

Visual Computing: Blending (Poisson & Graph Cut) SS 2016

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Perception



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Perception



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Perception



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Mach Band



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Von Polini 14:14, 30. Mär. 2008 (CEST) - selbst erstellt, CC BY-SA 3.0, https://de.wikipedia.org/w/index.php?curid=3444969



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Unisono01 - Bild selbst erstellt, Public Domain., PD-Schöpfungshöhe, https://de.wikipedia.org/w/index.php?curid=741732

Gaussian Kernel (blurring kernel)



$K = [-0.05, 0.25, 0.6, 0.25, -0.05]^{\top} [-0.05, 0.25, 0.6, 0.25, -0.05]$

$$G_i = (K * G_{i-1})_{\downarrow 2}, \quad i = 1, ..., N$$

Gaussian Kernel (blurring kernel)



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 $G_0 * K$

 $G_i = (K * G_{i-1})_{\downarrow 2}, \quad i = 1, ..., N$

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 $K = [-0.05, 0.25, 0.6, 0.25, -0.05]^{\top} [-0.05, 0.25, 0.6, 0.25, -0.05]$

 $G_0 * K$ $G_1 = (G_0 * K)_{\downarrow}$ $G_i = (K * G_{i-1})_{\downarrow 2}, \quad i = 1, ..., N$

Gaussian Kernel (blurring kernel)



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 $K = [-0.05, 0.25, 0.6, 0.25, -0.05]^{\top} [-0.05, 0.25, 0.6, 0.25, -0.05]$

$$G_0 * K$$

 $G_1 = (G_0 * K)_{\downarrow}$
 $G_i = (K * G_{i-1})_{\downarrow 2}, \quad i = 1, ..., N$
 $L_i = G_i - (K * G_i), \quad i = 0, ..., N - 1$

Laplacian Pyramid



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Multiresolution Blending

 $L_0^C = G_0^M L_0^S + (1 - G_0^M) L_0^T$

 $L_{N}^{C} = G_{N}^{M} L_{N}^{S} + (1 - G_{N}^{M}) L_{N}^{T}$

combine upsampled Laplacians to final composite image:

 $I^C =$



- build Laplacian pyramid for composite image with Gaussian G of map M and Laplacians L of source S and target T:

$$= \sum_{i=0}^{N} (L_i^C)_{\uparrow}$$

Multiresolution Blending





Multiresolution Blending







Difficult Example



(a)











Poisson/Gradient domain



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Problem Formulation

$\min_{I(x,y)\in\Omega} \iint \|\nabla I(x,y) - \nabla S(x,y)\|^2 \, dx \, dy$

s.t. I(x, y) = T(x, y) on $\partial \Omega$

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Problem Formulation

 $\min_{\substack{I(x,y)\in\Omega\\\Omega}}\iint_{\Omega}\|\nabla I(x,y)$

s.t. I(x, y) = T(x, y) on $\partial \Omega$

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$$\| -\nabla S(x,y) \|^2 dx dy$$

$\nabla^2 I(x, y) = \nabla^2 S(x, y) \text{ in } \Omega$ s.t. I(x,y) = T(x,y) on $\partial \Omega$

Problem Formulation

I(x+1, y) + I(x-1, y) + I(x, y+1) + I(x, y-1) - 4I(x, y)= S(x+1, y) + S(x-1, y) + S(x, y+1) + S(x, y-1) - 4S(x, y)

 $\left(\sum_{q\in\mathcal{N}(p)\cap\Omega}I(q)\right)+\left(\sum_{q\in\mathcal{N}(p)\cap\partial\Omega}I(q)\right)$



(3.14)



$$T(q) - 4I(x,y)$$

= S(x+1,y) + S(x-1,y) + S(x,y+1) + S(x,y-1) - 4S(x,y)

Results



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Diffi



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Graph Cut



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Solution to a Graph Cut Problem



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Min-Cut

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$|\mathcal{C}| = \sum w_{ij}$ $(i,j) \in C$

Images + Geometry

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References

 [CVFX] Computer Vision for Visual Effects, R.Radke, Poisson Image Editing Sec. 3.2. Graph-Cut Compositing Sec. 3.3.

To appear in the ACM SIGGRAPH '04 conference proceedings

Interactive Digital Photomontage

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Abstract

We describe an interactive, computer-assisted framework for combining parts of a set of photographs into a single composite picture, a process we call "digital photomontage." Our framework makes use of two techniques primarily: graph-cut optimization, to choose good seams within the constituent images so that they can be combined as seamlessly as possible; and gradient-domain fusion, a process based on Poisson equations, to further reduce any remaining visible artifacts in the composite. Also central to the framework is a suite of interactive tools that allow the user to specify a variety of high-level image objectives, either globally across the image, or locally through a painting-style interface. Image objectives are applied independently at each pixel location and generally involve a

In this paper, we look at how digital photography can be used to create photographic images that more accurately convey our subjective impressions - or go beyond them, providing visualizations or a greater degree of artistic expression. Our approach is to utilize multiple photos of a scene, taken with a digital camera, in which some aspect of the scene or camera parameters varies with each photo. (A film camera could also be used, but digital photography makes the process of taking large numbers of exposures particularly easy and inexpensive.) These photographs are then pieced together, via an interactive system, to create a single photograph that better conveys the photographer's subjective perception of the scene. We call this process digital photomontage, after the traditional process of combining parts of a variety of photographs to form a composite picture, known as photomontage.

• Photomontage: http://grail.cs.washington.edu/projects/photomontage/



Exercise

- case)
- Visualize Probability Distributions of BM and influence of sigma: potential libraries bokeh and seaborn.



• Poisson Image Blending (Additional: How to solve the problematic

Project

- Leap Motion / Intel Real Sense
- Oculus Rendering (GUI in VR)
- Photoscan Reconstruction

