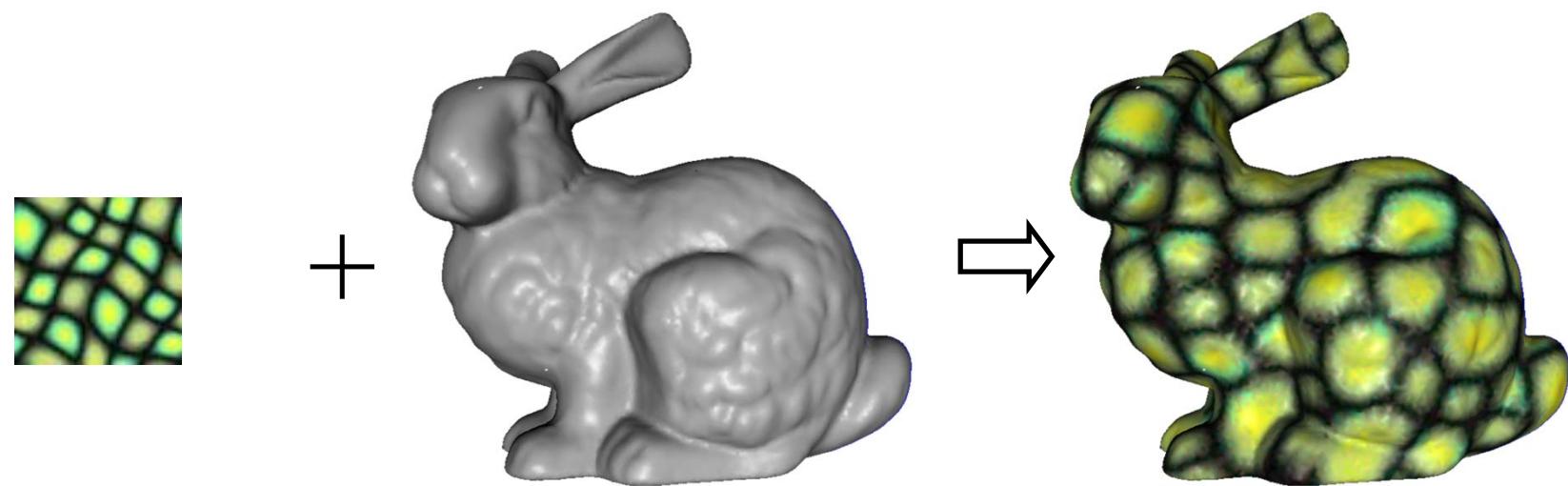


From input images

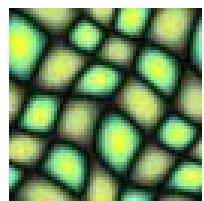


Manually designed

Texture Synthesis

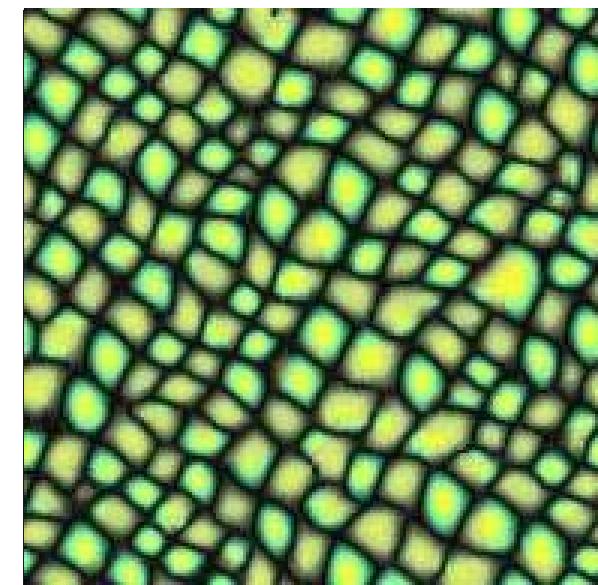


2D Texture Synthesis



Input
Sample

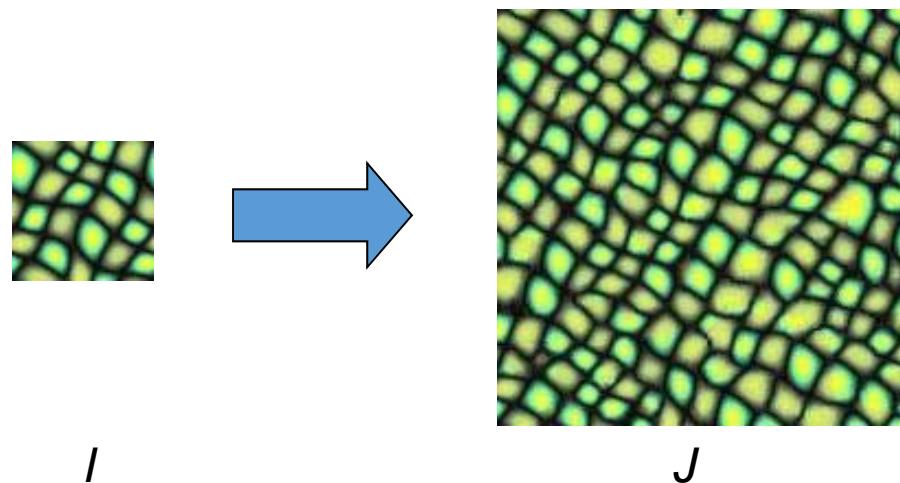
Texture Synthesis



Result

Texture Synthesis

- Given a sample texture I , the goal is to synthesize a new texture J that
 - looks like the input sample
 - no part of result is the duplicate of the sample



Applications



Texture Mapping



Texture Replacement



Special Effects

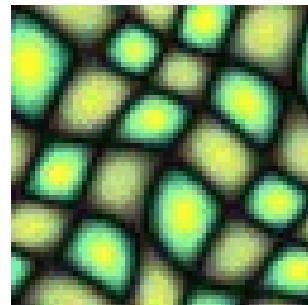
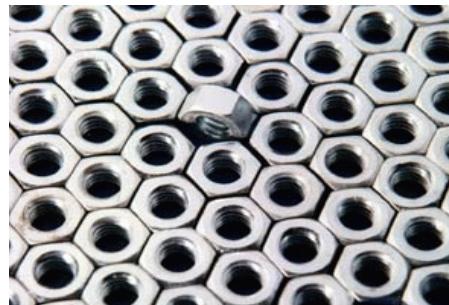


Object Removal / Hole Filling

1. Textures

What is texture?

- Characterized by appearance
- Elements arranged quasi-randomly



...
describing the response of that neuron—as a function of position—is perhaps the most functional description of that neuron. We seek a single conceptual and mathematical framework to describe the wealth of simple-cell receptive fields neurophysiologically^{1–3} and inferred⁴, especially if such a framework has the virtue of helping us to understand the function in a deeper way. Whereas no generic model exists (DOG), difference of offset Gaussians (DOG), difference of offset Gabor functions, and so on—can be expected to fit a simple-cell receptive field, we nonetheless have

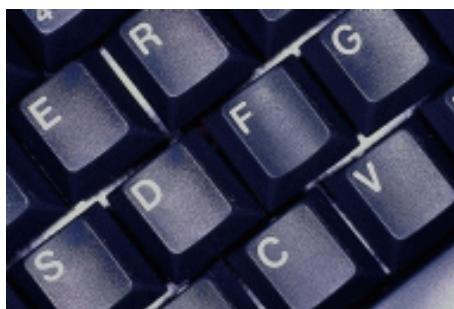
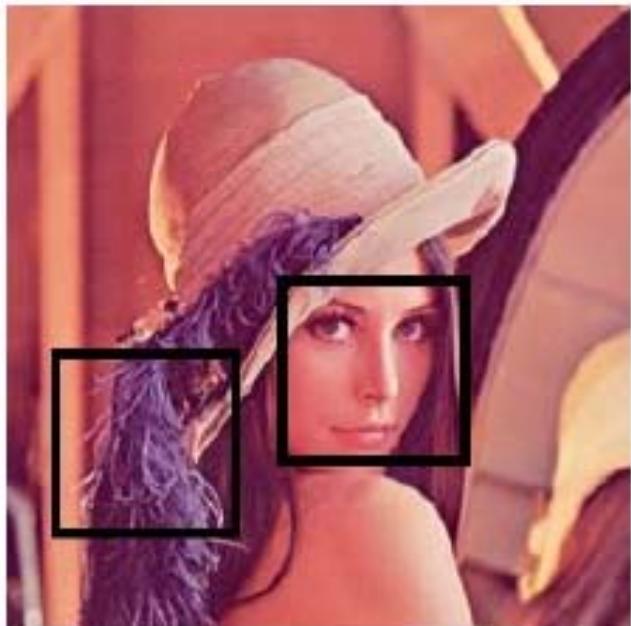
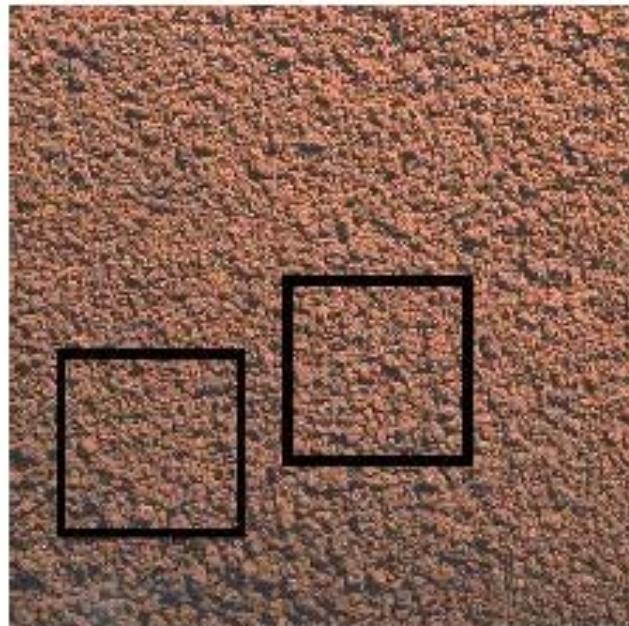


Image VS. Texture



(a)



(b)



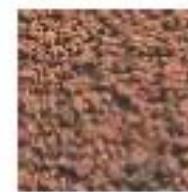
(a1)



(a2)



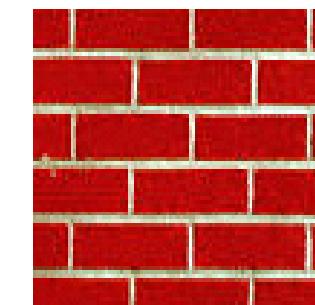
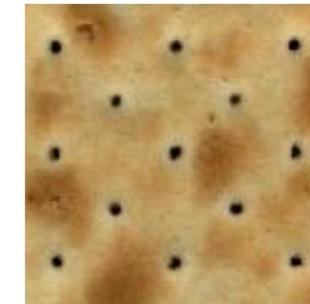
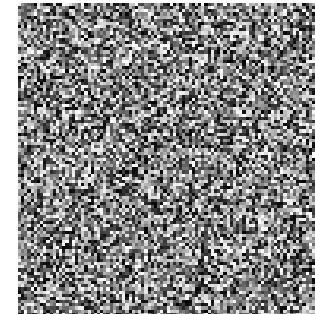
(b1)



(b2)

Texture

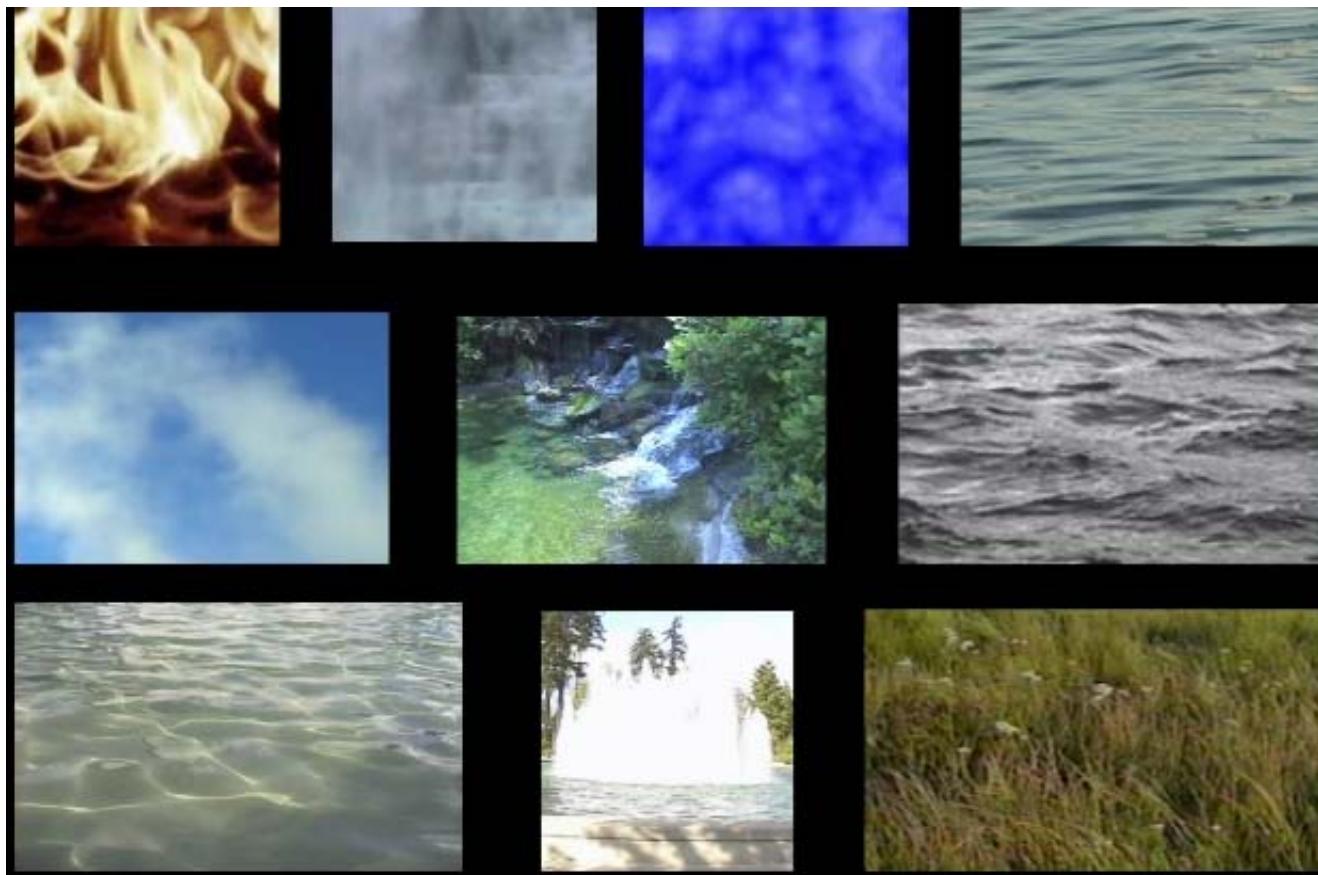
- Images containing repeating patterns
 - Locality: small parts look alike
 - Stochastic: they never look exactly the same



...
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as a function of position—is perhaps
functional description of that neuron.
seek a single conceptual and mathem
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leerer way. Whereas no generic mo
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ivative of a Gaussian, higher derivati
function, and so on—can be expect
imple-cell receptive field, we noneth

Dynamic Textures

- Appearance similar even as texture evolves



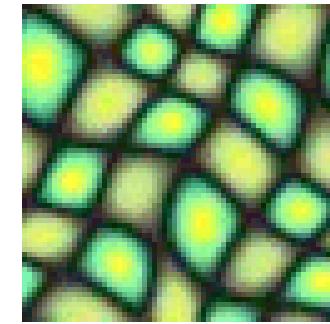
Classification (Attribution)



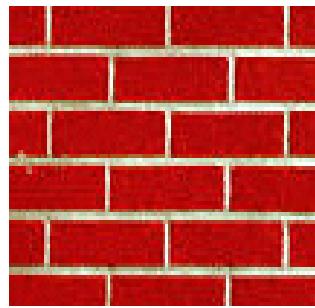
Isotropic



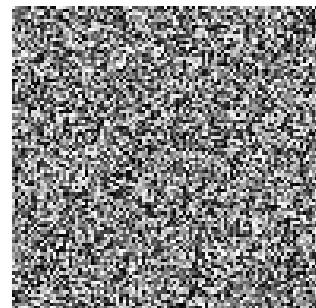
Anisotropic



Both?



Repeated



Stochastic



Both?

2. 2D Texture Synthesis

Texture Synthesis: A Simple Solution ... Tiling



small texture

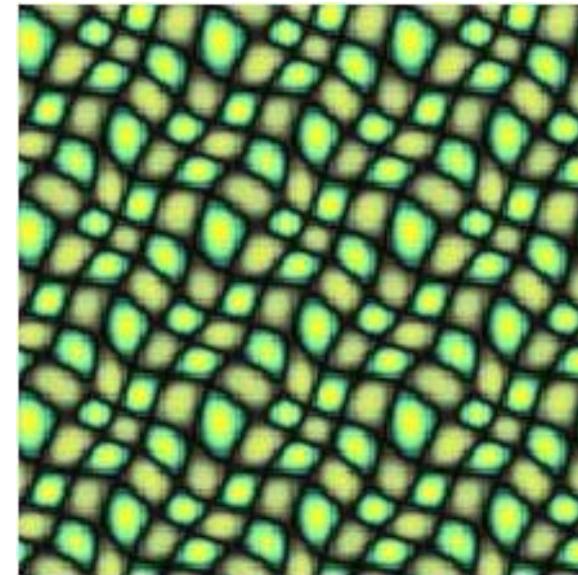
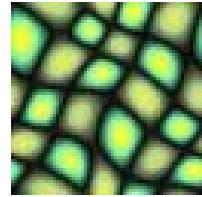


big texture?

Tiling is obvious!

Human eyes are sensitive to repeated pattern!

Tilable Texture

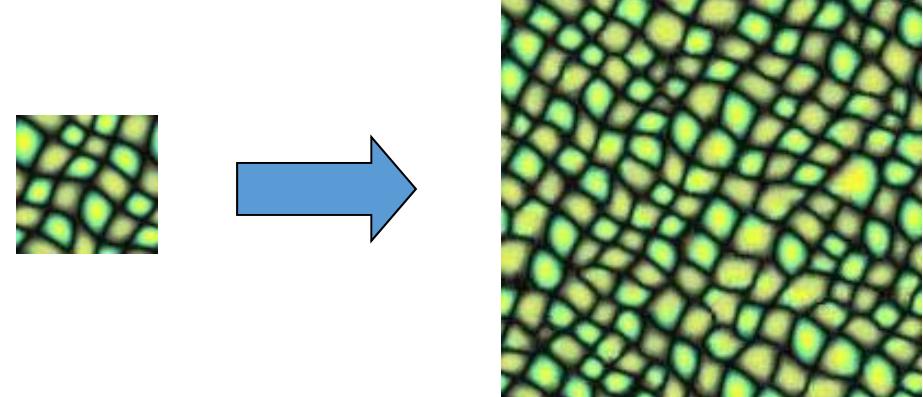


Tilable texture

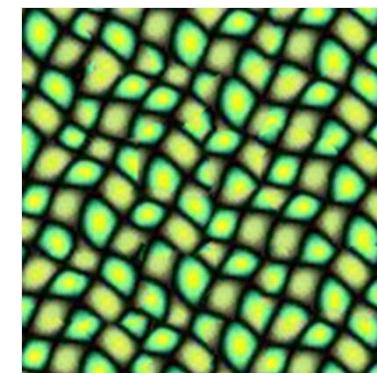
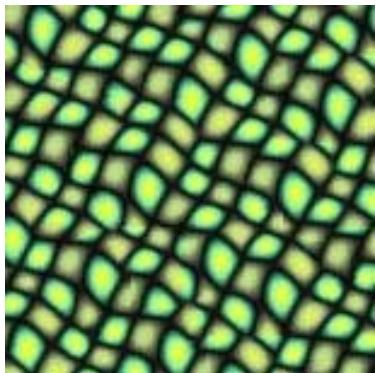
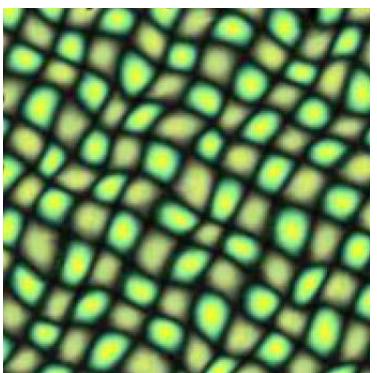
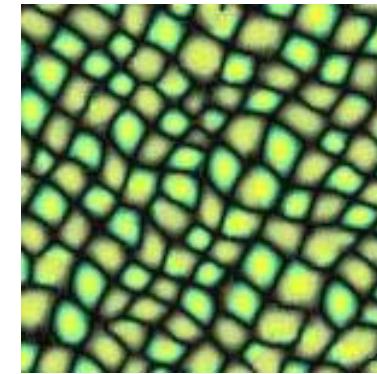
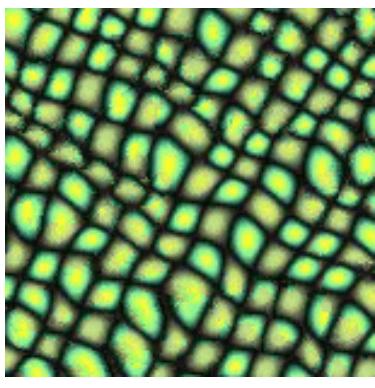
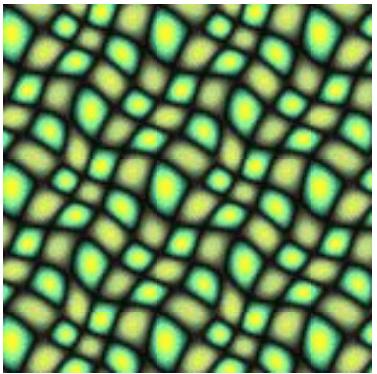
Simple tiling

Desirable Properties

- Result looks like the input
- Efficient
- General
- Easy to use
- Extensible



Ideal Solution?

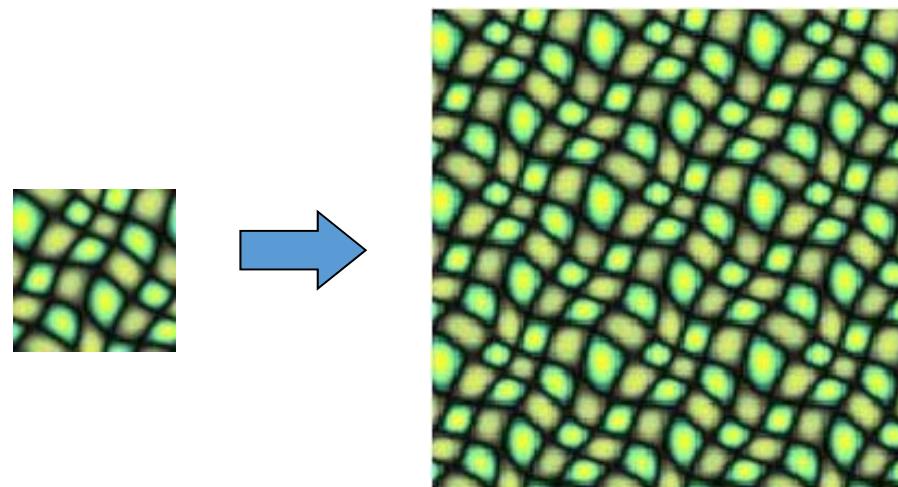


Challenges

- Texture analysis
 - how to capture the essence of texture?
- Need to model the whole spectrum
 - from repeated to stochastic texture
- This problem is at intersection of vision, graphics, statistics, and image compression

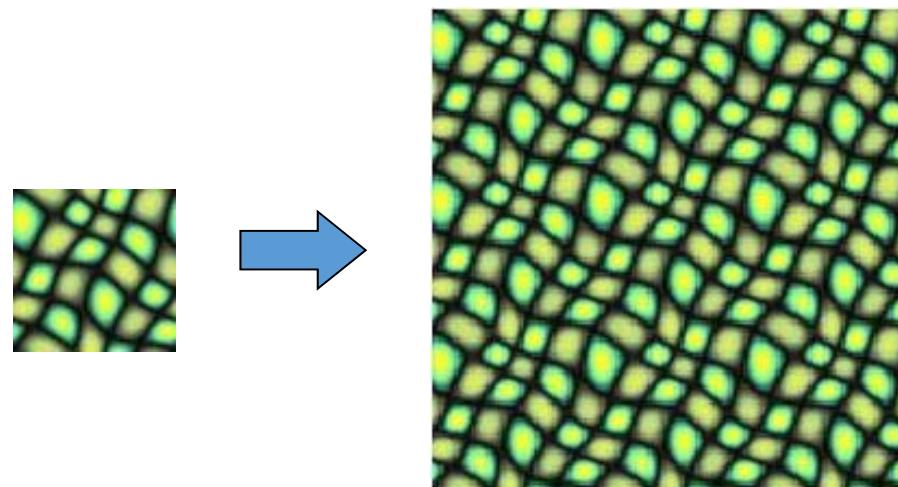
Example-based Techniques

- Parametric Techniques
 - Compute global statistics in feature space and sample images from texture ensemble directly
- Non-parametric Techniques
 - Estimate local conditional probability density function and synthesize pixels incrementally



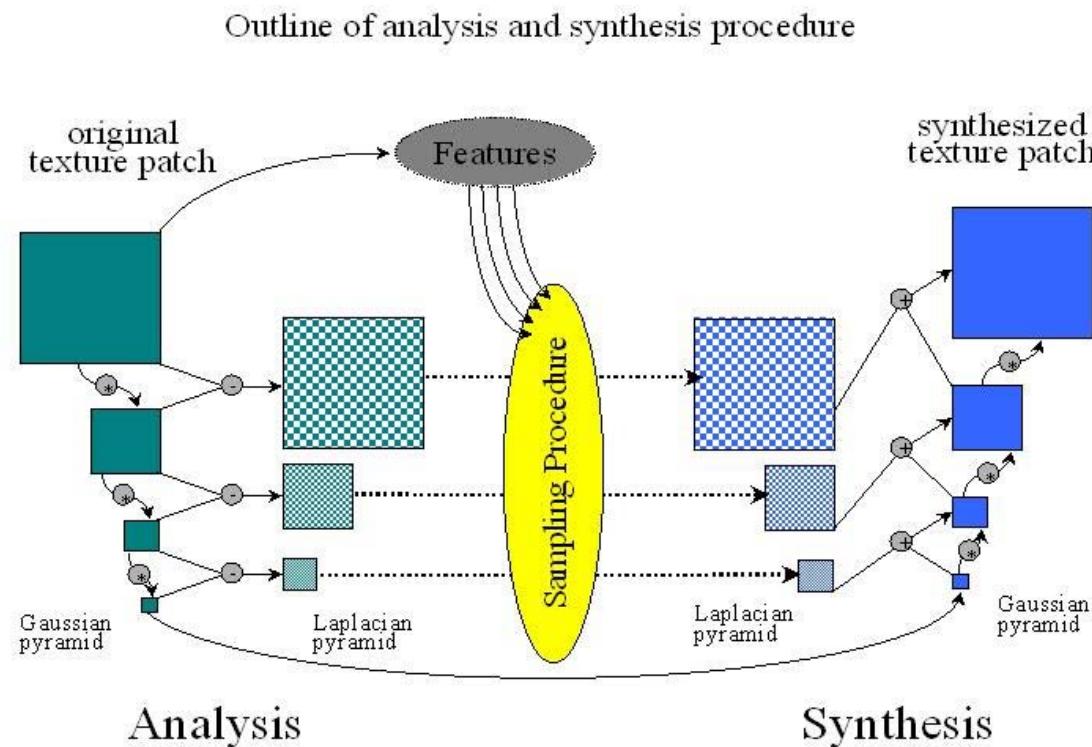
Example-based Techniques

- **Parametric Techniques**
 - Compute global statistics in feature space and sample images from texture ensemble directly
- Non-parametric Techniques
 - Estimate local conditional probability density function and synthesize pixels incrementally



Parametric Techniques

- Hypothesize a **mathematical model** for texture representation
- **Match** model parameters of input and output texture



Pyramid-Based Texture Analysis/Synthesis

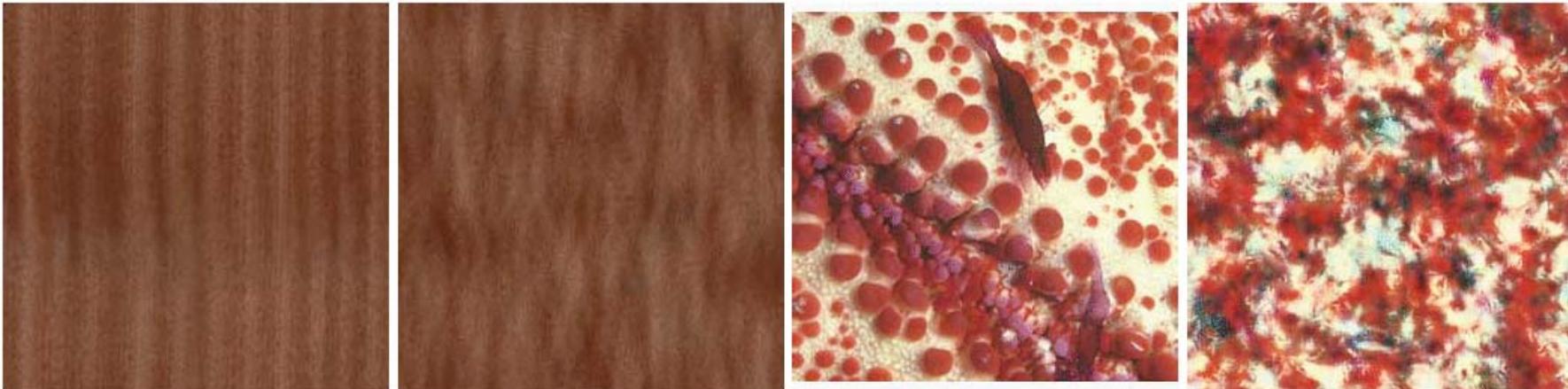
[Heeger & Bergen, Siggraph 1995]

- Initialize J to noise
- Create multiresolution pyramids for I and J
- Match the histograms of J 's pyramid levels with I 's pyramid levels
- Loop until convergence
- Can be generalized to 3D

Results

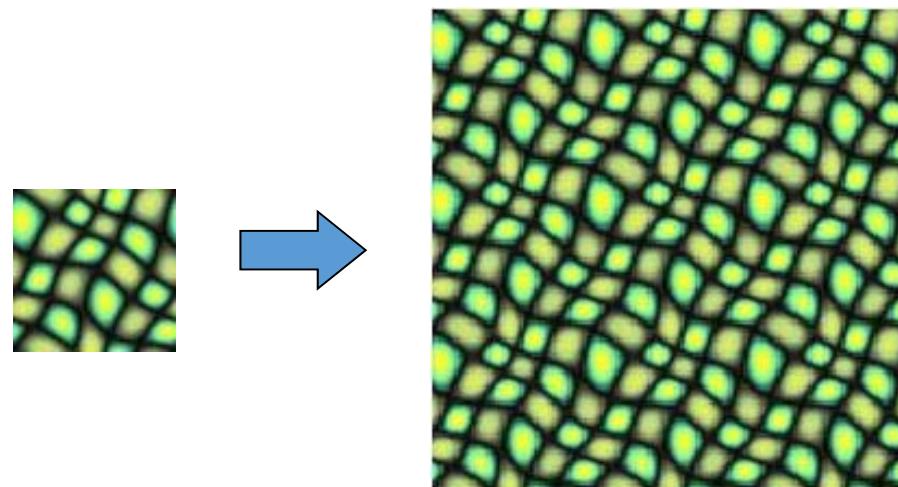


Failures



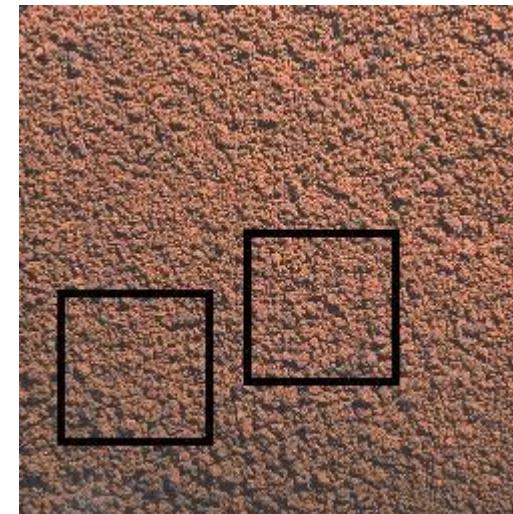
Example-based Techniques

- Parametric Techniques
 - Compute global statistics in feature space and sample images from texture ensemble directly
- Non-parametric Techniques
 - Estimate local conditional probability density function and synthesize pixels incrementally



Non-Parametric Techniques

- Synthesis by **copying** from the input
- Markov-Random Field Model
 - Pixel appearance depends only on neighborhood

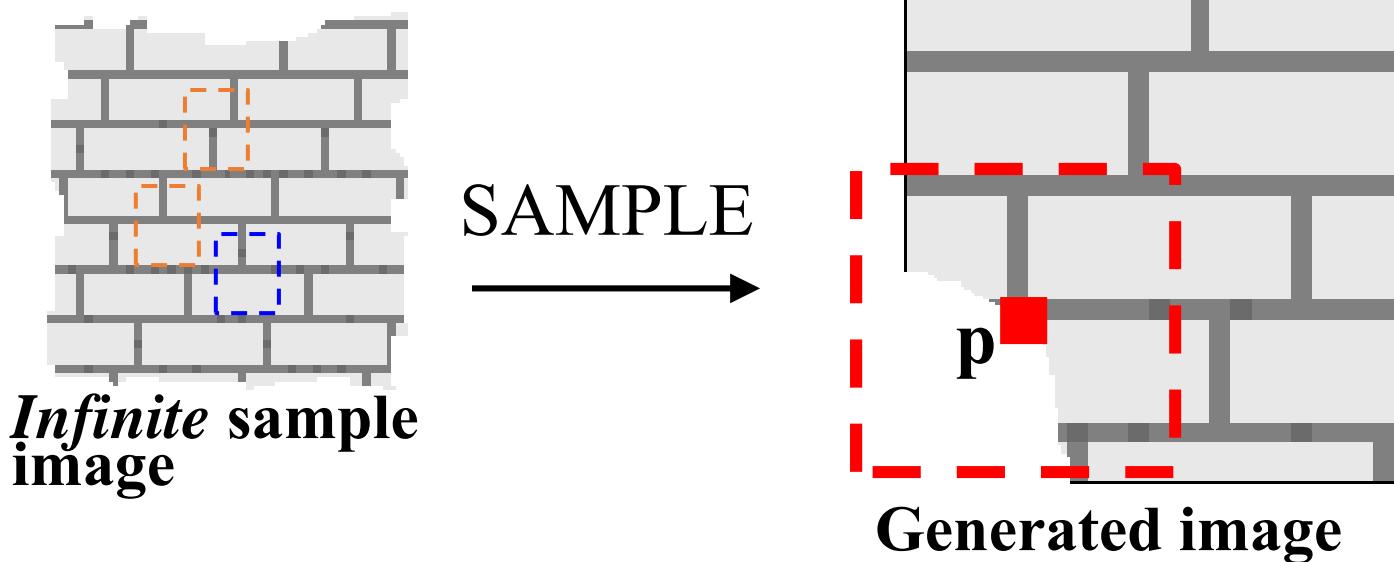


Markov Random Field

- [Shannon,'48] proposed a way to generate English-looking text using N-grams:
 - Assume a generalized Markov model
 - Use a large text to compute prob. distributions of each letter given N-1 previous letters
 - Starting from a seed repeatedly sample this Markov chain to generate new letters
 - Also works for whole words

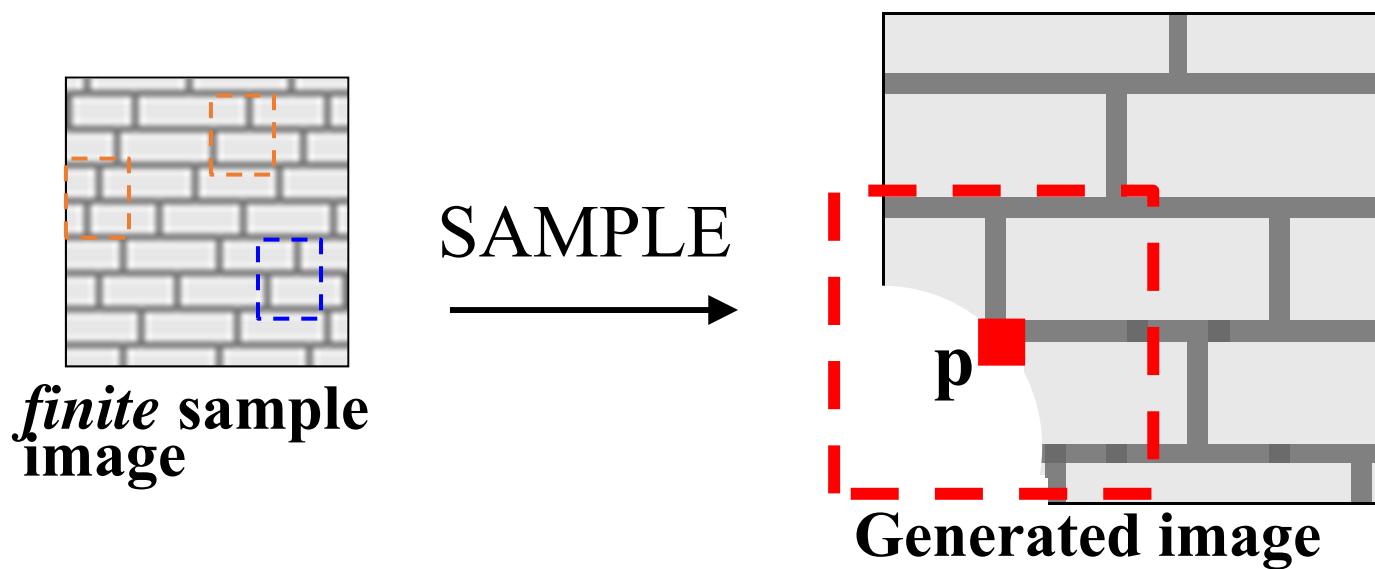
WE NEED TO EAT CAKE

Synthesizing One Pixel



- Assuming Markov property, what is conditional probability distribution of p , given the neighbourhood window?
- Instead of constructing a model, let's directly search the input image for all such neighbourhoods to produce a histogram for p
- To synthesize p , just pick one match at random

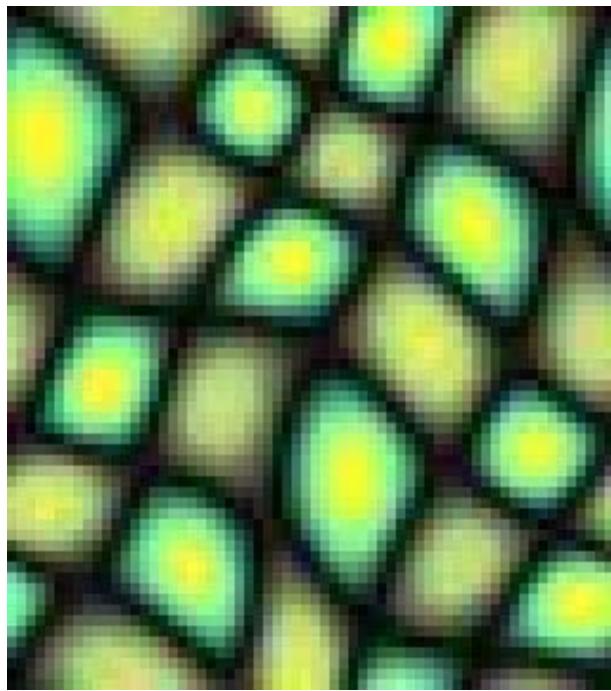
Really Synthesizing One Pixel



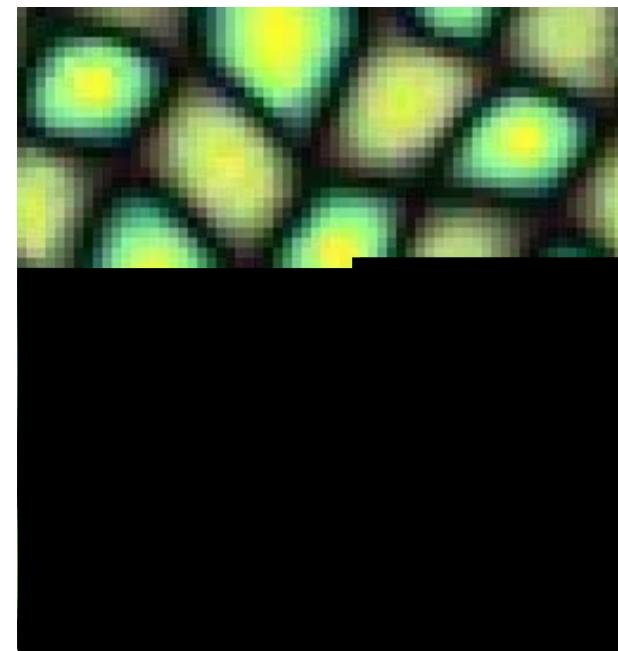
- However, since our sample image is finite, an exact neighbourhood match might not be present
- So we find the **best** match using SSD error (weighted by a Gaussian to emphasize local structure), and take all samples within some distance from that match

Pixel-Based Texture Synthesis

[Efros and Leung, ICCV 1999]



