

GrabCut

Interactive Foreground Extraction using Iterated Graph Cuts

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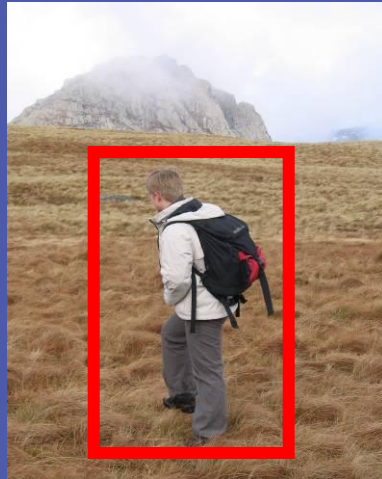


Microsoft Research Cambridge-UK

Photomontage



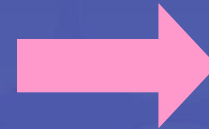
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Problem



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Fast &
Accurate ?



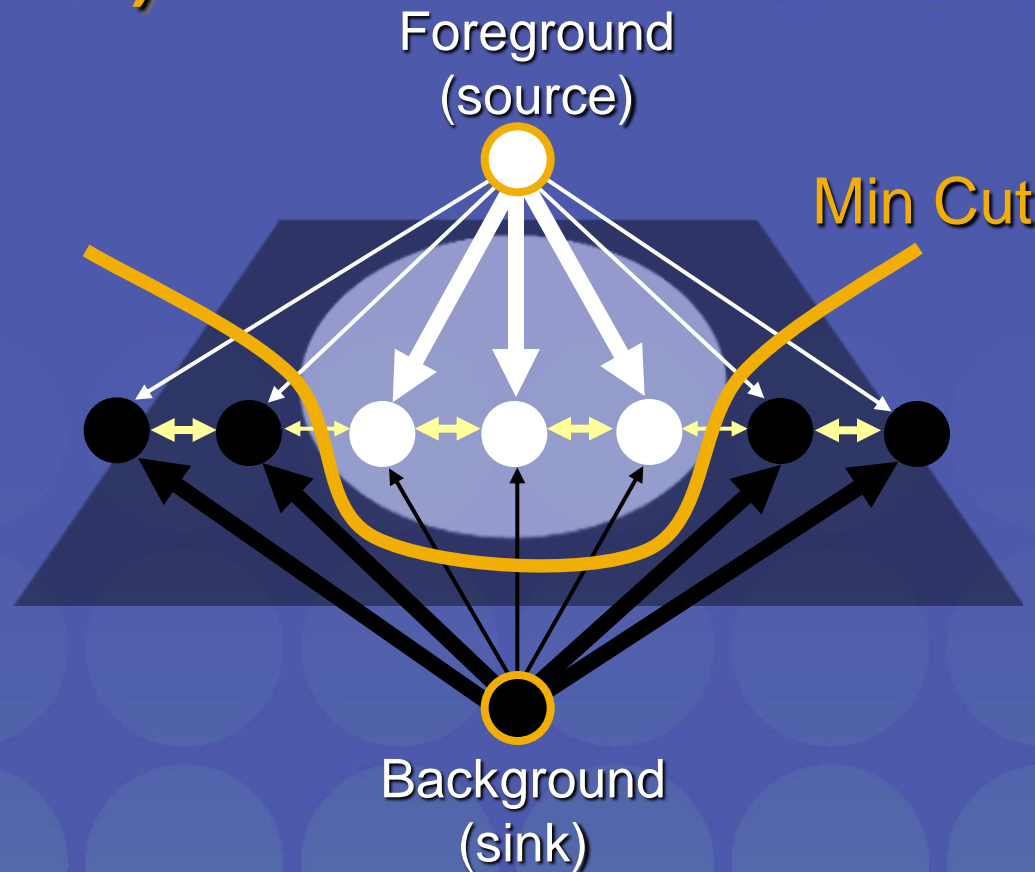
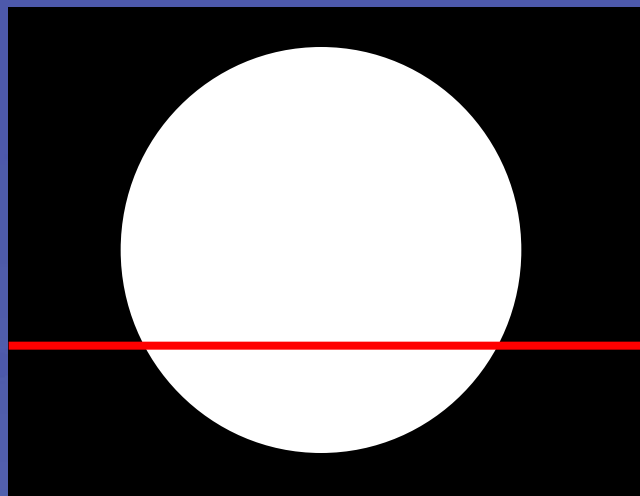
Graph Cuts

Boykov and Jolly (2001)



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Image



Cut: separating source and sink; Energy: collection of edges

Min Cut: Global minimal energy in polynomial time

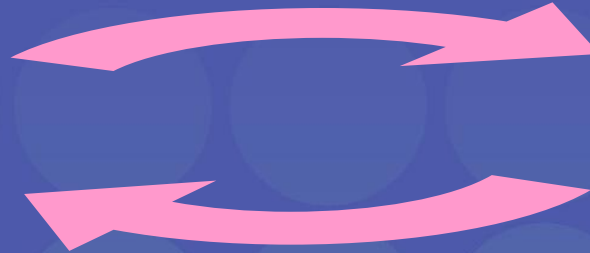
Iterated Graph Cut



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User Initialisation



**K-means for learning
colour distributions**

**Graph cuts to
infer the
segmentation**

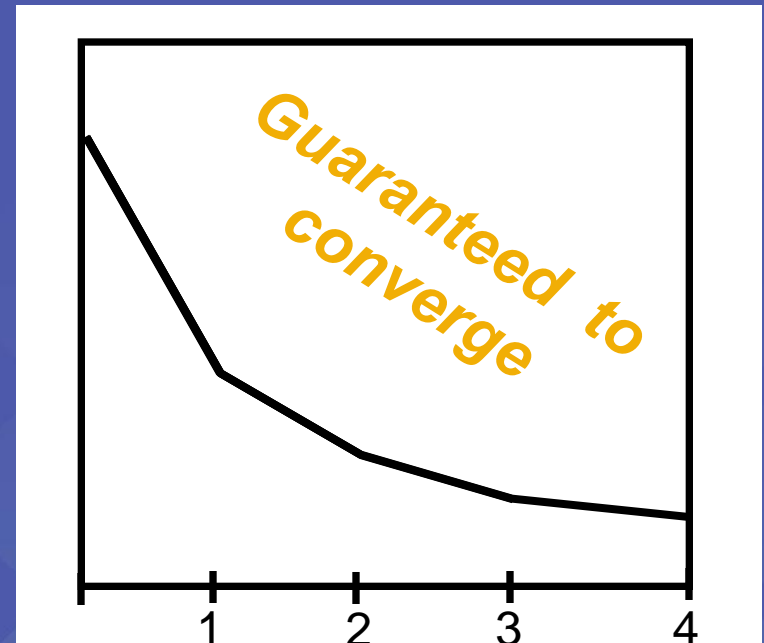
Iterated Graph Cuts



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Result

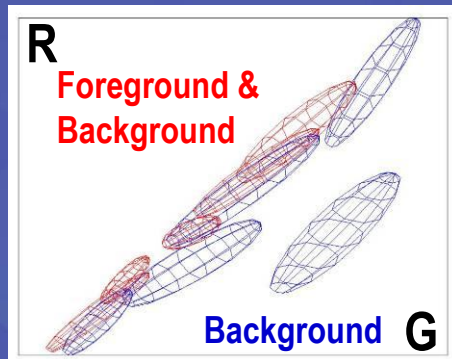


Energy after each Iteration

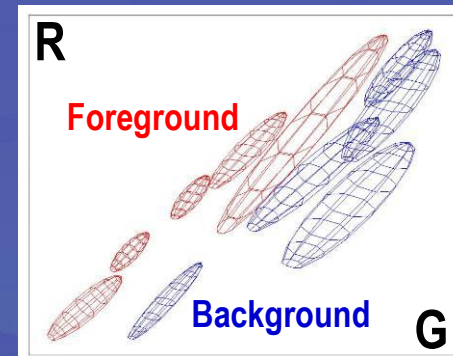
Colour Model



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Iterated
graph cut



Gaussian Mixture Model (typically 5-8 components)

Coherence Model



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An object is a coherent set of pixels:

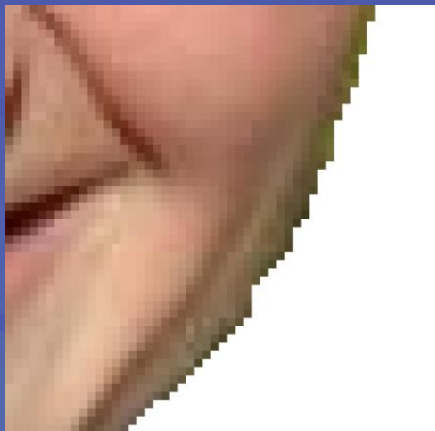


Blake et al. (2004): Learn jointly

Border Matting



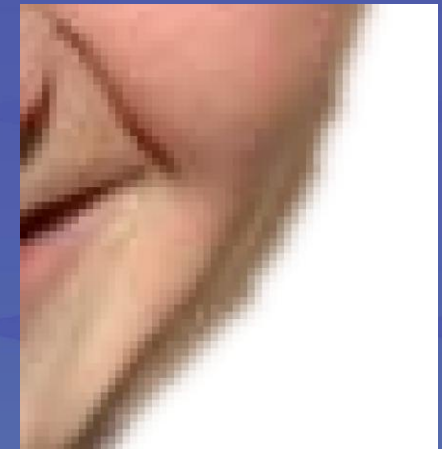
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Hard Segmentation



Automatic Trimap

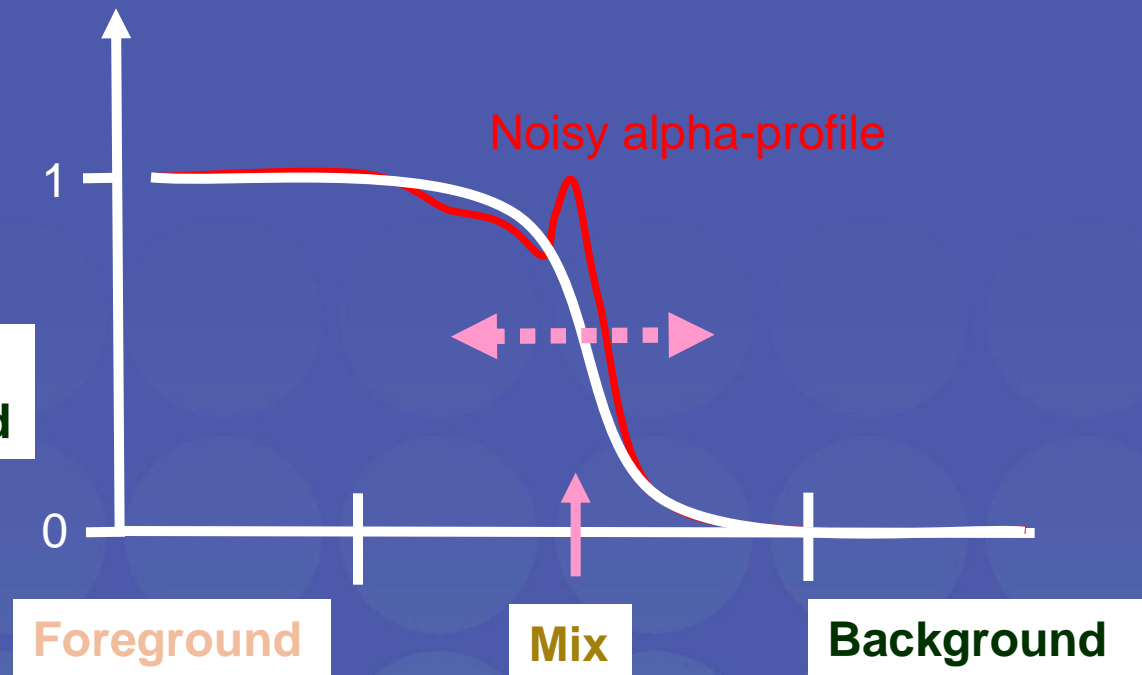
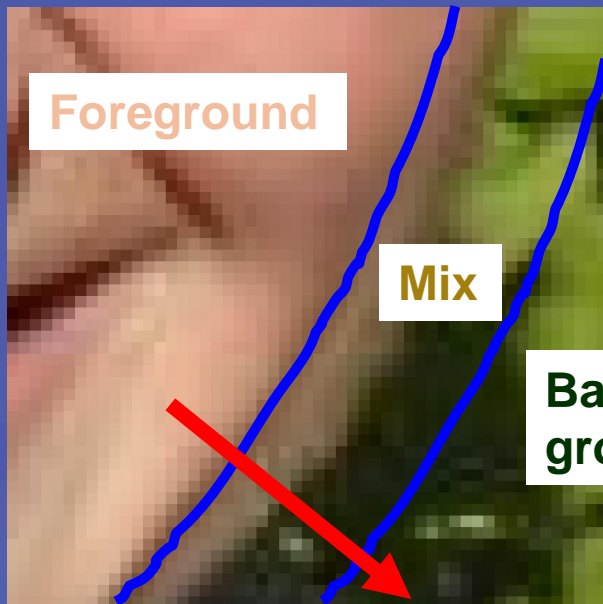


Soft Segmentation

Border Matting



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Fit a smooth alpha-profile with parameters

Comparison

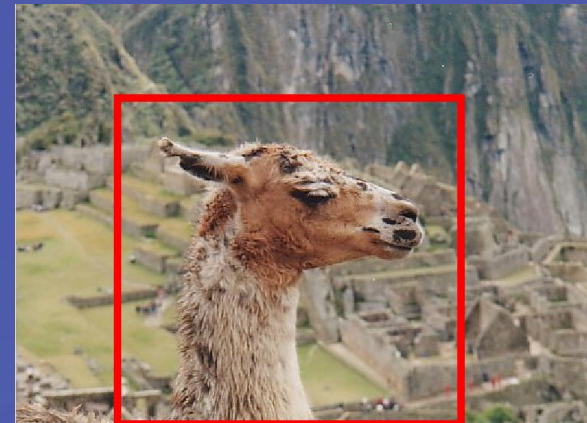
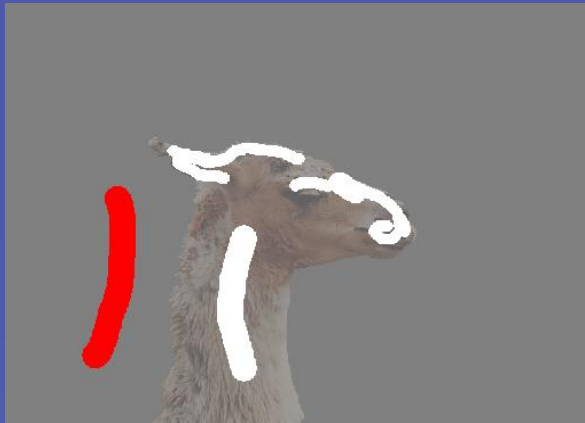


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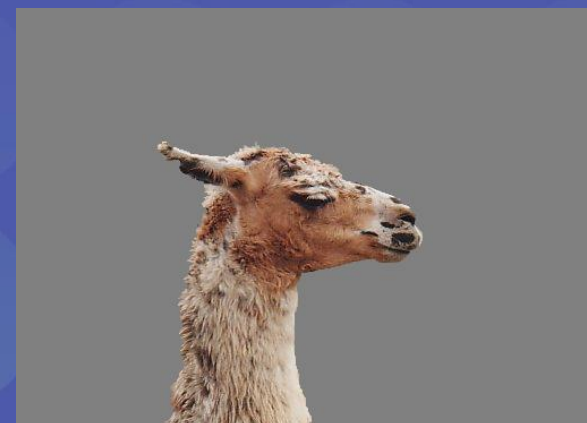
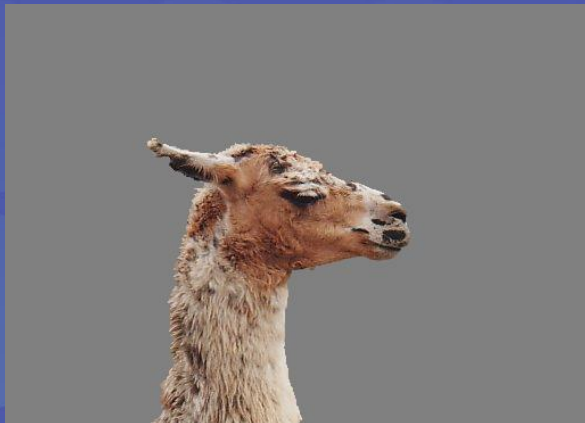
Boykov and Jolly (2001)

GrabCut

User
Input



Result



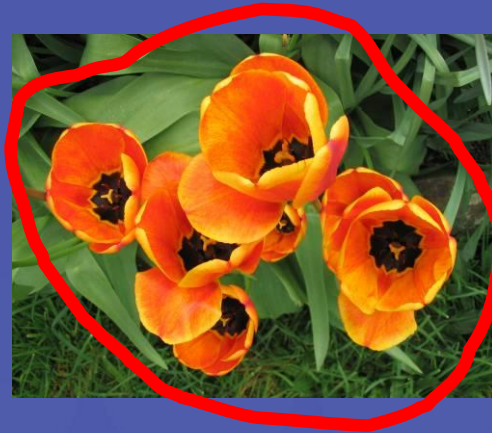
Error Rate: 0.72%

Error Rate: 0.72%

Moderately straightforward examples



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... GrabCut completes automatically

Difficult Examples



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Initial
Rectangle



Initial
Result



Difficult Examples



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No User
Interaction



Lazy Snapping

Lazy Snapping

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[†]Hong Kong University of Science and Technology

[‡]Microsoft Research Asia



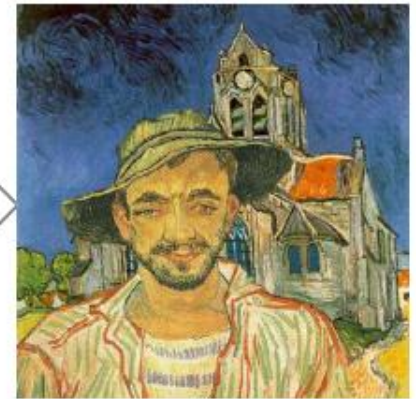
(a) Input image



(b) Object Marking



(c) Boundary editing



(d) Output composition

Lazy Snapping Overview

▣ Main Steps:

1. Mark strokes as foreground & background (object marking)
2. Perform Boykov & Jolly style graphcut
3. Boundary Editing

▣ Familiar Formulation

- Minimize Gibbs Energy:

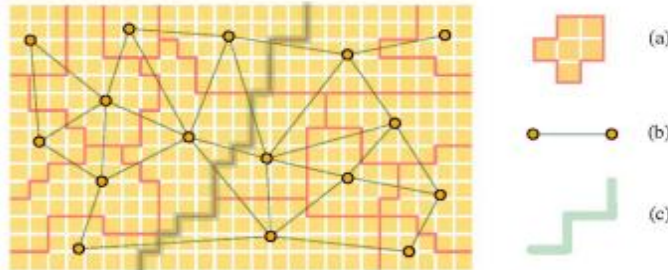
$$E(X) = \sum_{i \in \mathcal{V}} E_1(x_i) + \lambda \sum_{(i,j) \in \mathcal{E}} E_2(x_i, x_j)$$

- ▣ $E_1(X)$ – Cluster (K-means), encodes color similarity of node (likelihood energy)
- ▣ $E_2(X)$ – encodes energy due to gradient along object boundary (color gradient between two nodes)

Lazy Snapping—Pre-processing



Watershed Algorithm



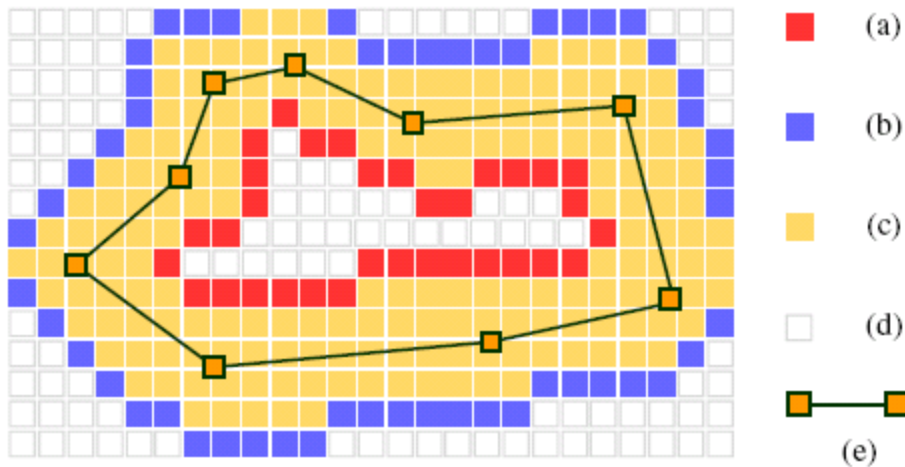
- Picture is presegmented with the Watershed Algorithm
- Graphcut works on segment instead of pixel level
- Segments are connected if one pixel within them is connected
- Results are available almost instantly

Lazy Snapping – Speedup

Image	Dimension	Nodes Ratio	Edges Ratio	Lag with Pre-segmentation	Lag without Pre-segmentation
Boy	(408, 600)	10.7	16.8	0.12s	0.57s
Ballet	(440, 800)	11.4	18.3	0.21s	1.39s
Twins	(1024, 768)	20.7	32.5	0.25s	1.82s
Girl	(768, 1147)	23.8	37.6	0.22s	2.49s
Grandpa	(1147, 768)	19.3	30.5	0.22s	3.56s

Boundary Editing

- Graph Cut is only step 1
- Allow user to edit boundary directly
- Formulate as Graph Cut
- Allow user to guide boundary regions w/ brush



(a)

(b)

User Study

▣ Results

- Easy of Use – Lazy Snapping 20% less mistakes
- Speed – Lazy Snapping 60% less time
- Accuracy – Lazy Snapping 60% less pixels wrong

▣ Feedback

- “Almost Magic”
- “Much Easier”

▣ Suggestions

- Get rid of 2 step process
- Make it easy to switch between graph cut strokes and boundary editing

More Results



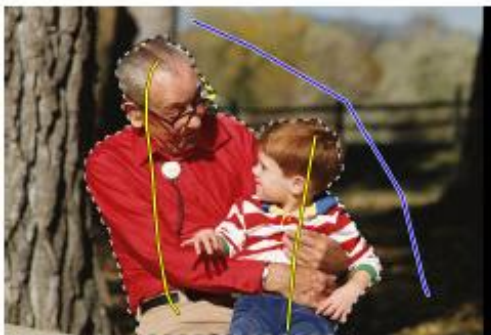
(a) Girl (4/2/12)



(b) Ballet (4/7/14)



(c) