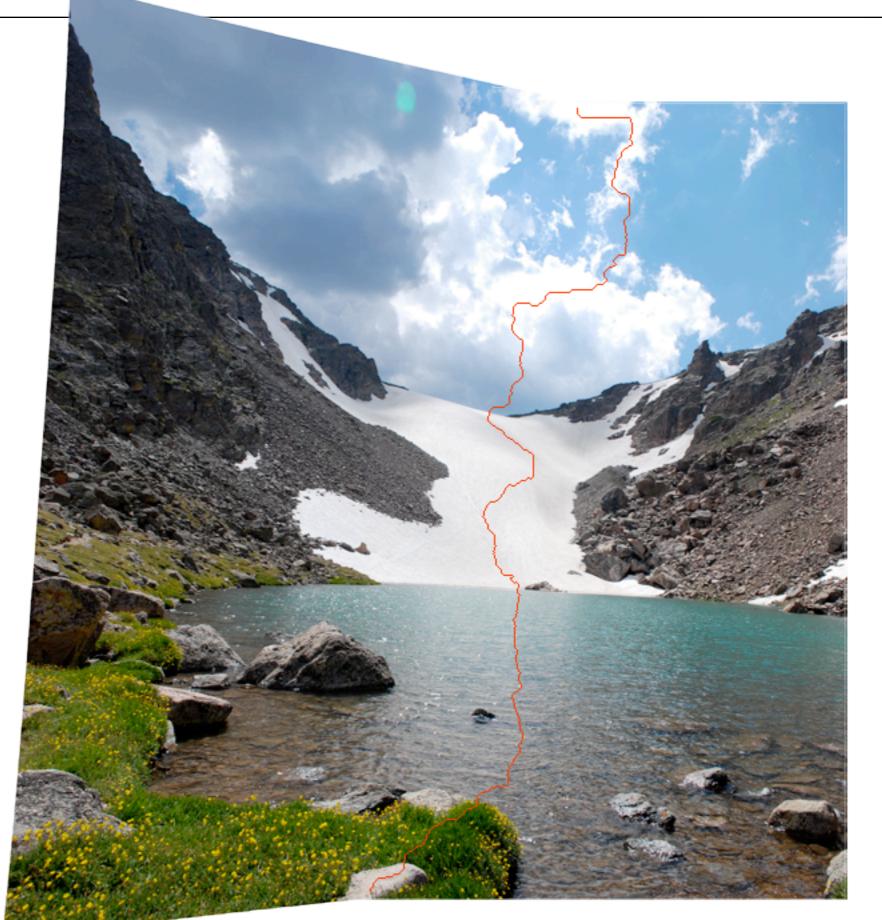
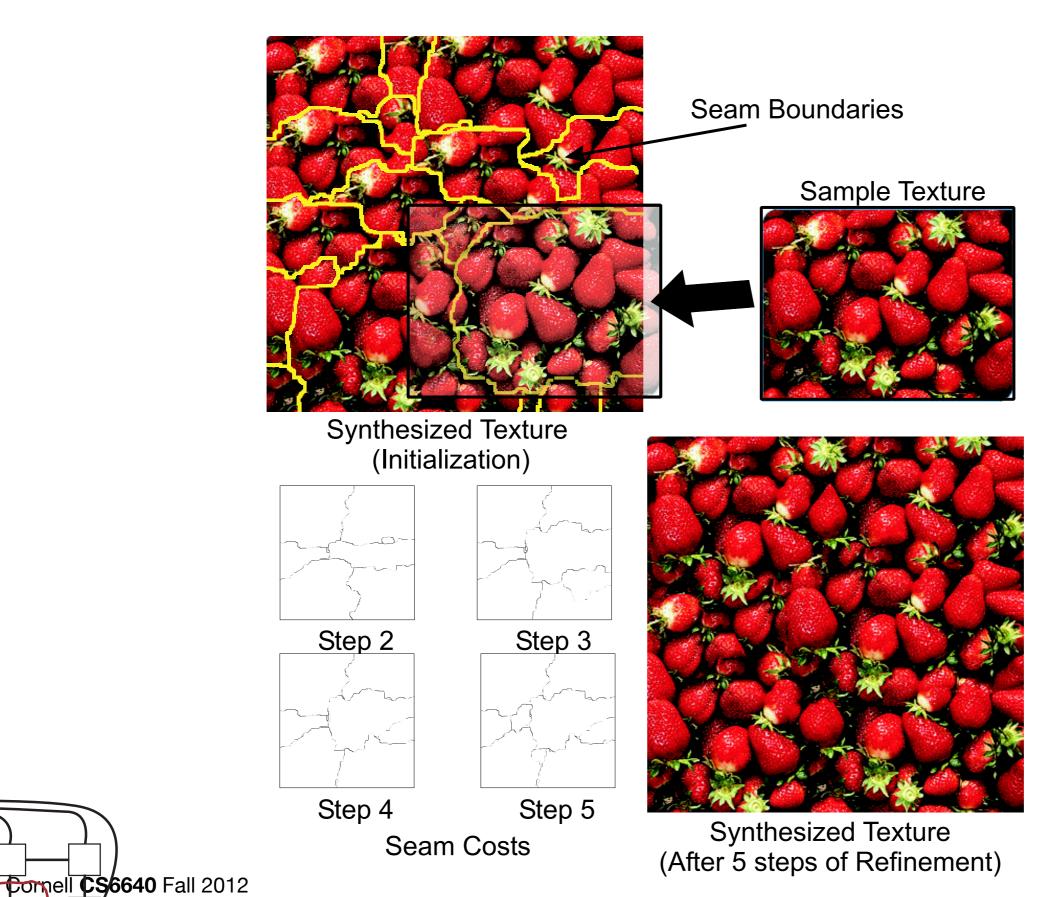
CS6640 Computational Photography

