

Artistic Stylization of Images and Video

Eurographics 2011

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<http://kahlan.eps.surrey.ac.uk/EG2011>

■ Texts



Strothotte &
Schlechtweg
ISBN: 1558607870

Gooch & Gooch
ISBN: 1568811330

Romero & Machado
ISBN: 3540728767

■ Web Bibliographies

<http://video3d.ims.tuwien.ac.at/~stathis/nprlib/index.php>

<http://isgwww.cs.uni-magdeburg.de/~stefans/npr/nprpapers.html>

<http://www.red3d.com/cwr/npr/> (dated)

■ Tutorials

SIGGRAPH 99 (Green et al.) – 2D/3D NPR

SIGGRAPH 02 (Hertzmann) – 2D NPR

SIGGRAPH 03 (Sousa et al.) – 2D/3D NPR

Eurographics 05,06 and...

SIGGRAPH 06 (Sousa et al.) – 3D NPR

SIGGRAPH 10 (McGuire) – 3D NPR for Games

■ Main Publication Forums

NPAR (Symposium on Non-photorealistic Animation)
Held in Annecy even years, at SIGGRAPH odd years.

IEEE Trans Visualization and Comp. Graphics (**TVCG**)

IEEE Computer Graphics and Applications (**CG&A**)

Eurographics and **Computer Graphics Forum**

SIGGRAPH, **SIGGRAPH Asia** and **ACM ToG**

EG Symposium on Rendering (**EGSR**)

ACM/EG Symposium on Computer Animation (**EGSA**)

Non-Photorealistic Rendering (NPR)

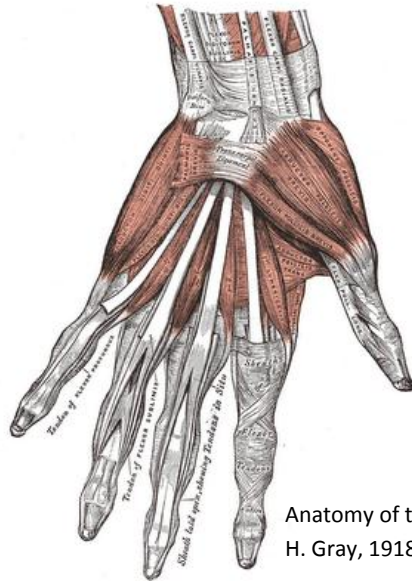
Coined by Salesin et al., 1994

Stylized Rendering

Aesthetic Rendering

Artistic Stylization

Artistic Rendering



Anatomy of the Human Body
H. Gray, 1918



Artistic Stylization

■ Why?