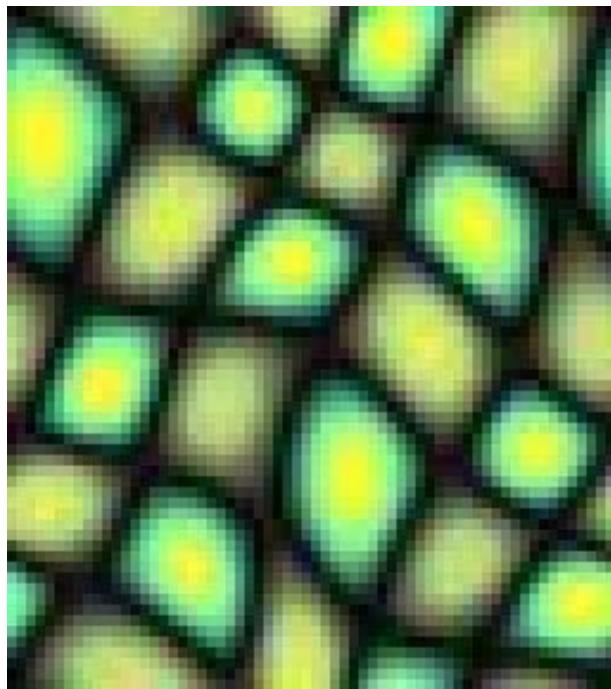
Input



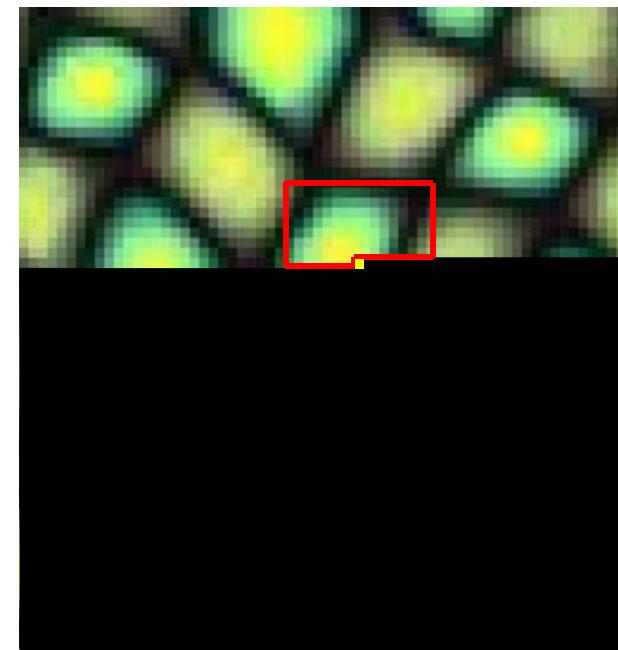
Output

Step n in Algorithm

Pixel-Based Texture Synthesis



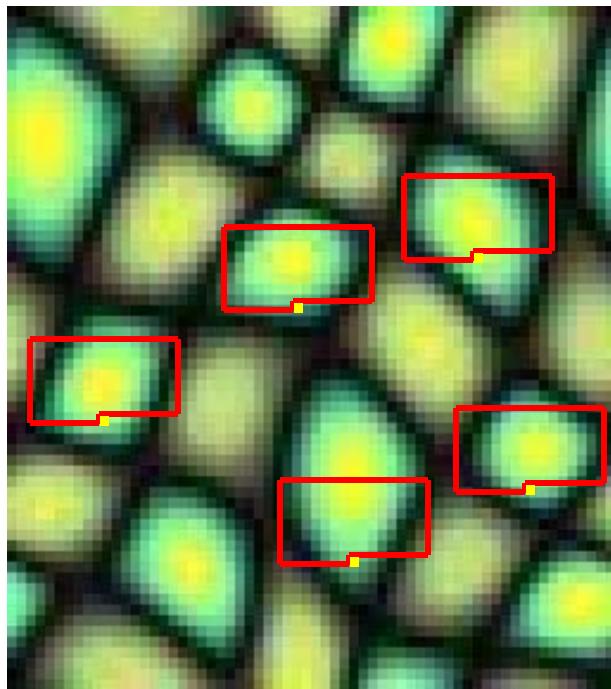
Input



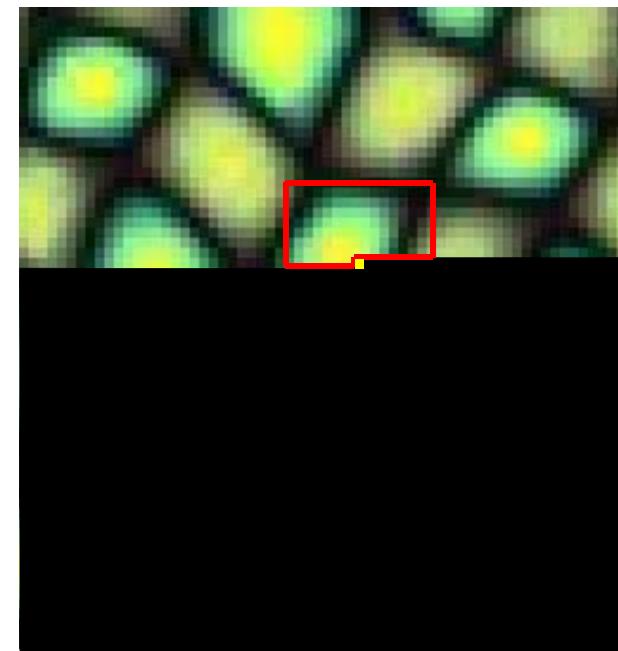
Output

Build L-shaped Neighborhood

Pixel-Based Texture Synthesis



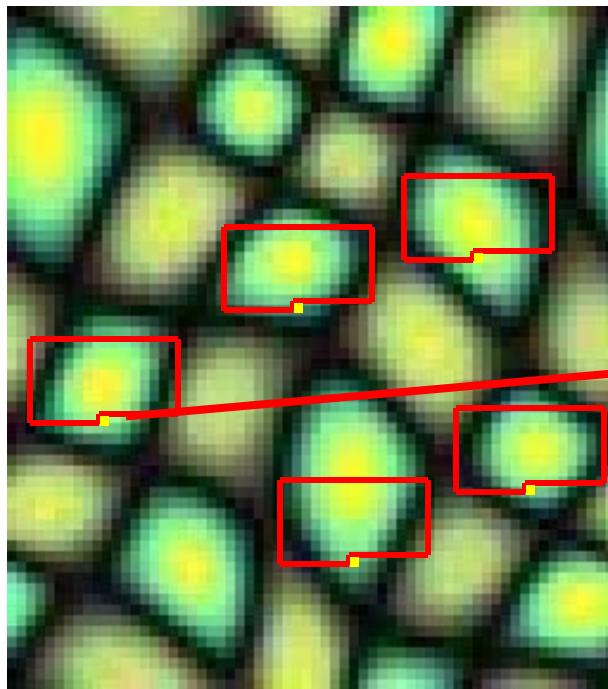
Input



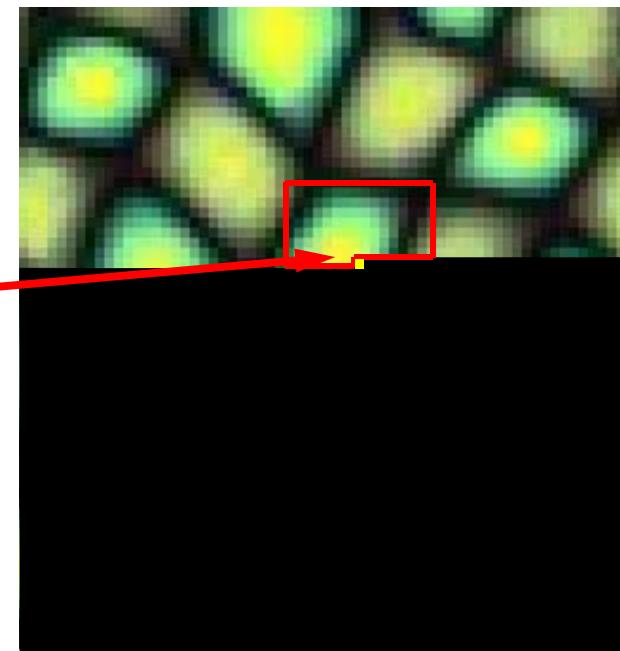
Output

Find Matching Neighborhood(s) in Input

Pixel-Based Texture Synthesis



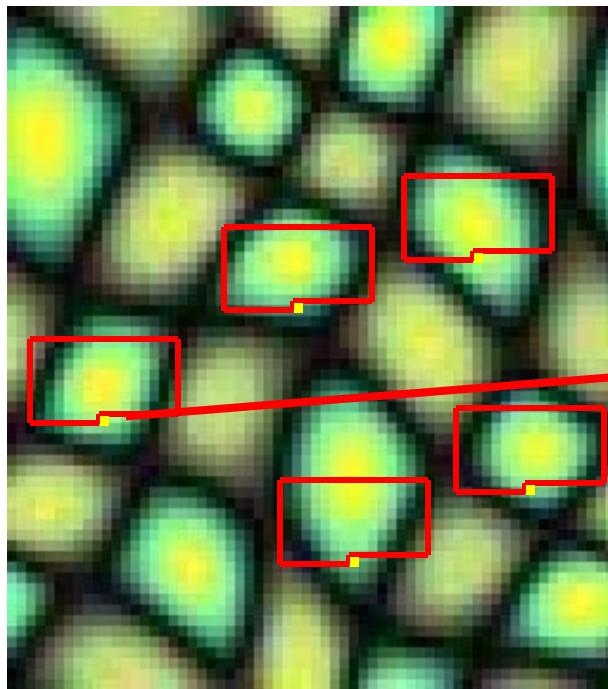
Input



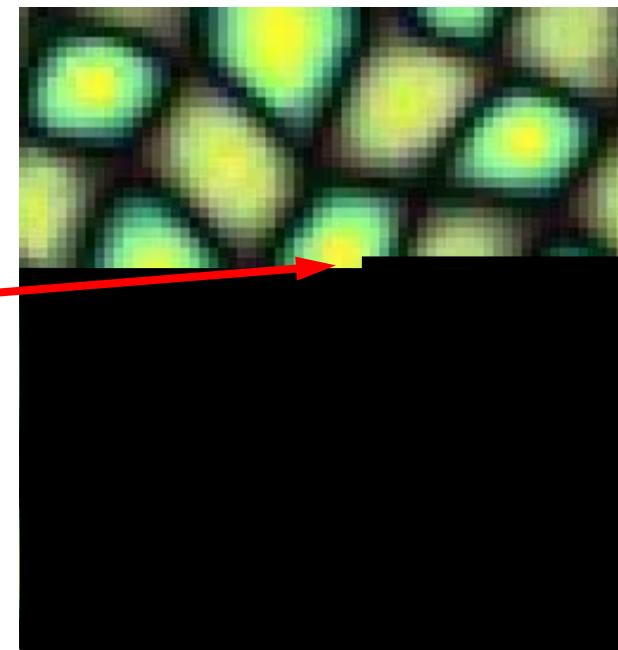
Output

Select *Best* Neighborhood from all Candidates

Pixel-Based Texture Synthesis



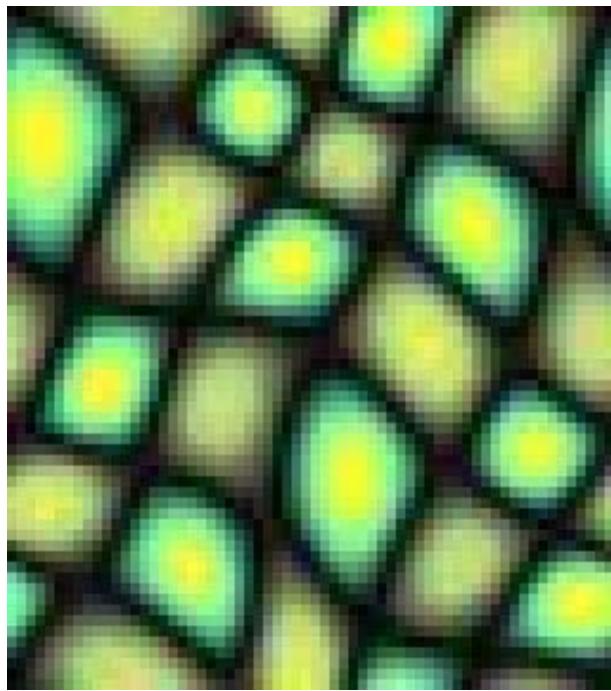
Input



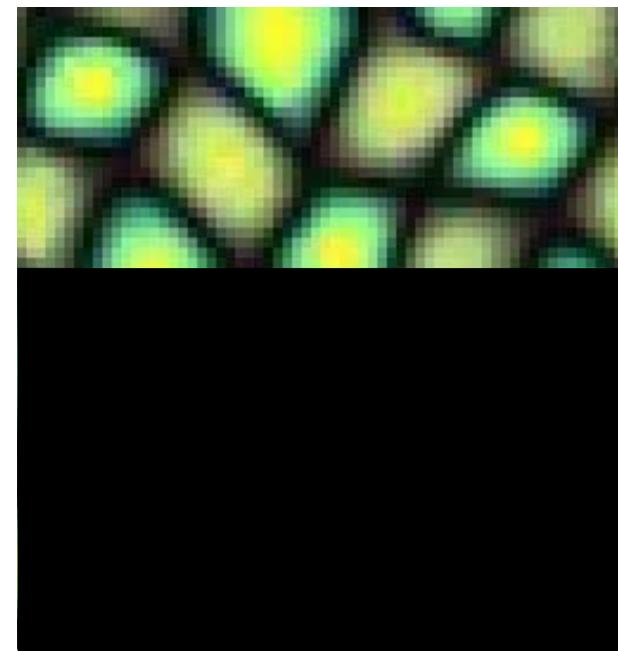
Output

Copy Selected Pixel to Output

Pixel-Based Texture Synthesis



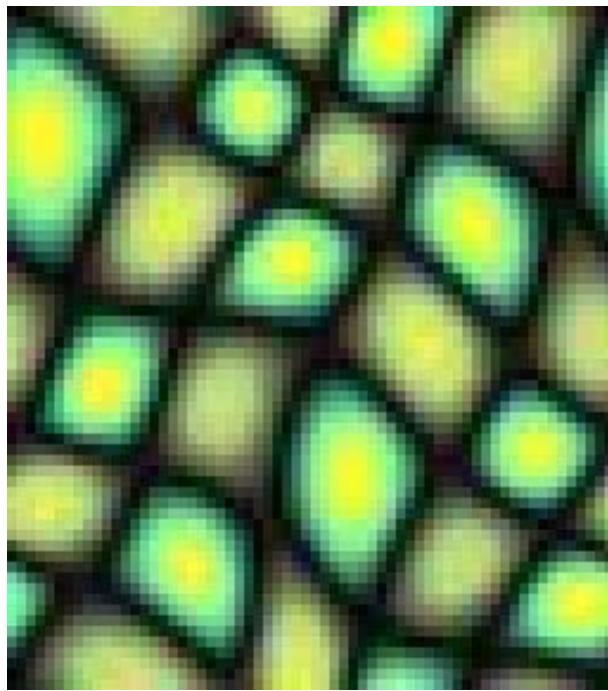
Input



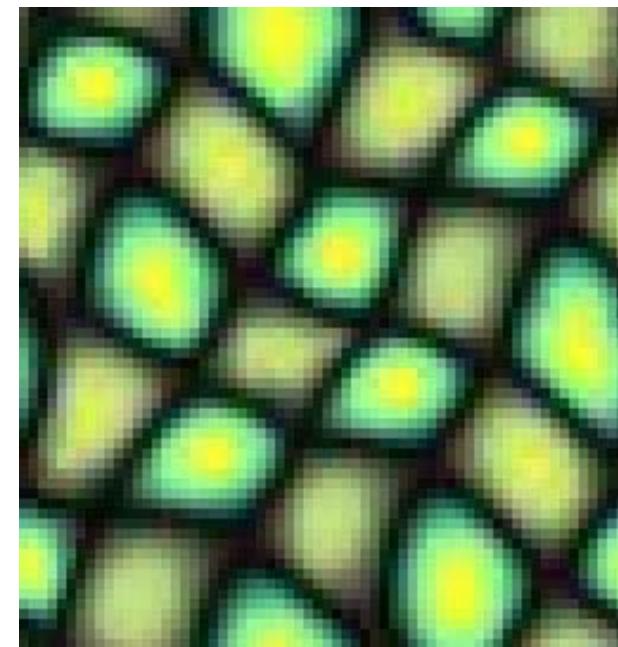
Output

Repeat for all Pixels in this Row...

Pixel-Based Texture Synthesis



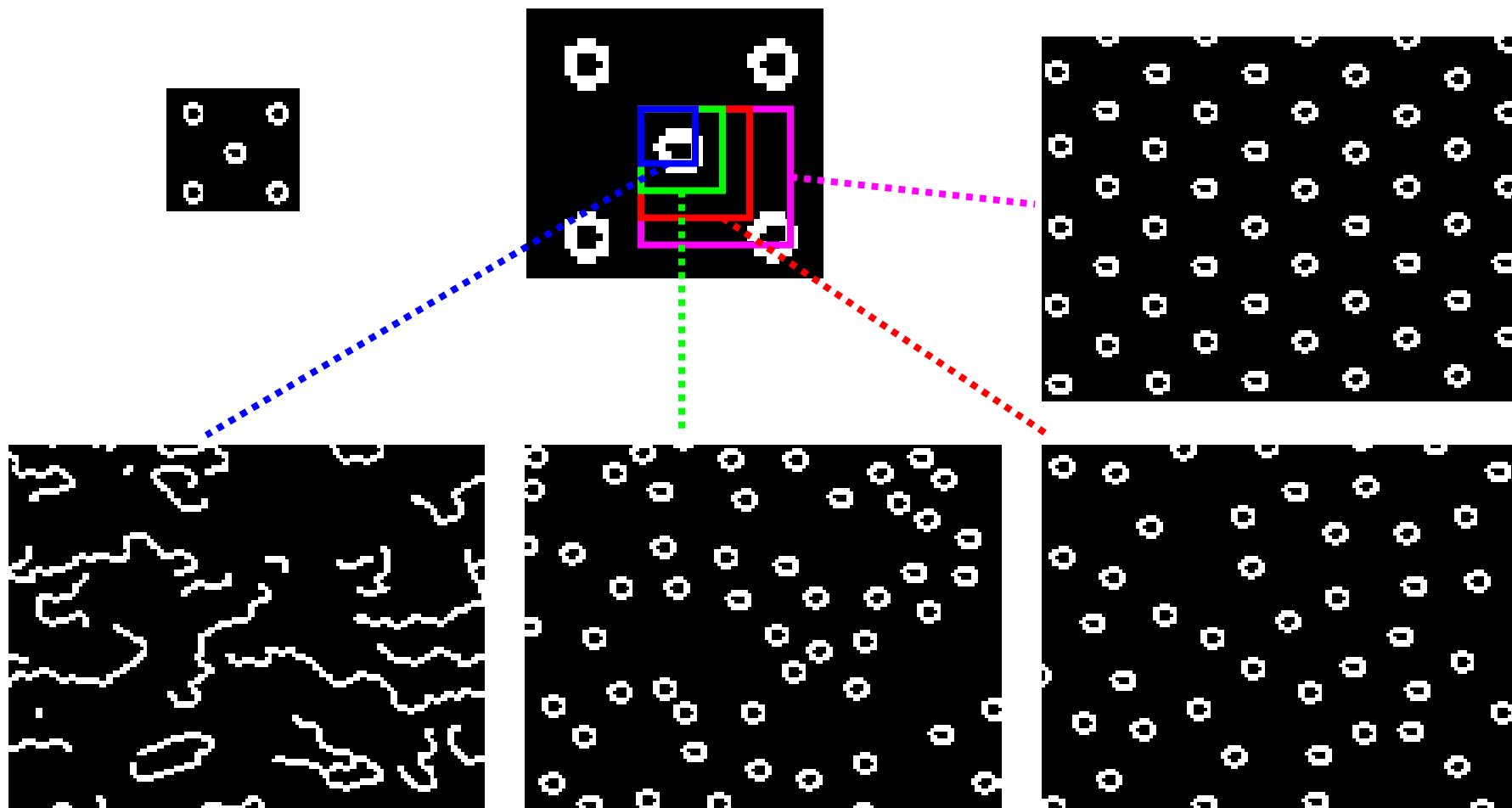
Input



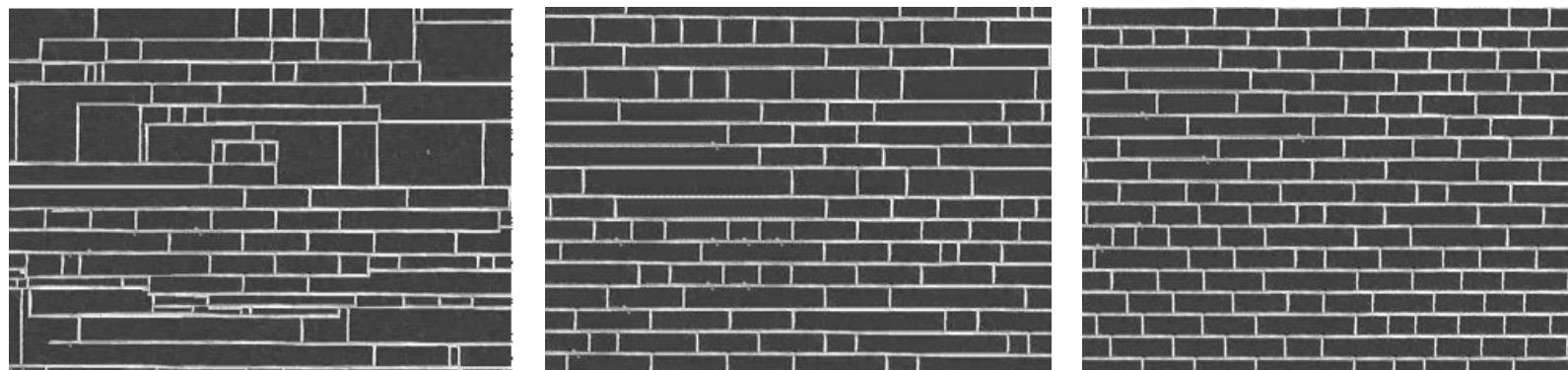
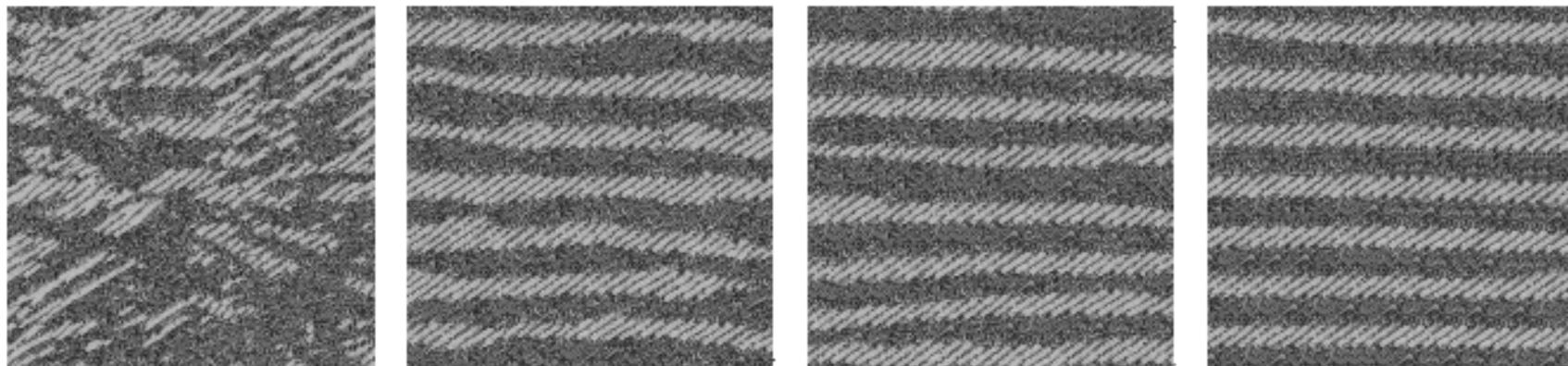
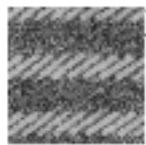
Output

...and all other Output Rows

Randomness Parameter



Results



Increasing window size

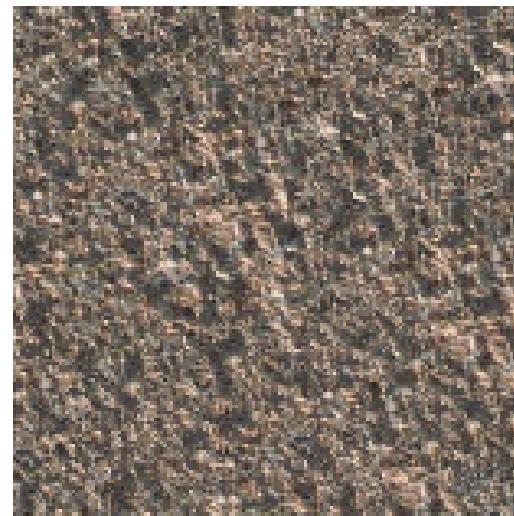


More Results

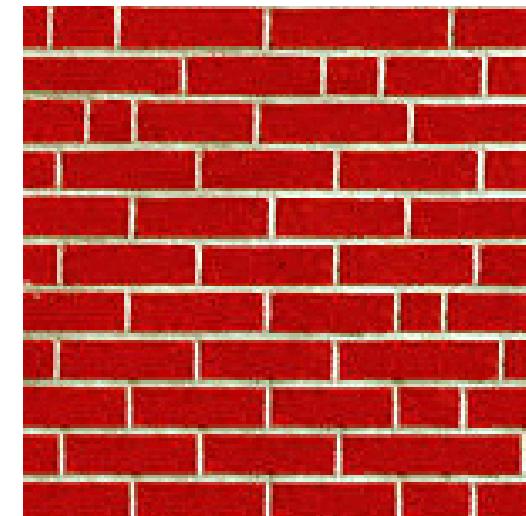
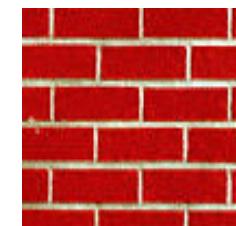
wood



granite



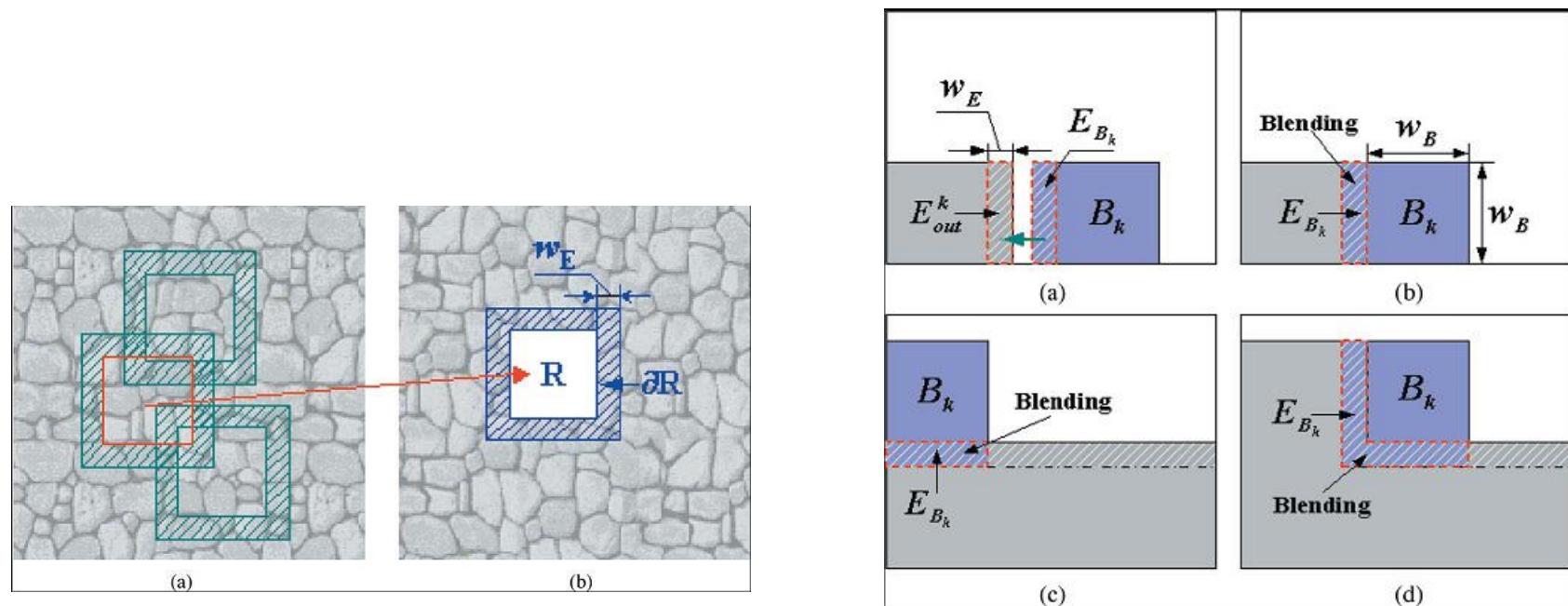
brick wall



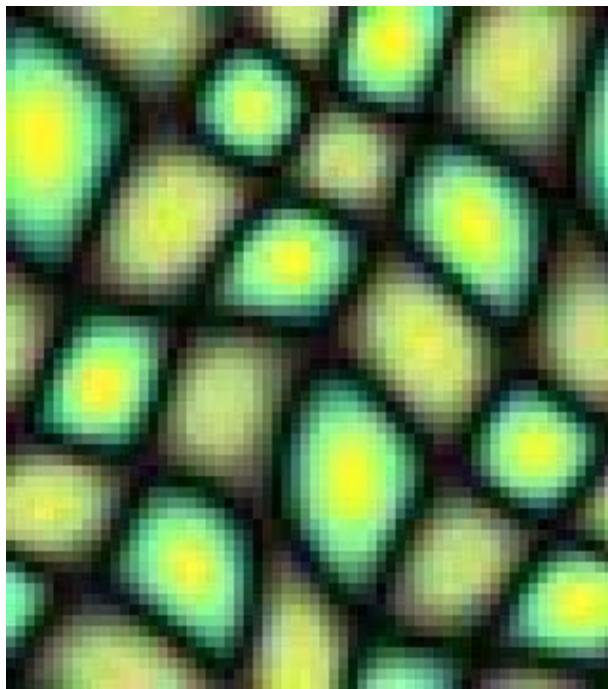
Patch-based Synthesis

[Liang et al. TOG 2002]

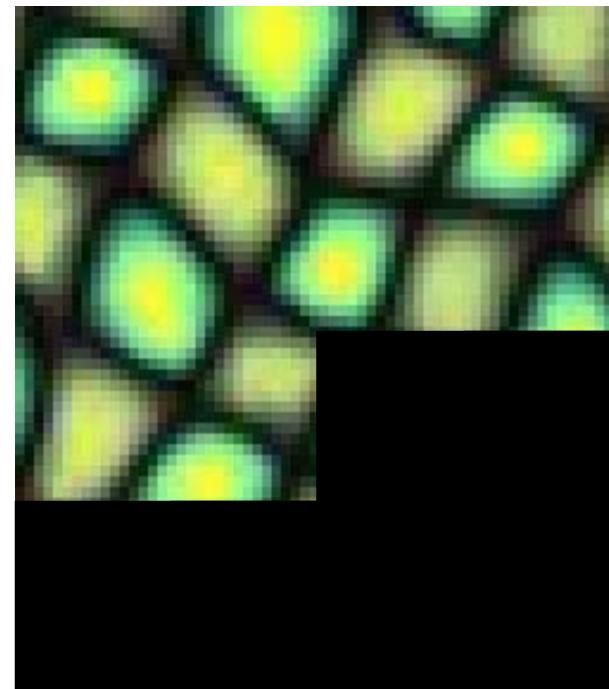
- Copy patches instead of single pixels



Patch-Based Texture Synthesis



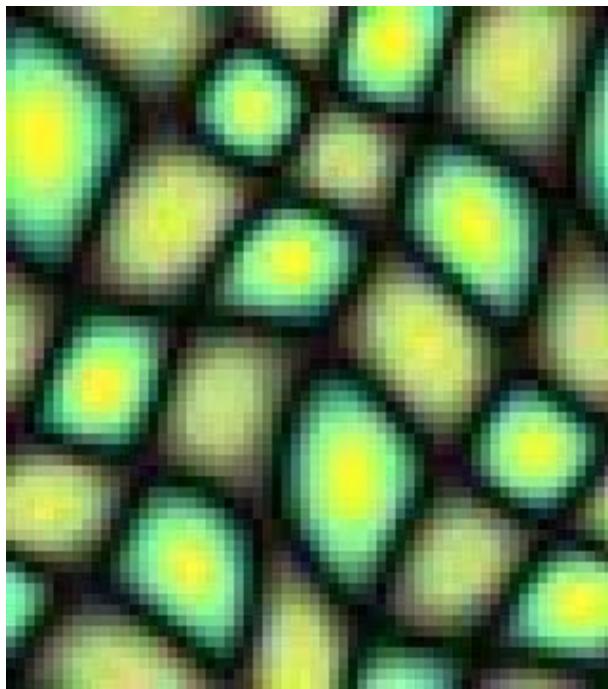
Input



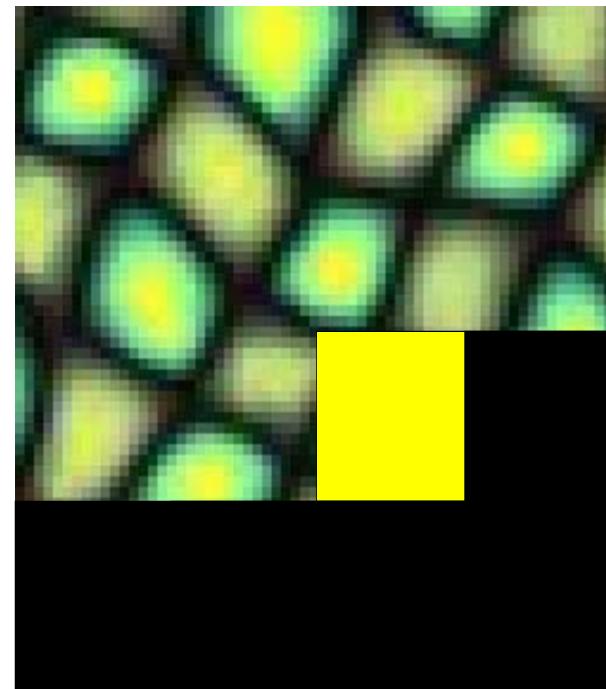
Output

Step n in Algorithm

Patch-Based Texture Synthesis



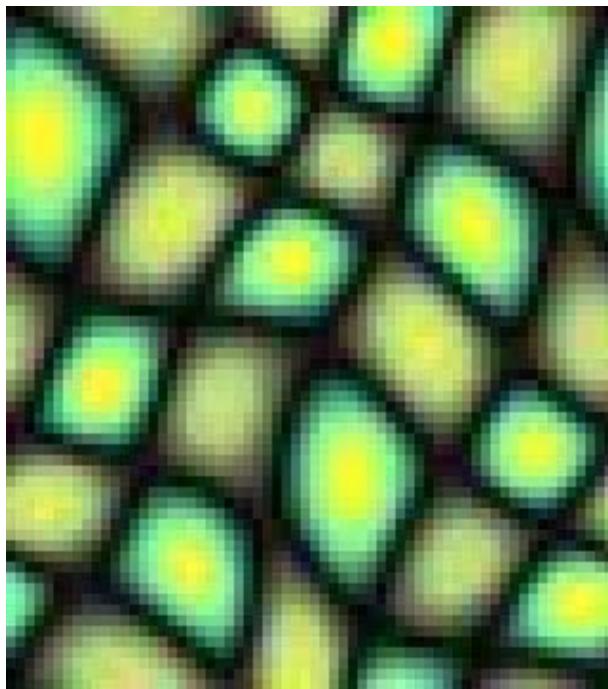
Input



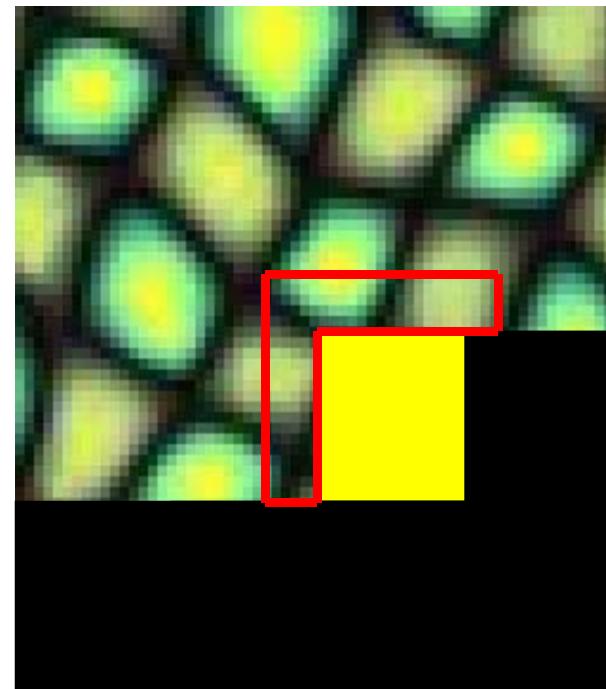
Output

Patch to be Synthesized in this Step

Patch-Based Texture Synthesis



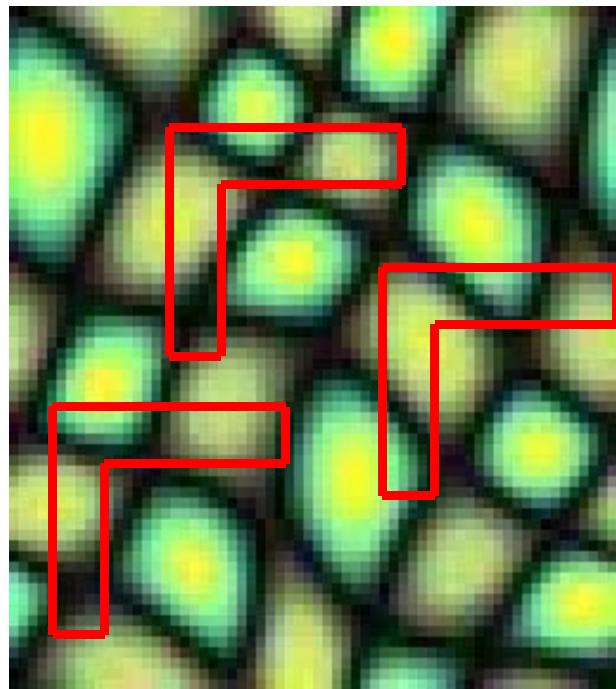
Input



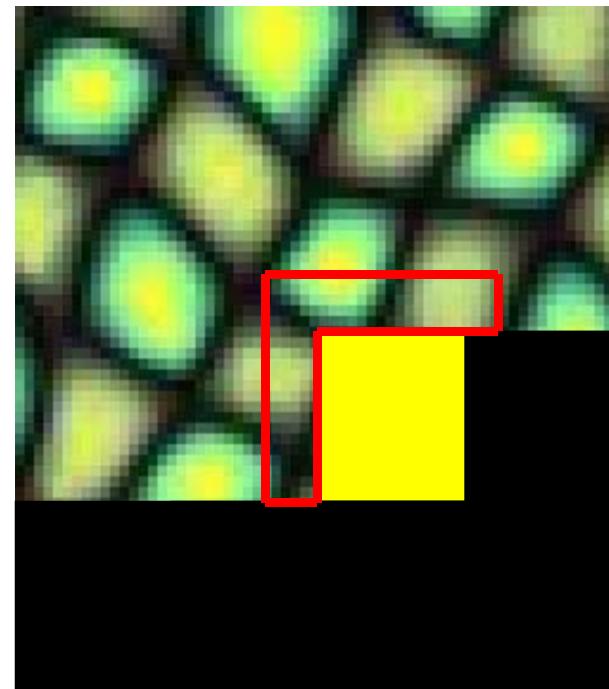
Output

Build Neighborhood (Image Mask)

Patch-Based Texture Synthesis



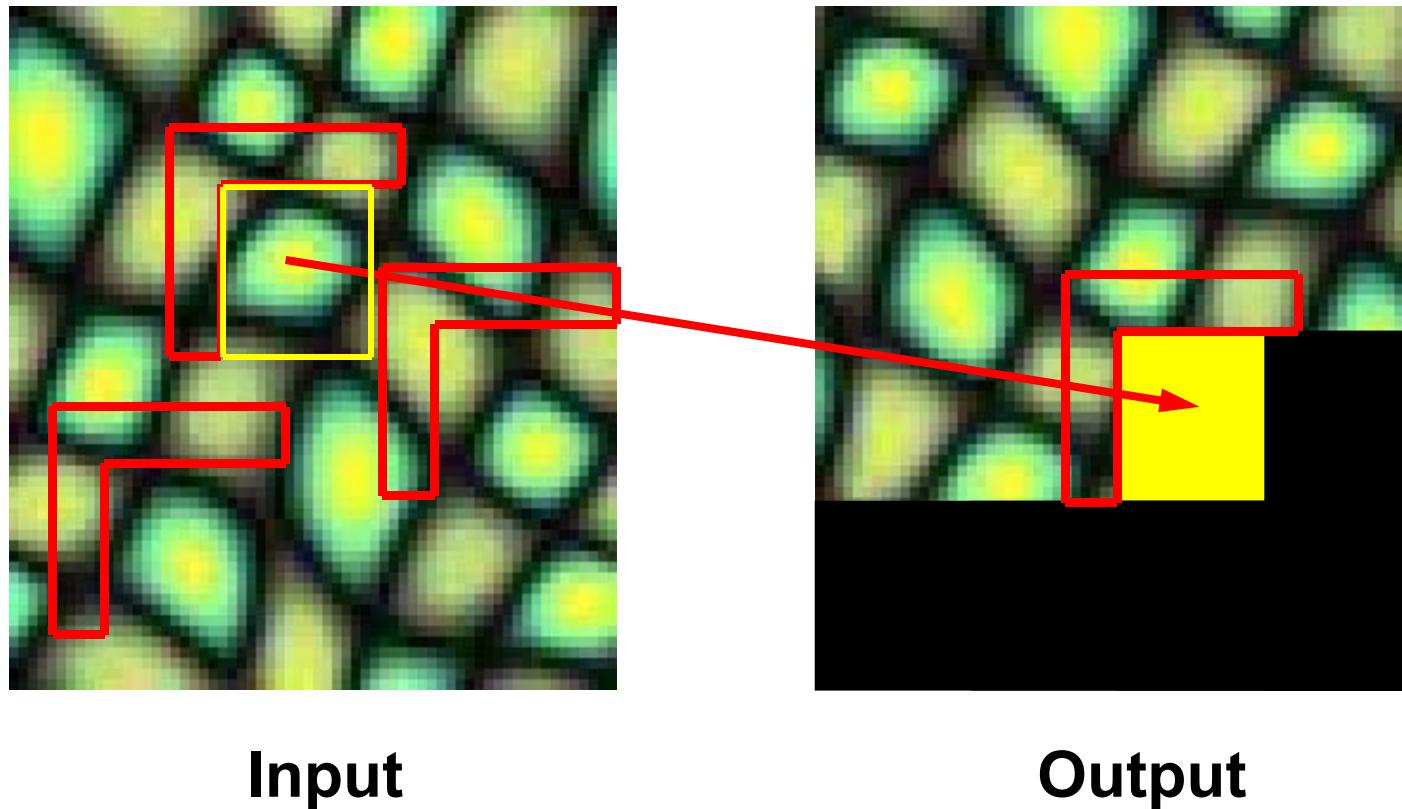
Input



Output

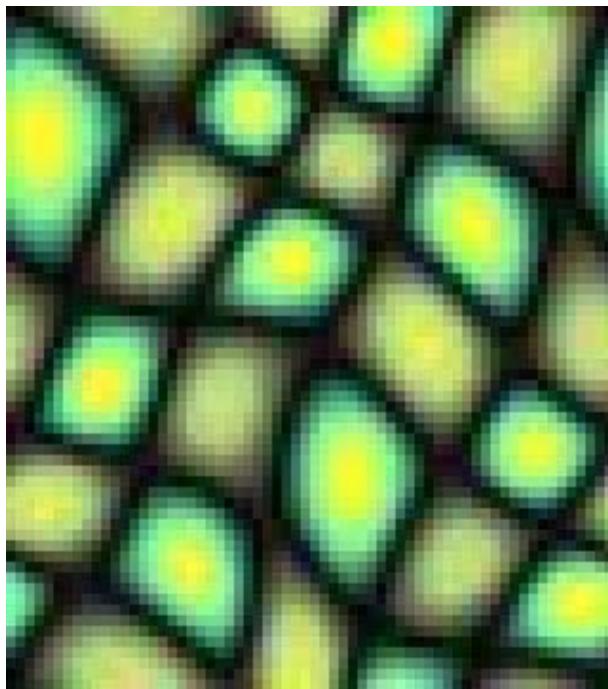
Find Matching Neighborhood(s) in Input

Patch-Based Texture Synthesis

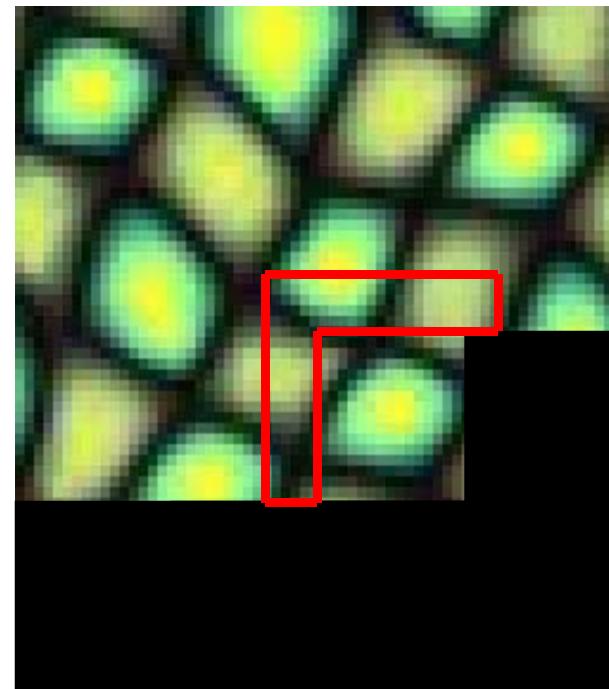


Select *Best* Neighborhood from all Candidates

Patch-Based Texture Synthesis



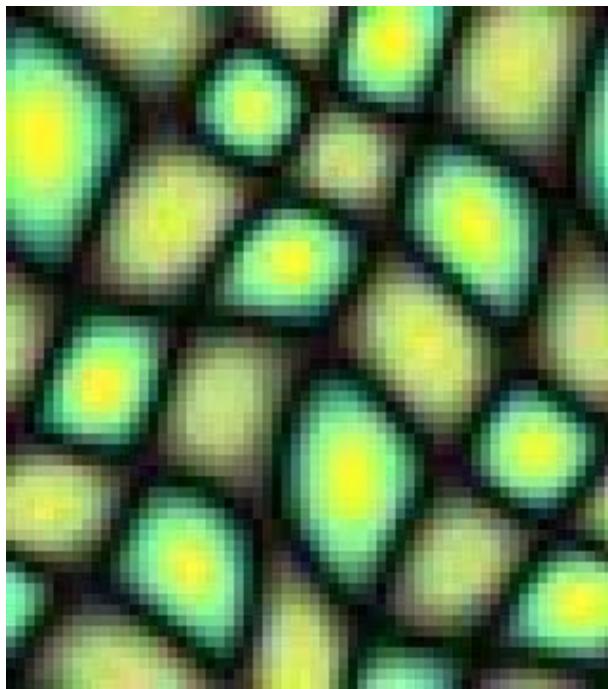
Input



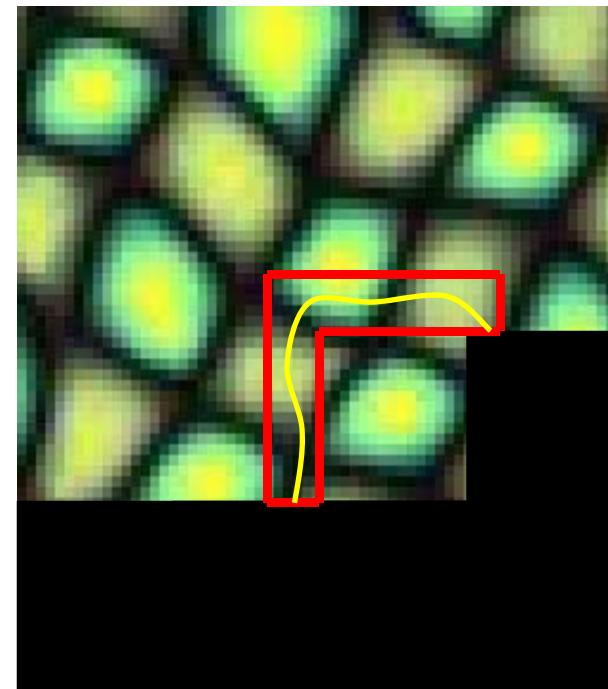
Output

Copy Selected Patch to Output

Patch-Based Texture Synthesis



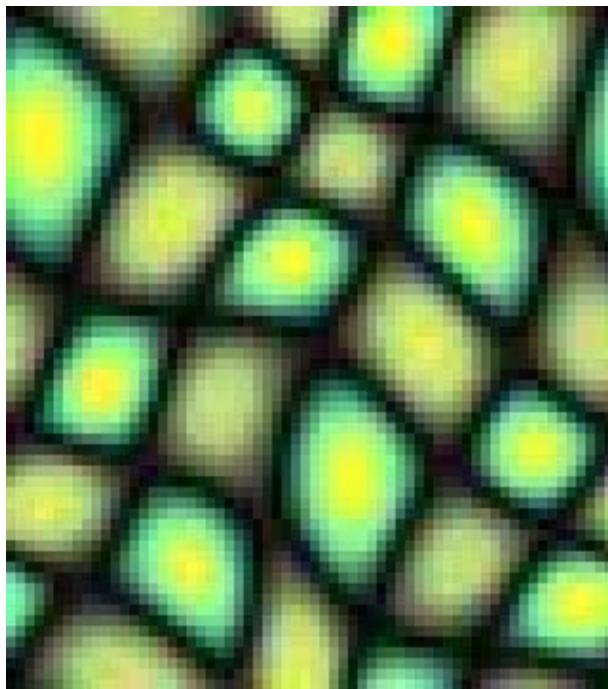
Input



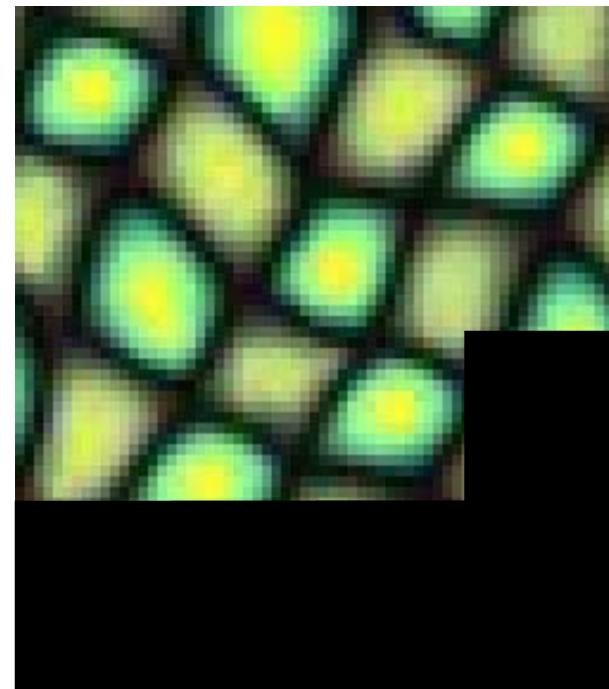
Output

Apply some *Overlap Repair Strategy* ...

Patch-Based Texture Synthesis



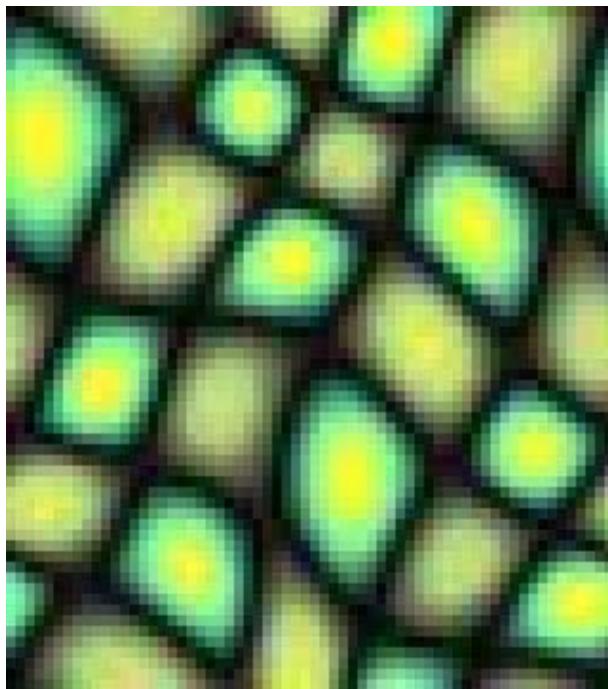
Input



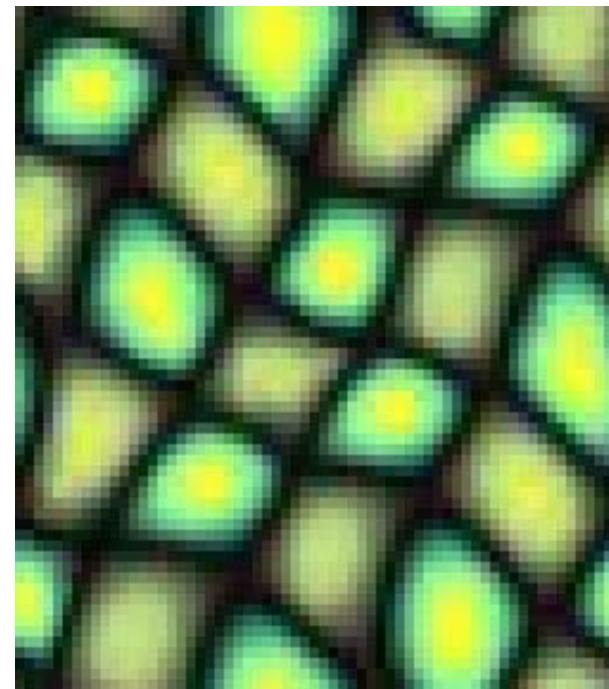
Output

... And we are done with this patch

Patch-Based Texture Synthesis



Input



Output

Repeat for all Patches in Output

Synthesis Result

— High-quality Texture



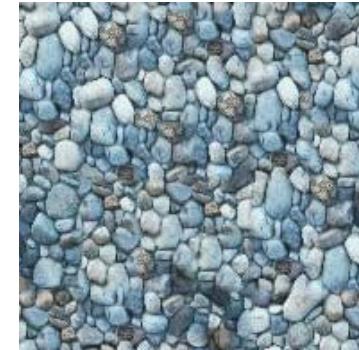
Input sample



Efros and Leung



Wei and levoy



Patch-based Sampling



Input sample



Efros and Leung



Wei and levoy



Patch-based Sampling

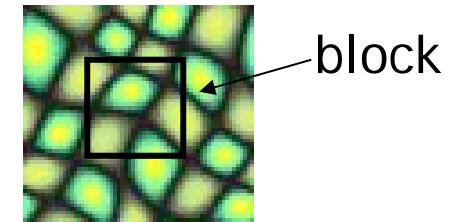
- Efros' algorithm has a tendency to grow garbage and Wei's TSVQ acceleration further aggravates this problem. In Contrast, patch-based sampling avoids growing garbage

Advantages

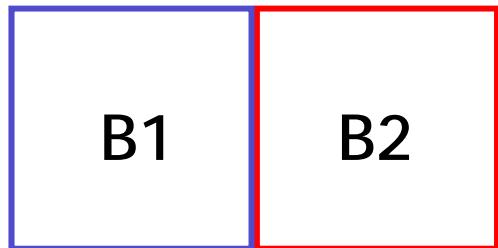
- Speed
 - Orders of magnitude faster than existing texture synthesis algorithm
 - Real-time synthesis
- Quality
 - Synthesize high-quality textures ranging from regular to stochastic
 - Avoid growing garbage
 - Synthesize subjectively better natural textures

Mincut: Graph-cut based

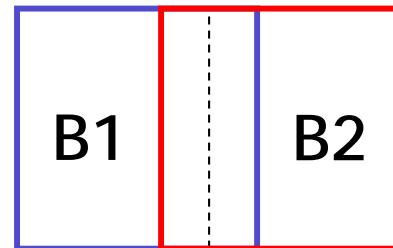
[Efros&Freeman, Siggraph 2002]



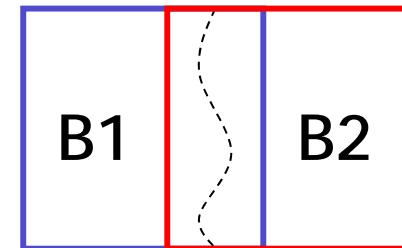
Input texture



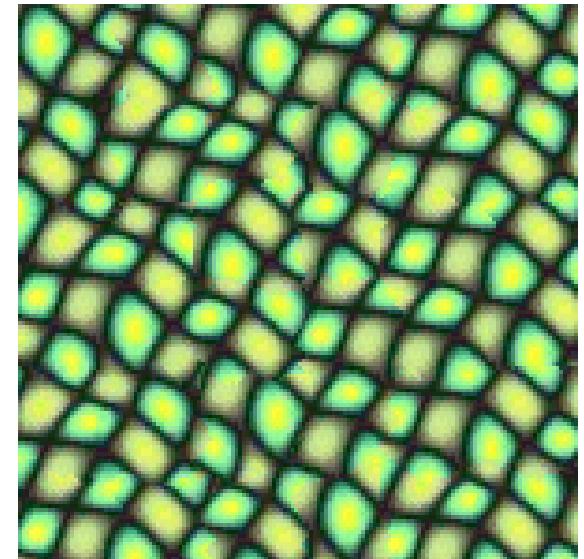
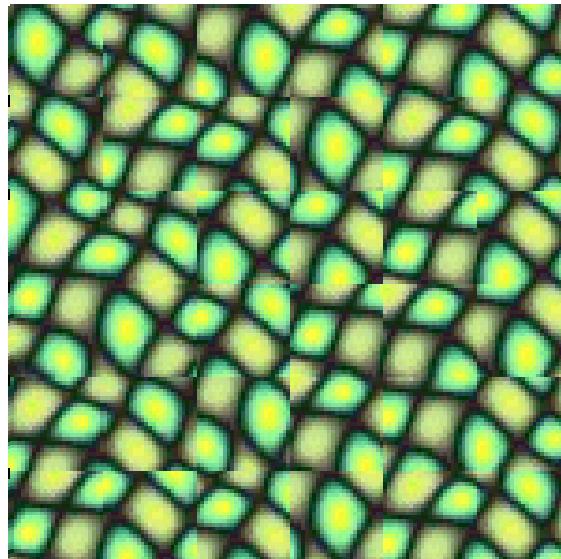
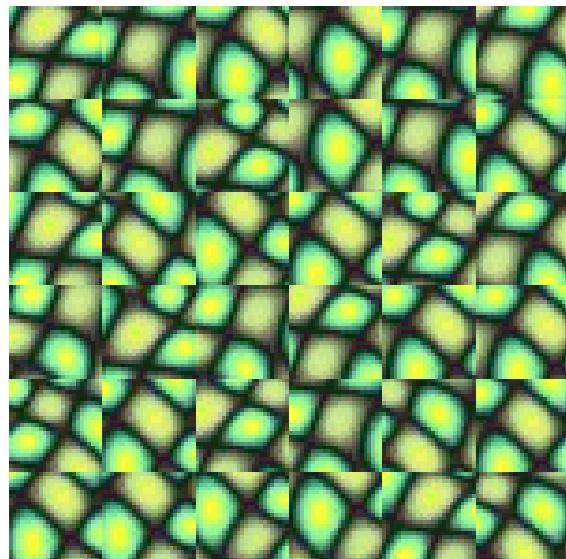
Random placement
of blocks



Neighboring blocks
constrained by overlap

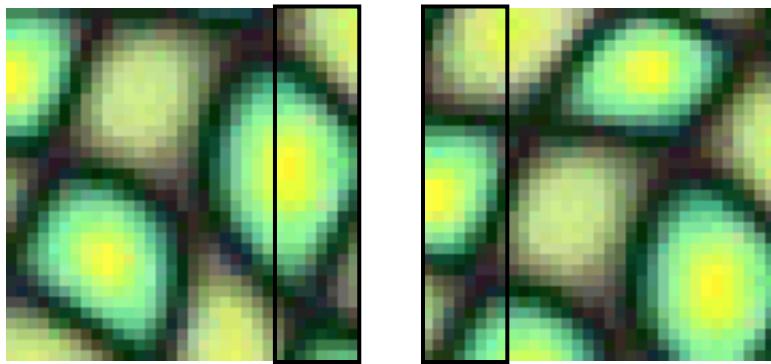


Minimal error
boundary cut

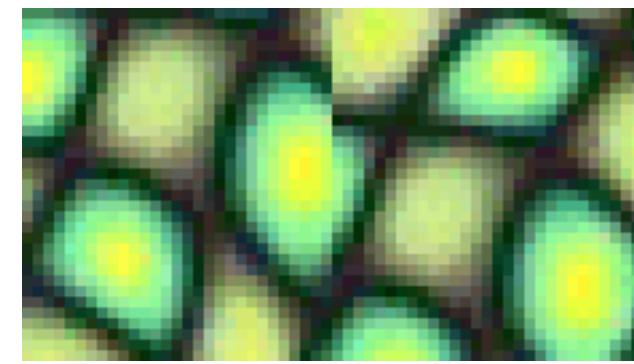


Minimal error boundary

overlapping blocks

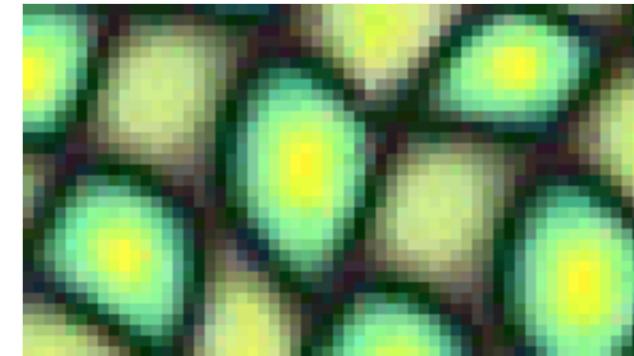


vertical boundary



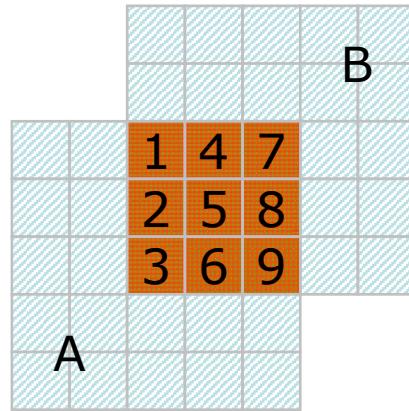
$$\left[\begin{array}{c} \text{block 1} \\ - \\ \text{block 2} \end{array} \right]^2 = \text{overlap error}$$

overlap error



min. error boundary

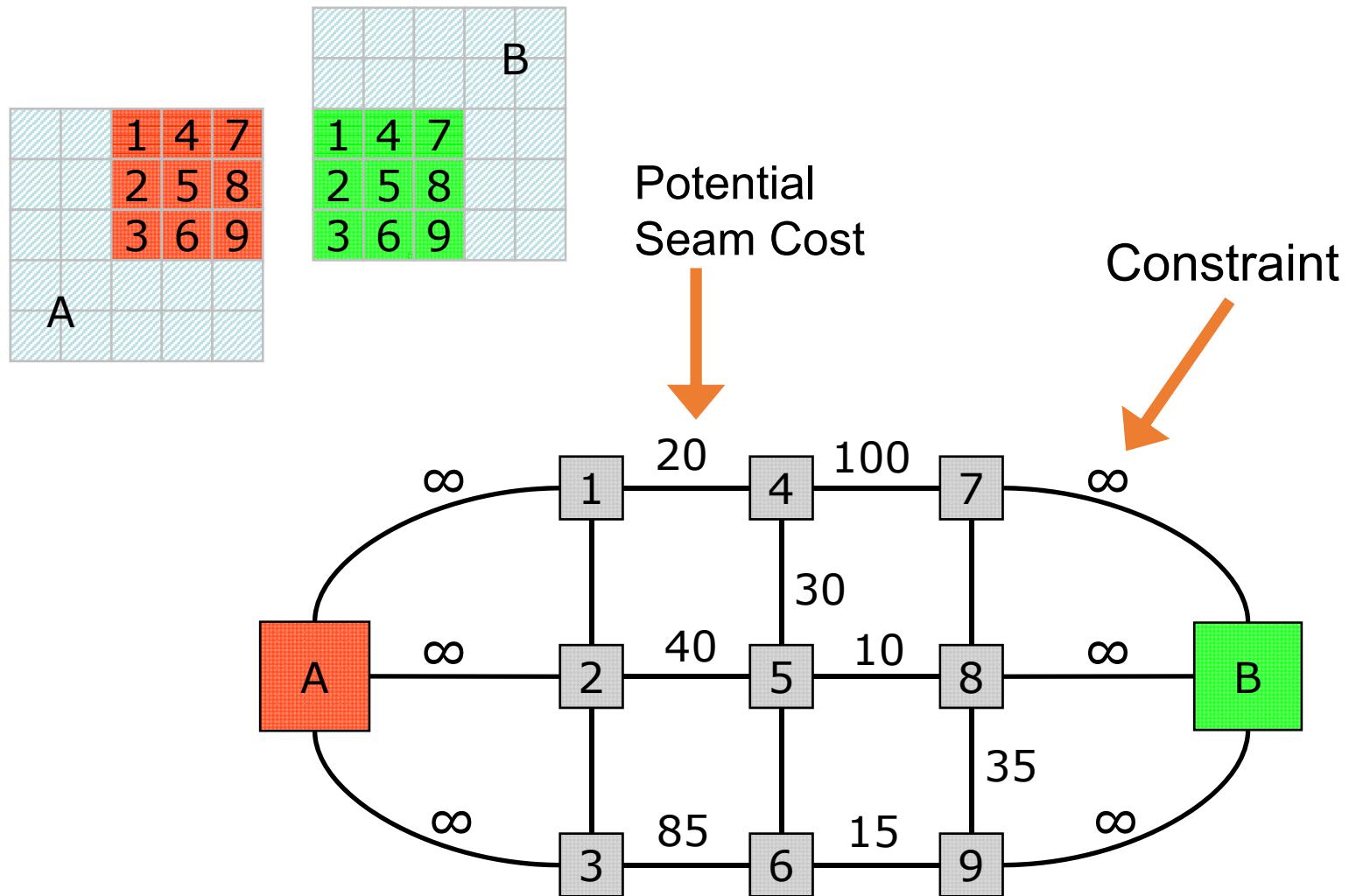
Seam Optimization



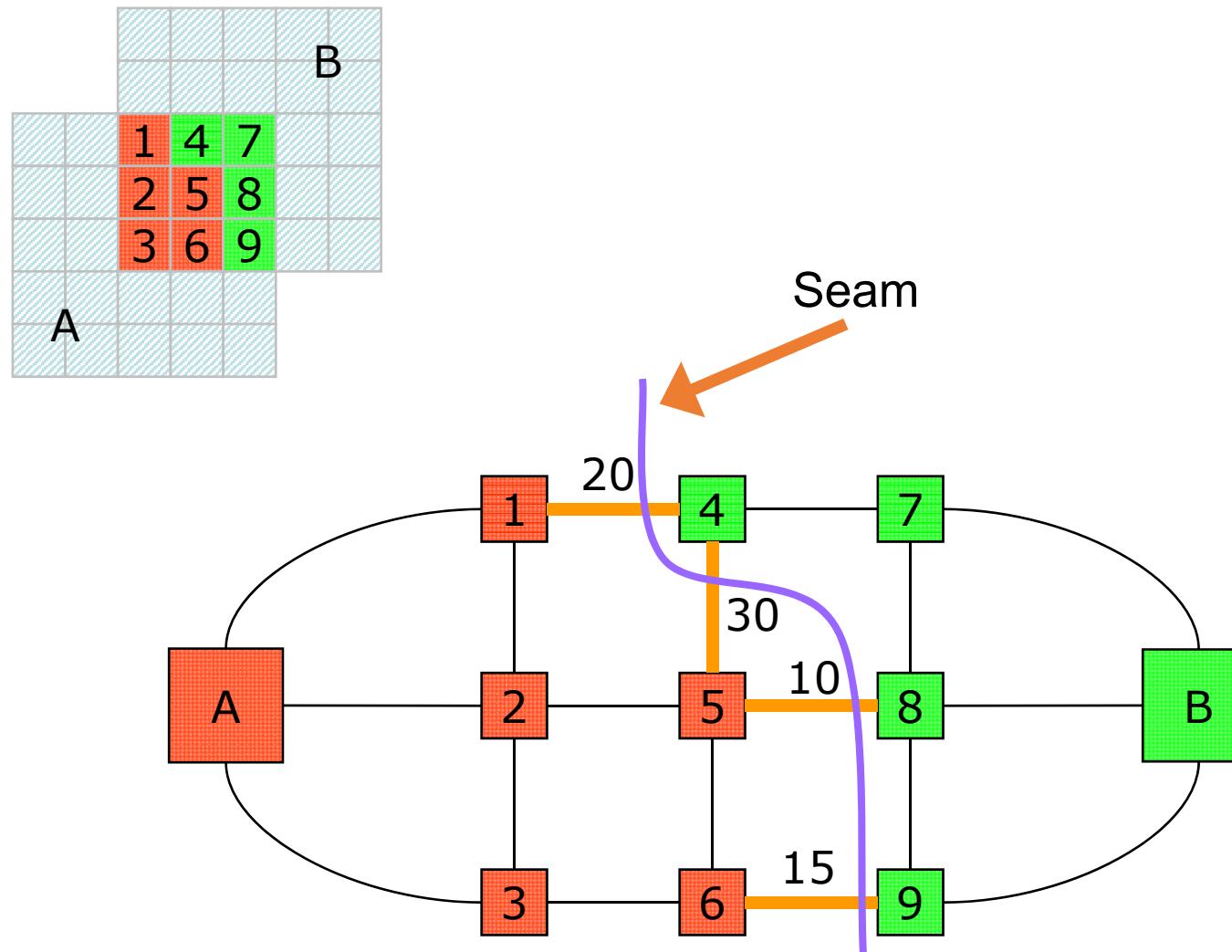
Construct graph such that:

Graph MinCut \Leftrightarrow Best Seam

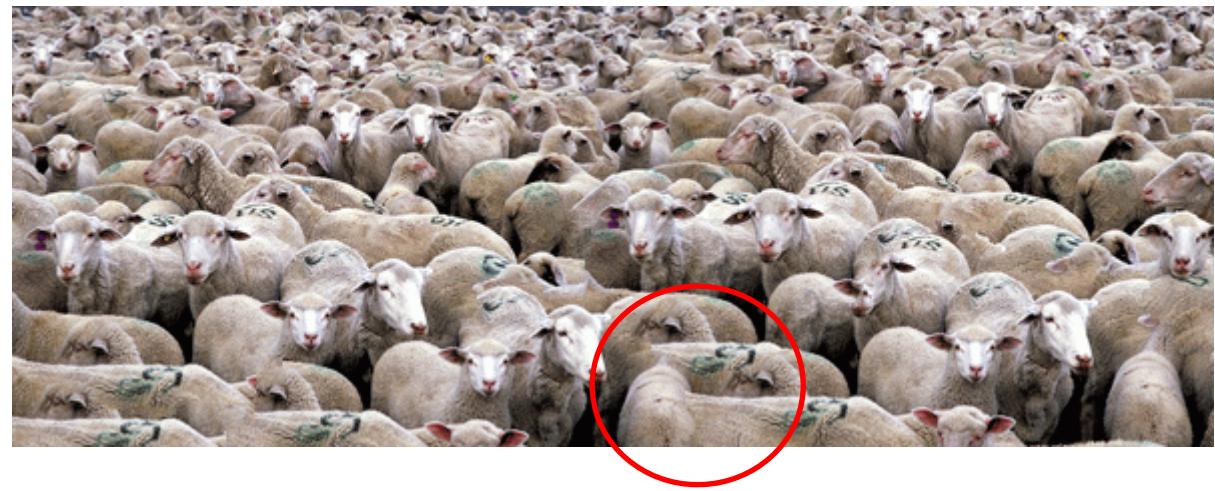
Seam Optimization



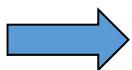
Seam Optimization



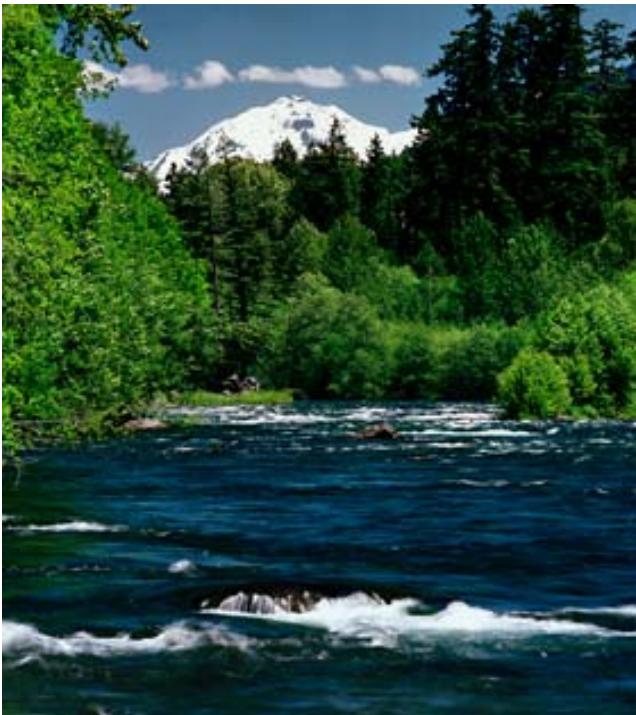
Results: Natural Images



Interactive Merging



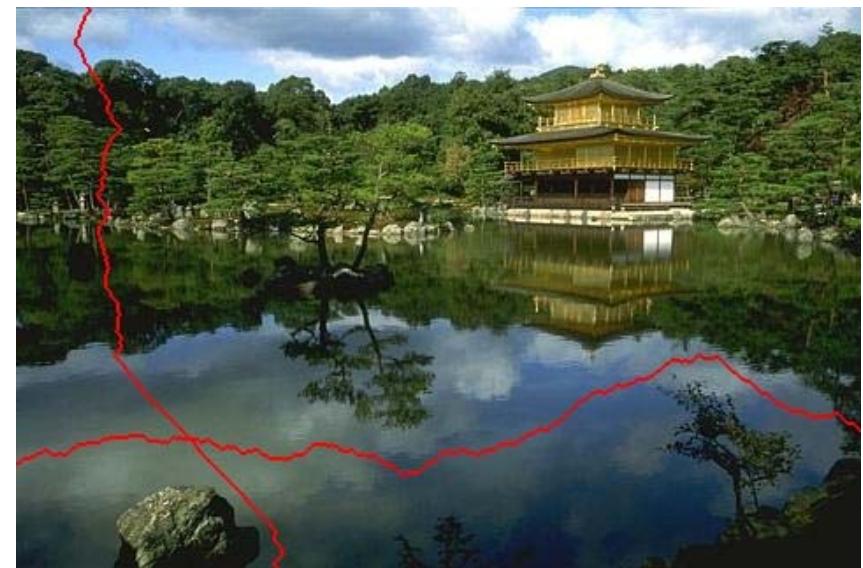
Results: Interactive Merging



Application: Seam Carving

[Avidan and Shamir, Siggraph 2007]

- Removing/adding unimportant seams
- Content-aware image resizing/retargetting



Seam Carving



Seam carving



Scaling



Cropping

Texton-based Synthesis

[Zhang et al., Siggraph 2003]

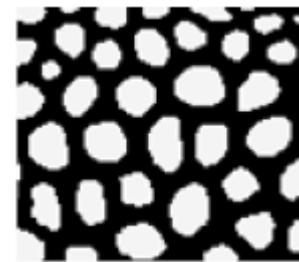
- Texton: texture element
- Texture elements don't break apart using texton synthesis



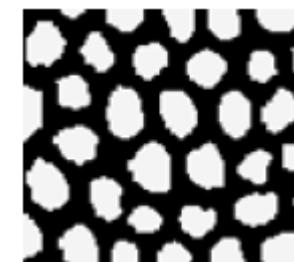
Texture sample



Synthesized texture



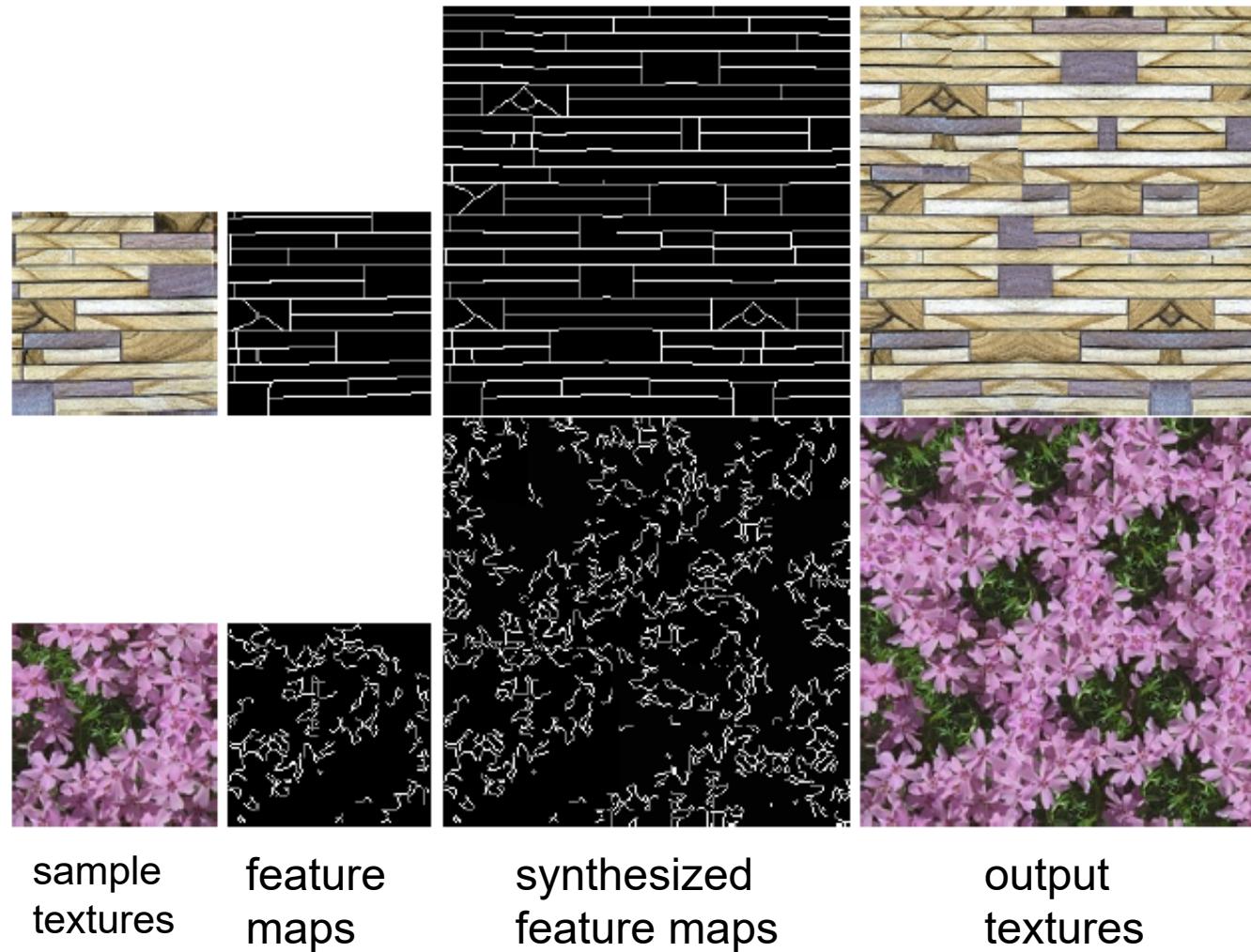
Texton mask



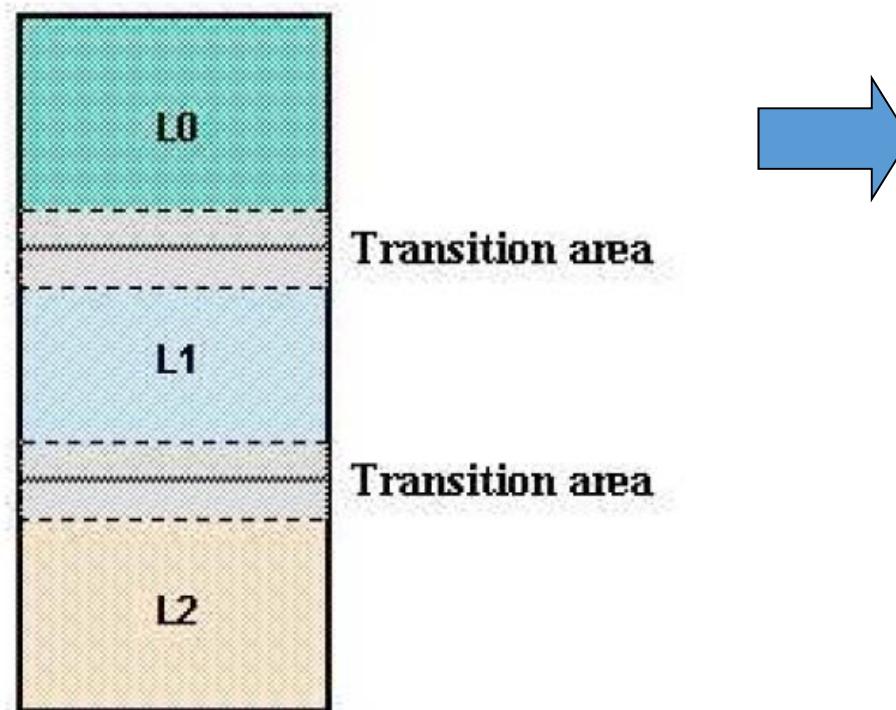
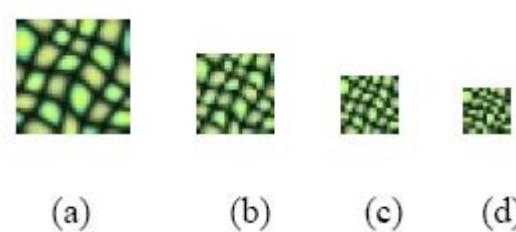
Synthesized texton

Feature Map + Texture Map

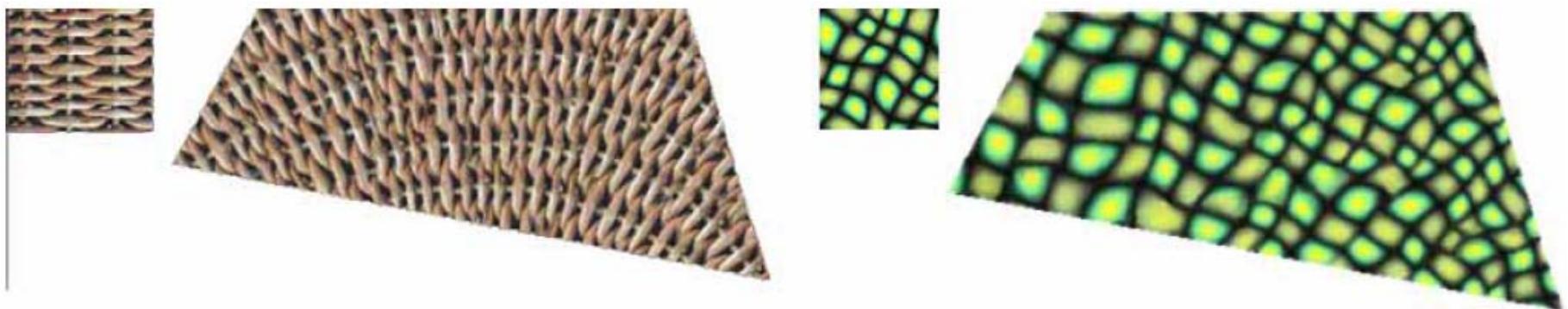
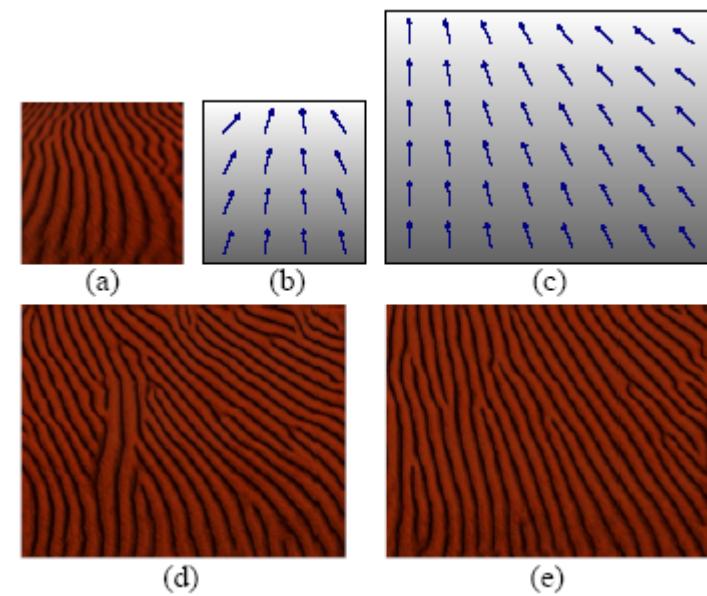
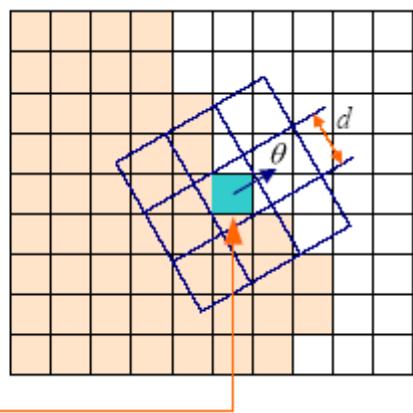
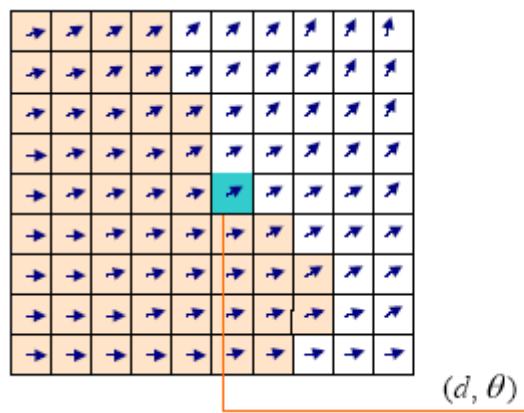
[Wu&Yu, Siggraph 2004]



Synthesis with Local Size Control

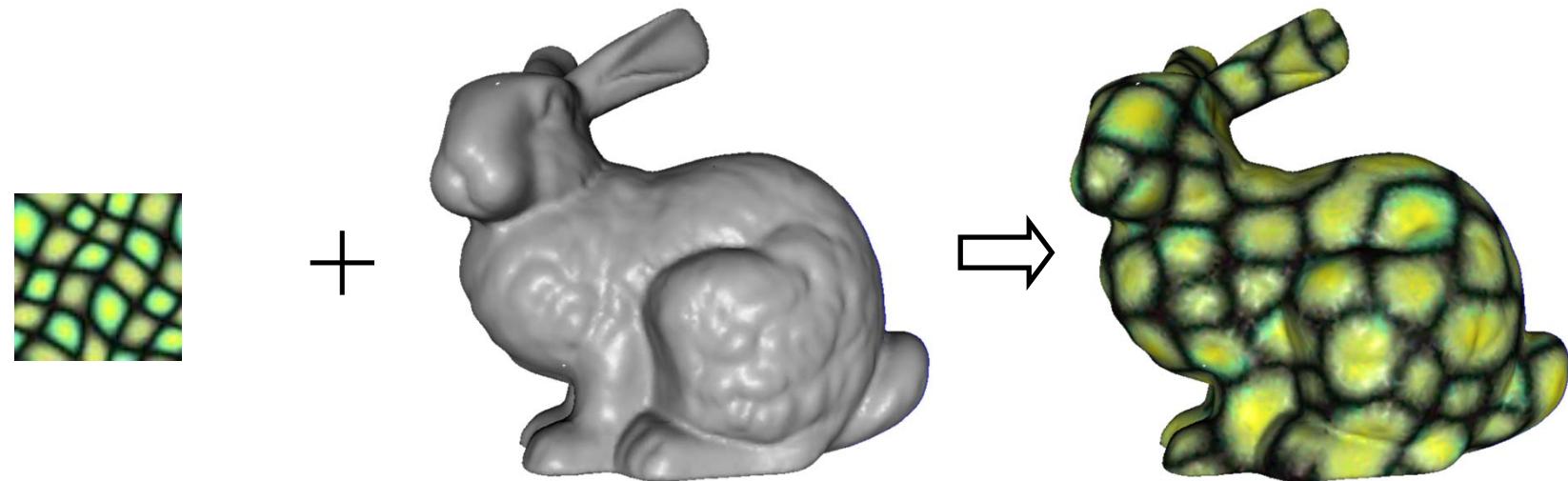


Synthesis with Vector Field Control

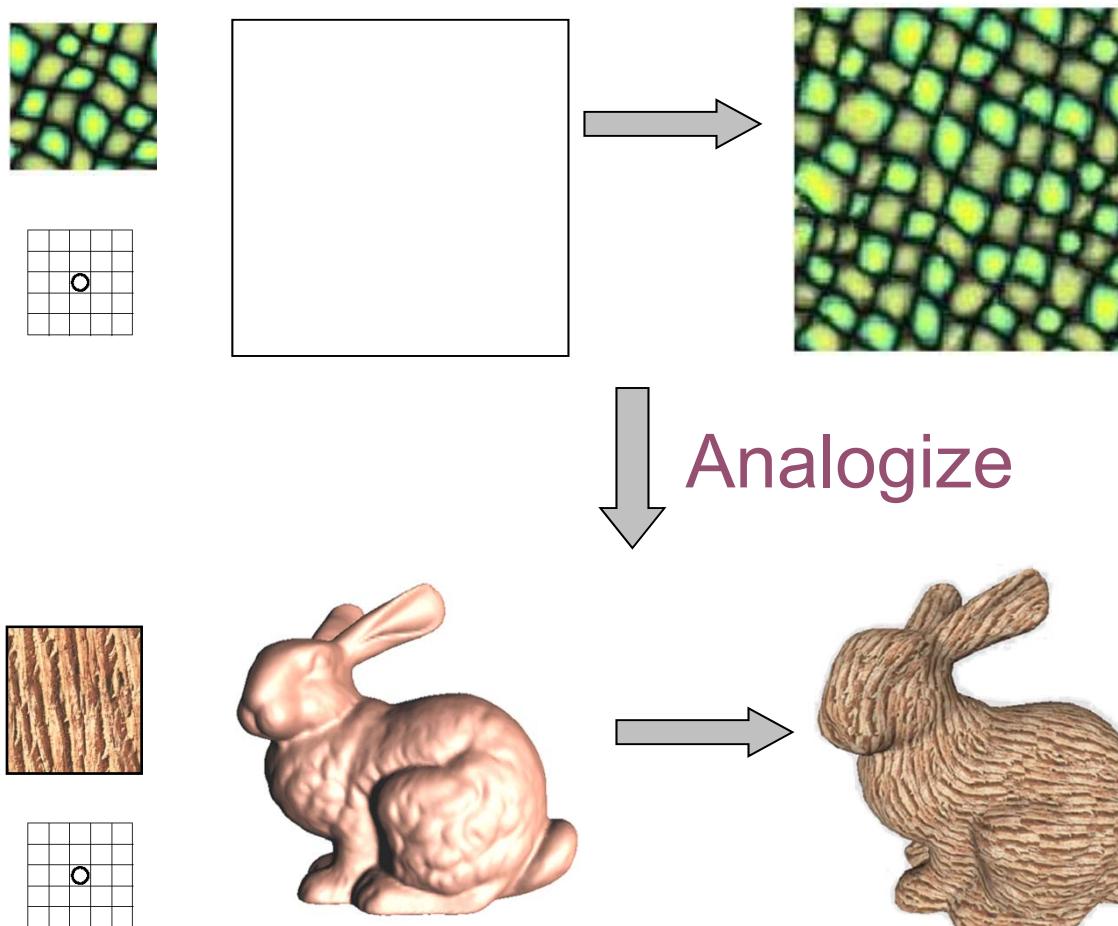


3. 3D Texture Synthesis

Texture Synthesis on Surfaces

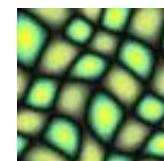


Analogy



Desired Properties

- Share advantages of 2D algorithm
 - Quality
 - Efficient
 - General
 - Easy to use
- Minimum distortion
- Minimum discontinuity



3D Texture Synthesis

- Procedural texture synthesis
- Sample-based texture synthesis
- Geometry synthesis

3D Texture Synthesis

- Procedural texture synthesis
- Sample-based texture synthesis
- Geometry synthesis

Procedural Textures

- Use 3D functions to define texture properties of objects
 - Non-trivial programming
 - Flexibility
 - Parametric control
 - Unlimited resolution, antialiasing possible
 - Low memory requirements
 - Low-cost visual complexity
 - Adapts to arbitrary geometry



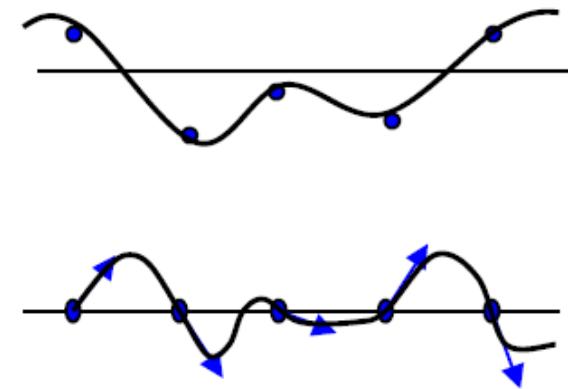
Procedural Textures

- **Analytic scalar function of world coordinates (x,y,z)**
- **Texturing: evaluation of function on object surface**
 - Ray tracing: 3D intersection point with surface
- **Textures of natural objects**
 - Similarity between different patches
 - Repetitiveness, coherence
 - Similarity on different resolution scales
 - Self-similarity
 - But never completely identical
 - Additional disturbances, turbulence, noise
- **Procedural texture function**
 - Mimics statistical properties of natural textures
 - Purely empirical approach
 - Looks convincing, but has nothing to do with material's physics

Perlin's Noise

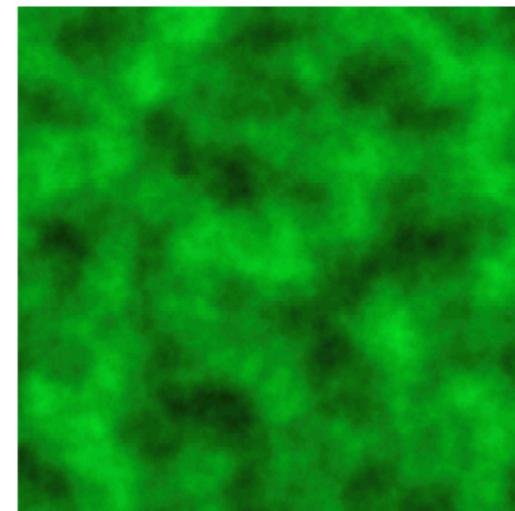
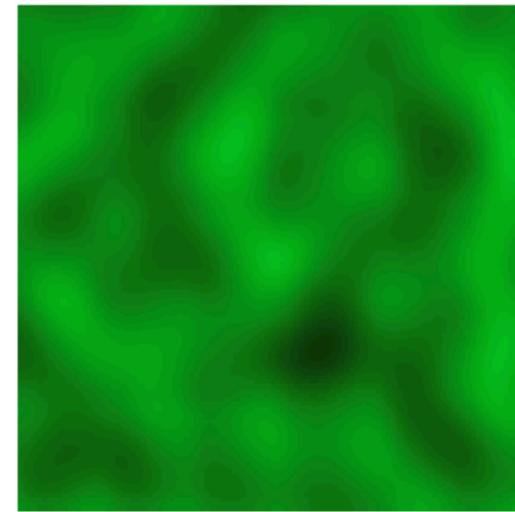
[Perlin, Siggraph 1985]

- **Noise(x,y,z)**
 - Statistical invariance under rotation
 - Statistical invariance under translation
 - Narrow bandpass limit in frequency
- **Integer lattice (i,j,k)**
 - Random number at each lattice point (i,j,k)
 - Look-up table or hashing function
 - Gradient lattice noise
 - Random gradient vectors
- **Evaluation at (x,y,z)**
 - Tri-linear interpolation
 - Cubic interpolation (Hermite spline → later)
- **Unlimited domain**
 - Lattice replicated to fill entire space
- **Fixed fundamental frequency of ~1 Hz over lattice**
- **Smooth interpolation of interim values**



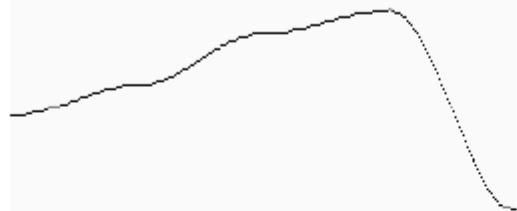
Turbulence Function

- **Noise function**
 - “White” frequency spectrum
- **Natural textures**
 - Decreasing power spectrum towards high frequencies
- **Turbulence from noise**
 - $\text{Turbulence}(x) = \sum_{i=0}^k \text{abs}(\text{noise}(2^i x) / 2^i)$
 - Summation truncation
 - $1/2^{k+1} < \text{size of one pixel}$ (band limit)
 - 1. Term: $\text{noise}(x)$
 - 2. Term: $\text{noise}(2x)/2$
 - ...
 - Power spectrum: $1/f$
 - (Brownian motion: $1/f^2$)

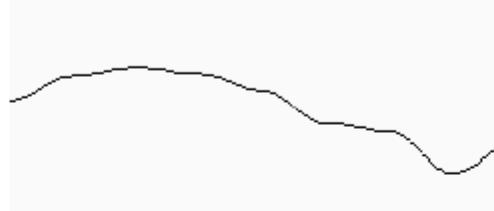


Synthesis of Turbulence (1D)

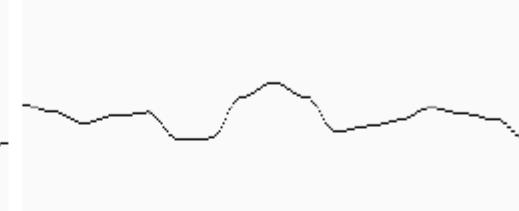
Amplitude : 128
frequency : 4



Amplitude : 64
frequency : 8



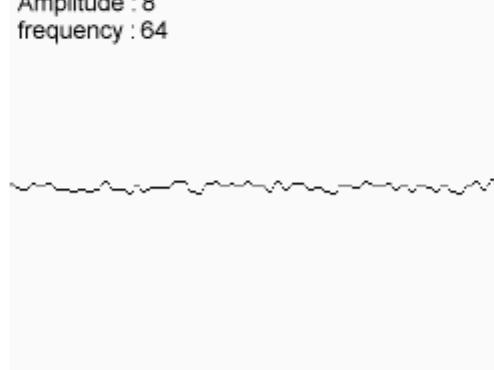
Amplitude : 32
frequency : 16



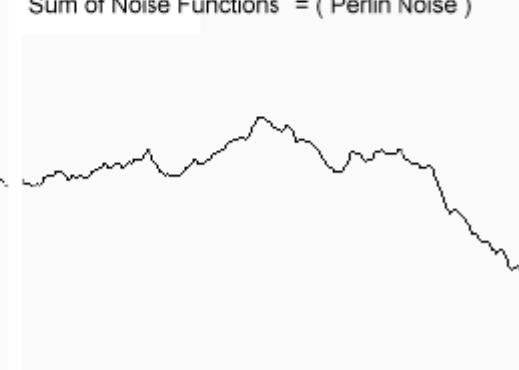
Amplitude : 16
frequency : 32



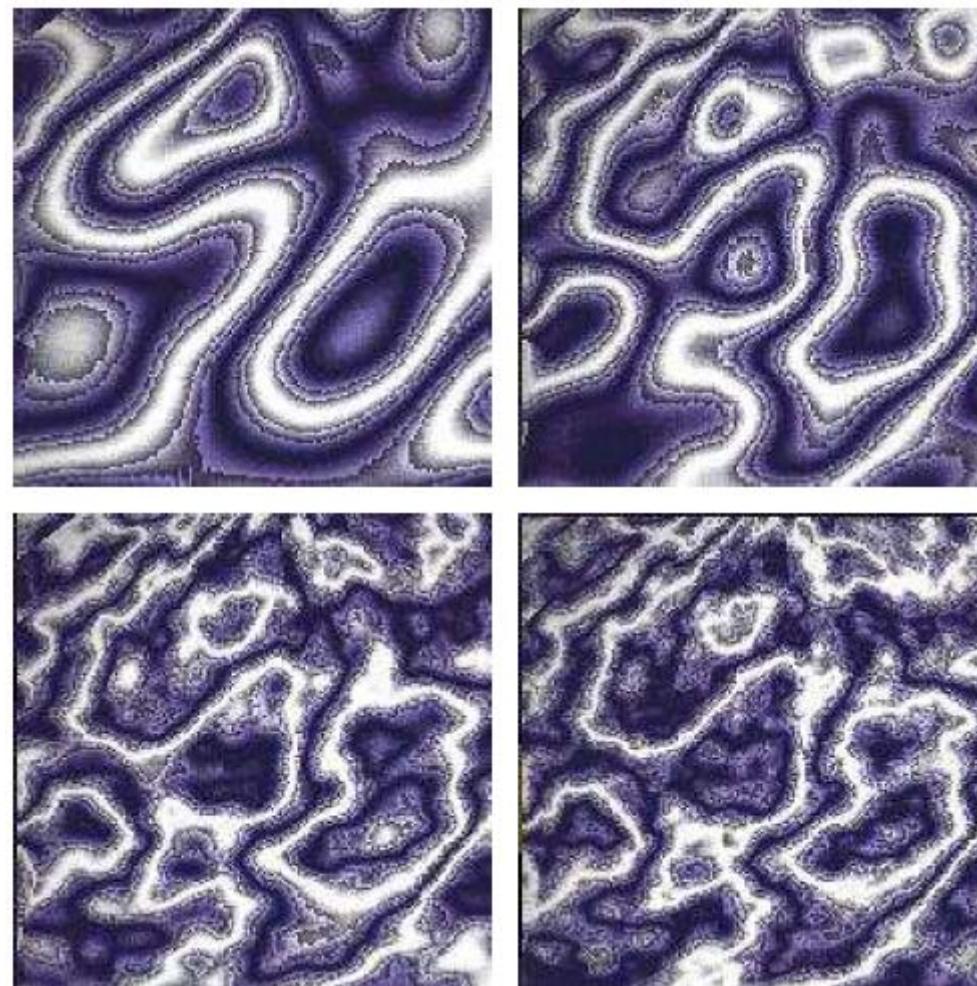
Amplitude : 8
frequency : 64



Sum of Noise Functions = (Perlin Noise)

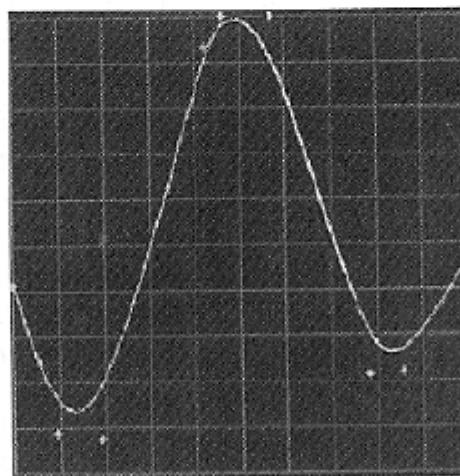
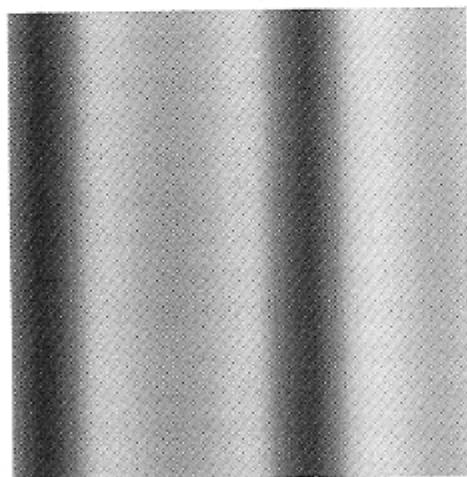


Synthesis of Turbulence (2D)



Example: Marble Texture Function

- **Overall structure: alternating layers of white and colored marble**
 - $f_{\text{marble}}(x,y,z) := \text{marble_color}(\sin(x))$
 - `marble_color` : transfer function
- **Realistic appearance: simulated turbulence**
 - $f_{\text{marble}}(x,y,z) := \text{marble_color}(\sin(x + \text{turbulence}(x,y,z)))$
- **Moving object: turbulence function also transformed**



Reaction Diffusion Based Method

[Turk, Siggraph 1991]

- Two-chemical reaction-diffusion system

$$\frac{\partial a}{\partial t} = F(a, b) + D_a \nabla^2 a$$

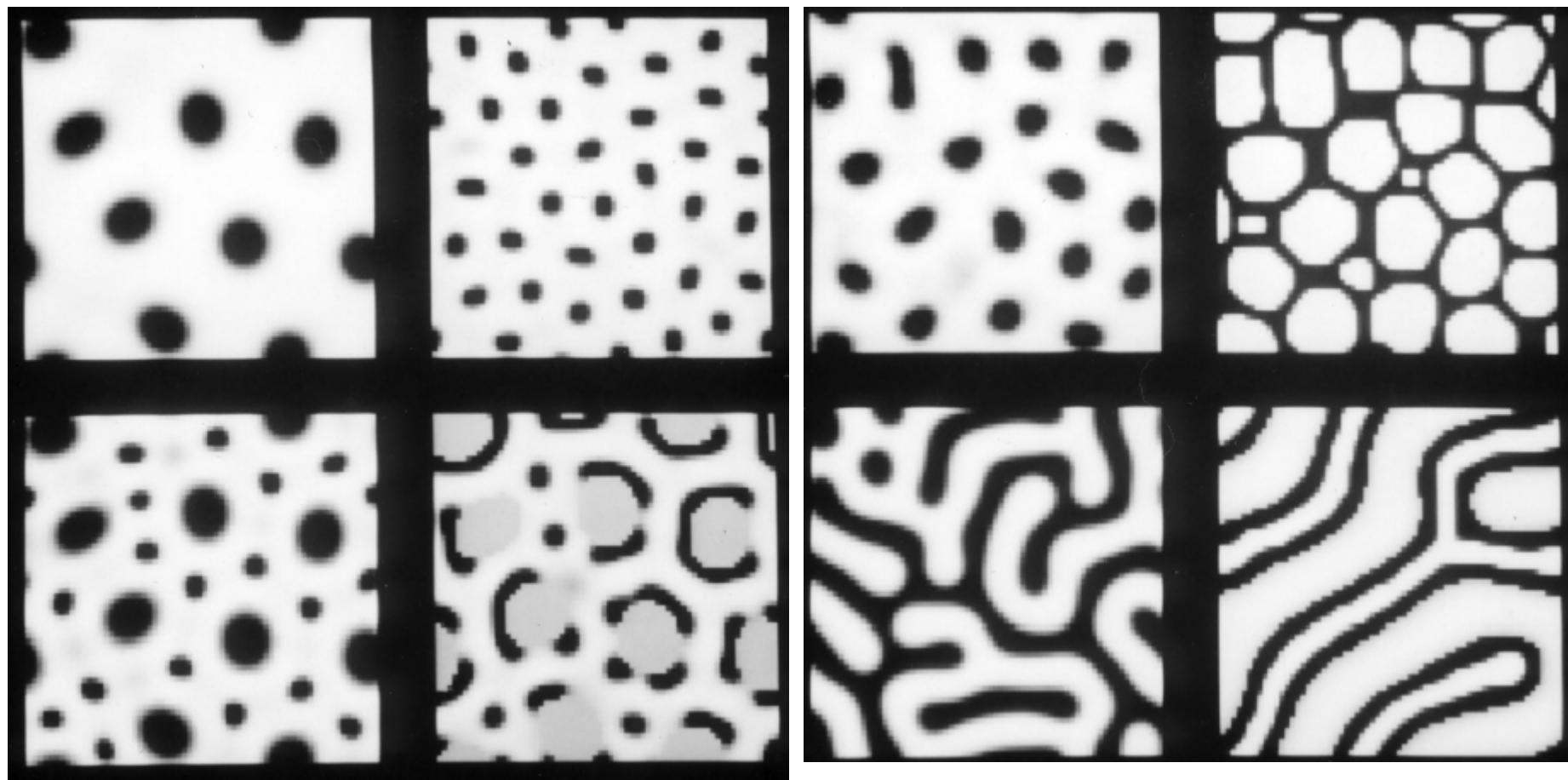
$$\frac{\partial b}{\partial t} = G(a, b) + D_b \nabla^2 b$$

- Discrete computation

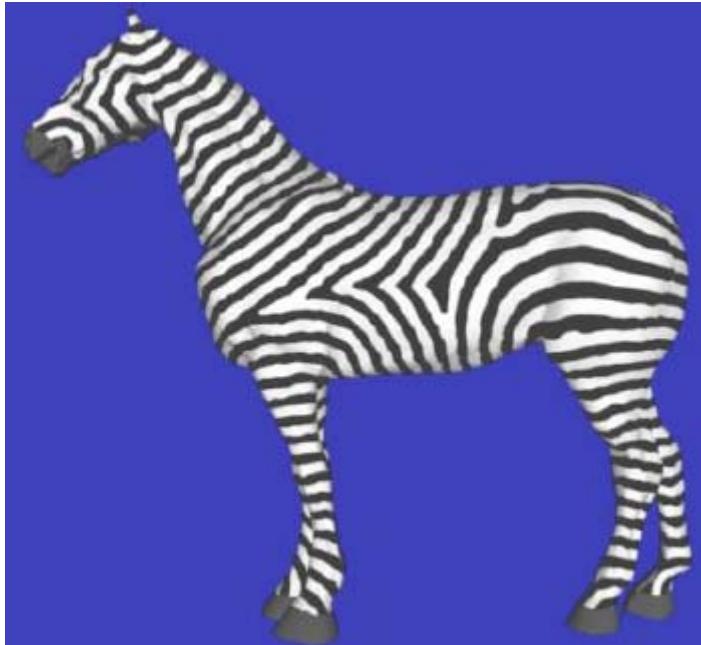
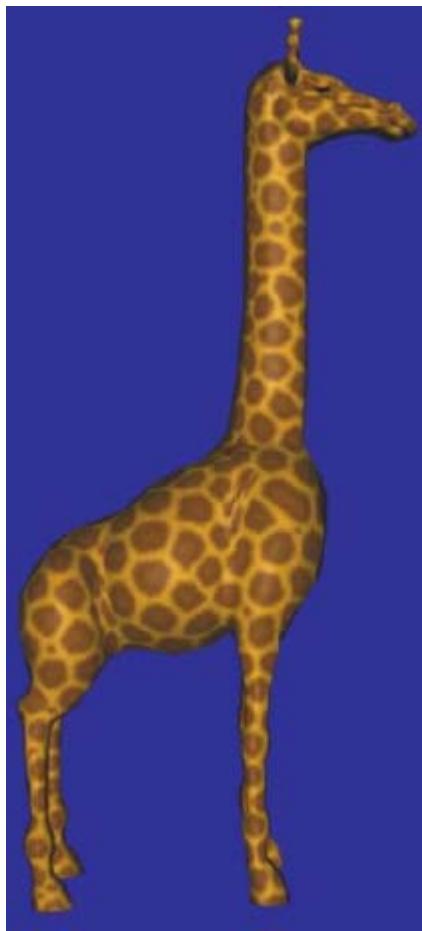
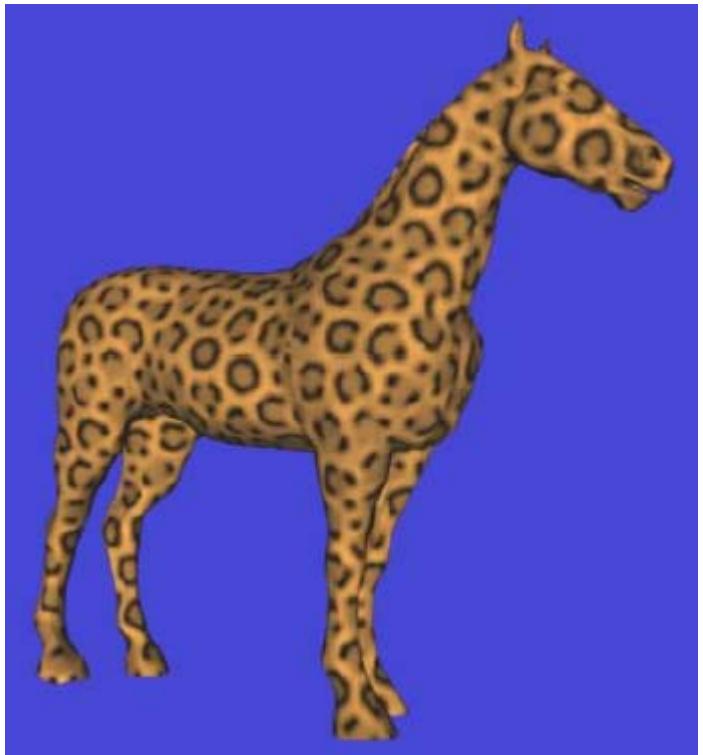
$$\Delta a_{ij} = s (16 - a_{ij} b_{ij}) + D_a (a_{i+1,j} + a_{i-1,j} + a_{i,j+1} + a_{i,j-1} - 4a_{ij})$$

$$\Delta b_{ij} = s (a_{ij} b_{ij} - b_{ij} - \beta_{ij}) + D_b (b_{i+1,j} + b_{i-1,j} + b_{i,j+1} + b_{i,j-1} - 4b_{ij})$$

Examples of Reaction-Diffusion



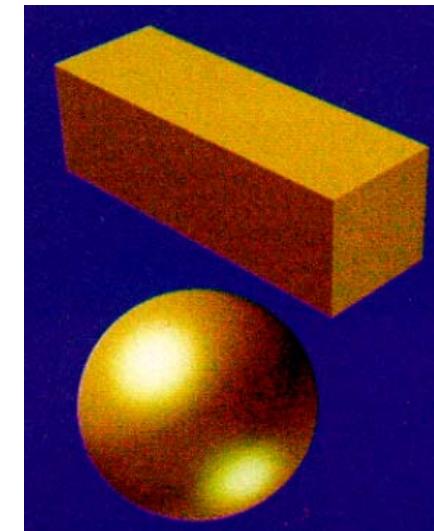
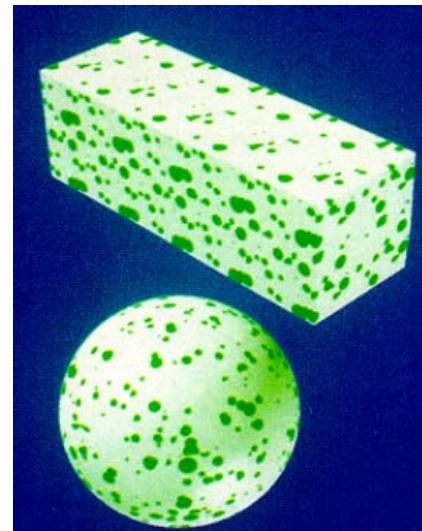
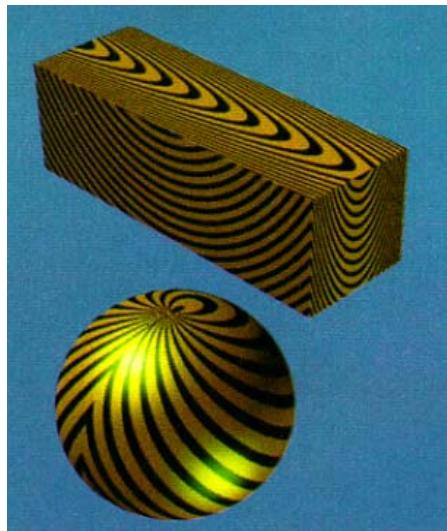
Results



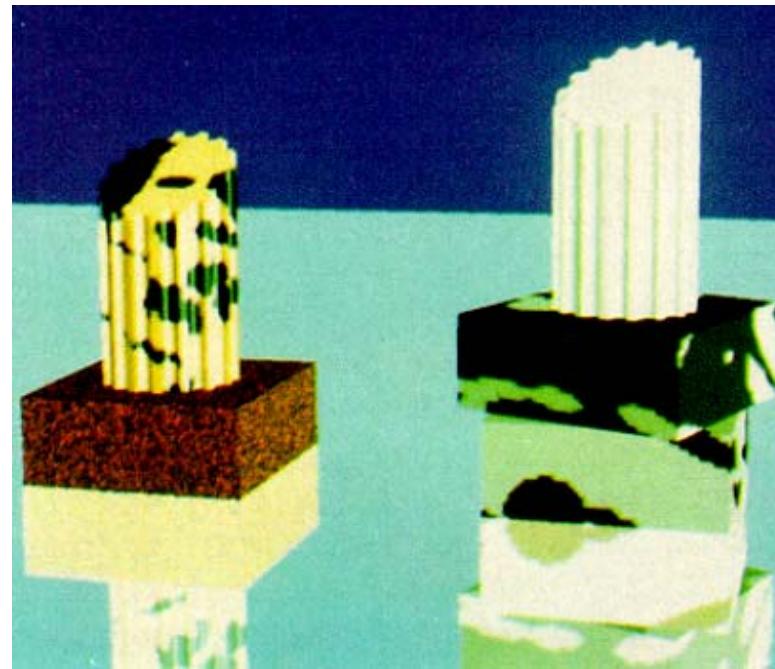
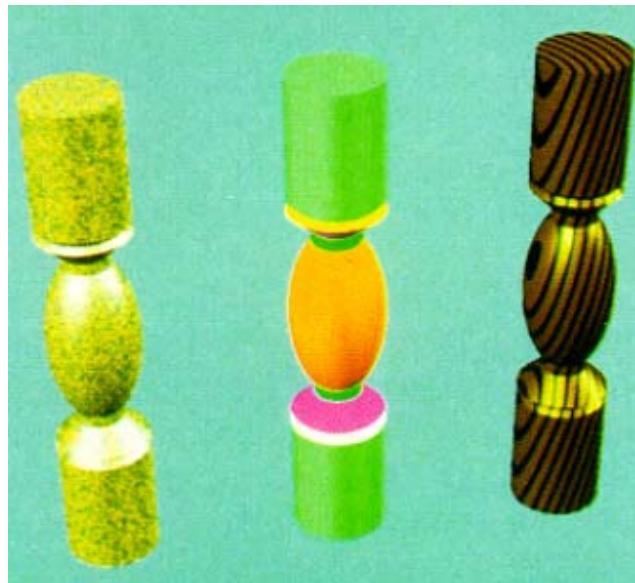
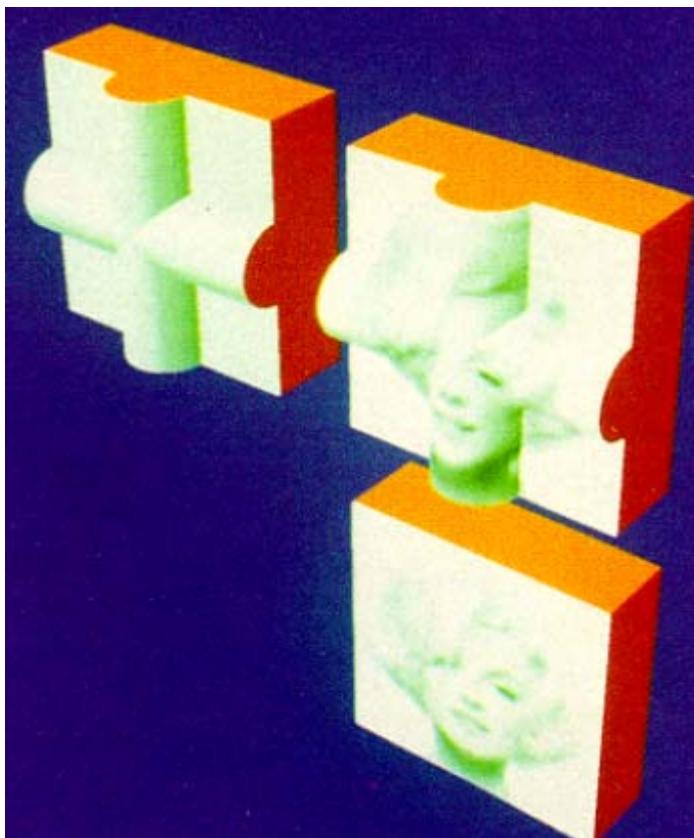
Solid Textures

[Peachey, Siggraph 1985]

- Solid texture functions in 3D space
- Nonhomogeneous materials
 - wood and stone



Examples

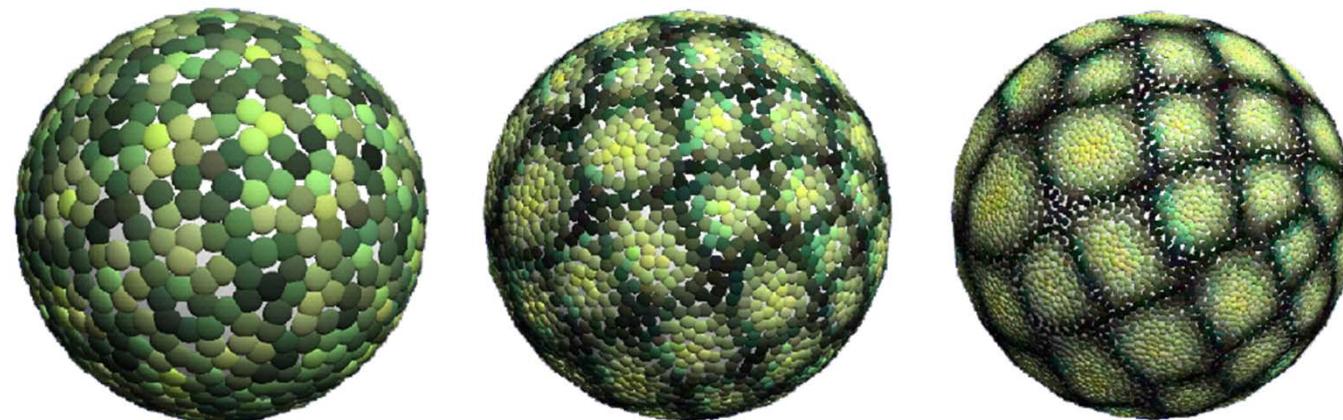


3D Texture Synthesis

- Procedural texture synthesis
- Sample-based texture synthesis
- Geometry synthesis

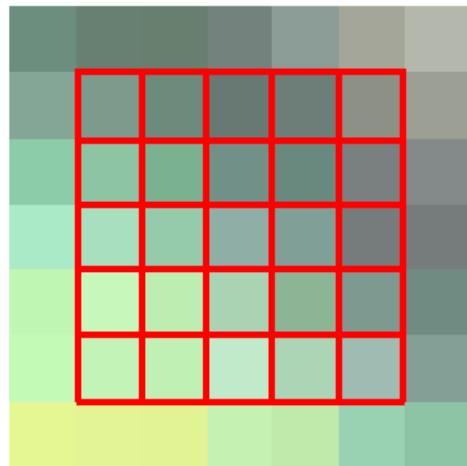
Methodology

- Synthesize a surface texture by **coloring mesh vertices**
- Extensions from 2D texture synthesis
- Key issues
 - Resampling
 - Local flattening
 - Size
 - Orientation

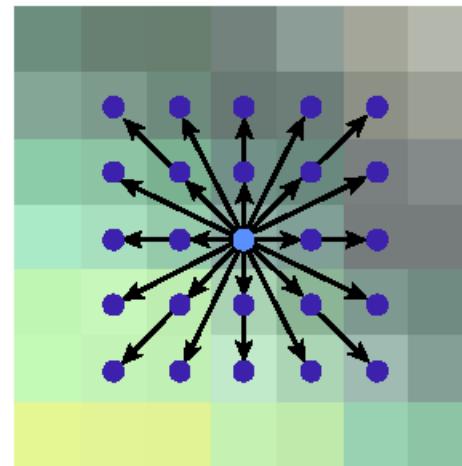


Vertex colors on three levels in the mesh hierarchy

Neighborhood Sampling on Surface

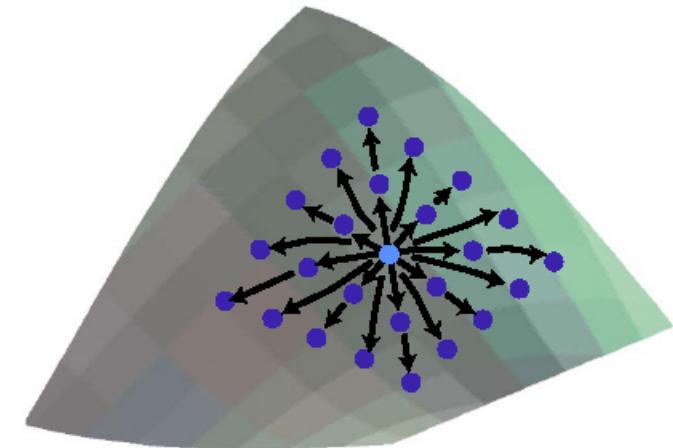


(a)



(b)

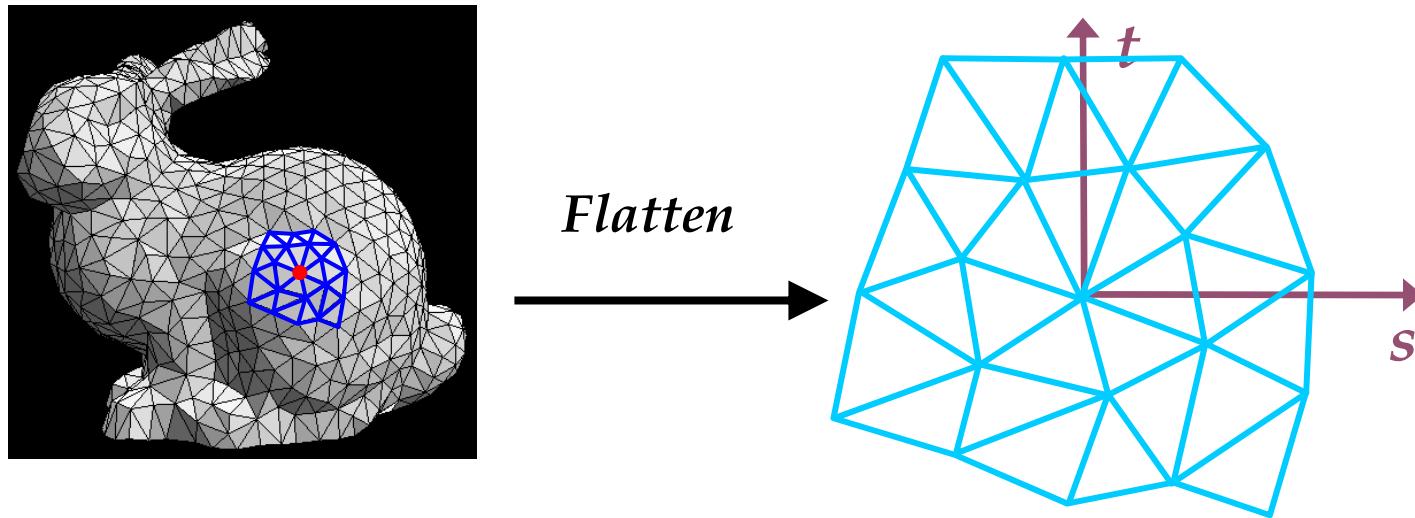
A pixel in image



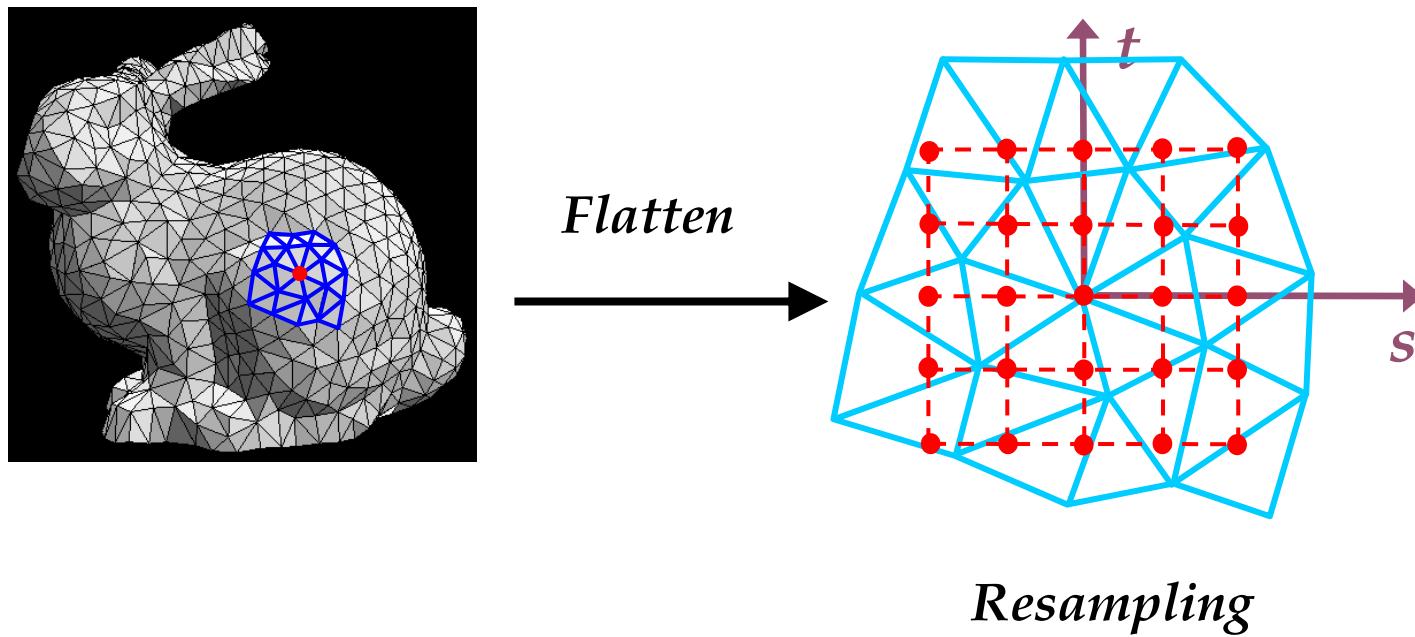
(c)

A vertex on surface

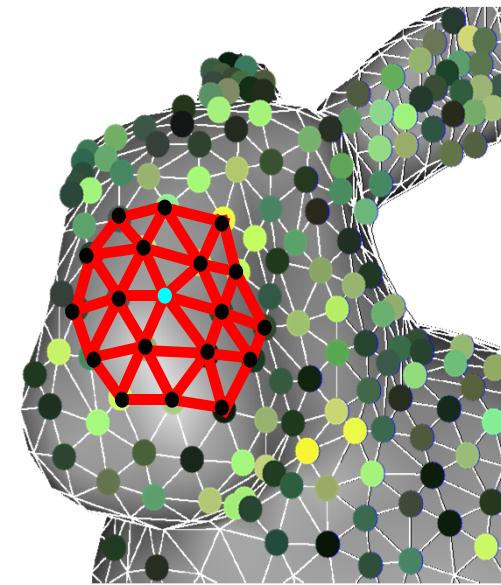
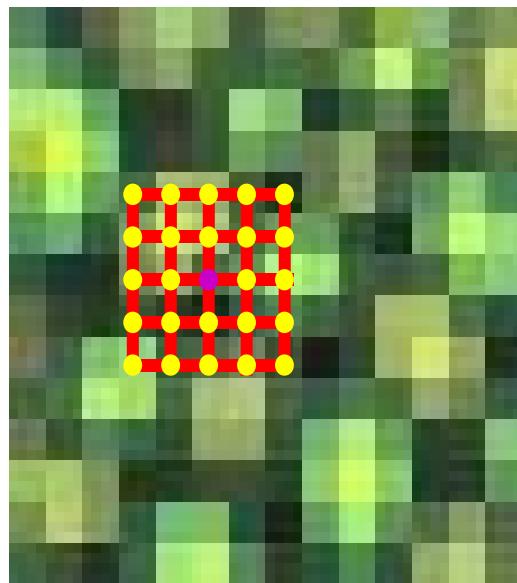
Resampling



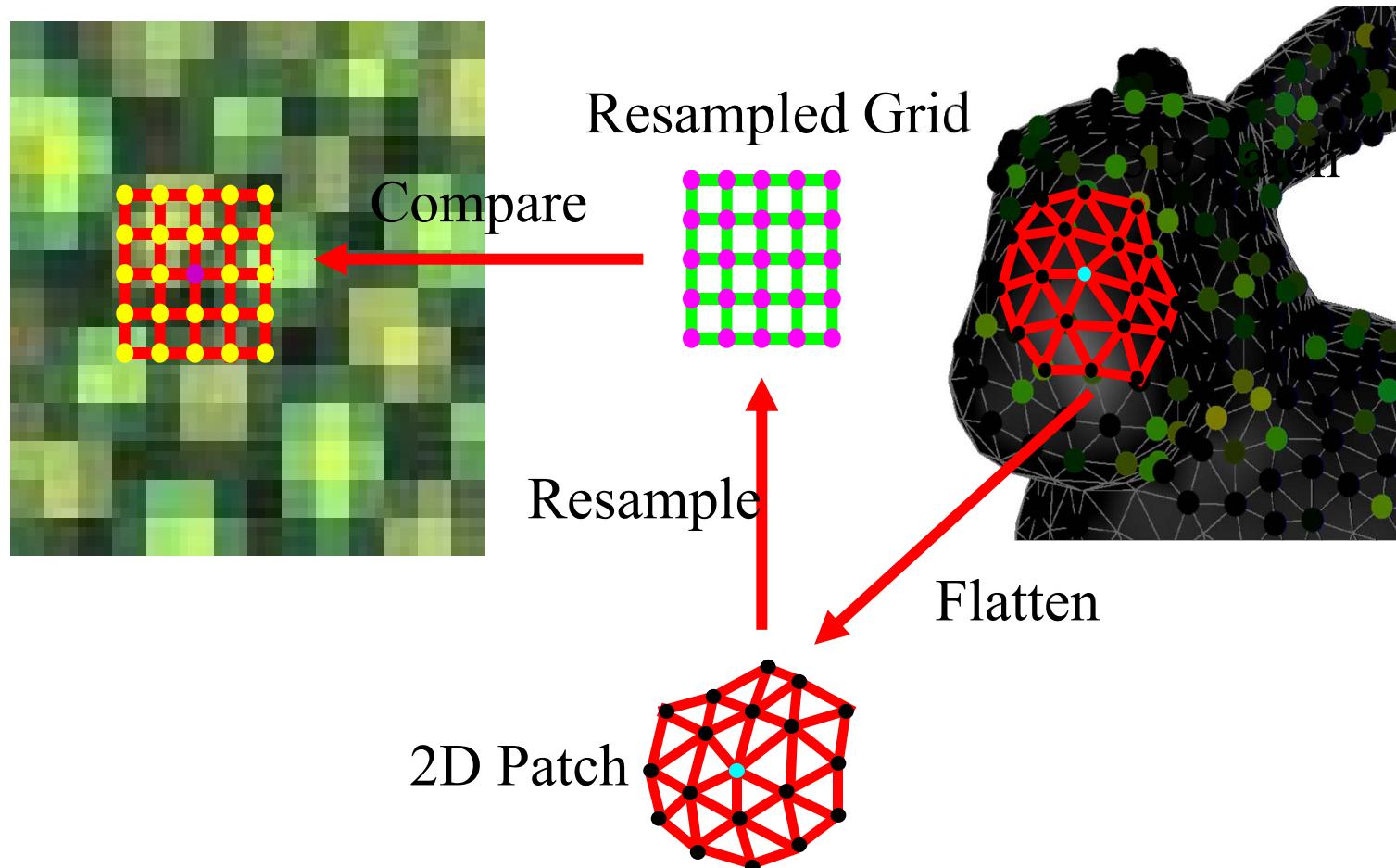
Resampling



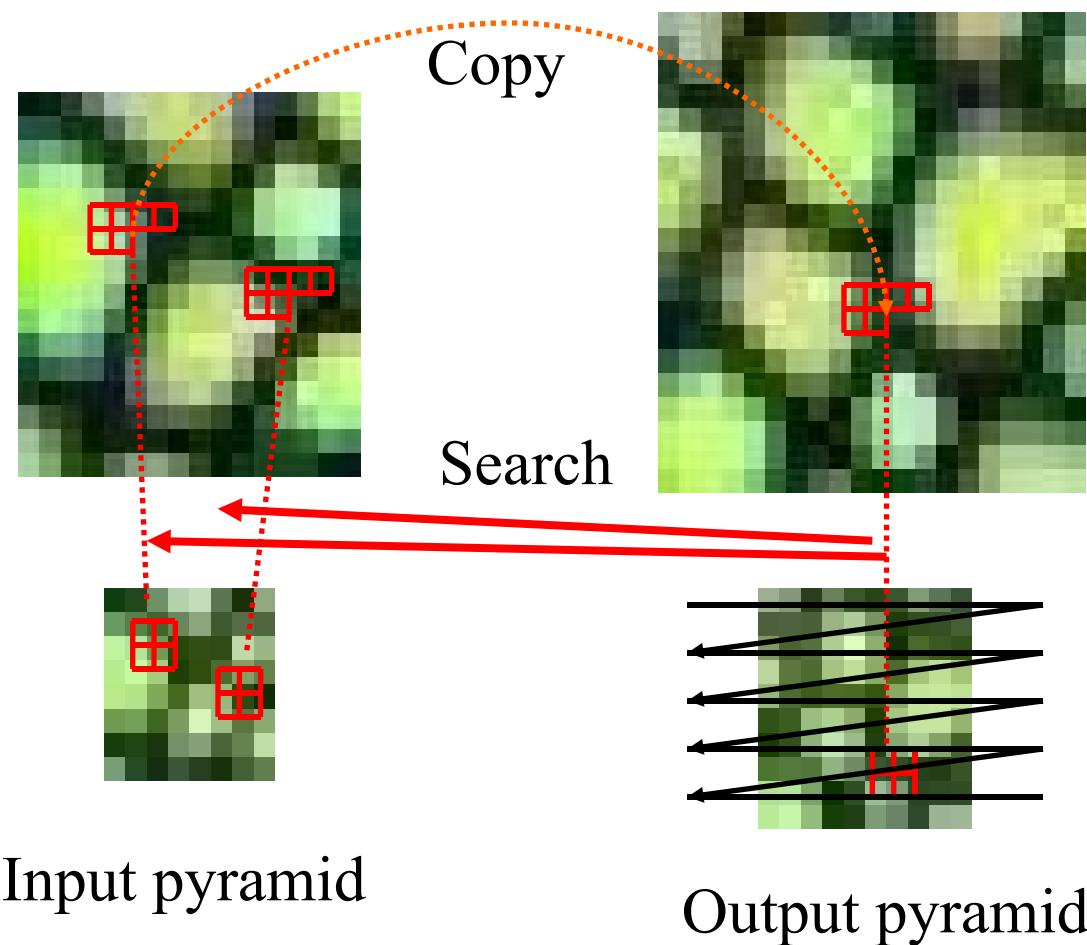
Neighborhood Comparison



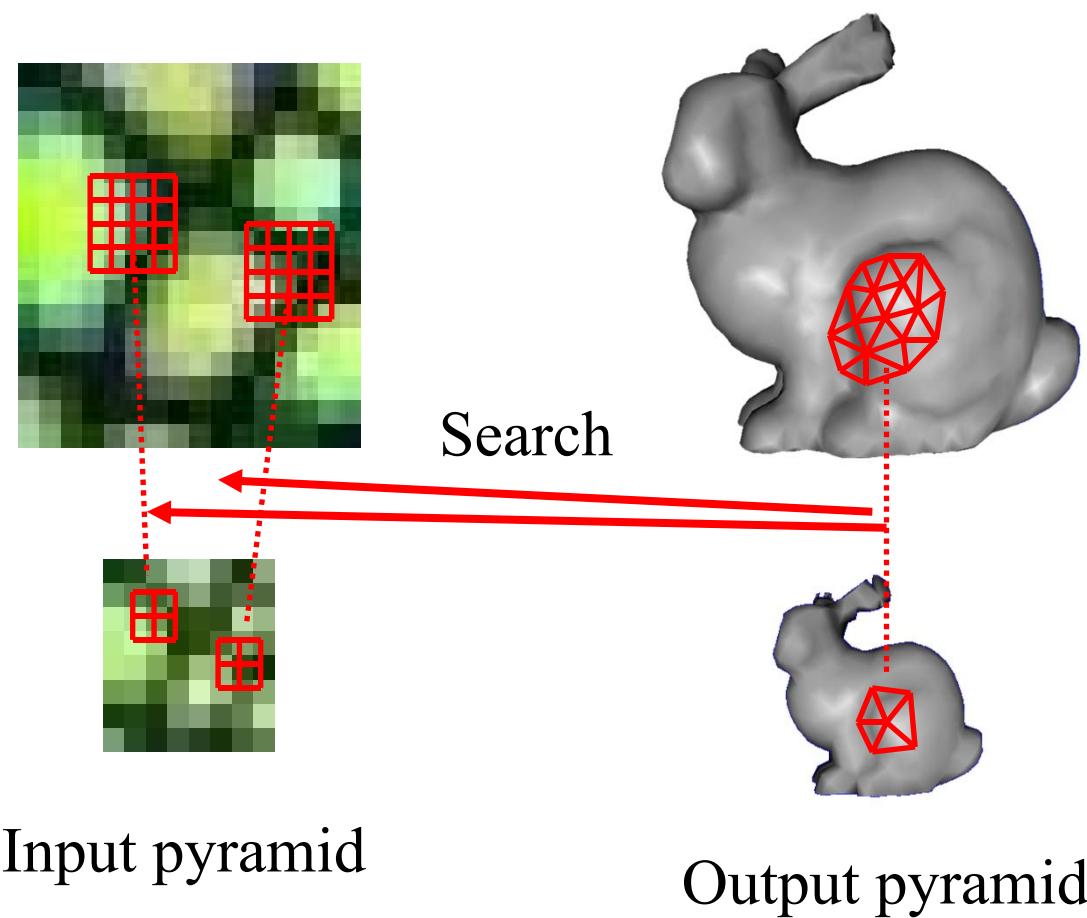
Mesh Neighborhood



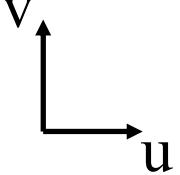
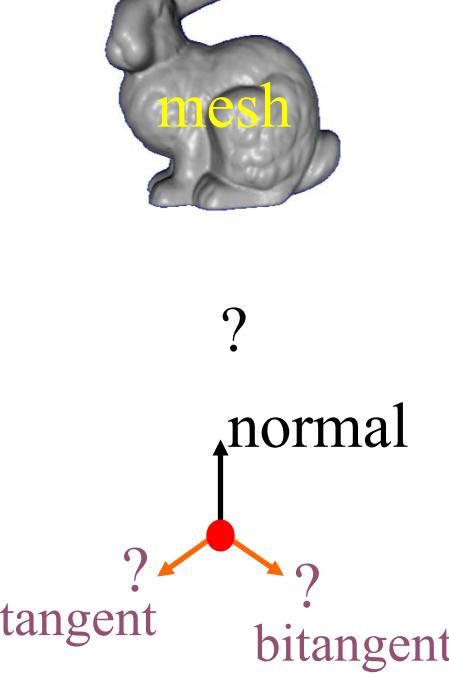
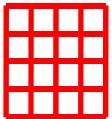
Recap: Texture Synthesis by Neighborhood Search



Surface Texture Synthesis by Neighborhood Search



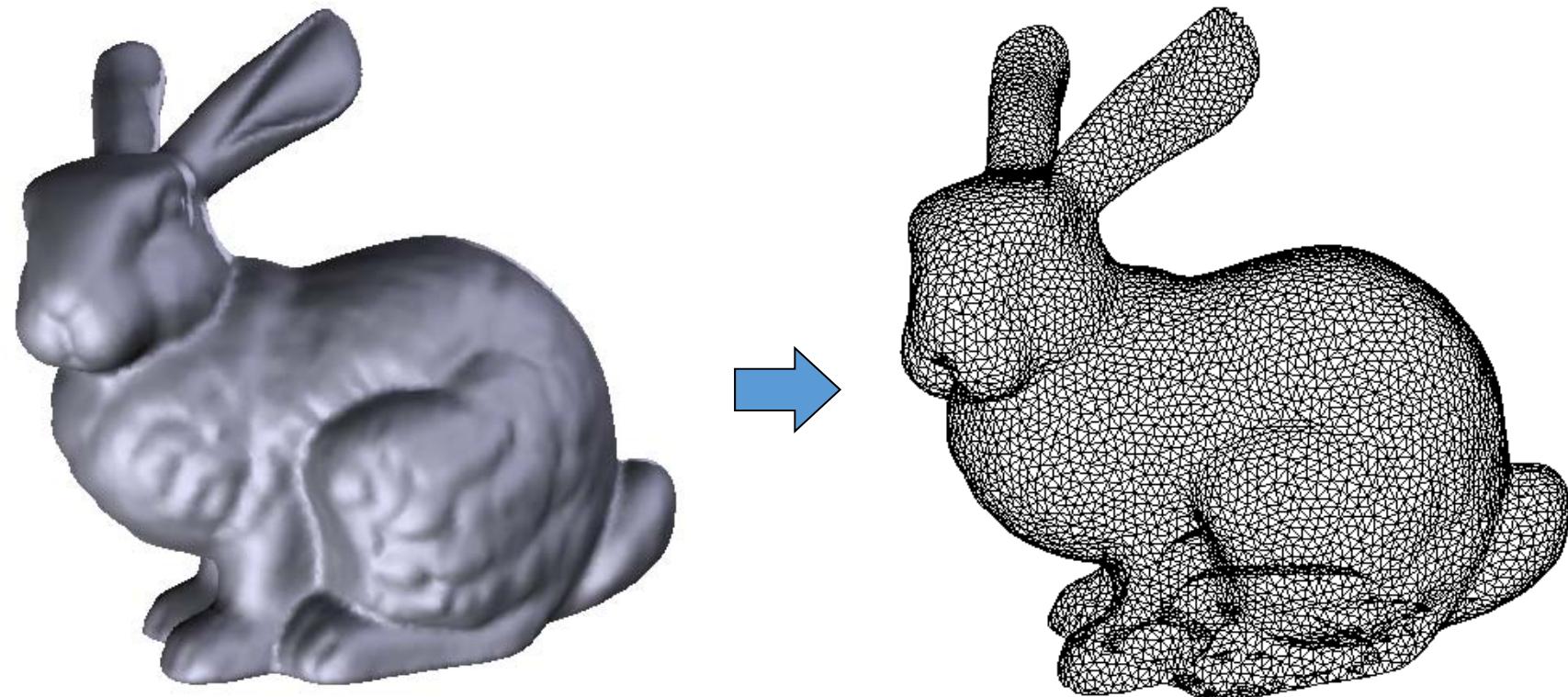
Differences between 2D and 3D

	image	mesh	Solution
Pixels/Vertices	grid	?	mesh re-tiling [Turk'92]
Local Orientation (Vector field)			user-specified relaxation random
Synthesis Order	scanline	?	random
Neighborhood			flattening/resampling

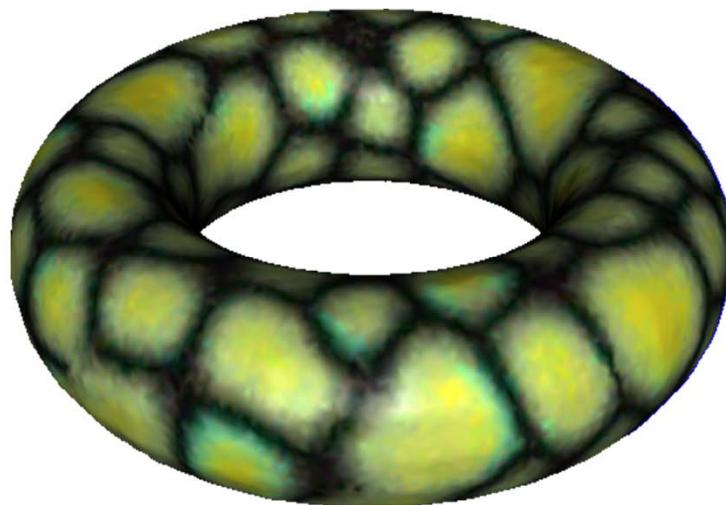
Sampling on Surface: Retiling

[Turk, Siggraph 1992]

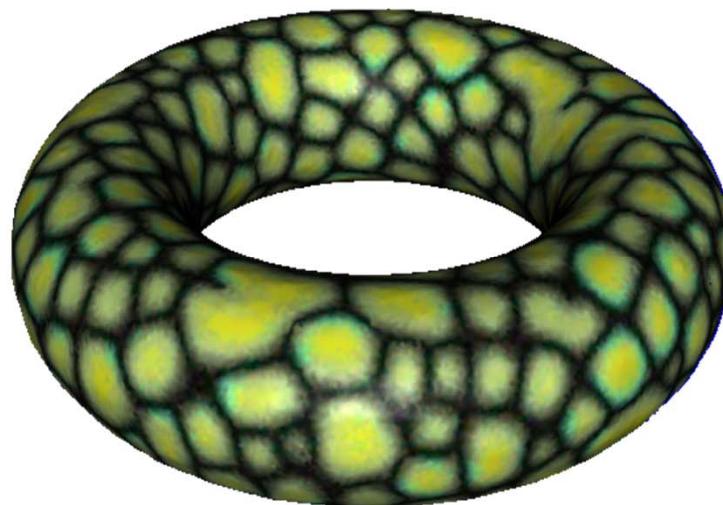
- Distribute a set of vertices over surface as uniformly as possible



Retiling Density



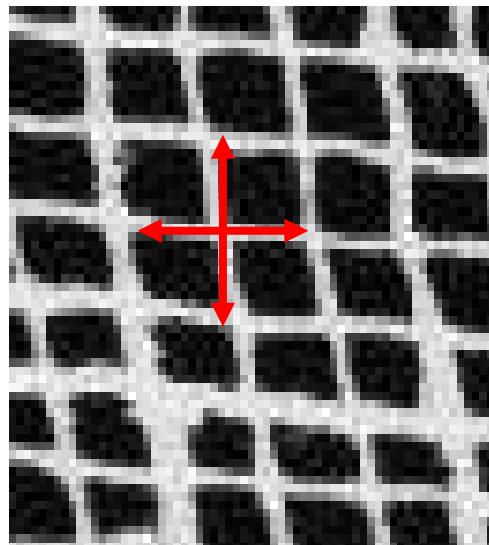
24576 vertices



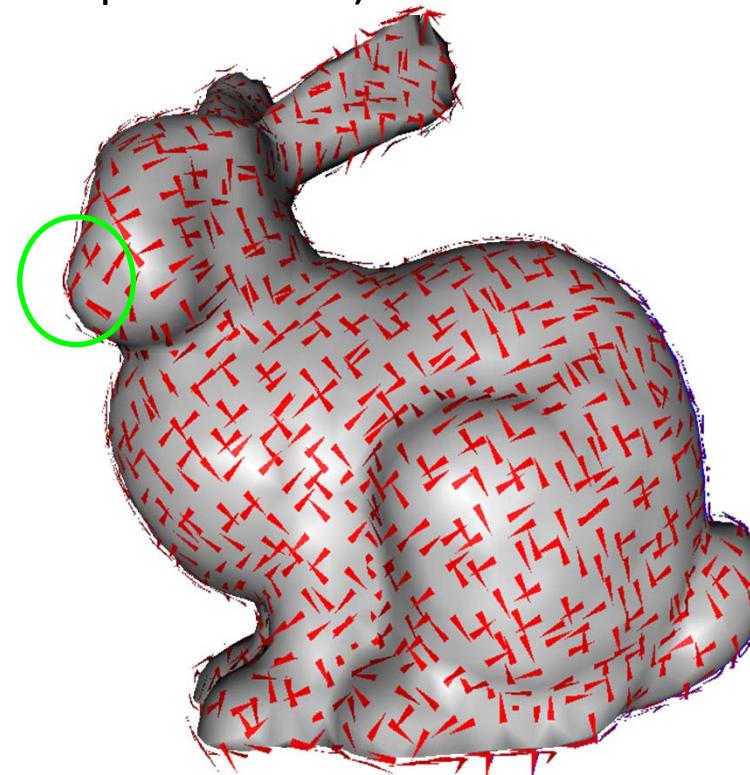
73728 vertices

Texture Orientation

- Methods for orienting textures
 - user-specified
 - random (for isotropic textures)
 - smooth or symmetric (for anisotropic textures)
 - by relaxation

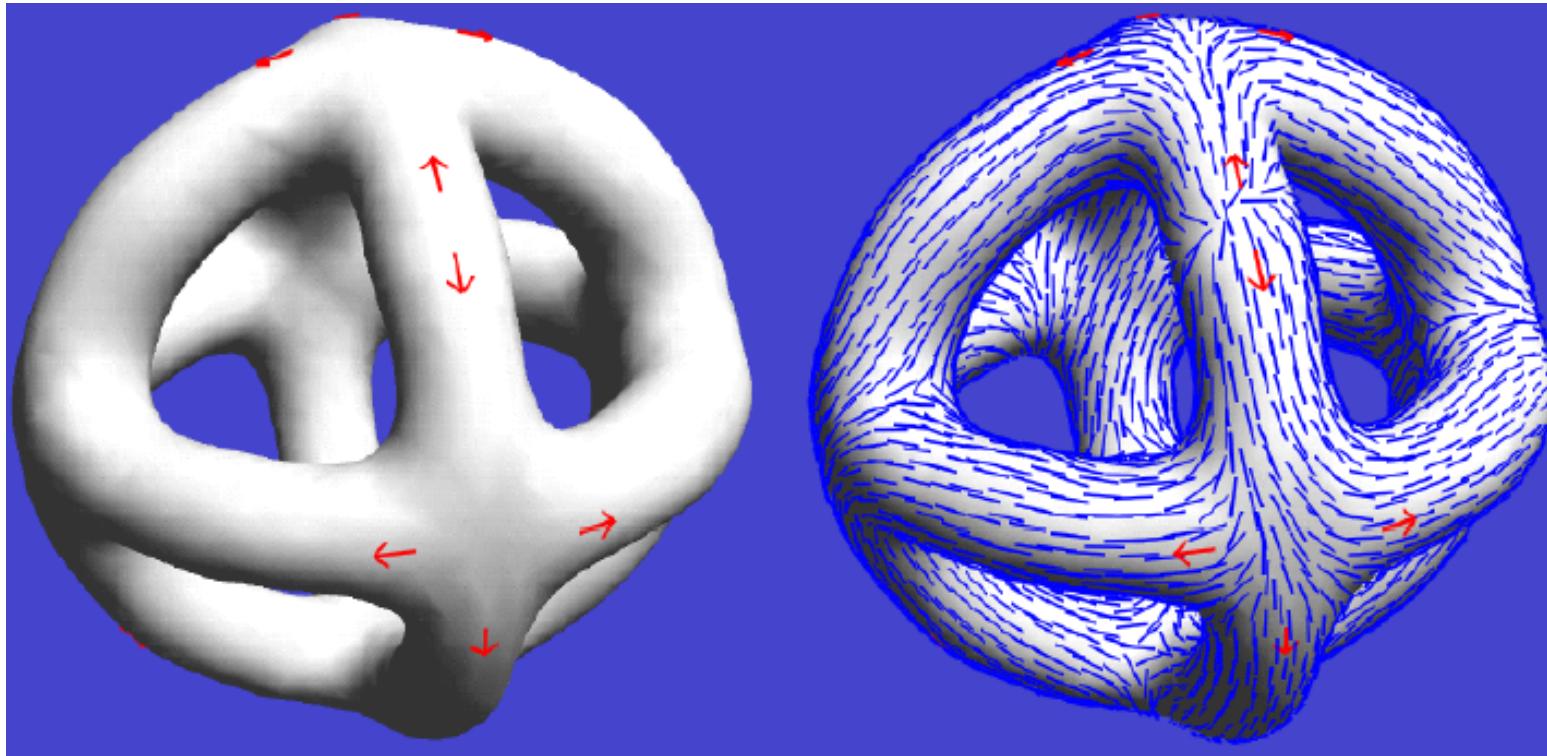


4-way symmetric texture



4-way symmetric vector field

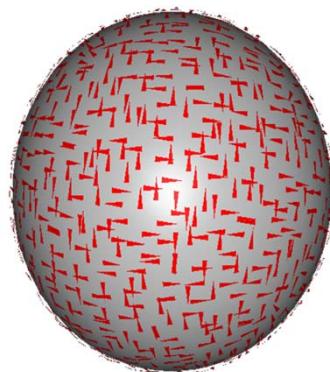
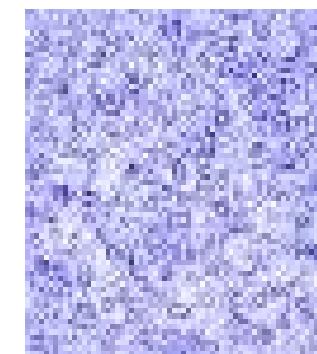
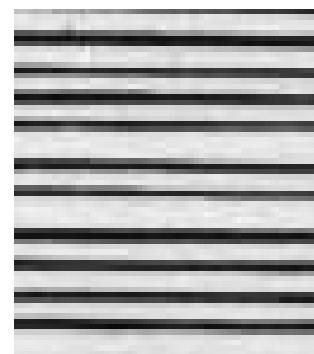
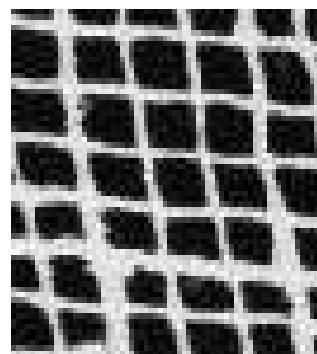
User Specified Vector Field



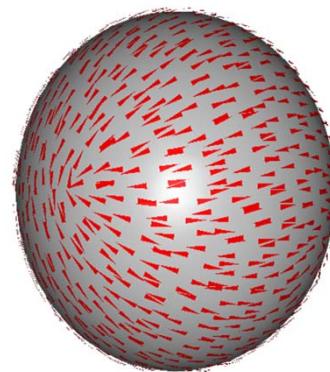
User specified vectors
on some vertices

Interpolated vector field
on all vertices

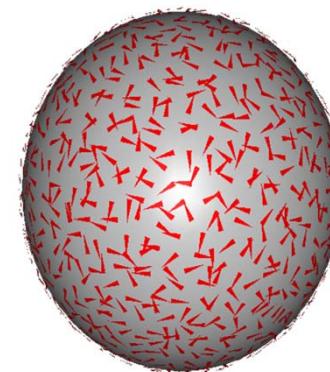
Texture Orientation Examples



4-way symmetric

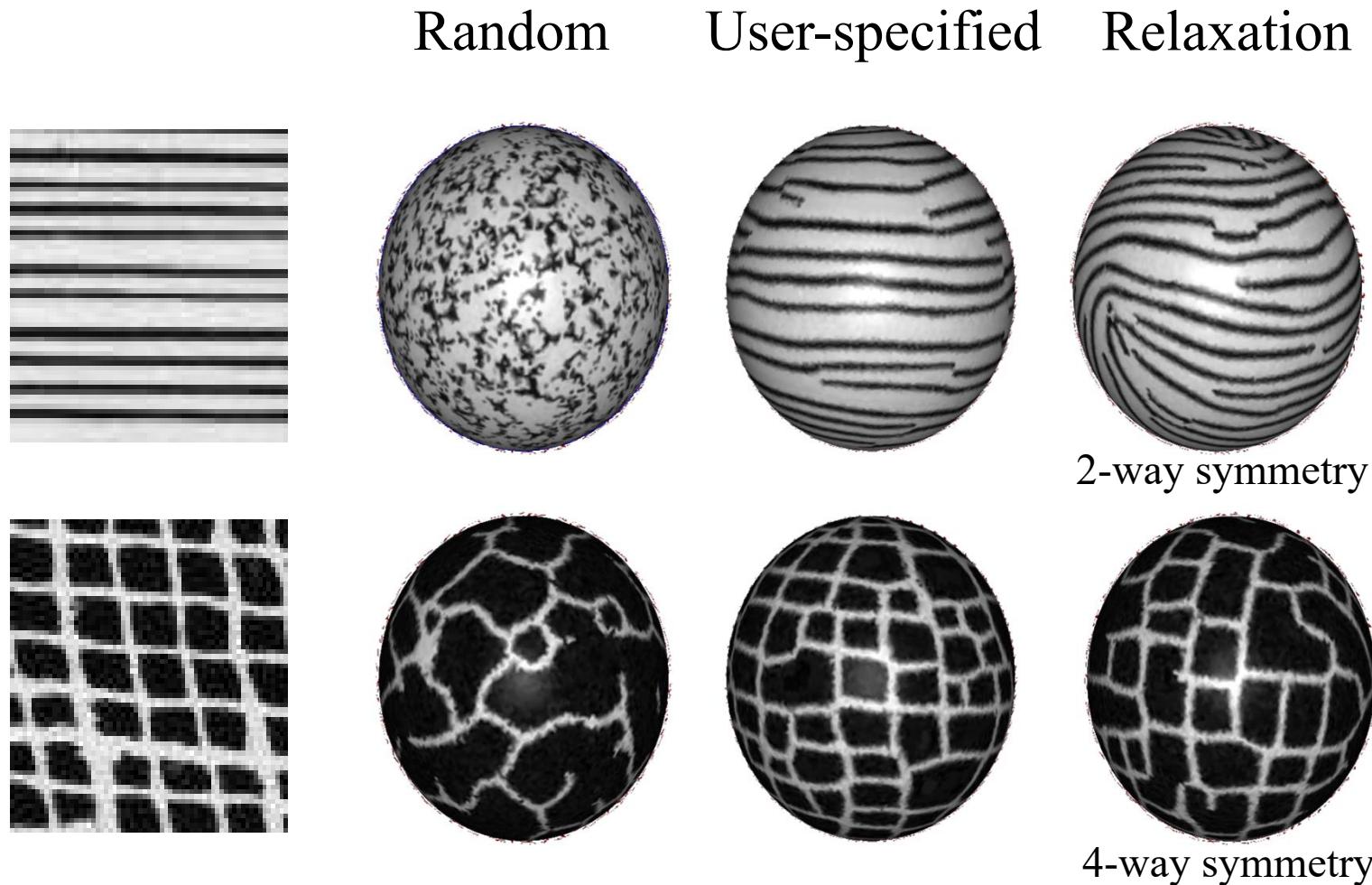


2-way symmetric



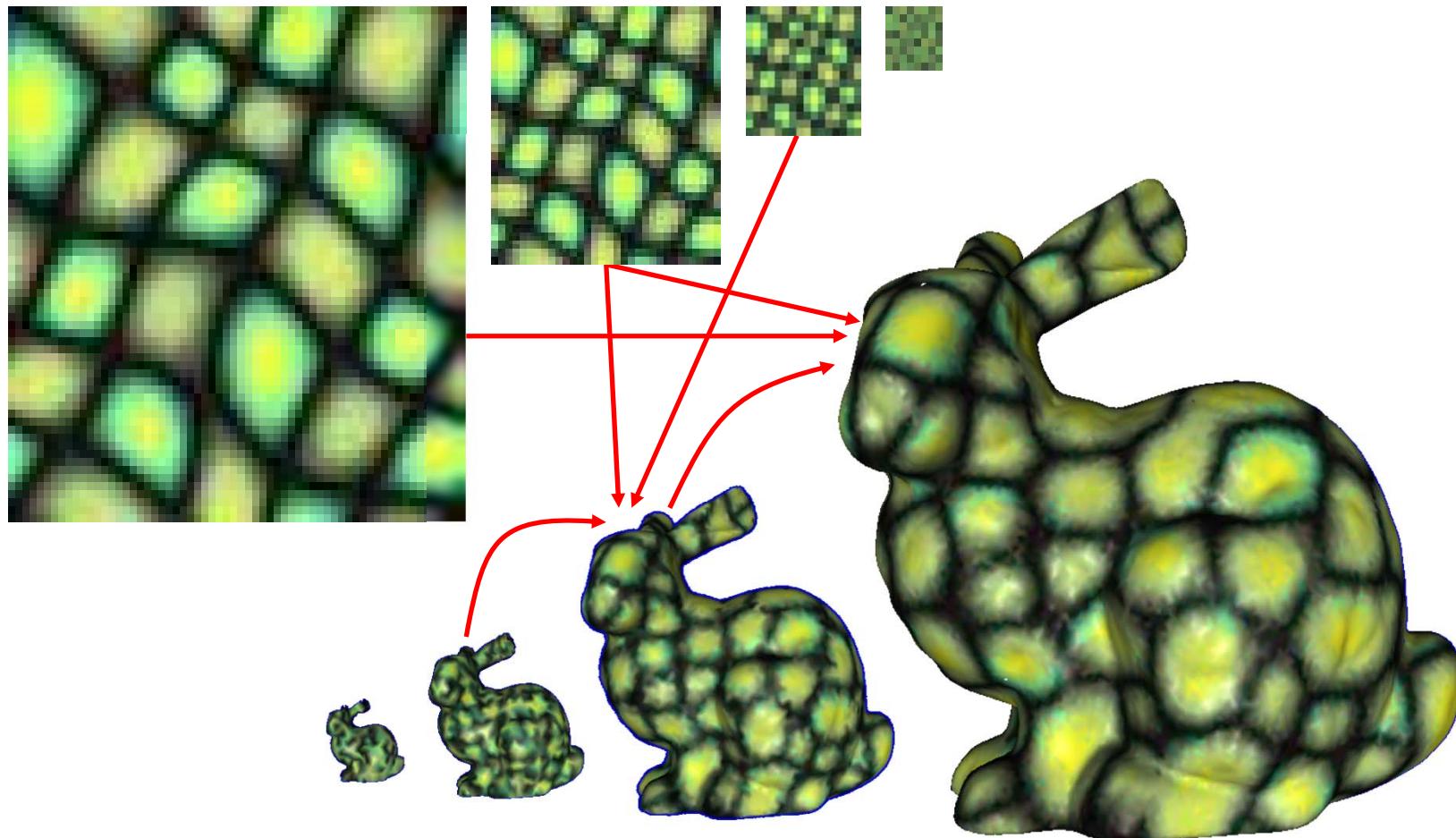
Random

Results: Other Orientations

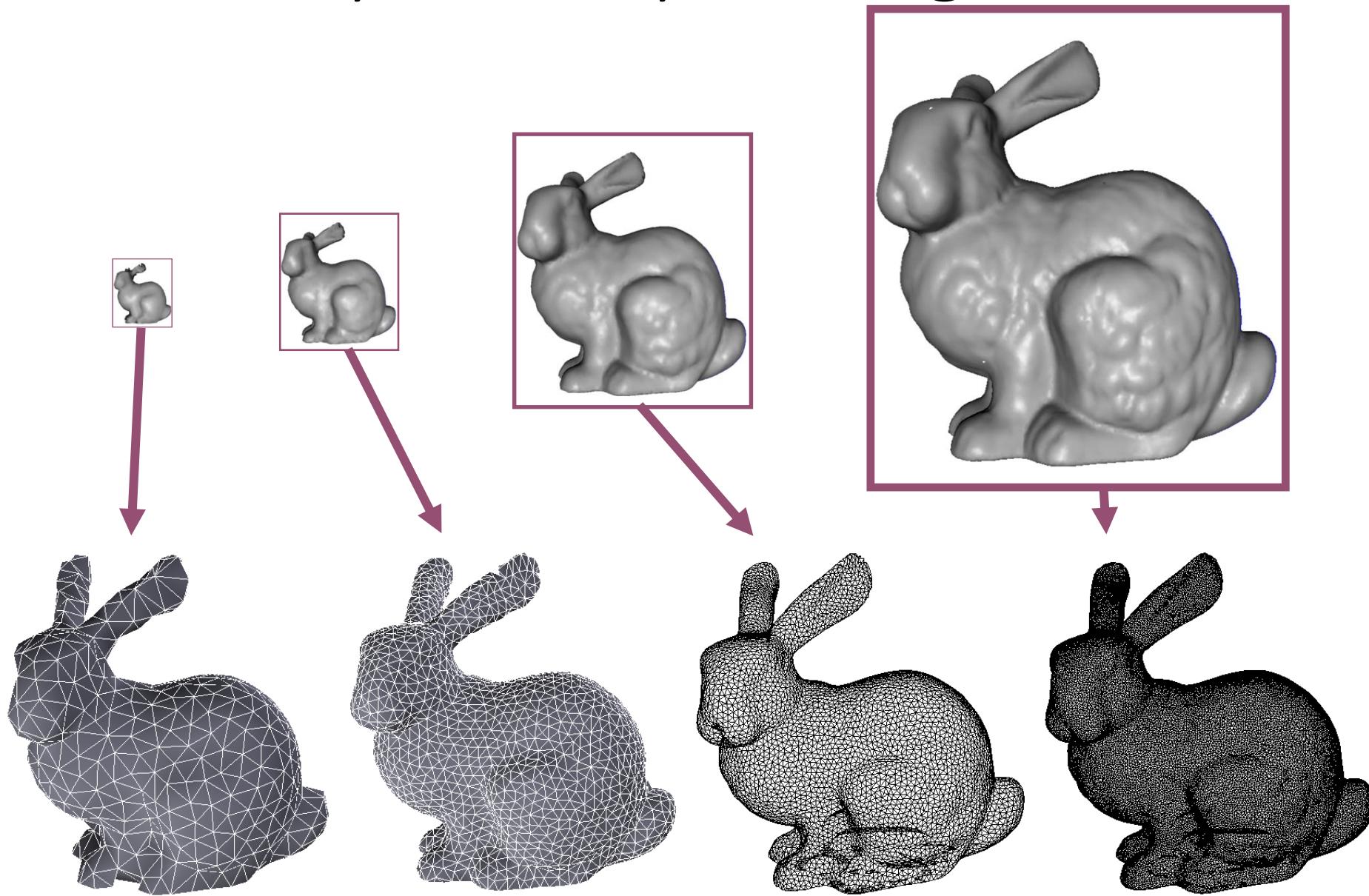


Multiresolution Synthesis

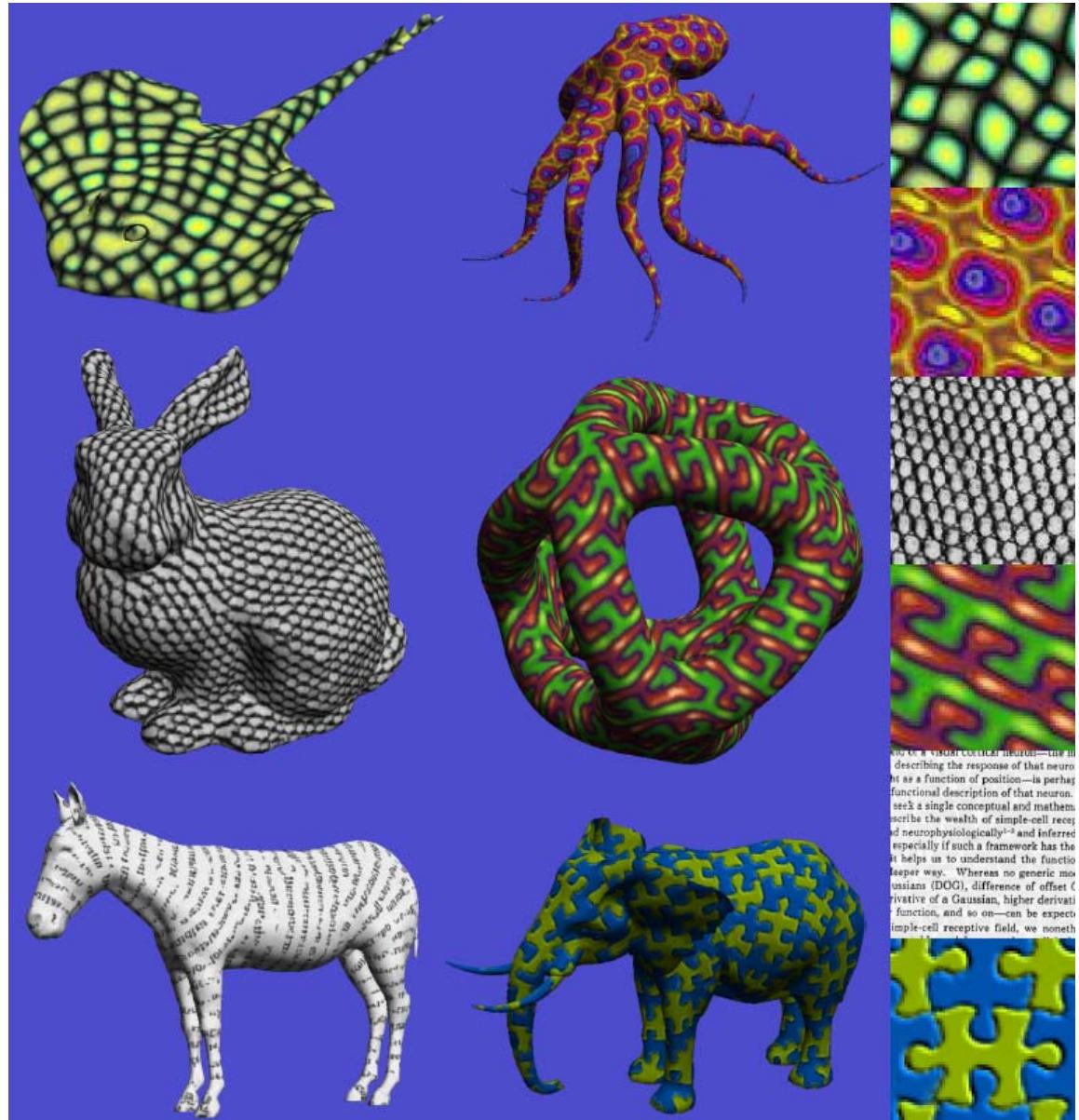
[Wei&Levoy, Siggraph 2001]



Mesh Pyramid by Retiling



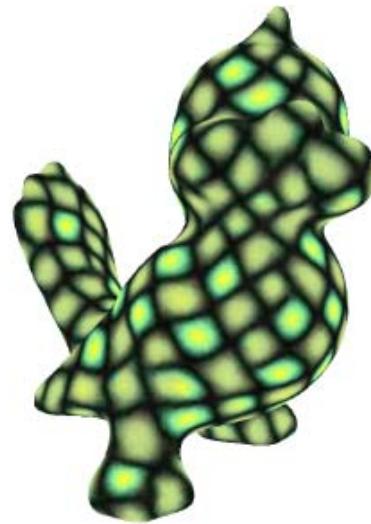
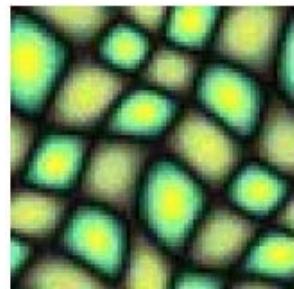
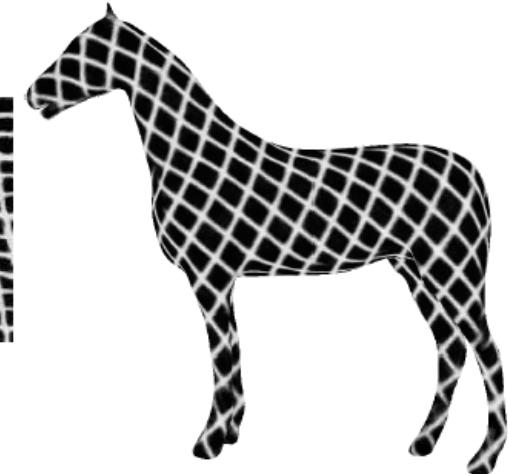
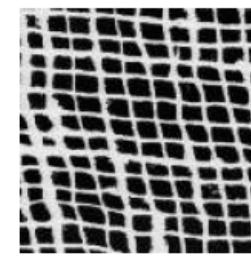
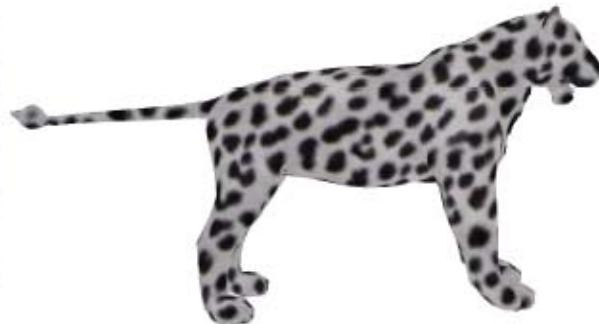
Examples



and of a visual cortical neuron—the in describing the response of that neuro it as a function of position—is perhaps functional description of that neuron. seek a single conceptual and mathem describe the wealth of simple-cell rece and neurophysiologically^{1–2} and inferred especially if such a framework has the it helps us to understand the functio deeper way. Whereas no generic moians (DOG), difference of offset C derivative of a Gaussian, higher derivati function, and so on—can be expect simple-cell receptive field, we nonetheless



Results

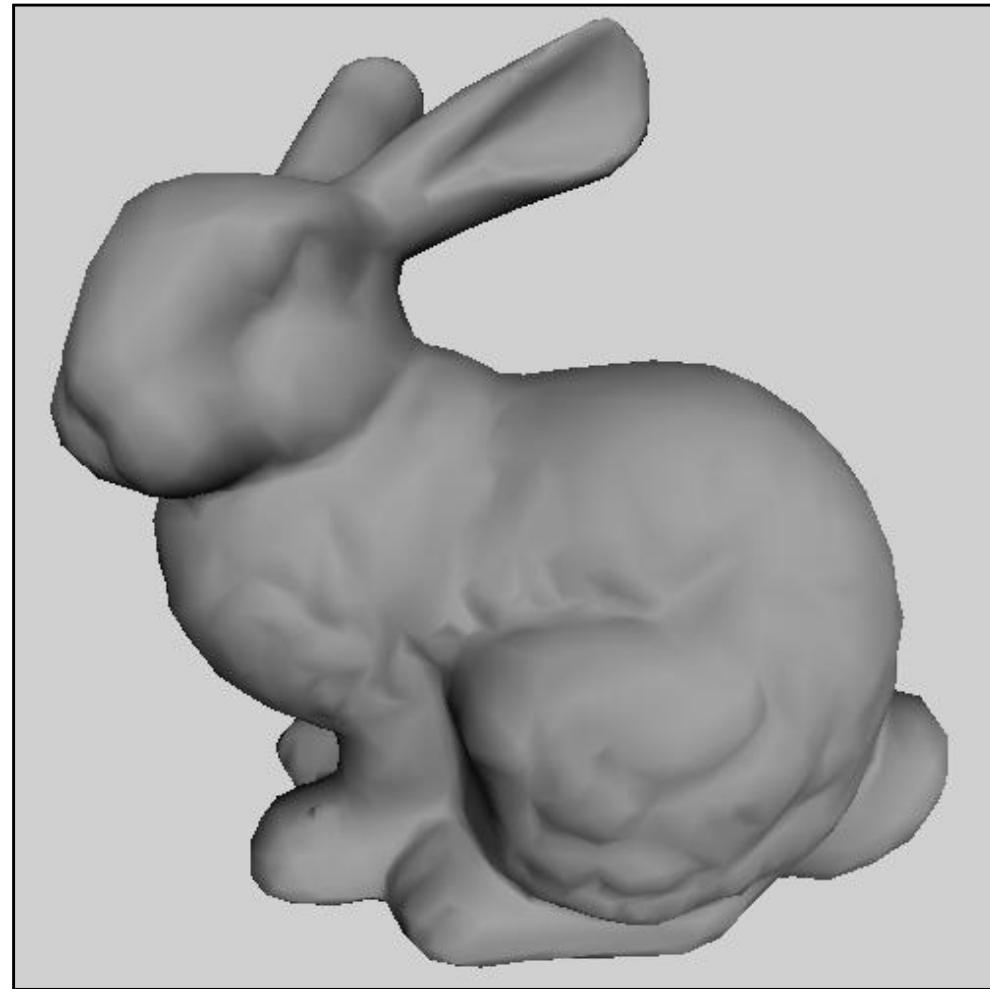


Patch-based Synthesis

[Praun et al., Siggraph 2001]

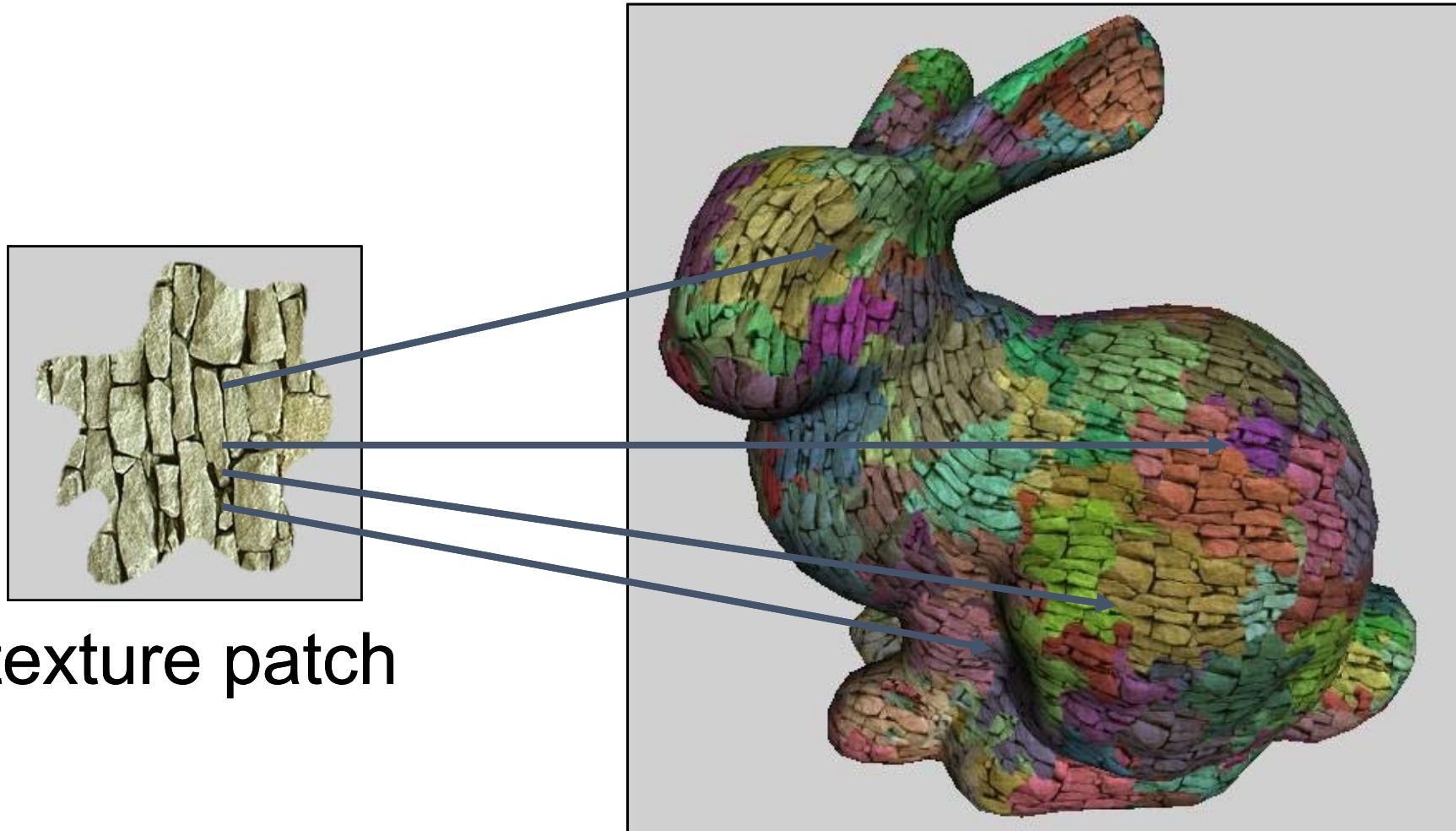


texture patch



surface

Key Idea: Patch Pasting



texture patch

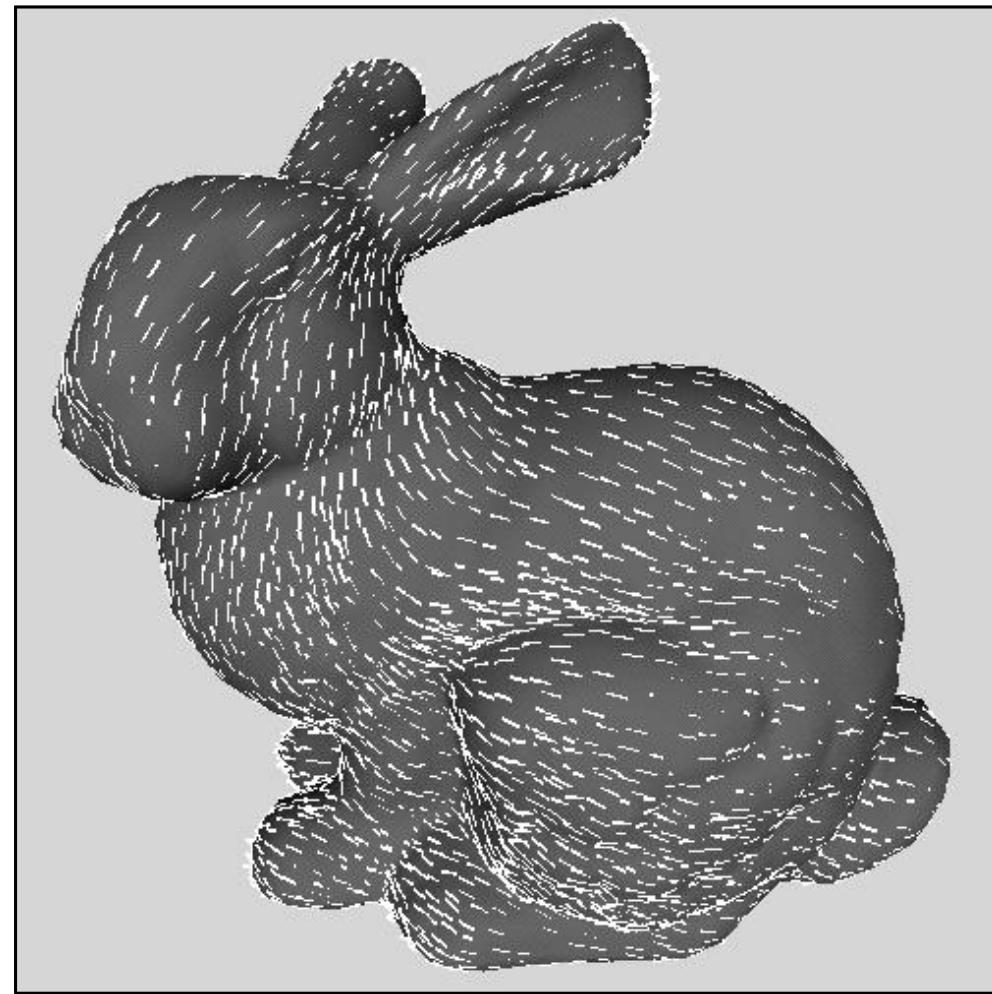
“lapped textures”

surface

Algorithm

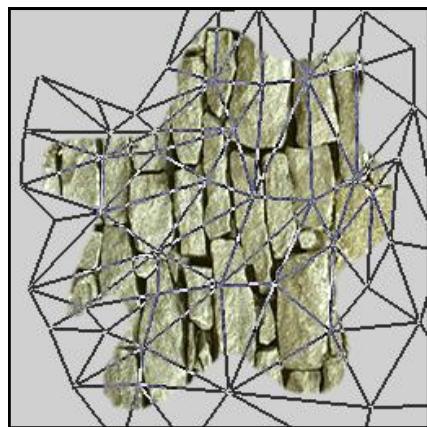


texture patch

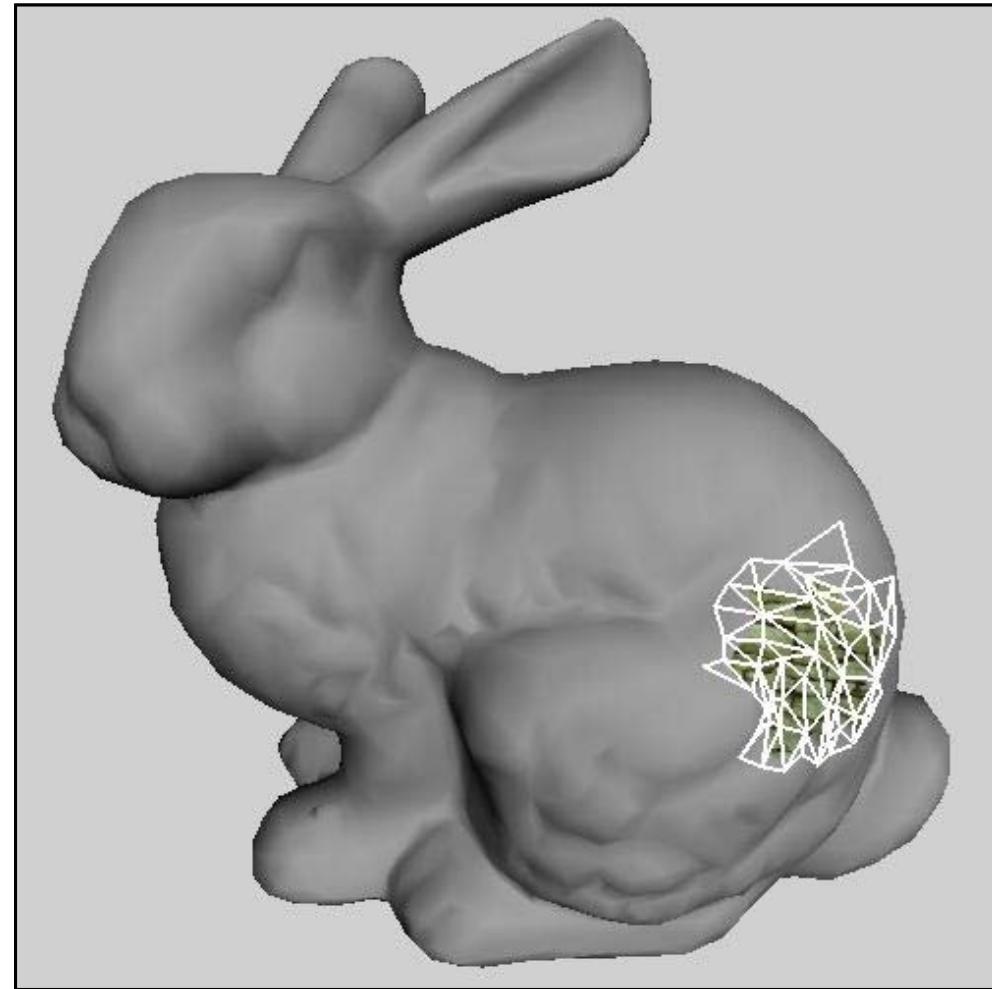


surface

Algorithm

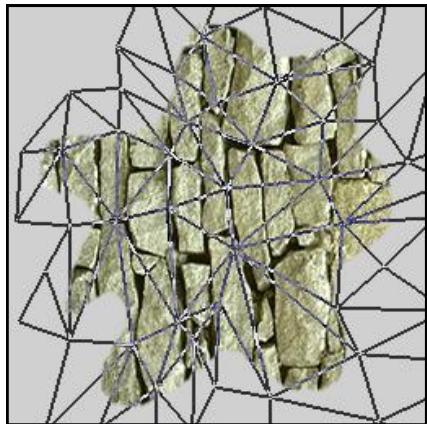


texture patch

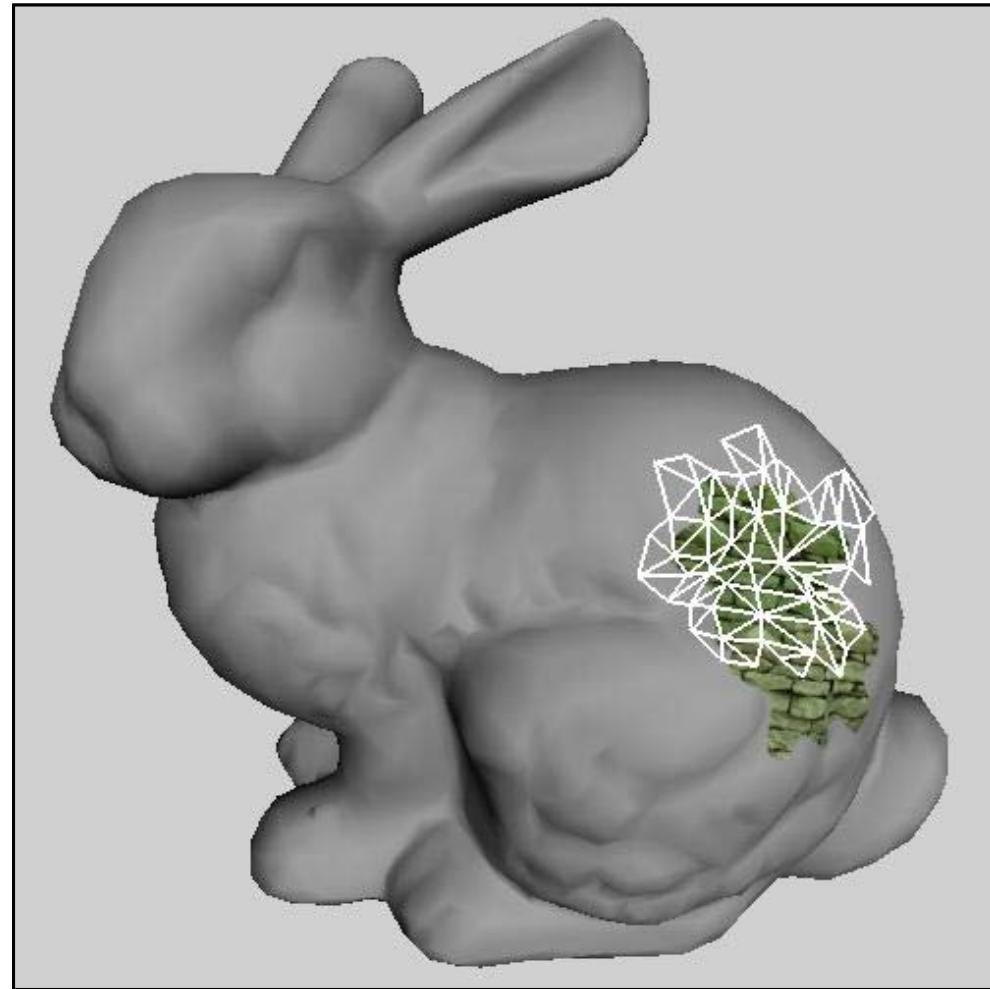


surface

Algorithm



texture patch



surface

Algorithm

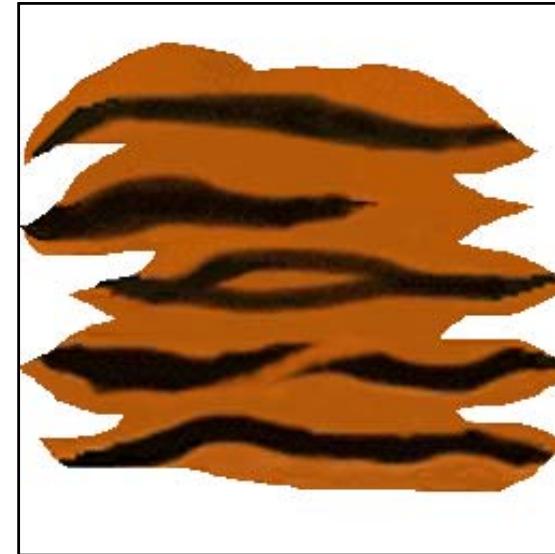
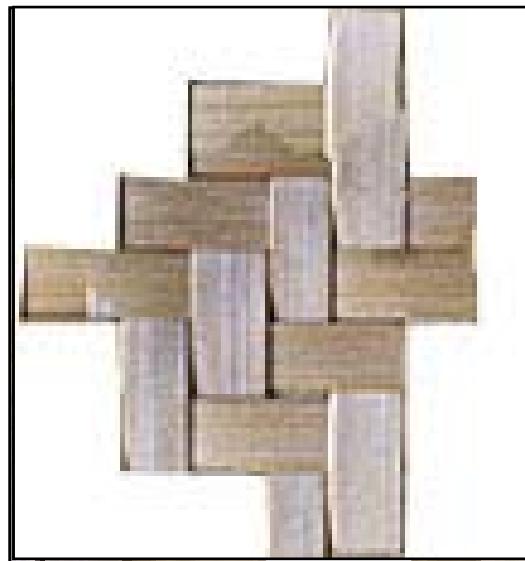


texture patch

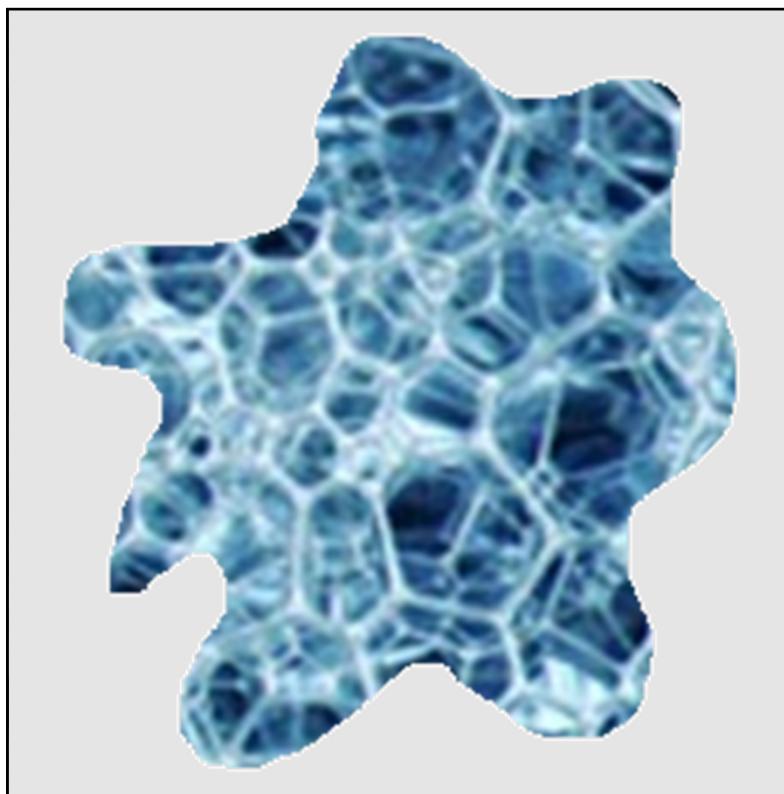


surface

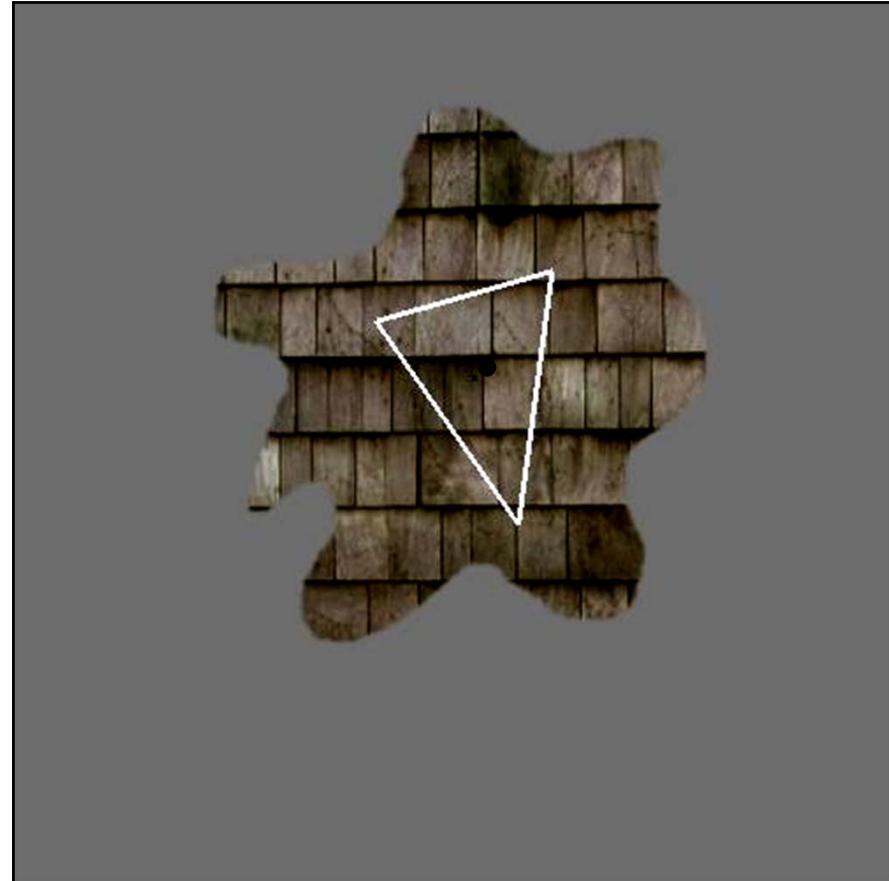
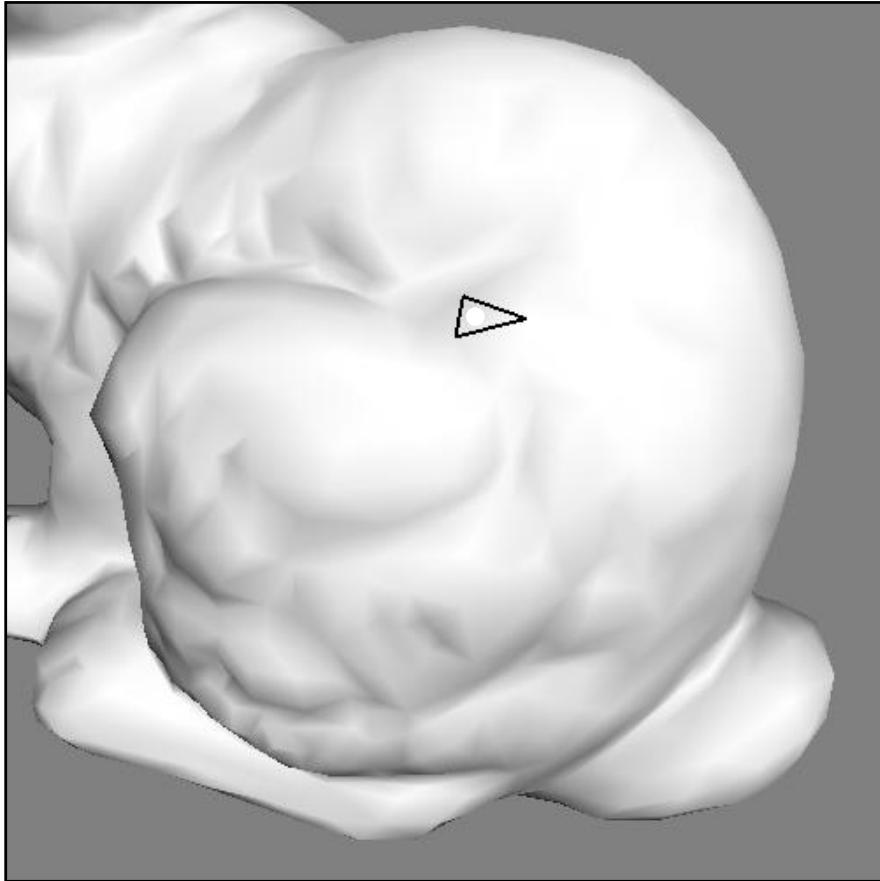
Texture Patch Creation



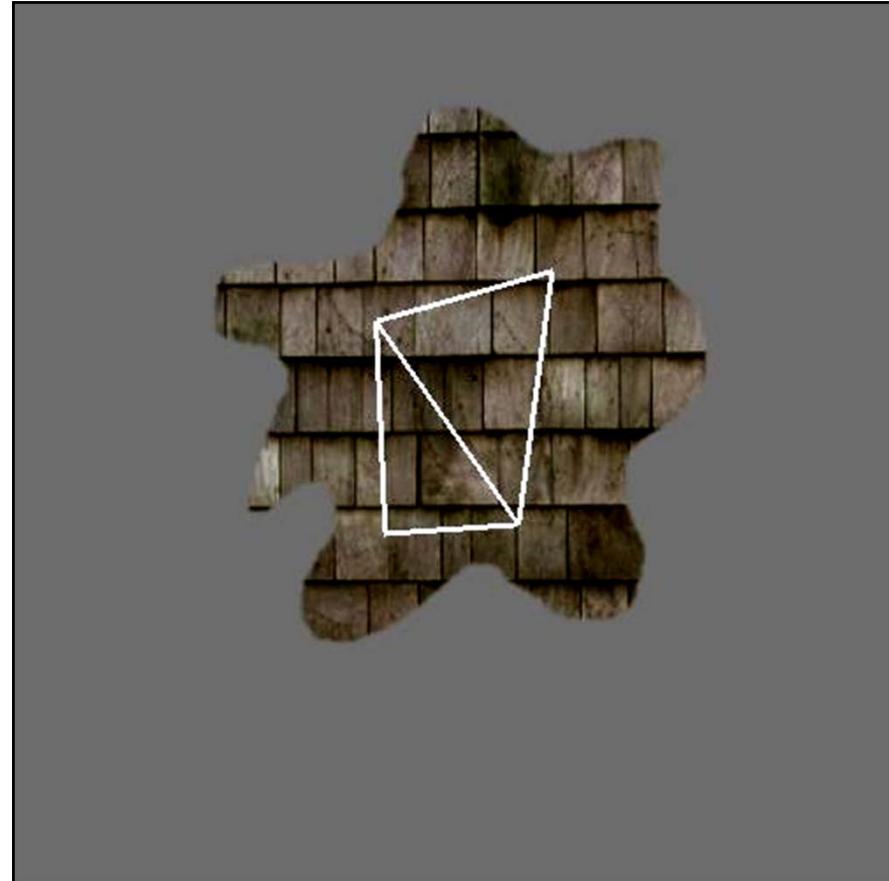
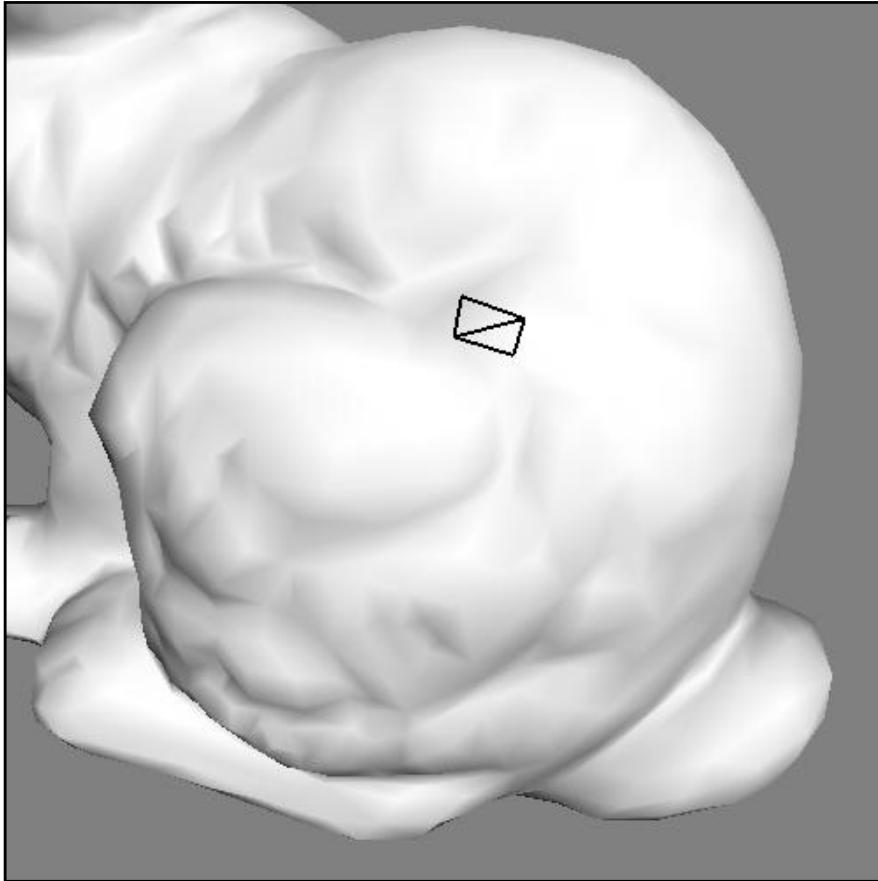
Less Structure → Splotch



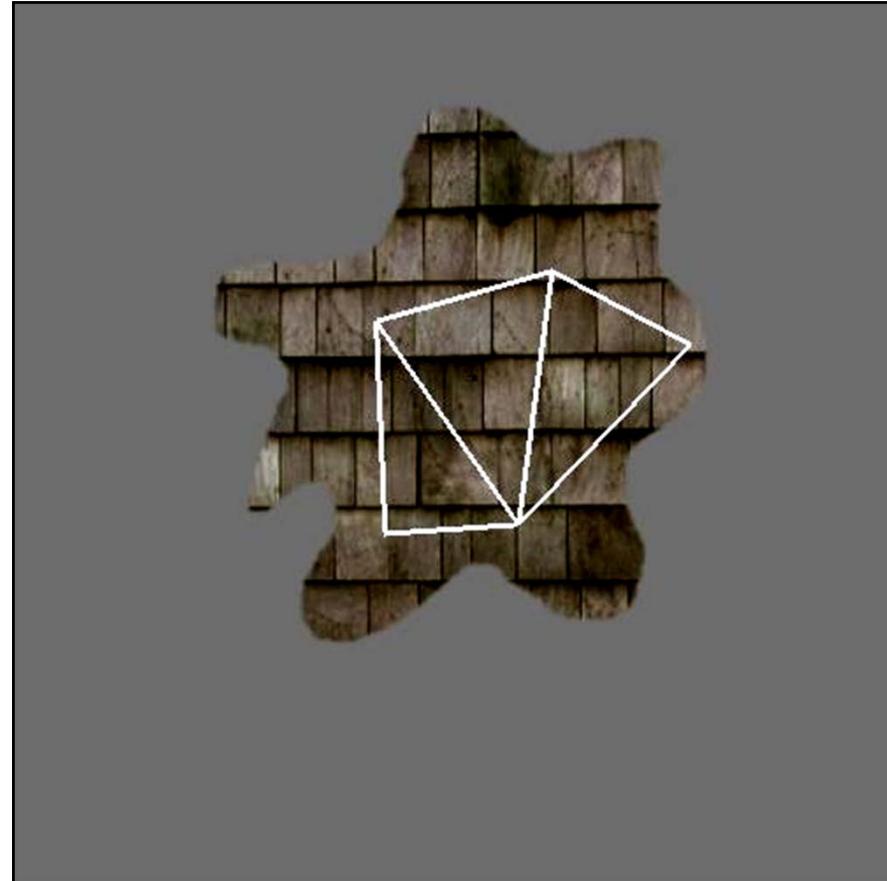
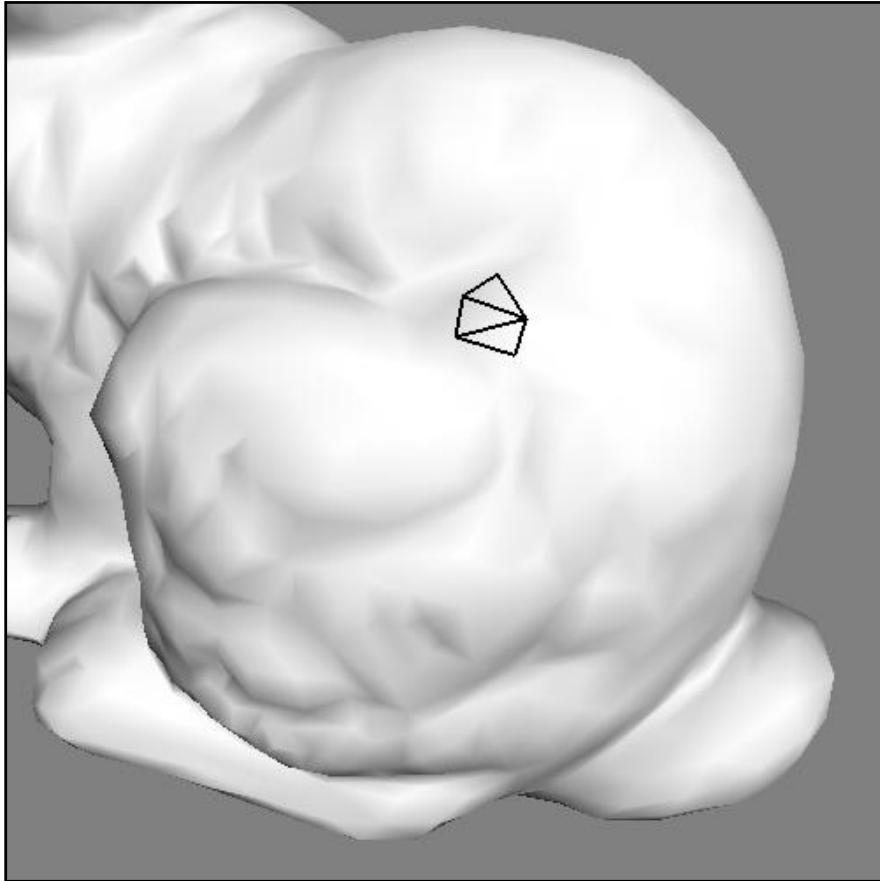
Patch Growth



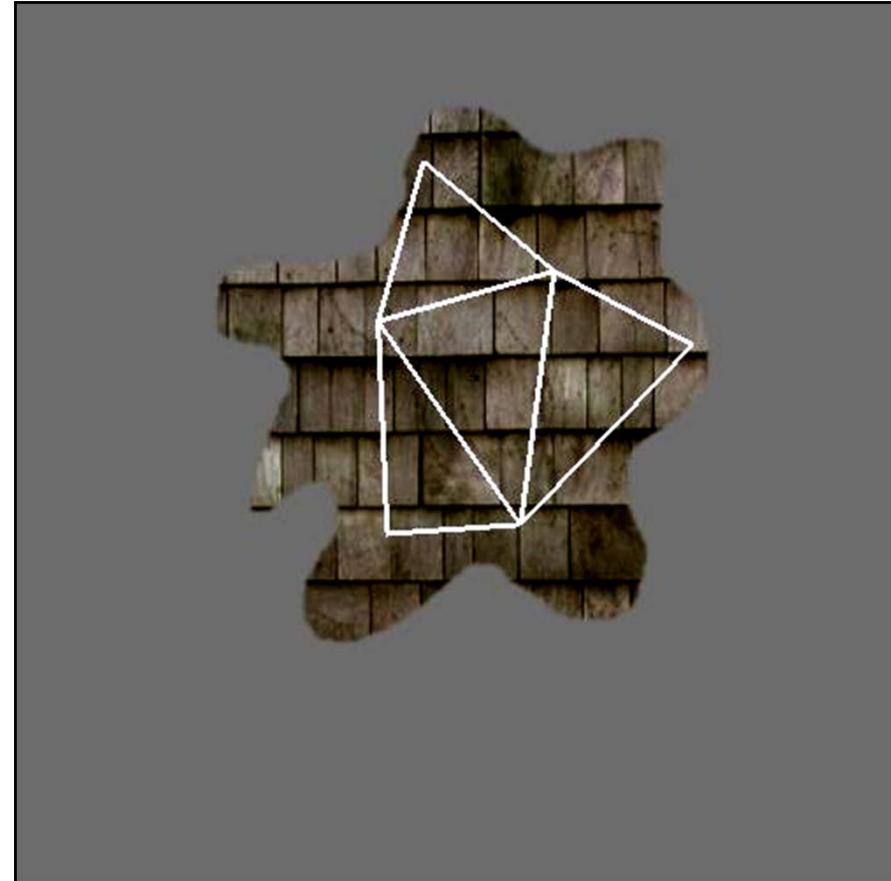
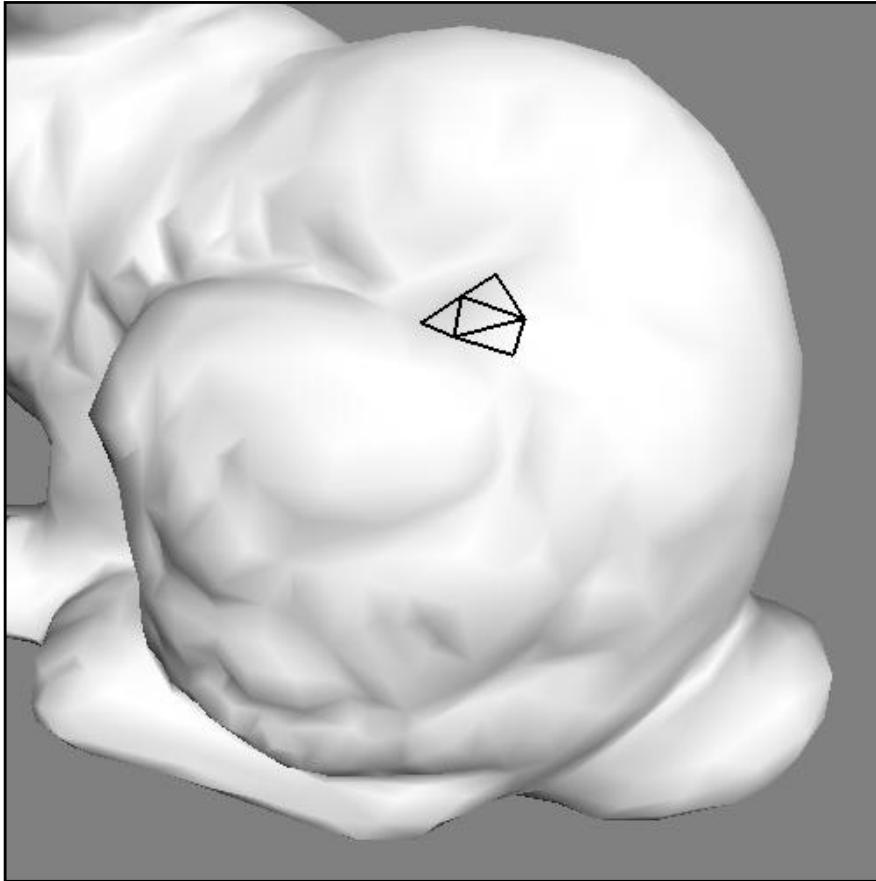
Patch Growth



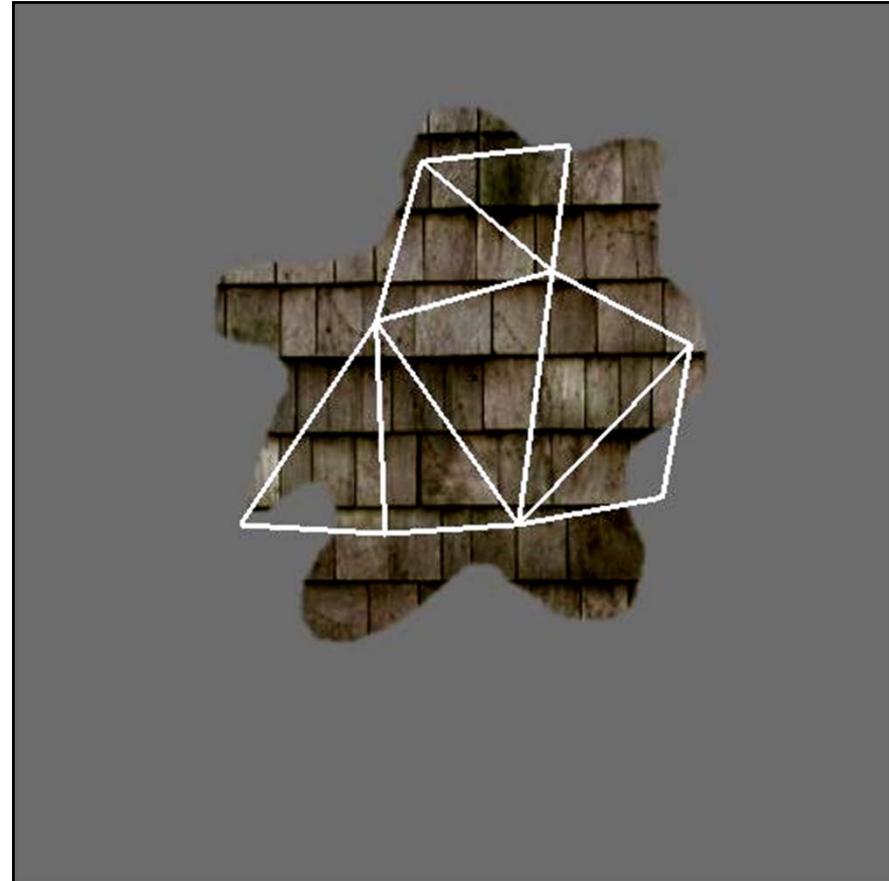
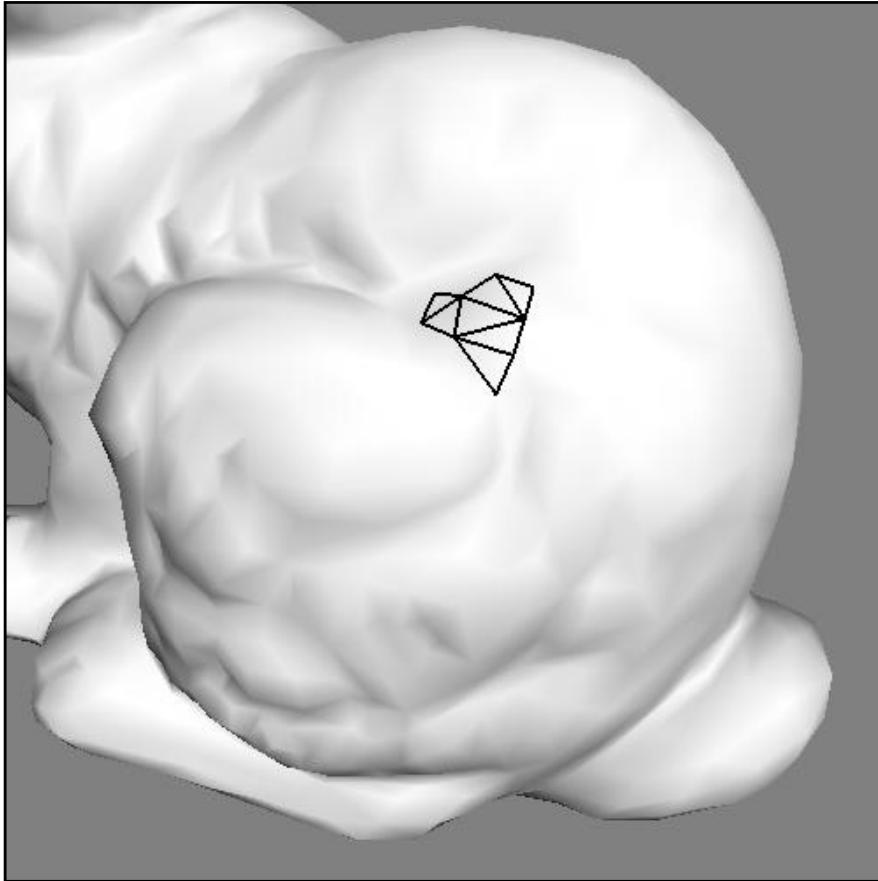
Patch Growth



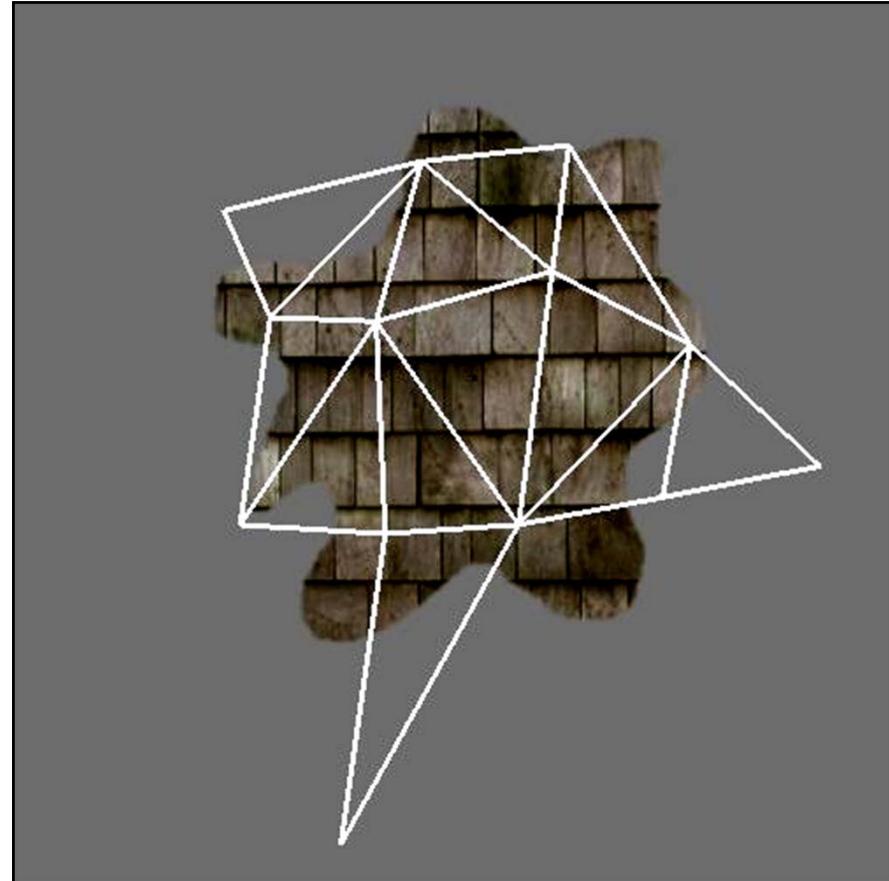
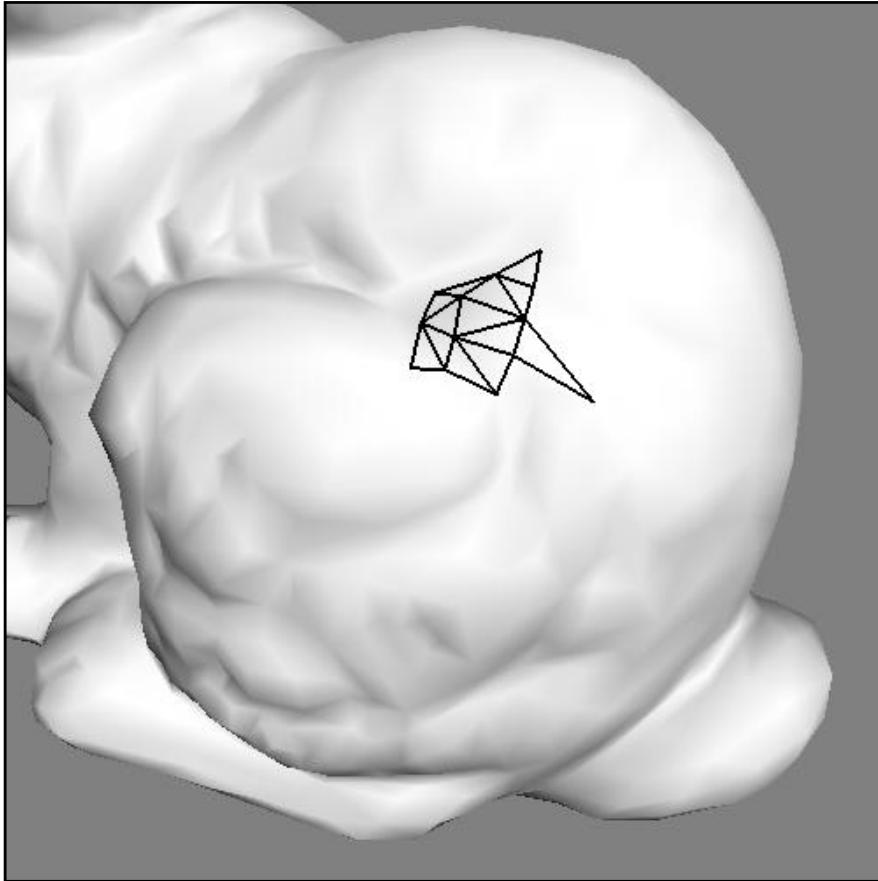
Patch Growth



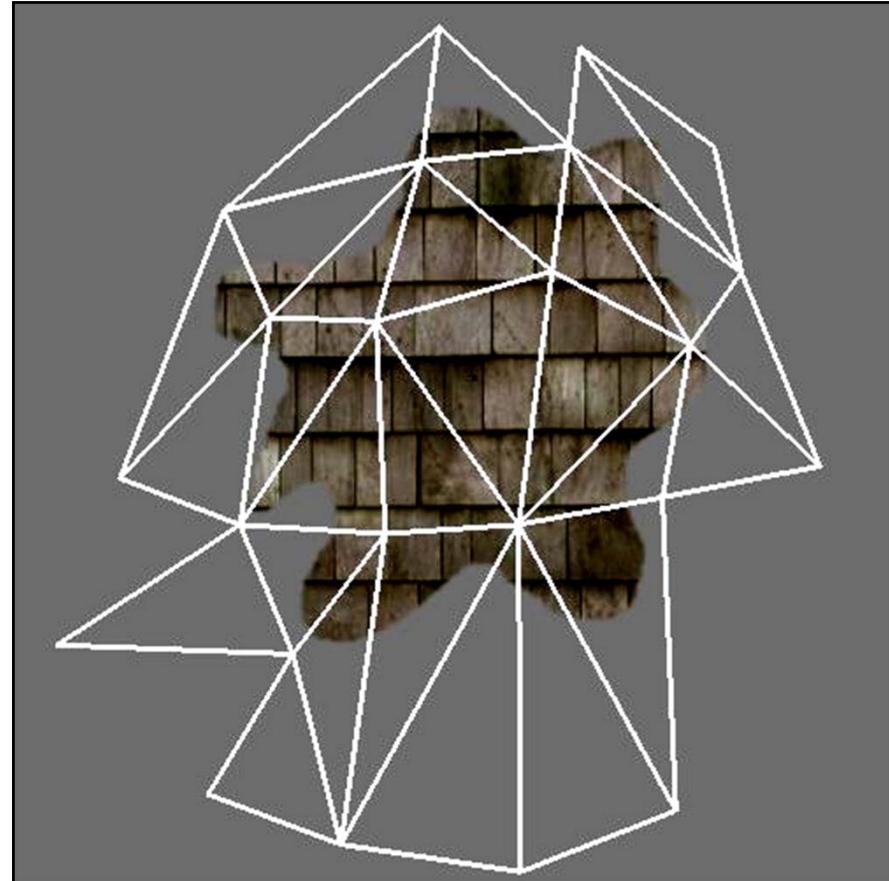
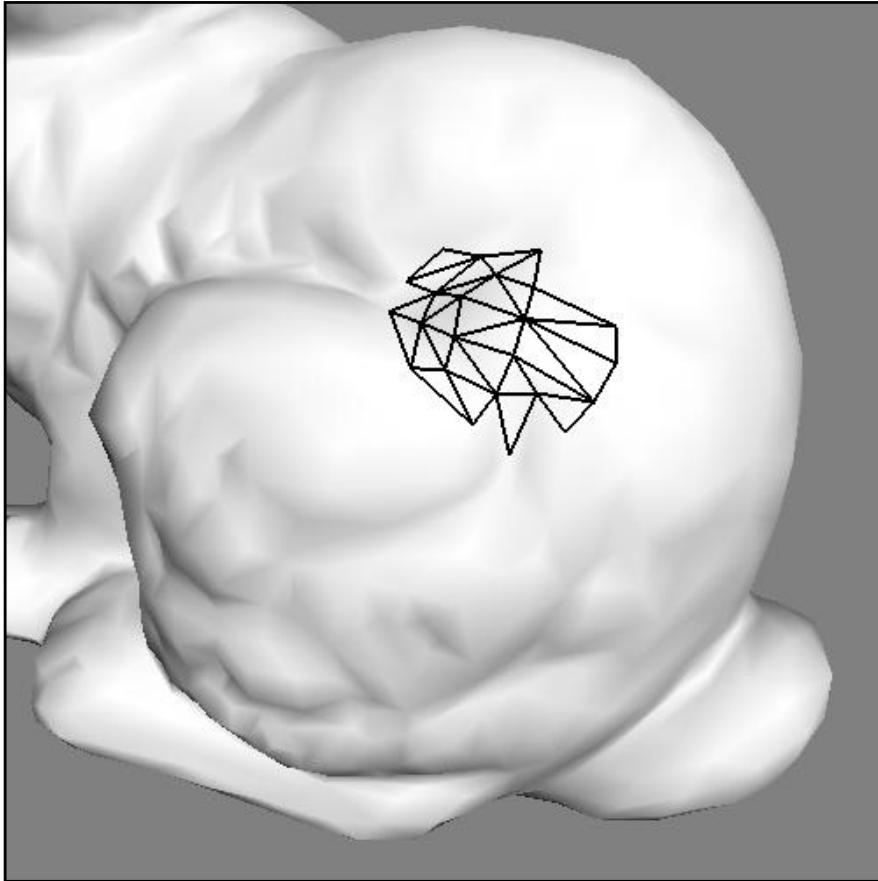
Patch Growth



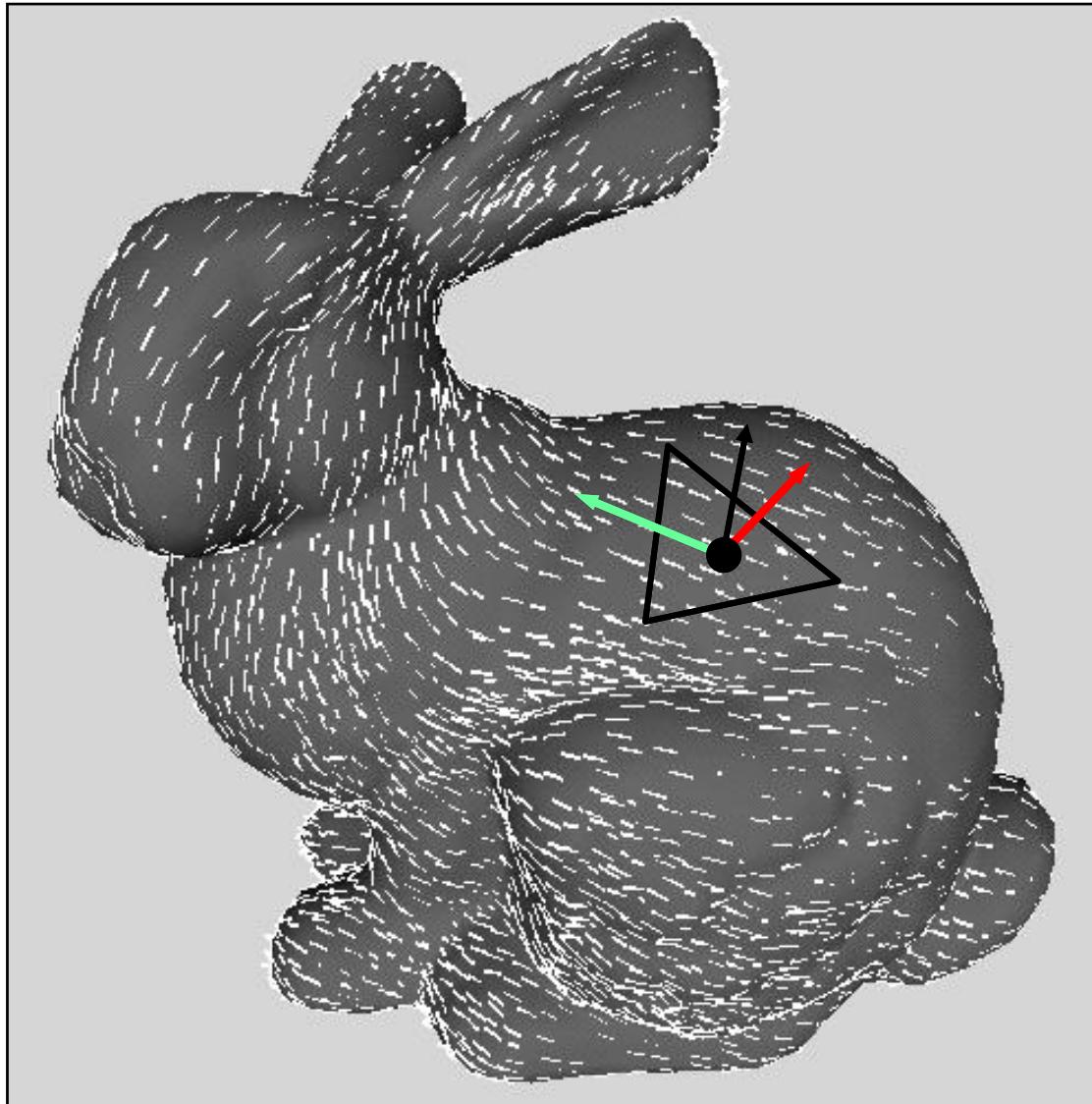
Patch Growth



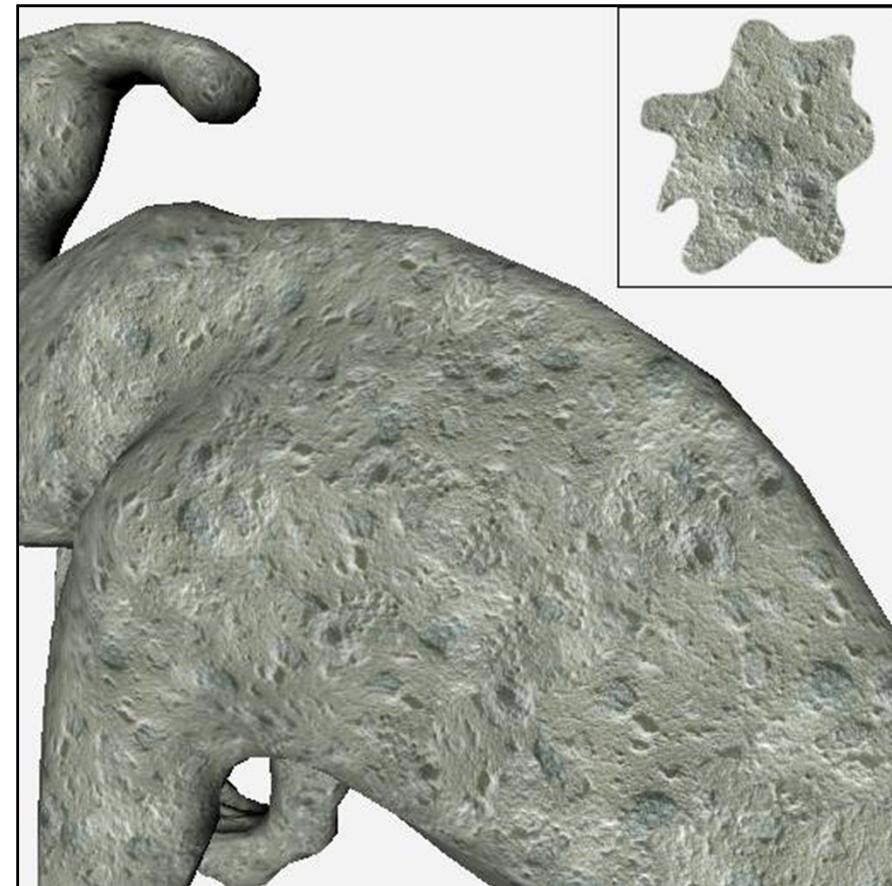
Patch Growth



Tangential Vector Field

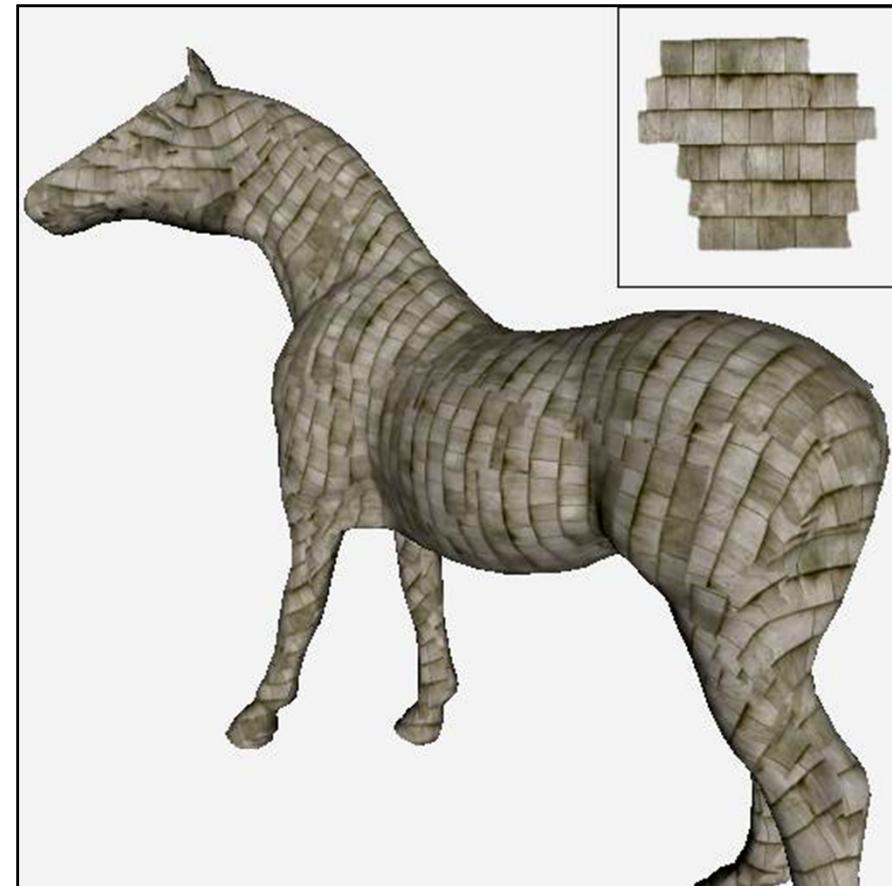
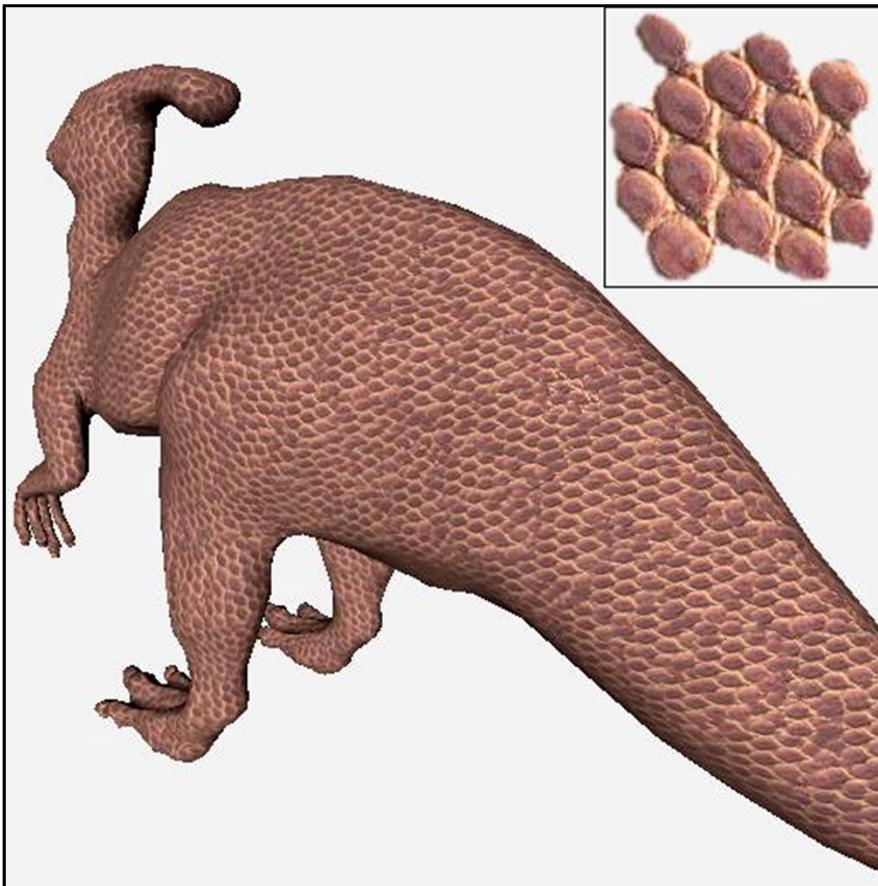


Results: Splotches

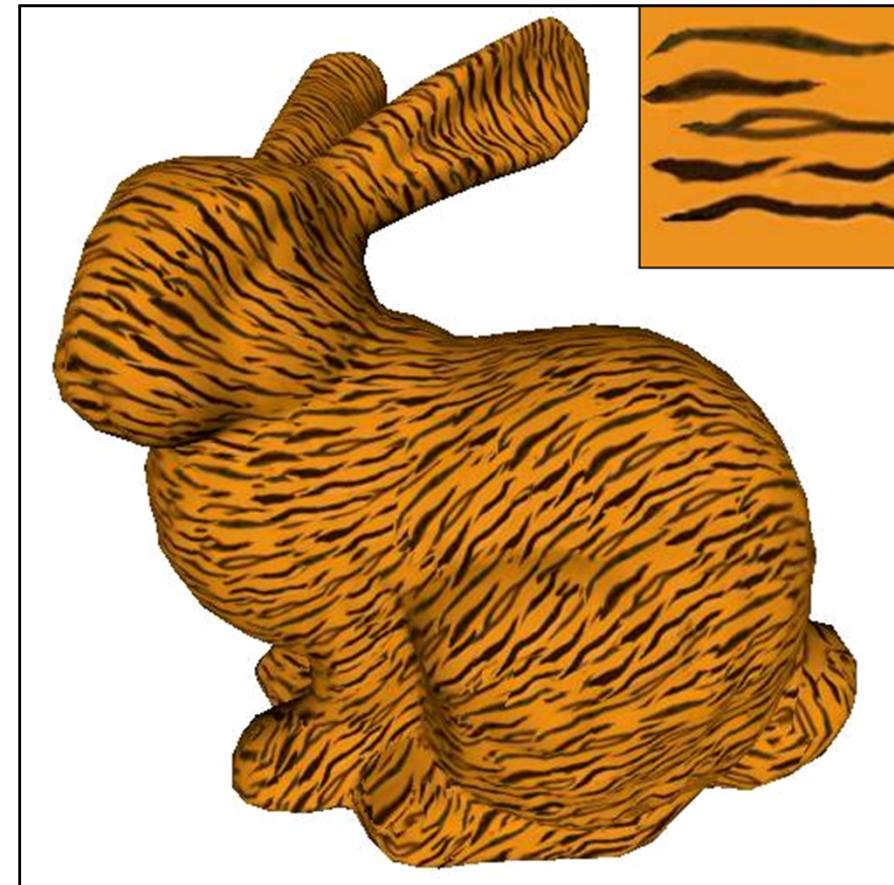
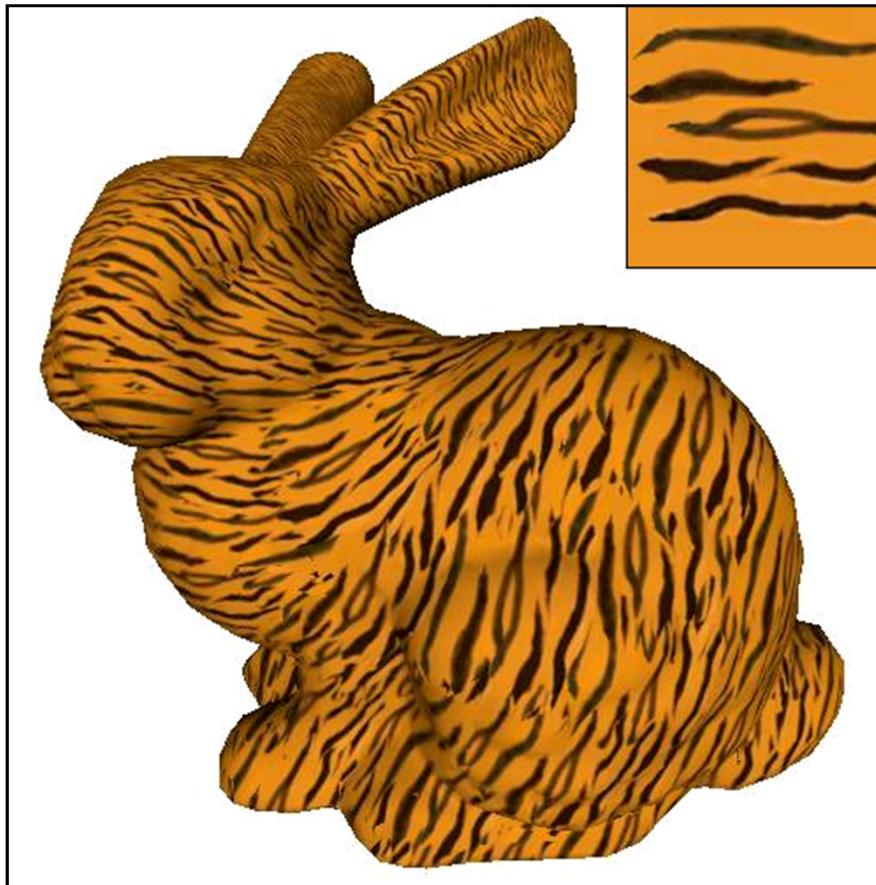


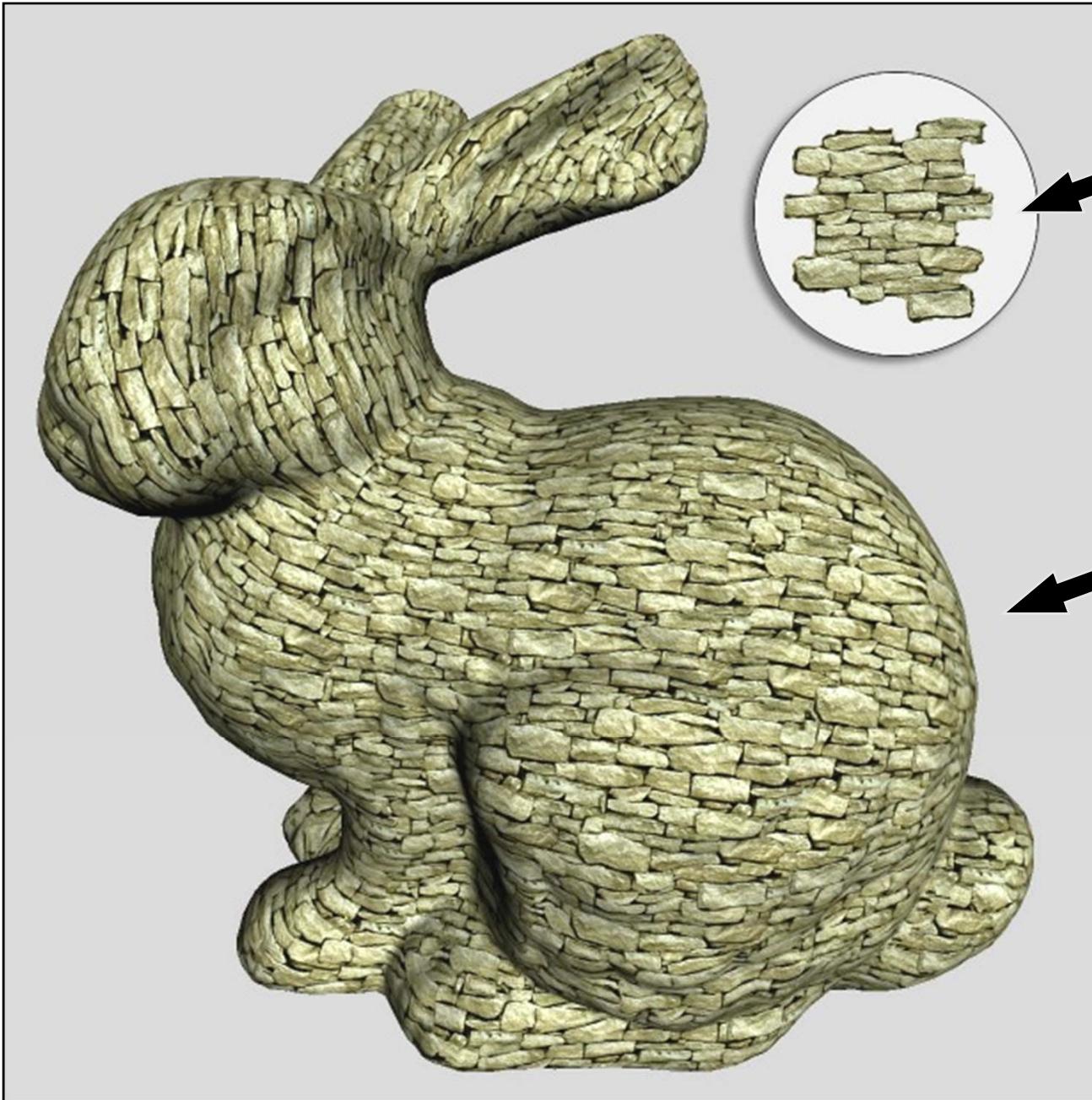
(completely automatic: no direction field)

Results: Anisotropic



Controlling Direction and Scale





256 x 256
texture
(282 times)

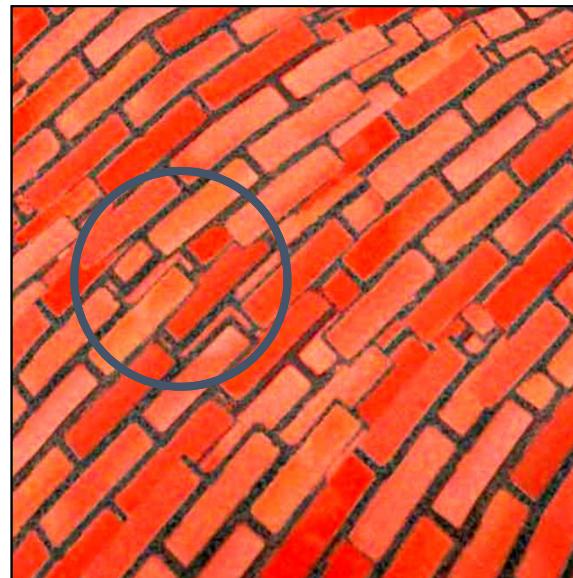
15,000 faces

25 frames
per sec!

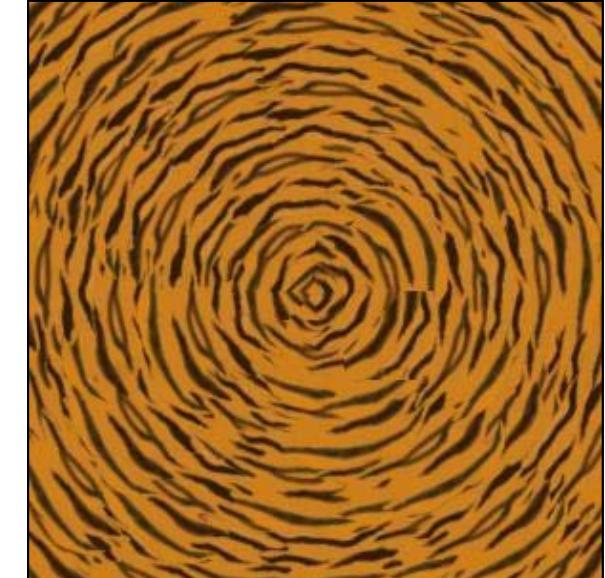
Limitations



low-frequency
components



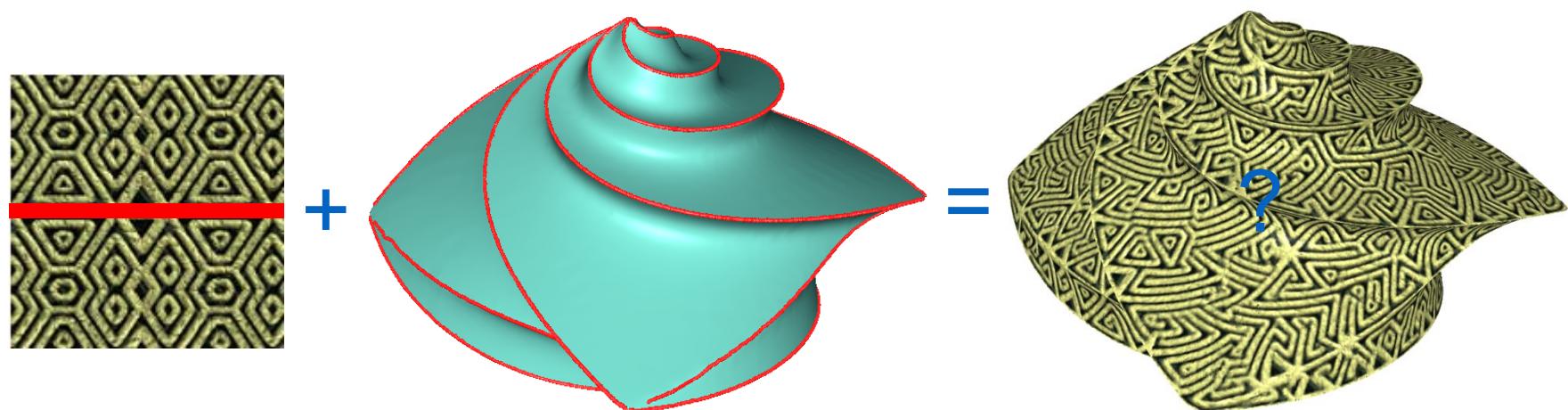
boundary
mismatches



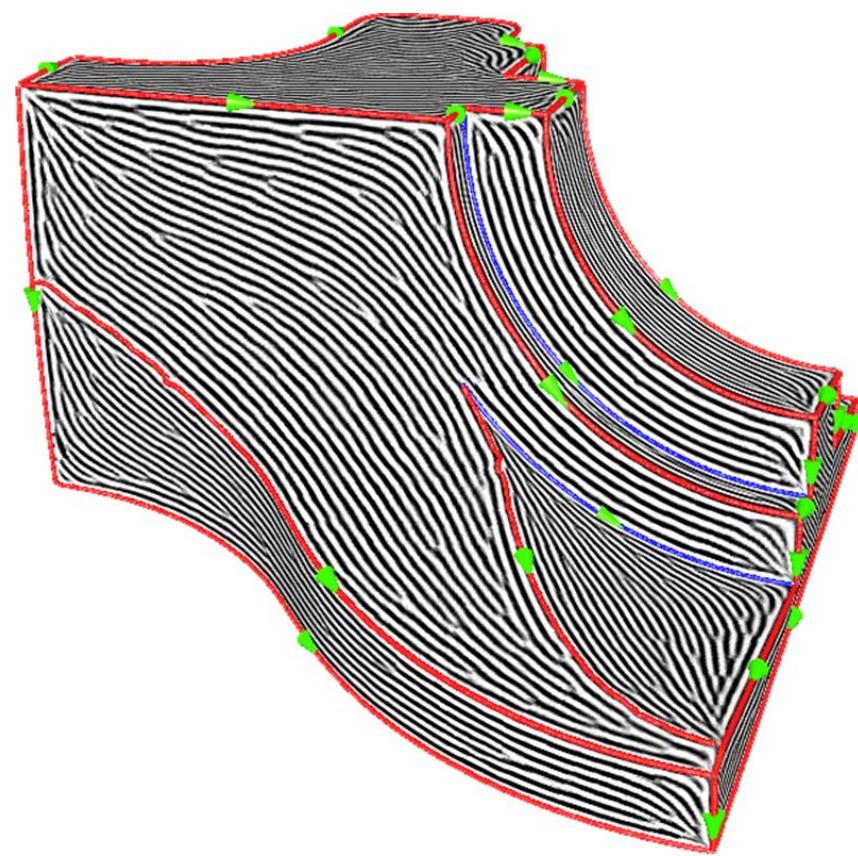
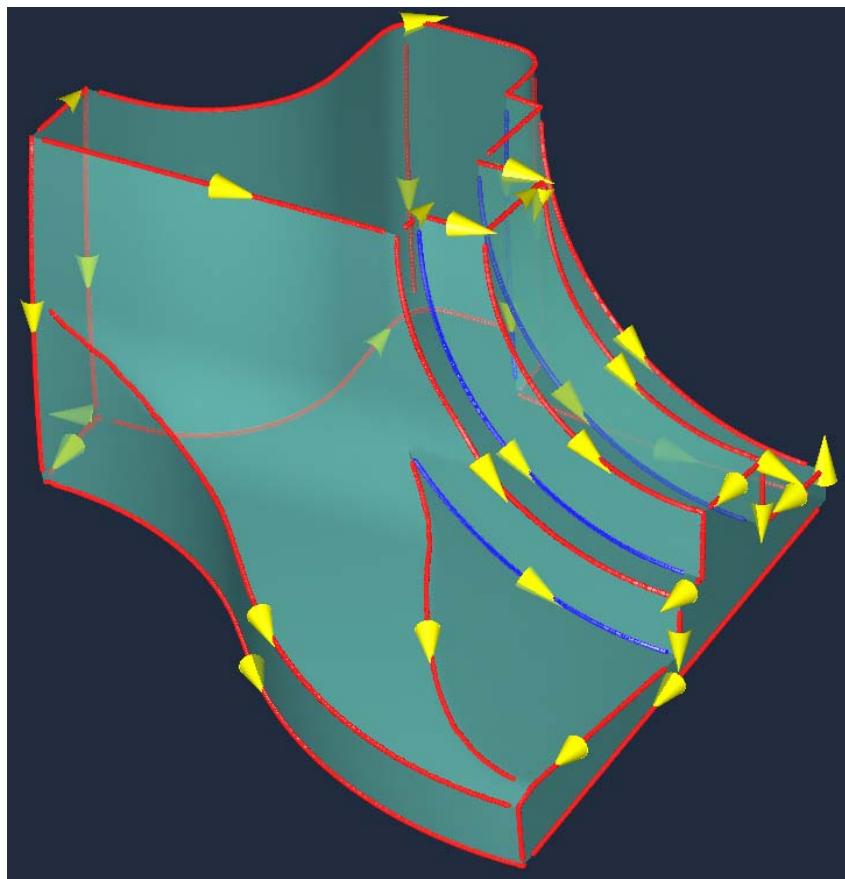
direction field
singularities

Feature-aligned Texture Synthesis

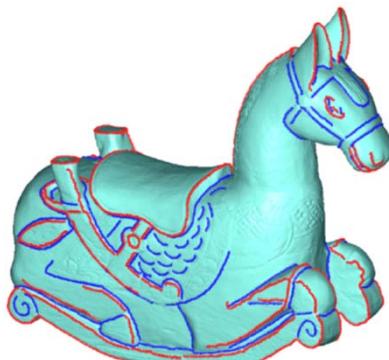
[Xu et al., Siggraph Asia 2009]



Curve Orientation and Vector Field

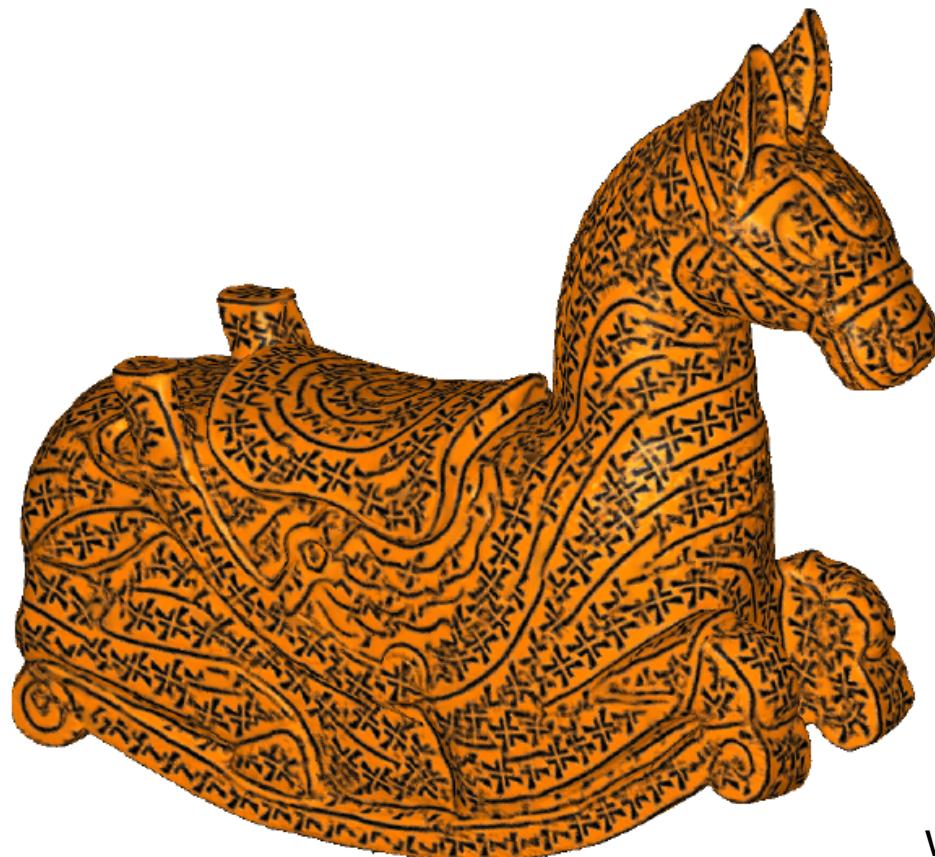
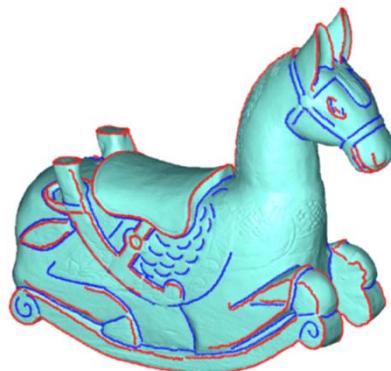


Results



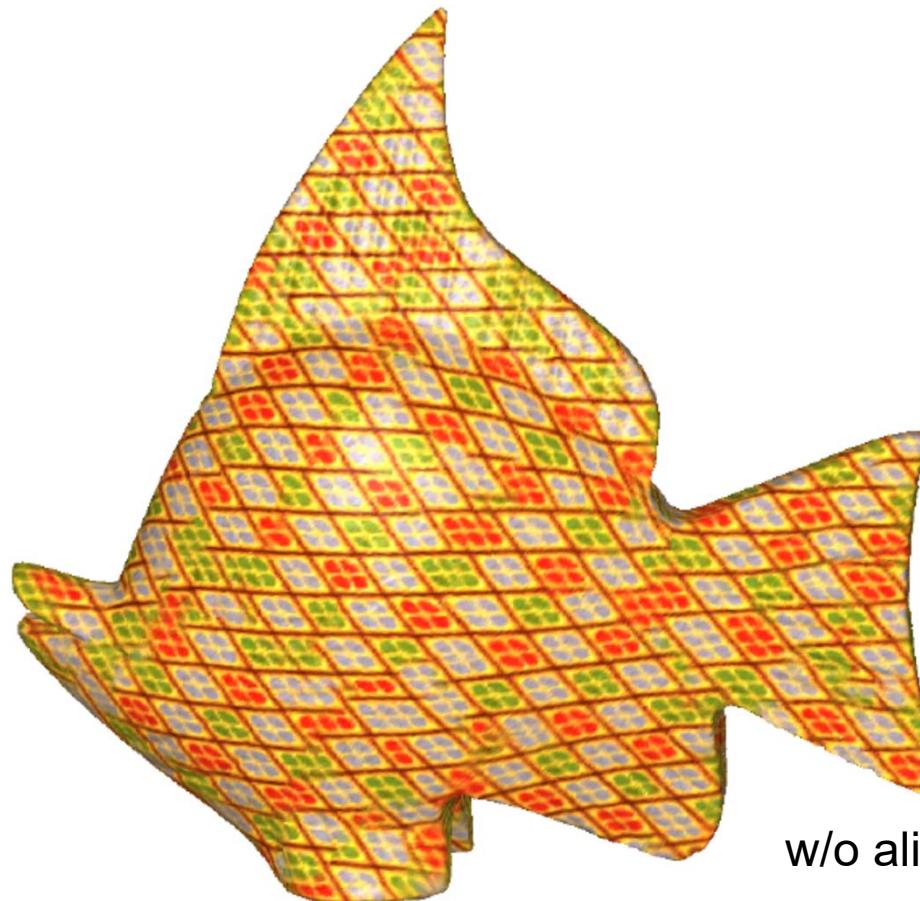
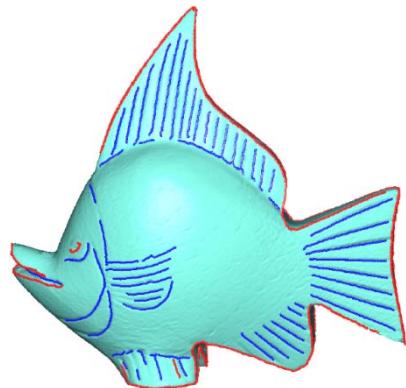
w/o alignment

Results



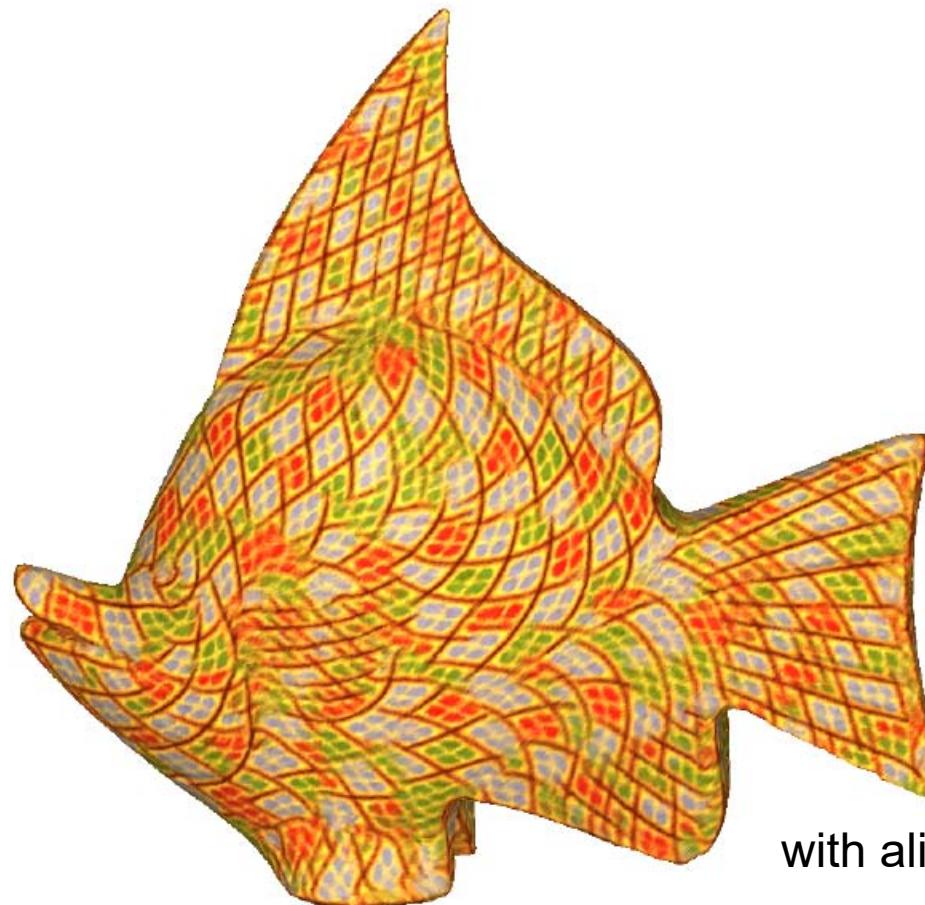
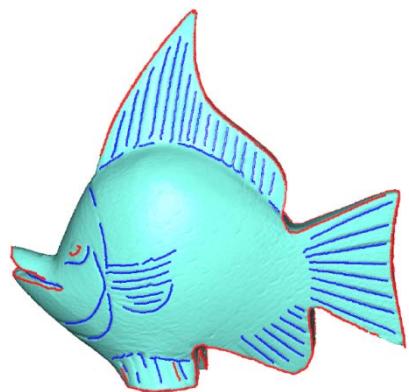
with alignment

Results



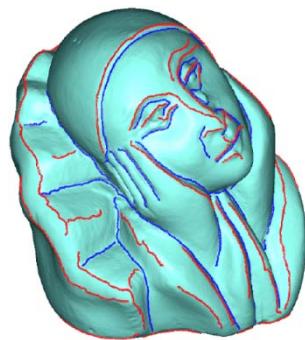
w/o alignment

Results



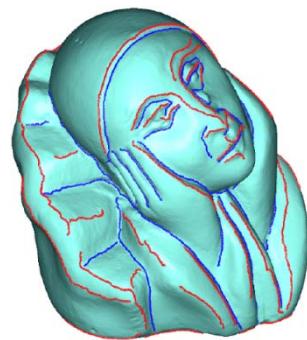
with alignment

Results



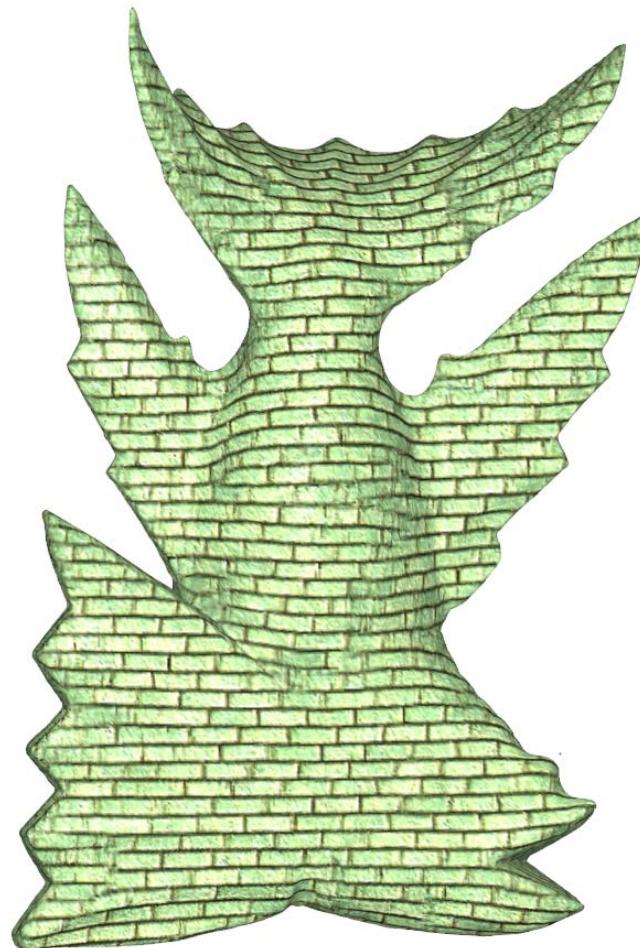
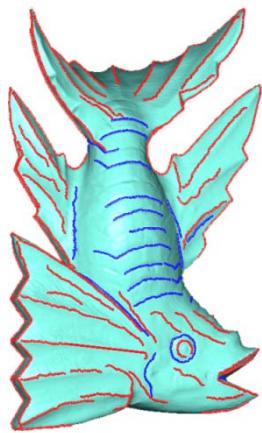
w/o alignment

Results



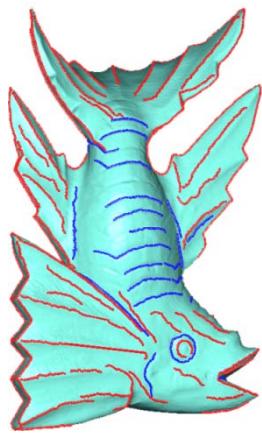
with alignment

Results



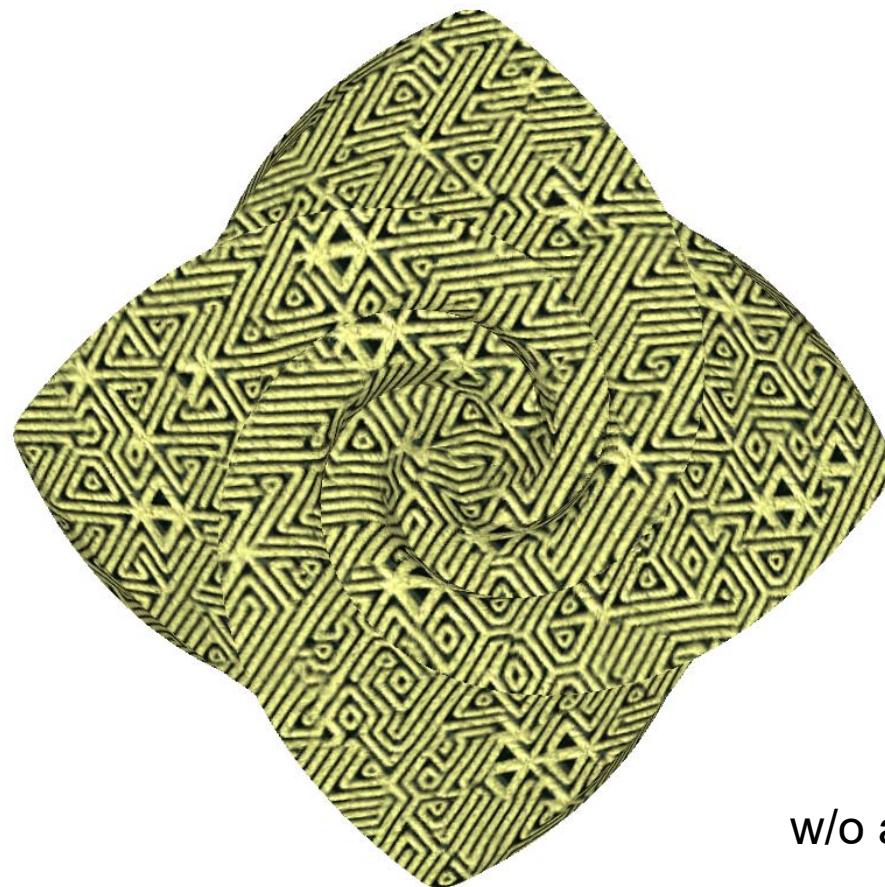
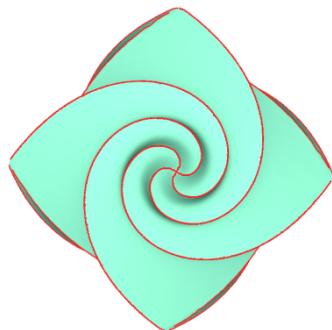
w/o alignment

Results



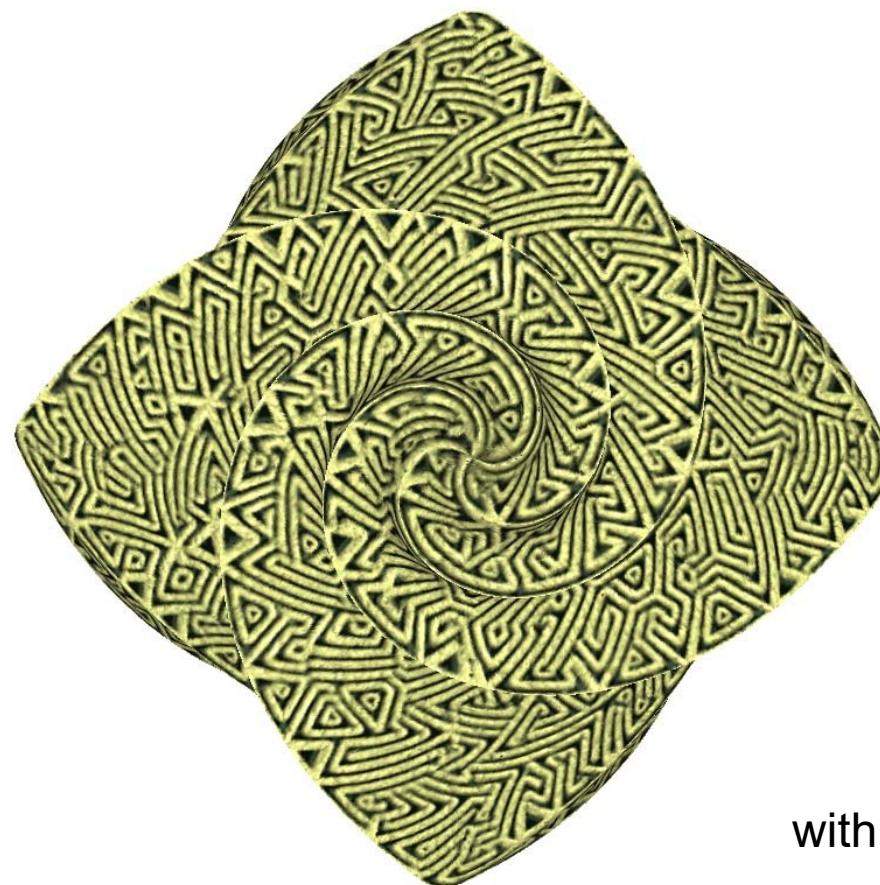
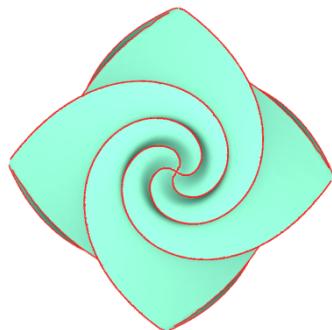
with alignment

Results



w/o alignment

Results



with alignment

Progressively-Variant Texture Synthesis

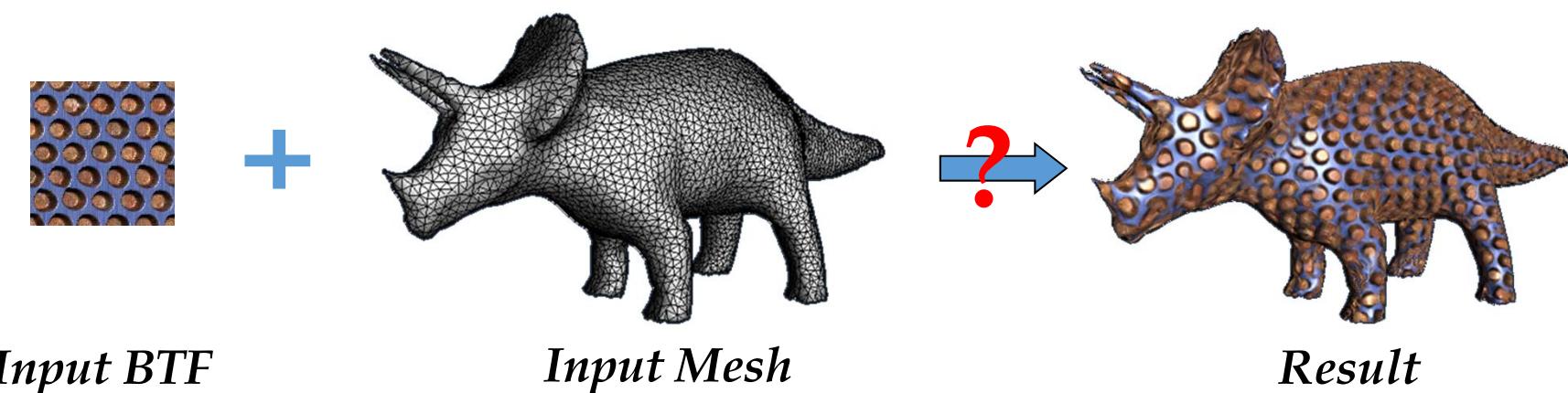
[Zhang et al., Siggraph 2003]



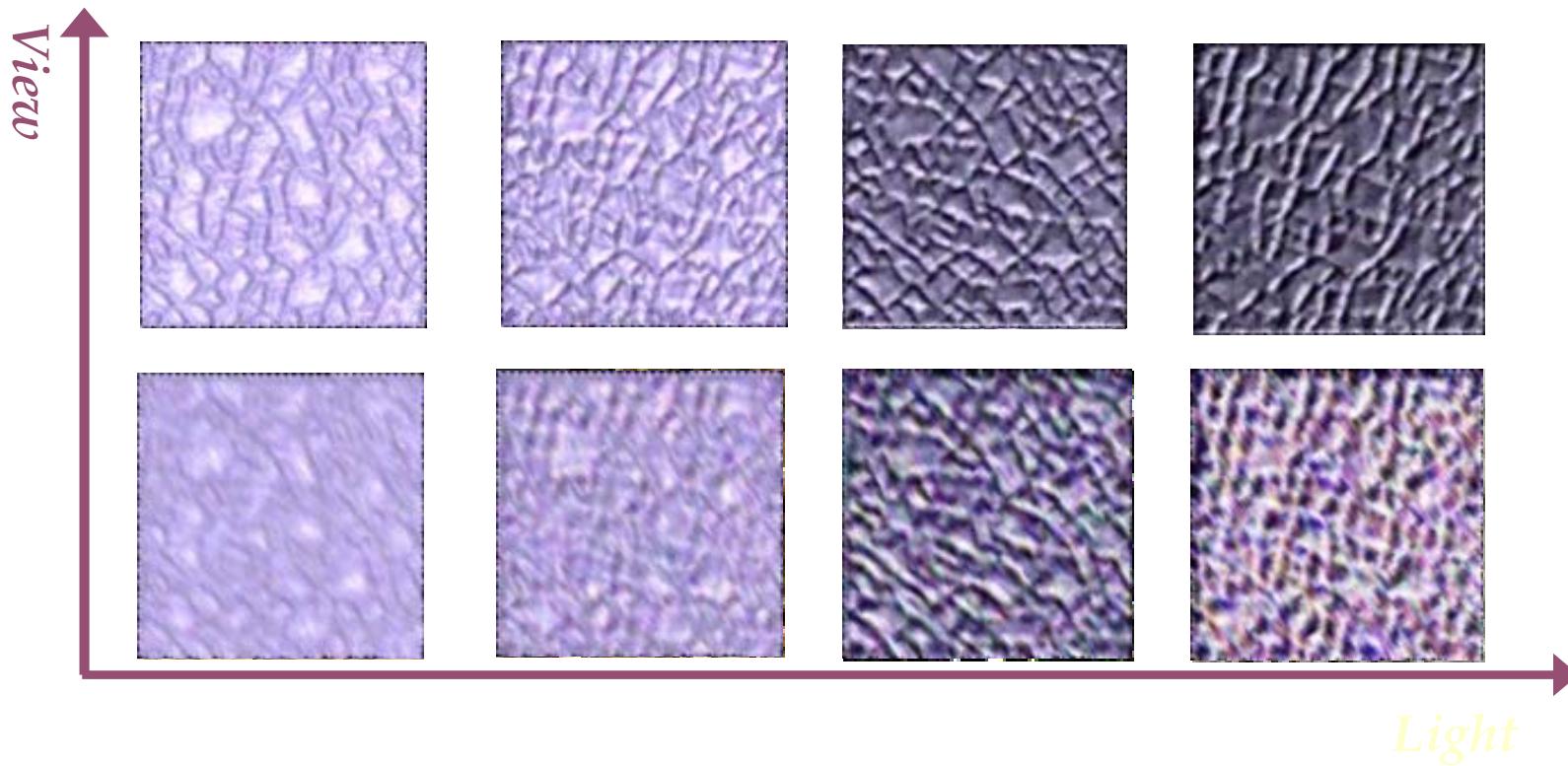
BTF Synthesis

[Tong et al., Siggraph 2002]

- Bidirectional Texture Functions (BTF): A collection of images of the same surface under different lighting and viewing directions.
 - 6D Function ($x, y, l_\theta, l_\varphi, v_\theta, v_\varphi$)
 - Dense Sampling in Viewing/Lighting Directions
 - Capturing Appearance of Real World Surface

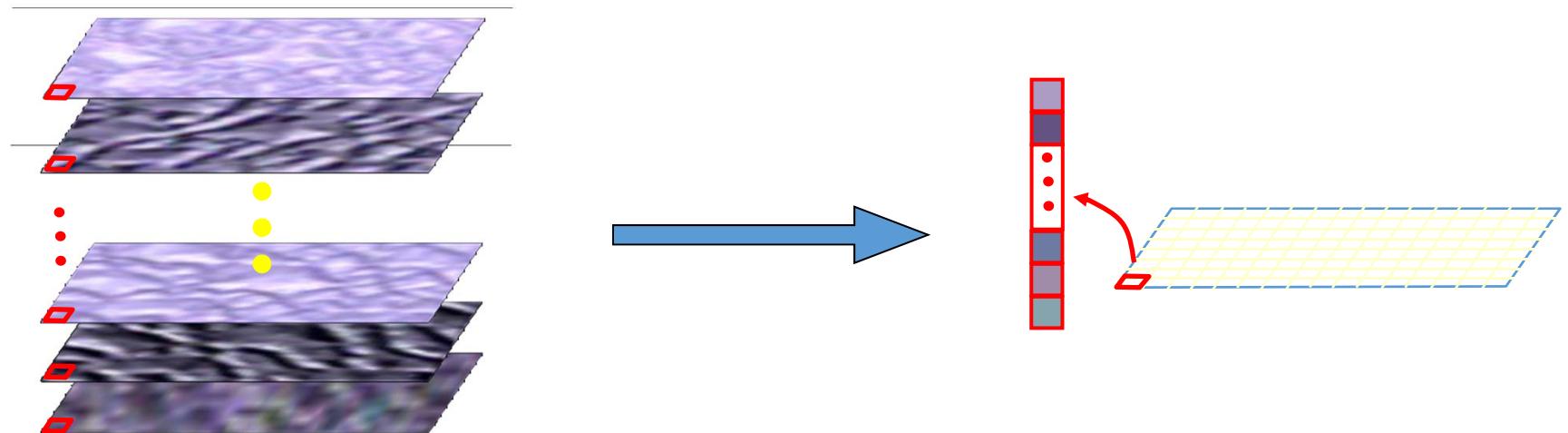


Real World Texture from CuRet



- Geometry Details (Mesostructure) on Surface
- Self-Occlusion, Self-Shadow, and Specularity

Treating BTF as a 2D Texture Map

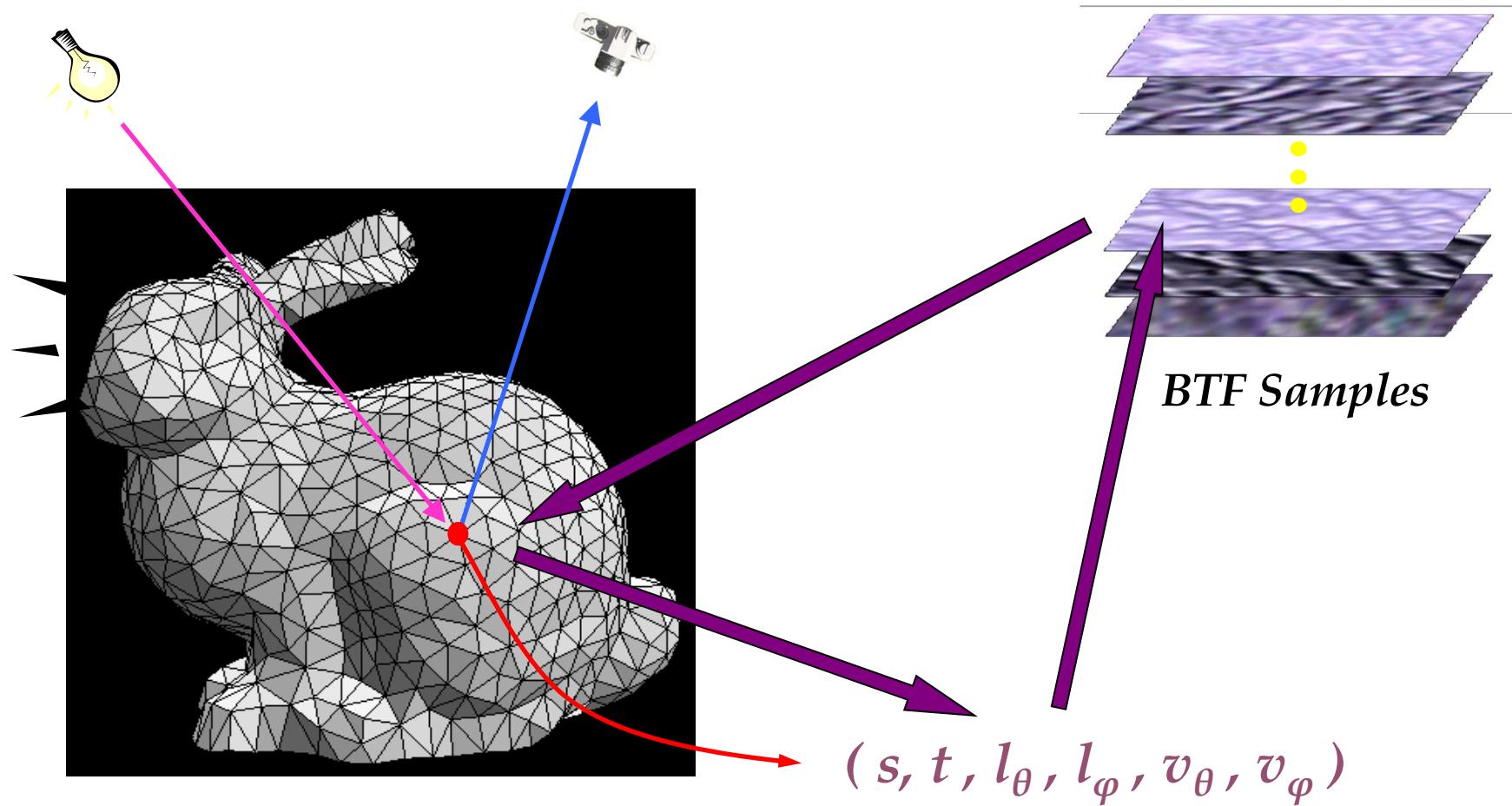


$$12 \times 5 \times 12 \times 5 = 3600$$

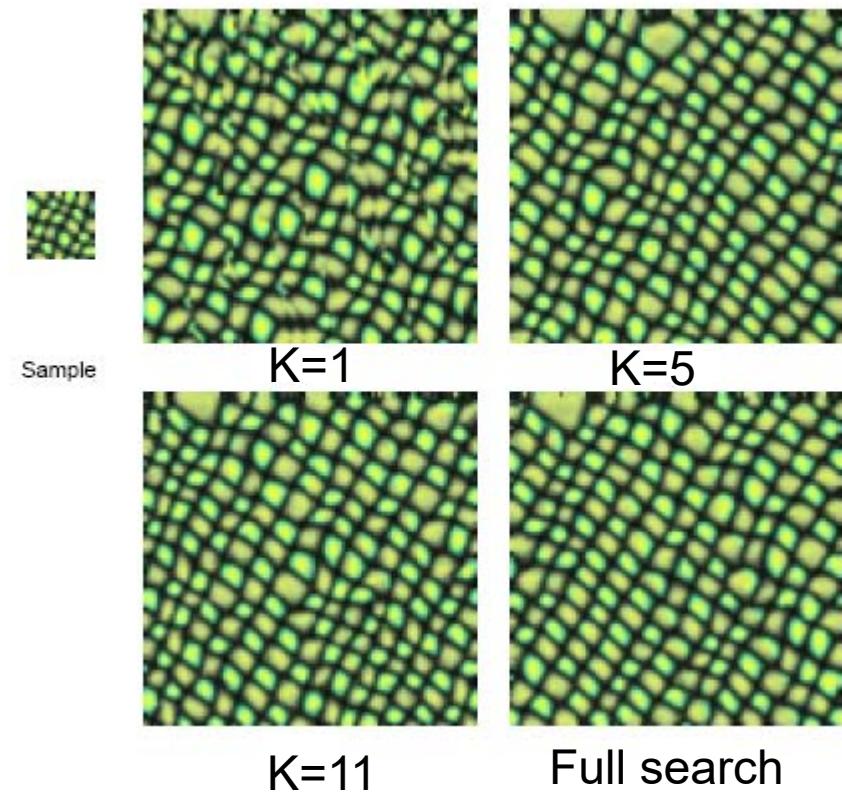
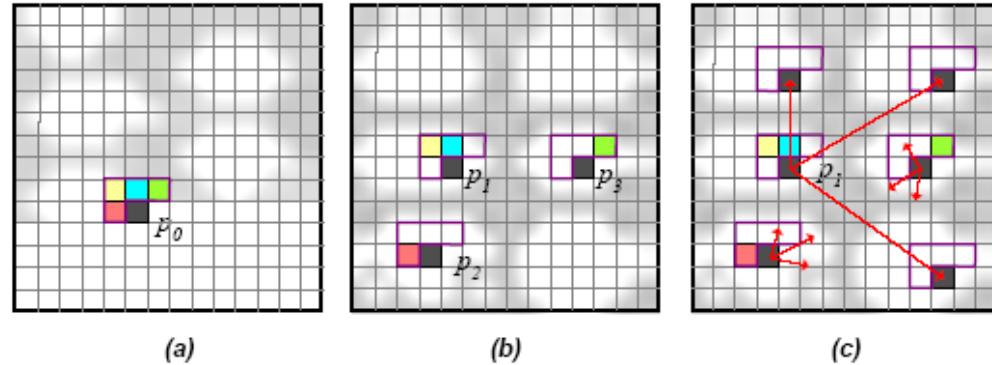
10800-dimensional vector

Surface Texton

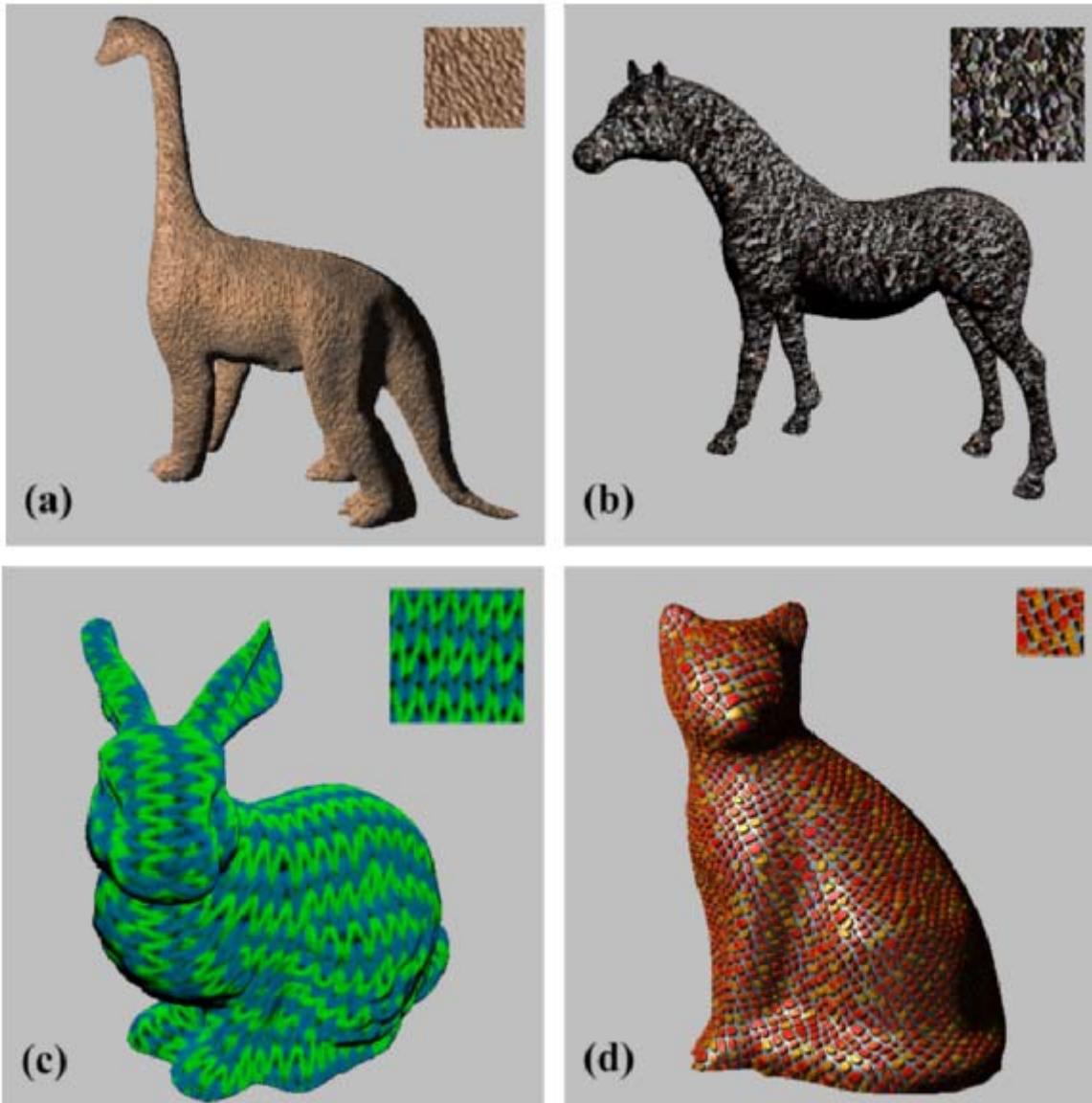
Surface Texton Map & Rendering



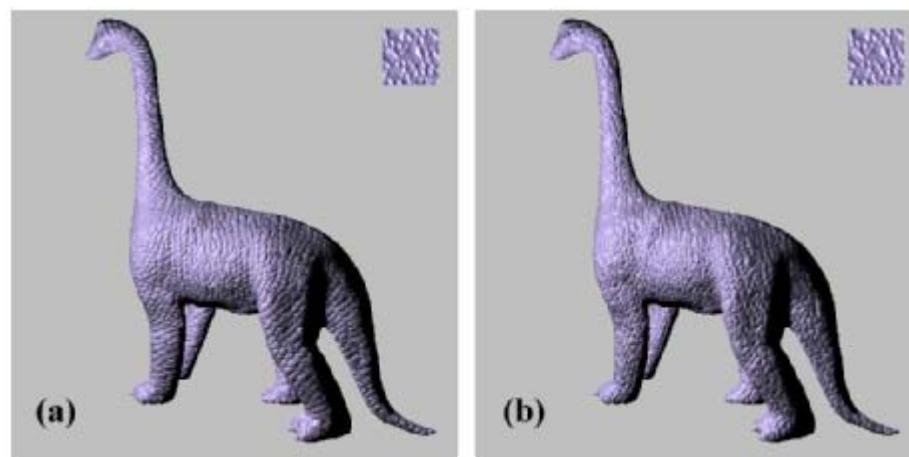
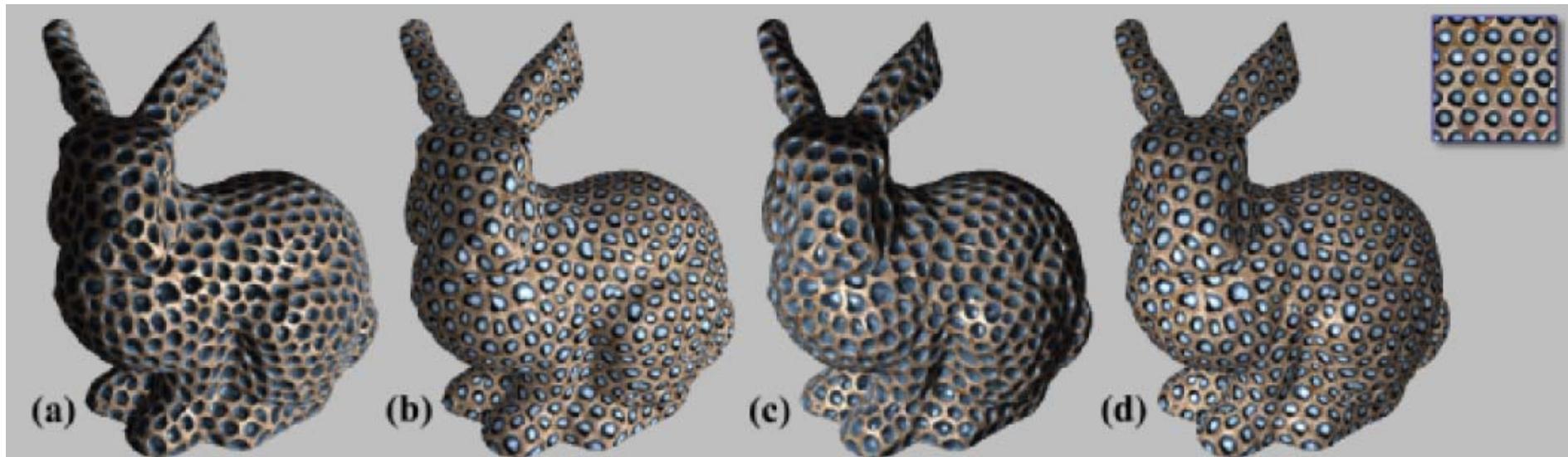
K-Coherent Search



Examples



Comparison



3D Texture Synthesis

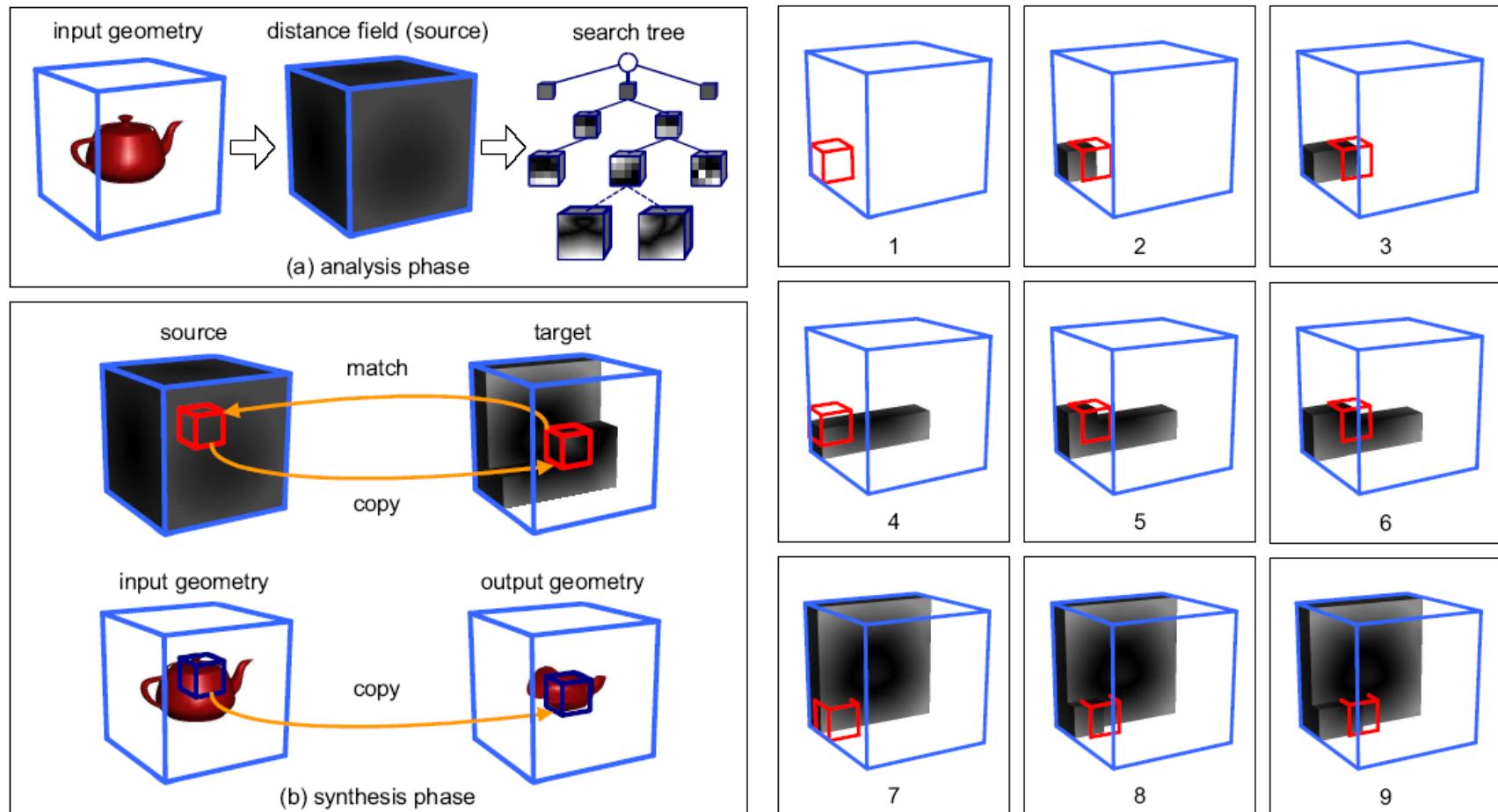
- Procedural texture synthesis
- Sample-based texture synthesis
- **Geometry synthesis**

Geometry Synthesis

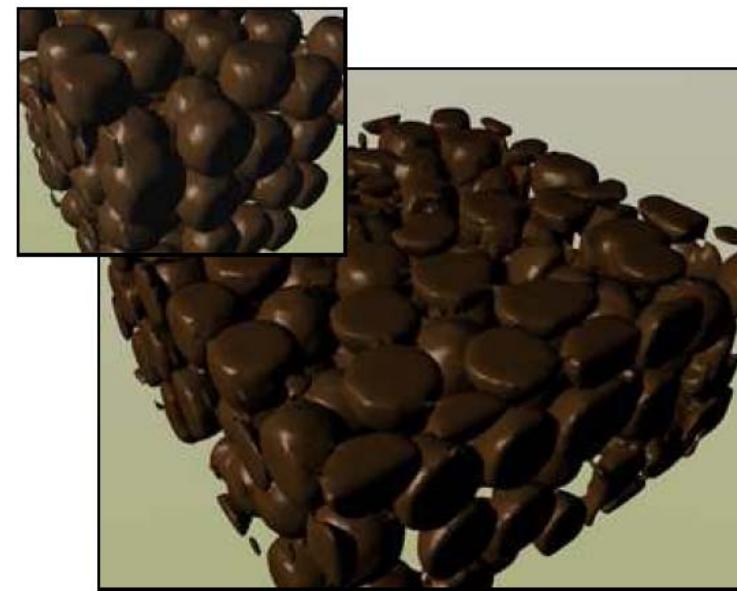
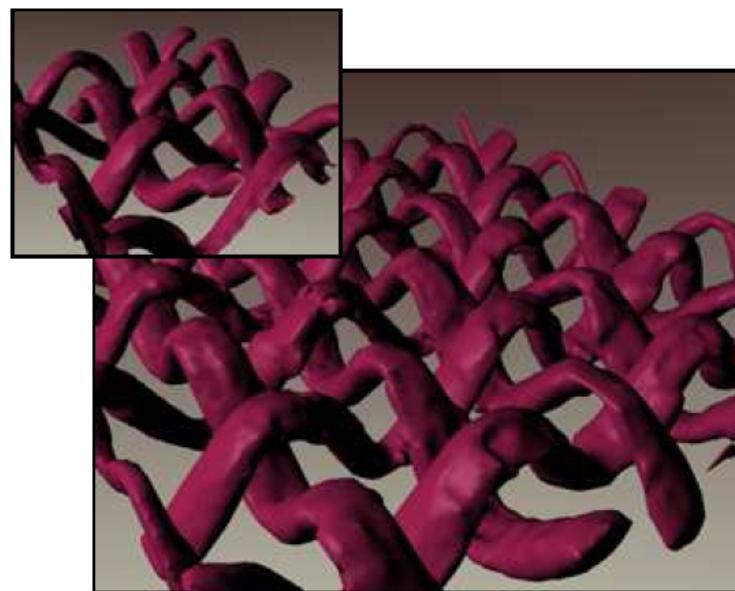
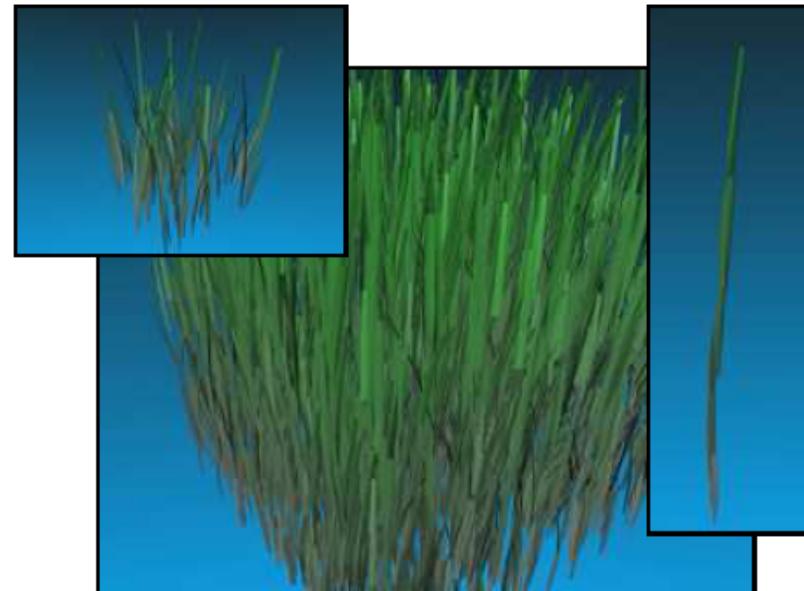
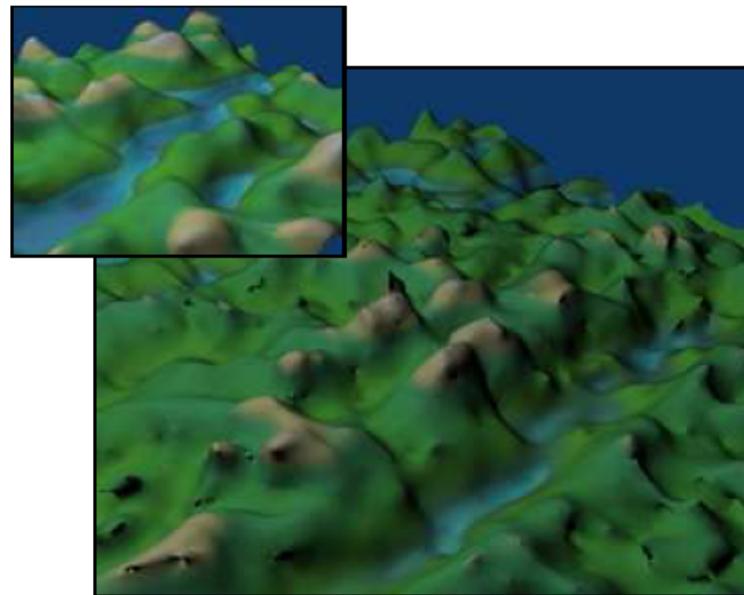
- Generating geometry over surfaces by texture (geometry) samples
- Methods:
 - 3D distance field based method
 - Image analogies extended to volumes
 - Mesh-based geometric texture synthesis technique

3D Distance Field based Method

[Lagae et al., TR 2004]

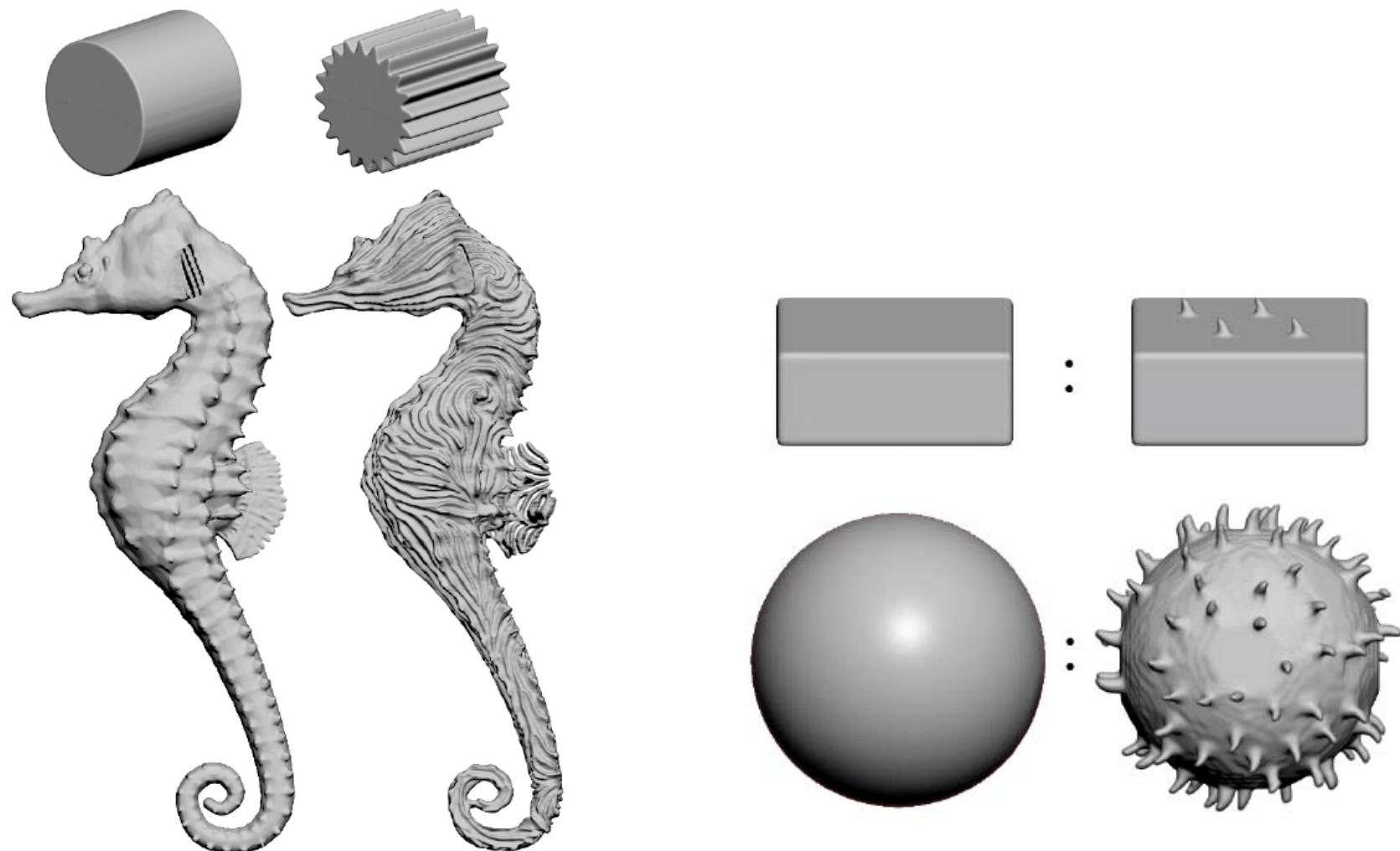


Results



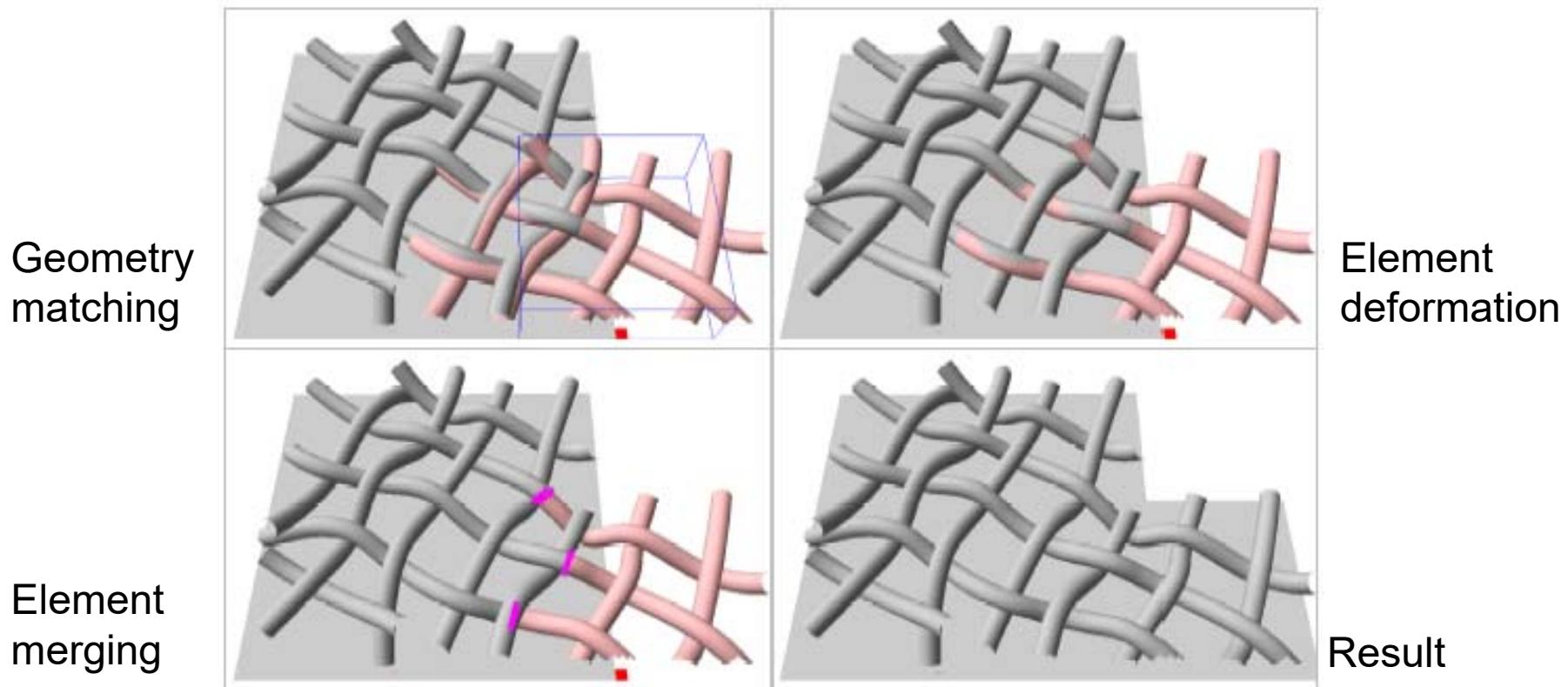
Geometry Analogy

[Bhat et al., SGP 2004]

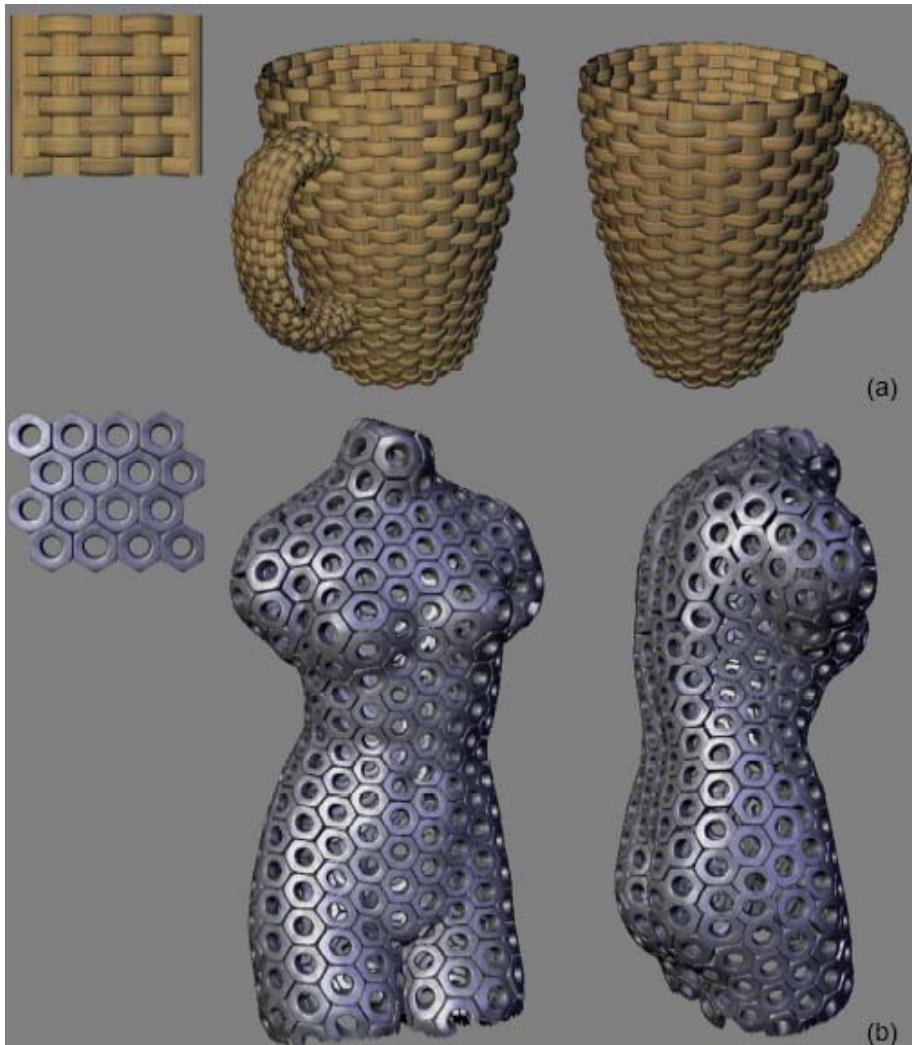
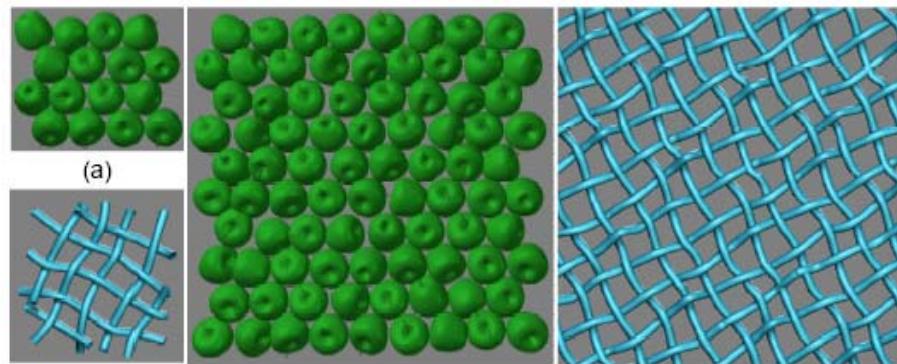


Mesh Quilting

[Zhou et al., Siggraph 2006]



Results



Summary: Texture Synthesis

- An important topic on content generation (2D/3D) textures or repeated geometries
- A well-studied topic
- Many applications





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谢谢！