13. Graph Cut Optimization

Stitching a wide-angle view

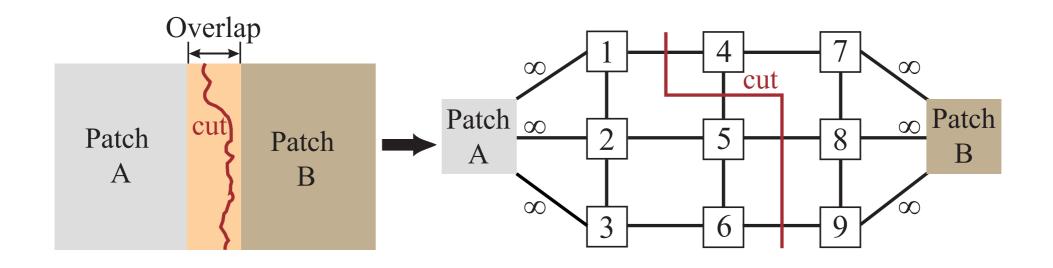


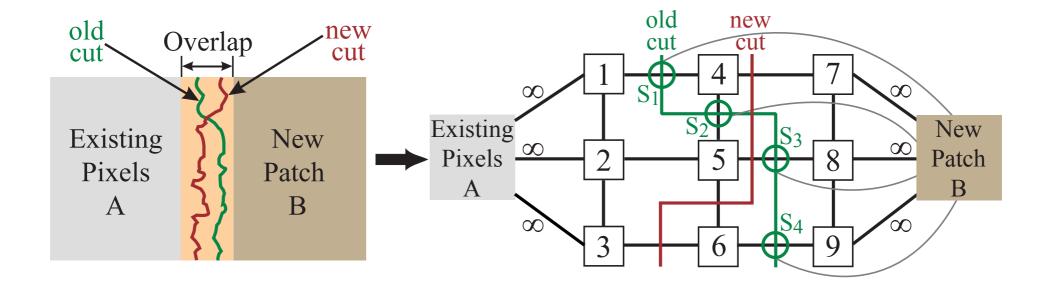
Texture synthesis with graph cuts



[Kwatra et al. 2005]

Accounting for existing seams





[[]Kwatra et al. 2005]

stical neuron—the m sponse of that neuro: of position—is perhar.

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escribe the wealth of simple-cell recept id neurophysiologically¹⁻³ and inferred especially if such a framework has the it helps us to understand the function leeper way. Whereas no generic most ussians (DOG), difference of offset C rivative of a Gaussian, higher derivati function, and so on—can be expect imple-cell receptive field, we noneth approximate the tree weatur of small

describing the response of that neuron ht as a function of position—is perhar functional description of that neuron.

seek a single conceptual and mathem.

that n'describing the response of that neurophysiologically¹⁻³ and and mat as a function of position—is perhally if such a framework ple-cefunctional description of that neuron. us to understand the ³ and issek a single conceptual and mathr way. Whereas no gen work 1 scribe the wealth of simple-cell ians (DOG), difference of d neurophysiologically¹⁻³ and ivative of a Ga response of that especially if such a framework functionnetion of position—i o geneit helps us to understand the funcional description of that ce of oeeper way. Whereas no generick a single conceptual and igher dussians (DOG), difference of a function of position—is per an be divative of a Gaussian, higher donal description of that neur he response of so on—can be a single conceptual and math iscribing the response of that ne the wealth of simple-cell ruune was a function of position—is perbphysiologically¹⁻³ and infe

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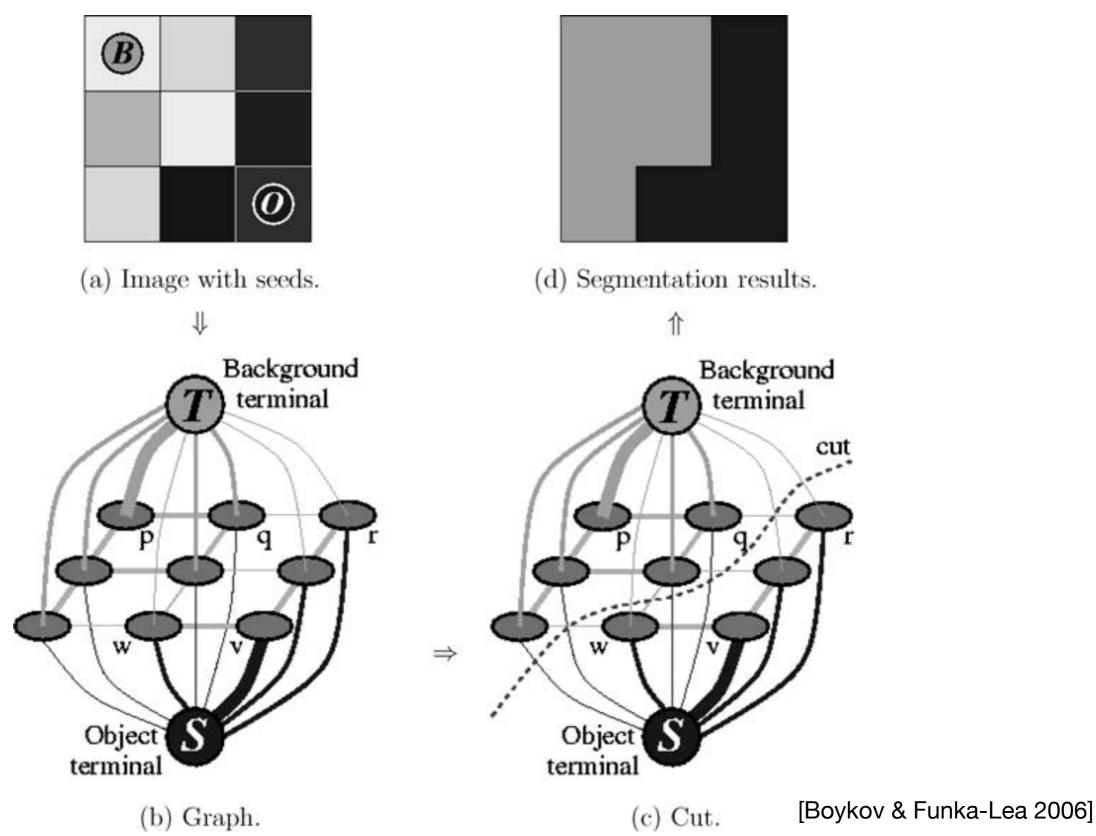




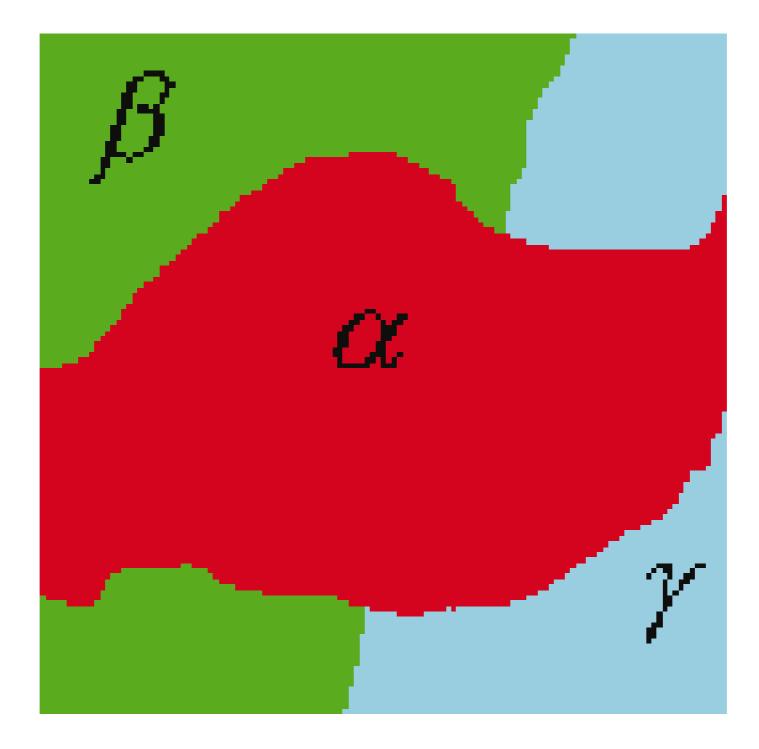




Segmentation with graph cuts



a-Expansion



Cornell CS6640 Fall 2012

[Boykov et al. 2001]

- 1. Start with an arbitrary labeling f
- 2. Set success := 0
- For each label α ∈ L
 3.1. Find f̂ = arg min E(f') among f' within one α-expansion of f
 3.2. If E(f̂) < E(f), set f := f̂ and success := 1
 If success = 1 goto 2
 Return f

Multi-way cuts: Photomontage



Figure 1 From a set of five source images (of which four are shown on the left), we quickly create a composite family portrait in which everyone is smiling and looking at the camera (right). We simply flip through the stack and coarsely draw strokes using the *designated source* image objective over the people we wish to add to the composite. The user-applied strokes and computed regions are color-coded by the borders of the source images on the left (middle).

Multi-way cuts: Photomontage