Comprehension

Communication

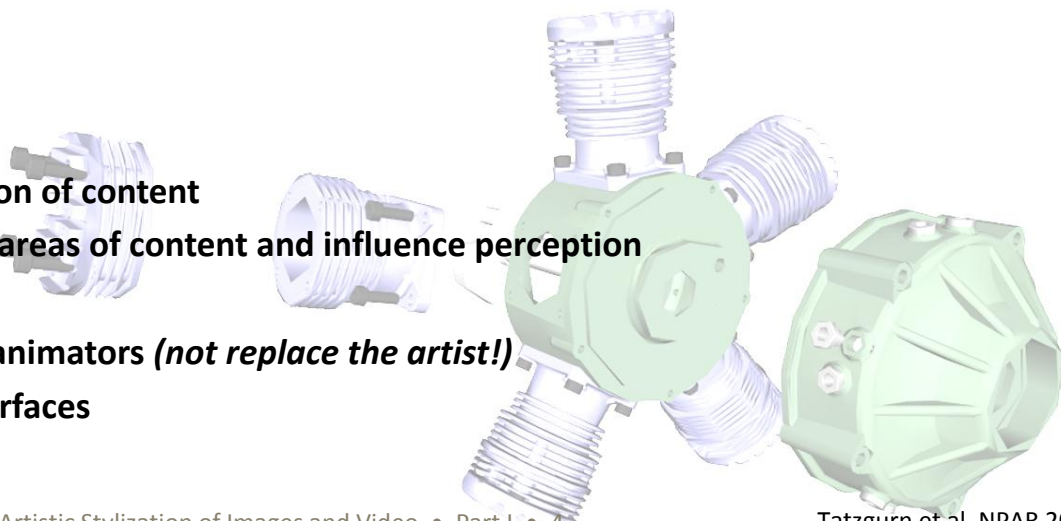
Visualization

Aesthetics

Animation

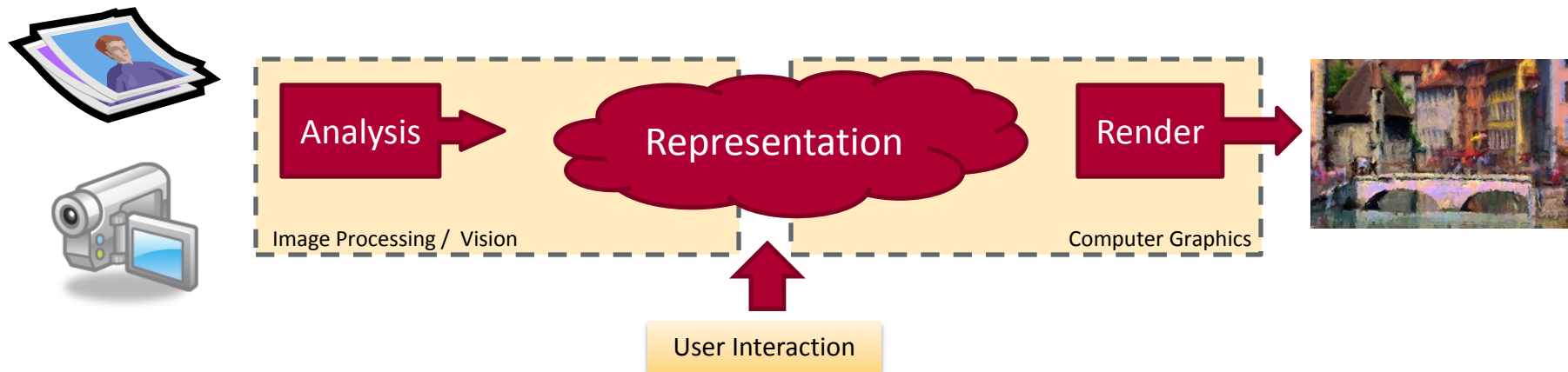
■ Artistic Stylization can

- Simplify and structure the presentation of content
- Selectively guide attention to salient areas of content and influence perception
- Learn and emulate artistic styles
- Provide assistive tools to artists and animators (*not replace the artist!*)
- Help us to design effective visual interfaces

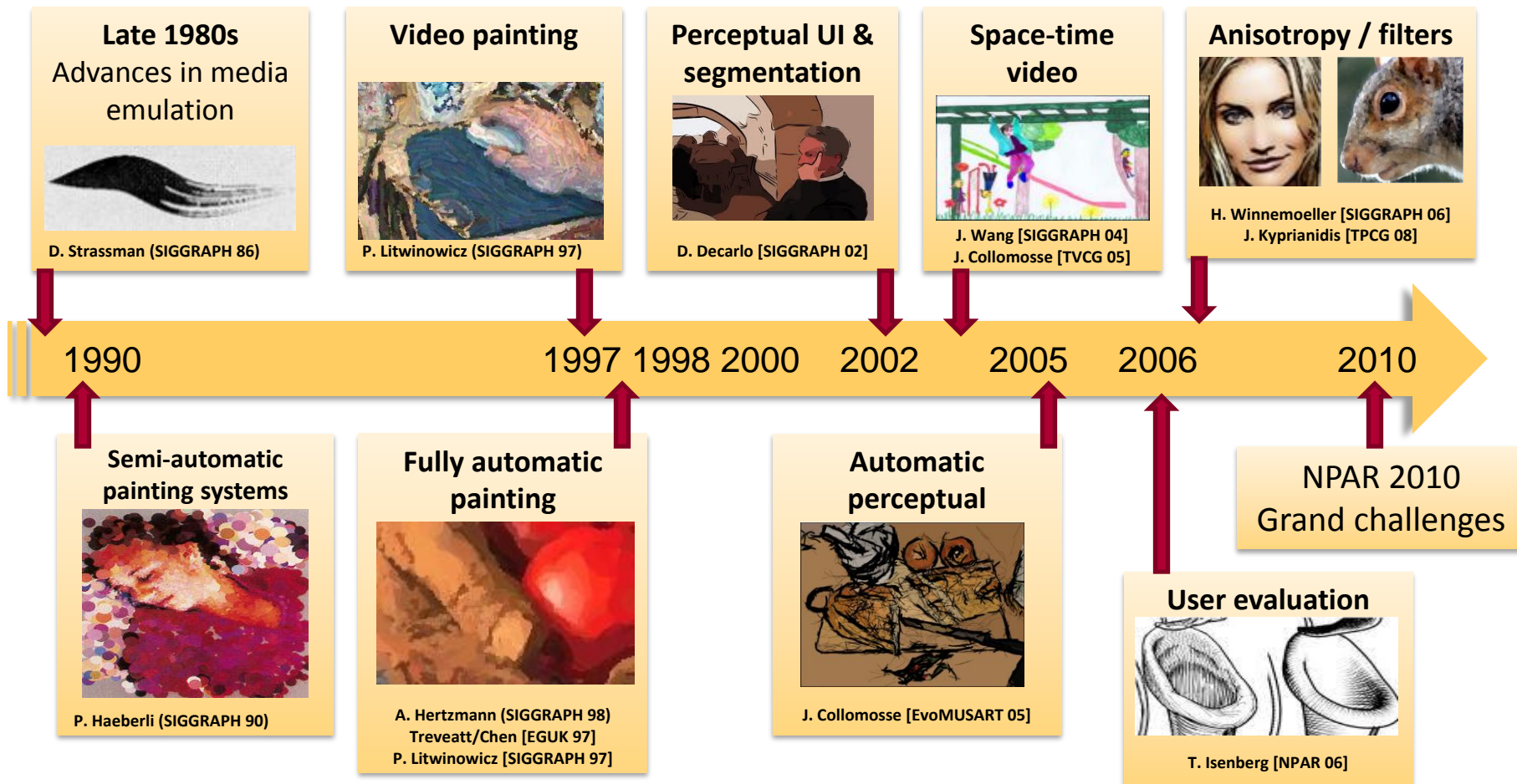


Artistic Stylization

- Rendering real images/video footage in to pseudo-artistic styles
- Convergence of Computer Vision, Graphics (and HCI)



- Visual analysis enables new graphics. Graphical needs motivate new vision.



Rendering process is guided by...

emulation

Perceptual UI &
segmentation

User
subconscious
interaction

Space-time
video

Anisotropy / filters

User conscious interaction

Low-level image
processing

Higher level computer
vision

Direct Anisotropic filtering

1990

1997 1998 2000

2002

2005

2006

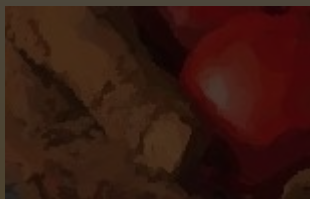
2010

Semi-automatic
painting systems



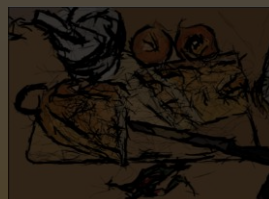
P. Haerberli [SIGGRAPH 90]

Fully automatic
painting



A. Hertzmann [SIGGRAPH 98]
Treveatt/Chen [EGUK 97]
P. Litwinowicz [SIGGRAPH 97]

Automatic
perceptual



J. Collomosse [EvoMUSART 05]

User evaluation



T. Isenberg [NPAR 06]

NPAR 2010
Grand challenges

Rendering process is guided by...

emulation

Perceptual UI &
segmentation

User
subconscious
interaction

Part III: Anisotropy and Filtering
(70 min)

User conscious interaction

Low-level image
processing

Higher level computer
vision

Image filtering

BREAK!

1990

1997 1998 2000 2002

2010

Part I: Classical algorithms (30 min)

Part IV: Future
Challenges (20 min)

Part II: Vision for Stylisation (60 min)

P. Haerberli [SIGGRAPH 90]

A. Hertzmann [SIGGRAPH 98]
Treveatt/Chen [EGUK 97]
P. Litwinowicz [SIGGRAPH 97]

J. Collomosse [EvoMUSART 05]

T. Isenberg [NPAR 06]

Artistic Stylization of Images and Video

Part I – Classical Algorithms /
Stroke Based Rendering
Eurographics 2011

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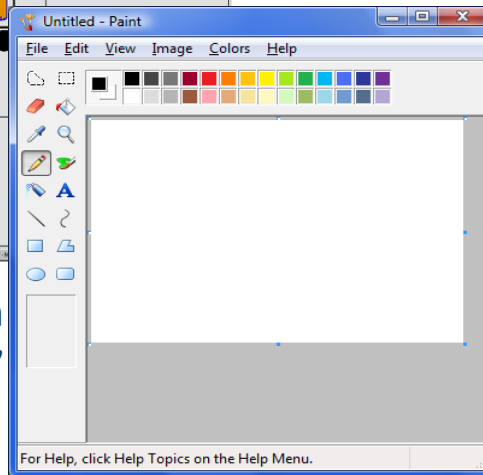
- **Paint by numbers: Abstract image representations**
P. Haeberli, SIGGRAPH 1990
- **Almost Automatic Computer Painting**
P. Haggerty, IEEE CG & A 1991
- **Orientable Textures for Image based Pen-and-Ink Illustration**
D. Salisbury et al., SIGGRAPH 1997
- **Processing images and video for an impressionist effect**
P. Litwinowicz, SIGGRAPH 1997
- **Statistical techniques for the automated synthesis of non-photorealistic images**
S. Treavett and M. Chen, Eurographics UK 1997.
- **Automatic Painting based on Local Source Image Approximation**
Shiraishi and Yamaguchi, NPAR 2000.
- **Painterly Rendering with Curved Strokes of Multiple Sizes**
A. Hertzmann, SIGGRAPH 1998.
- **Paint by Relaxation**
A. Hertzmann, CGI 2001
- **Fast Paint Texture**
A. Hertzmann, NPAR 2002

- Early painting systems lacked appropriate UI for rich digital painting

Xerox superpaint (1980s)

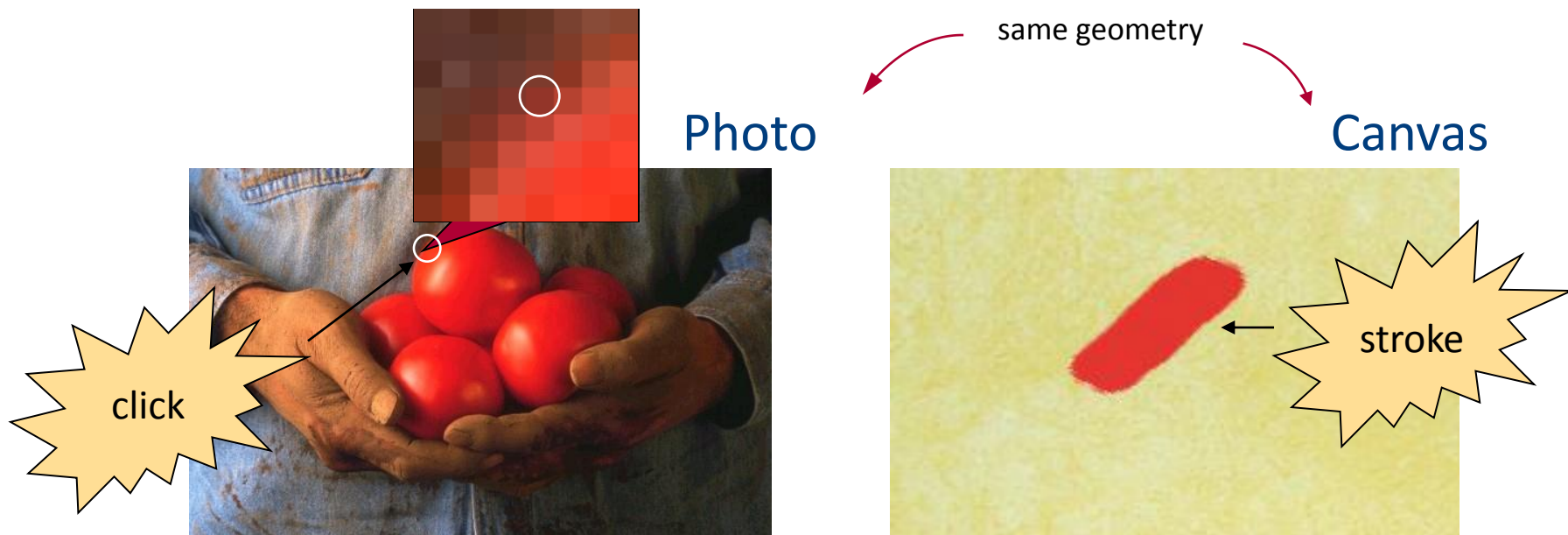


Windows Vista Paint 2007





- **Stroke based rendering (SBR)**
- **Painting is a manually ordered list of strokes, placed interactively.**
- **Stroke attributes sampled from the photo.**



- Stroke colour and orientation are sampled from the source image
- Stroke order and scale are user-selected
- Addition of RGB noise generates an impressionist effect

Photo credit: Haeberli '90.



Paintings with / without orientable strokes



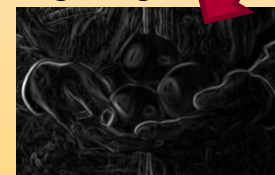
Sobel Edge
detection

-1	-2	-1
0		0
1	2	1

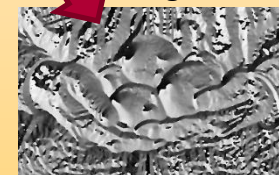
$$E(I) = \left[\left(\frac{\partial I}{\partial x} \right)^2 + \left(\frac{\partial I}{\partial y} \right)^2 \right]^{\frac{1}{2}}$$

$$\theta(I) = \arctan \left(\frac{\partial I}{\partial y} / \frac{\partial I}{\partial x} \right)$$

Edge Mag.



Edge orient.



Orientation

- **More stylised orientation effects with a manually defined orientation field**



Orientation field



Painterly Rendering



Paint by numbers: Abstract Image Representations

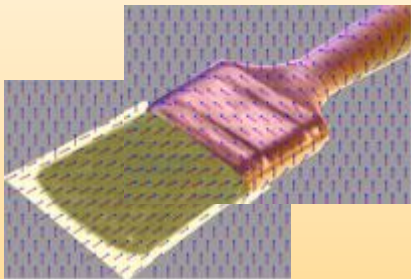
Haeberli. (1990)



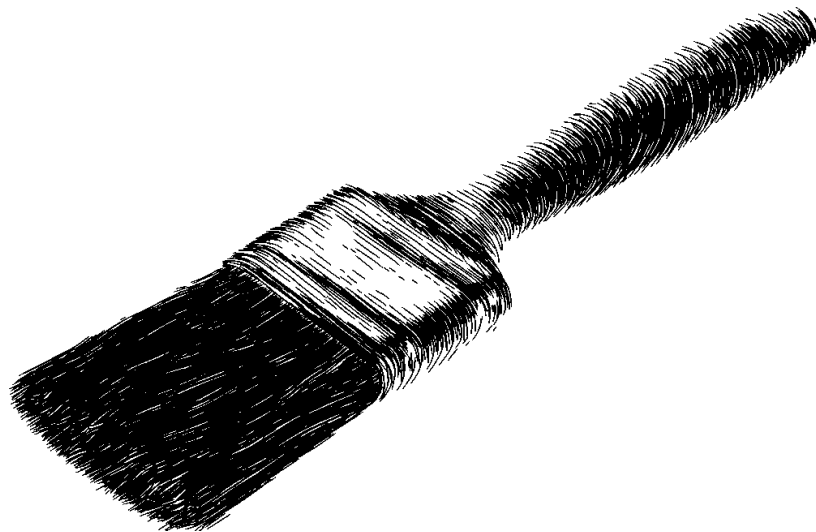
All tutorial code at <http://kahlan.eps.surrey.ac.uk/EG2011>



- Very similar system for pen-and-ink rendering of photos
- User defined orientation field.
 - Regions manually drawn and marked up with orientation
- Stroke (line) placement automatic. Strokes clipped to keep within regions.



**Manually defining regions of
the orientation field**



- Stroke colour and orientation are sampled from the source image
 - Stroke order and scale ~~are user selected~~
 - Scale sampled from Sobel edge magnitude
 - Regularly place strokes. Order of strokes randomly generated
- } Fully automated

Photo credit: Haeberli '90.

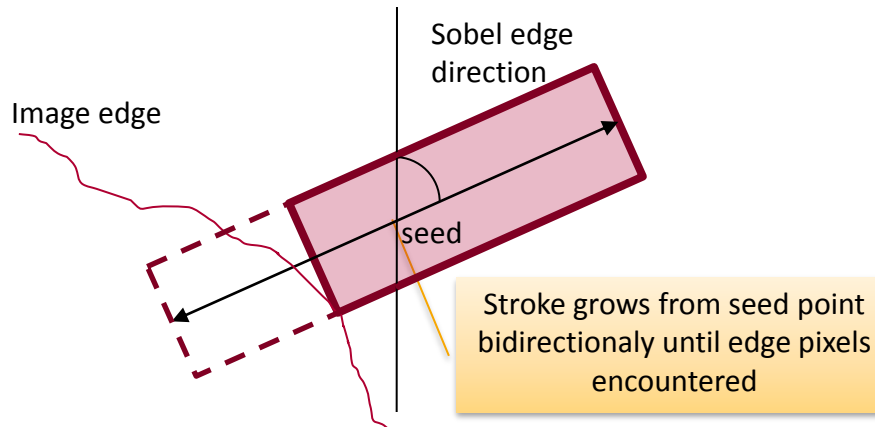


Interactive (Haeberli)



**Loss of detail
in important
regions**

Pseudo-random (as Haggerty)



No clipping



Clipping





- **Common recipe for SBR in the 1990s**
 - Sobel edge detection on blurred image
 - Regular seeding of strokes on canvas
 - Scale strokes inverse to edge magnitude
 - Orient strokes along edge tangent
 - Place strokes in a specific way using this data
- **An interesting alternative uses 2nd order moments within local window to orient strokes.**
 - Extended to multi-scale strokes by Shiraishi and Yamaguchi (NPAR 2000)





- **2D zero-moments for greyscale image $I(x,y)$**

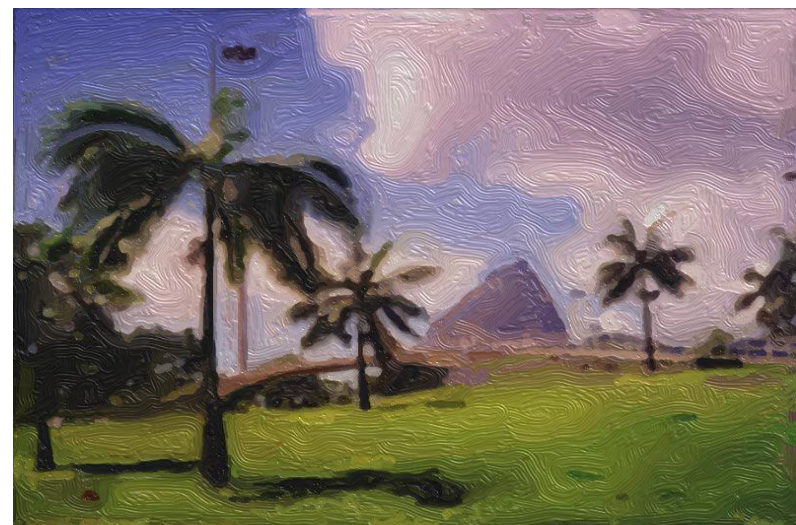
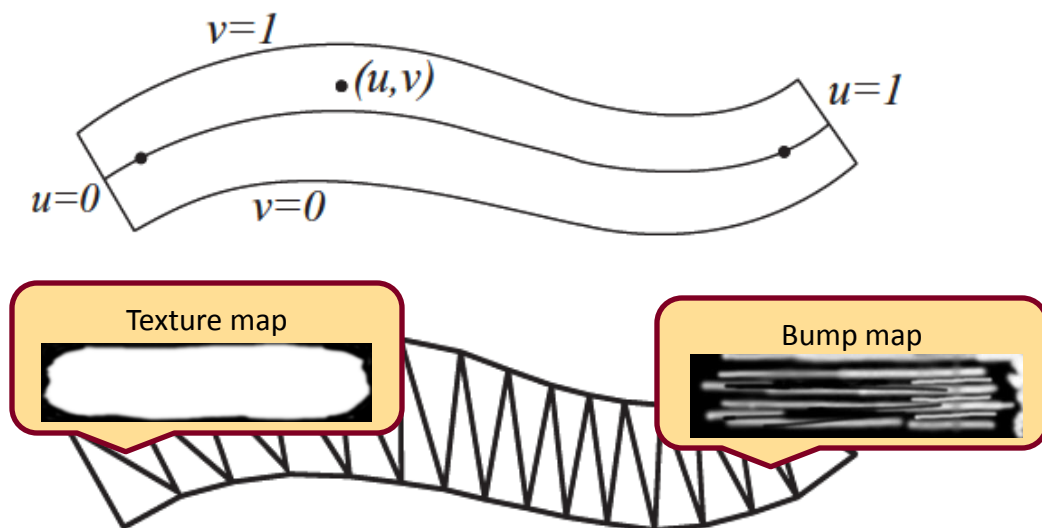
$$M_{lm} = \sum \sum x^l y^m I(x,y).$$



- The canvas is built up in layers from coarse to fine
 - Analysis window scale, and stroke scale are varied in proportion



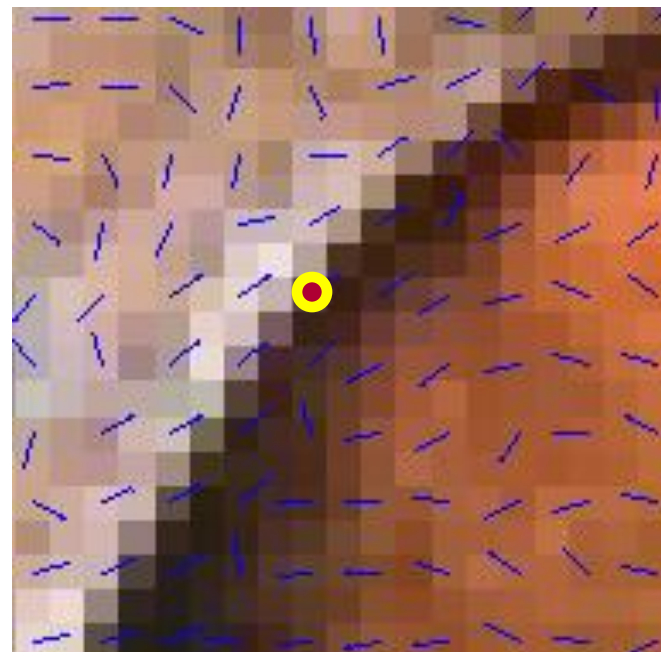
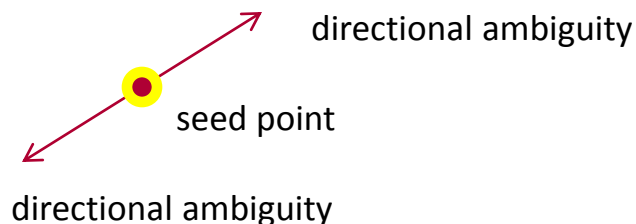
- Artists do not paint with uniformly shaped short strokes (pointillism excepted!)
- Two key contributions (1998)
 - Multi-layer (coarse to fine) painting
 - Painting using β -spline strokes
- Spline strokes can be bump mapped for an improved painterly look (NPAR 2002)





- Greedy algorithm for stroke placement
- Regularly sample the canvas to seed strokes
- Build a list of control points for each stroke by “hopping” between pixels*

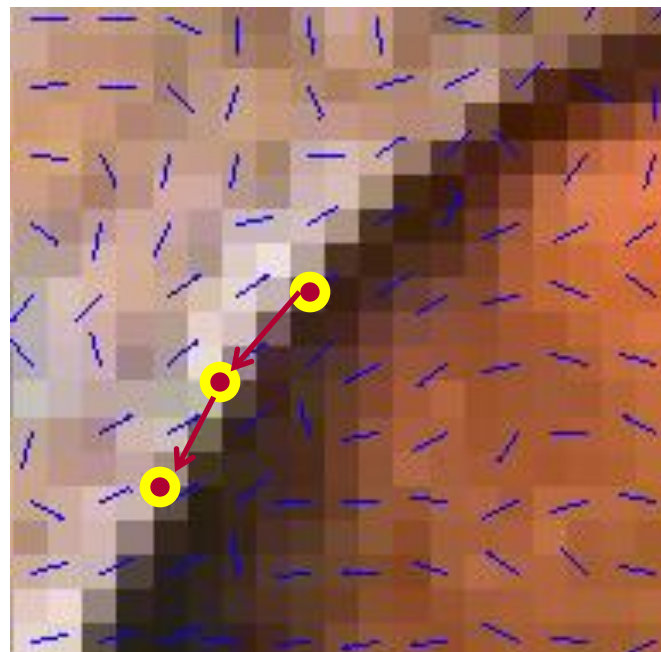
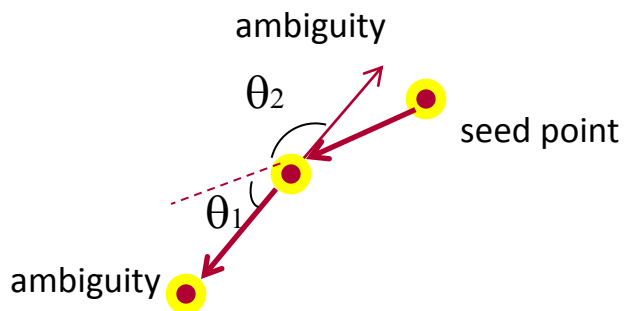
- 1) Pick a direction arbitrarily
(some implementations explore both)



* In practice, best to use float coordinates and interpolate edge orientation

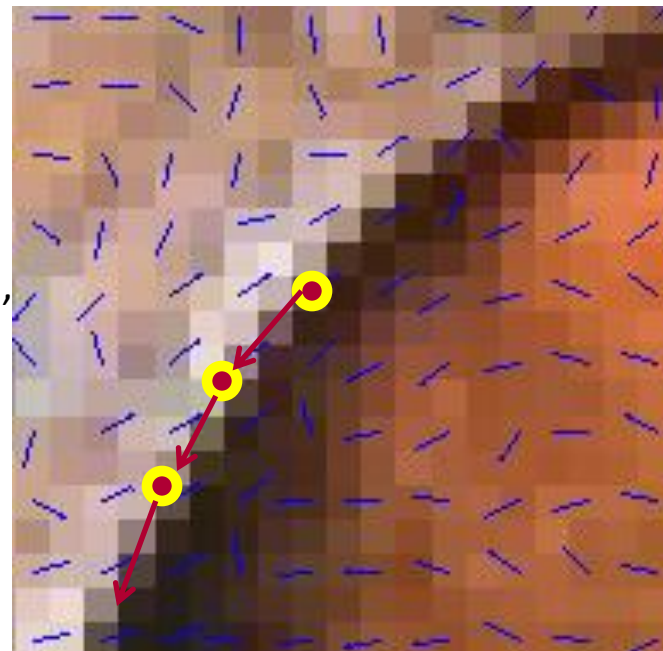
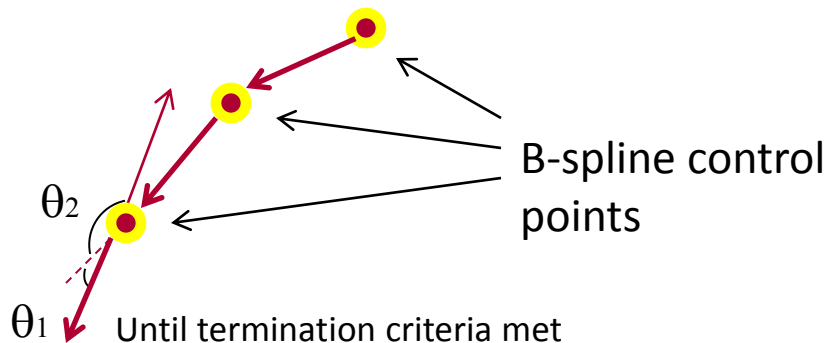
- Greedy algorithm for stroke placement
- Regularly sample the canvas to seed strokes
- Build a list of control points for each stroke by “hopping” between pixels*

- 2) Make another hop, resolving directional ambiguity by hopping in the direction of min θ



* In practice, best to use float coordinates and interpolate edge orientation

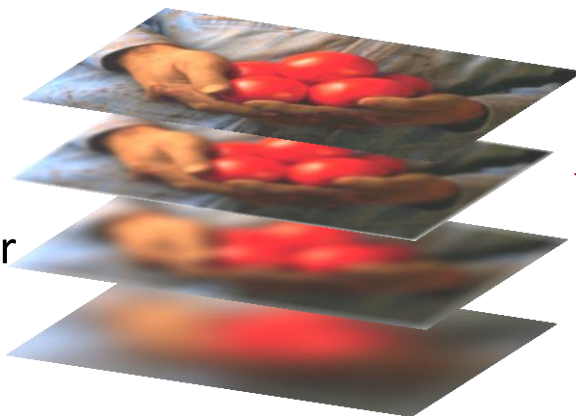
- Greedy algorithm for stroke placement
 - Regularly sample the canvas to seed strokes
 - Build a list of control points for each stroke by “hopping” between pixels*
- 3) Keep hopping until end land on a pixel whose RGB colour differs ($>$ threshold) from mean colour of stroke, or the stroke length is $>$ a second threshold.



* In practice, best to use float coordinates and interpolate edge orientation

- Painting is laid down in multiple layers (coarse to fine)
- Band-pass pyramid (= differenced layers of low-pass)
- Strokes from early layers are visible in final layer

- Paint coarsest layer with large strokes
- Paint next layer with smaller strokes
 - Only paint regions that differ between the layers
 - Use RGB difference



Compositing order





■ Tips and tricks

- Non-linear diffusion* instead of Gaussian blur sharpens the painting – preserves edges and accuracy of edge orientation.
- Build Gaussian pyramid at octave intervals, $\sigma=(1,2,4,8)$. 4 layers sufficient.
- Stroke thickness also at octave intervals
- Low-pass filter the hop direction θ



* “Scale-Space and Edge Detection using Anisotropic Diffusion”. P. Perona and J. Malik. *PAMI* 12:629–639. 1990.



- Global Optimization to Iteratively Produce “Better” Paintings



Hertzmann 1998
(Greedy stroke placement)



Hertzmann 2001
(Global stroke optimization)

- How to define the optimality of a painting 'P' derived from a photo 'G'

$$E(P) = E_{app}(P) + E_{area}(P) + E_{nstr}(P) + E_{cov}(P)$$

$$E_{app}(P) = \sum_{(x,y) \in \mathcal{I}} w_{app}(x,y) ||P(x,y) - G(x,y)||$$

$$E_{area}(P) = w_{area} \sum_{S \in P} \text{Area}(S)$$

$$E_{nstr}(P) = w_{nstr} \cdot (\text{number of strokes in } P)$$

$$E_{cov}(P) = w_{cov} \cdot (\text{number of empty pixels in } P)$$

Weighted sum of Heuristics

Painting similar to photo - weighted

Stroke area ("paint used by artist")

Number of strokes

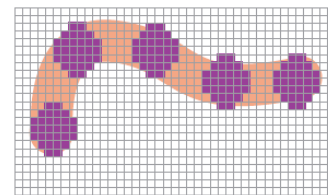
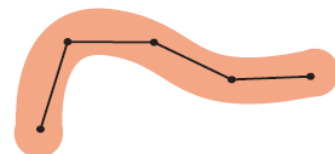
Fraction of canvas covered by strokes

- Weighting w_{app} is derived from a Sobel edge magnitude (or user defined)
- The right strokes in the right place will minimize the energy function $E(P)$

Internal energy External energy



- Strokes selected at random and modified by local optimization to minimize $\mathbf{E(P)}$
- Strokes modelled as active contours (“snakes”)
 - ... but energy has no 1st/2nd order derivative terms
 - $\mathbf{E(P)}$ is approximated under control points



$$E(P) = E_{app}(P) + E_{area}(P) + E_{nstr}(P) + E_{cov}(P)$$

$$E_{app}(P) = \sum_{(x,y) \in \mathcal{I}} w_{app}(x,y) ||P(x,y) - G(x,y)||$$

$$E_{area}(P) = w_{area} \sum_{S \in P} \text{Area}(S)$$

$$E_{nstr}(P) = w_{nstr} \cdot (\text{number of strokes in } P)$$

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Weighted sum of Heuristics

Painting similar to photo - weighted

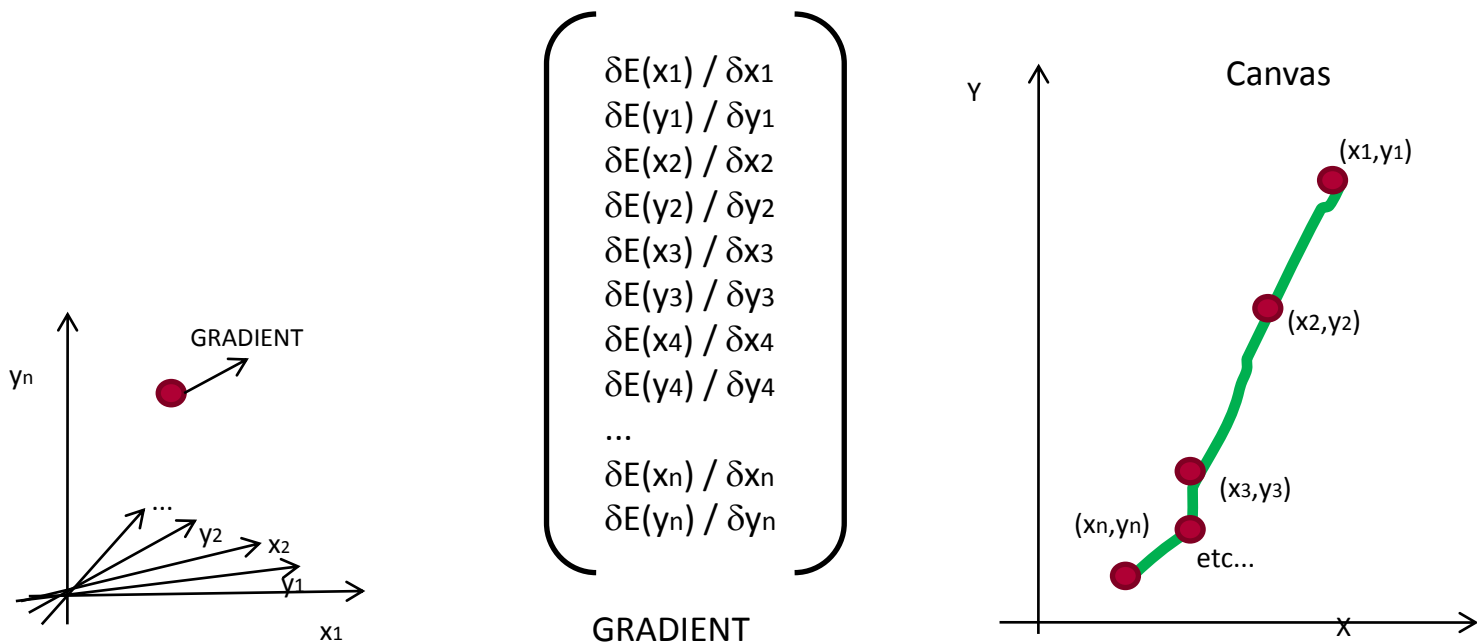
Stroke area (“paint used by artist”)

Number of strokes

Fraction of canvas covered by strokes



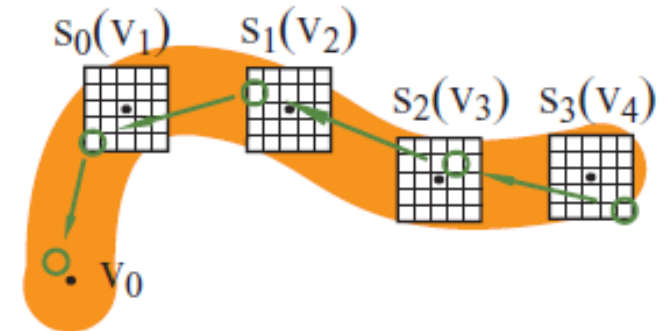
- Simplest solution (gradient descent)
 - Can be unstable for this weighted heuristic function



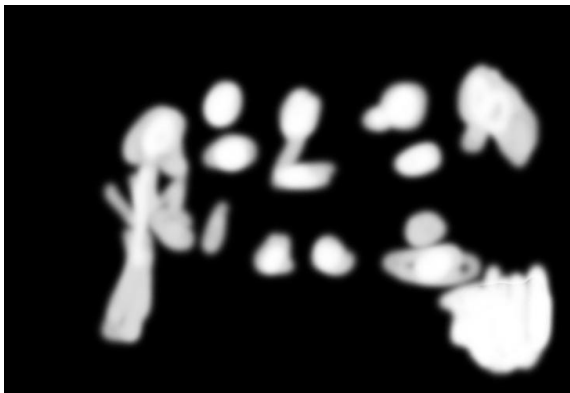


- Dynamic programming solution (Amini et al. '90)
 - Move each control point to obtain locally optimal position (5x5)
 - $E(P)$ at control point dependent only on current v_i and previous v_{i-1}

$$\begin{aligned}s_0(v_1) &= \min_{v_0} e_0(v_0) + e_0(v_1) + e_1(v_0, v_1) \\s_1(v_2) &= \min_{v_1} s_0(v_1) + e_0(v_2) + e_1(v_1, v_2) \\&\vdots \\s_{i-1}(v_i) &= \min_{v_{i-1}} s_{i-2}(v_{i-1}) + e_0(v_i) + e_1(v_{i-1}, v_i)\end{aligned}$$



- Sobel magnitude can be replaced with a manually sketched mask to alter emphasis





- Quick Start: OpenGL **research code** for bump-mapped paint strokes
 - Strokes as Catmull-Rom (interpolating) splines
 - Bump mapping via Multi-texturing (can be disabled)
 - Dependency on OpenCV to load images (can substitute this trivially)
 - Code used in “Empathic Painting” Collomosse et al. NPAR 2006

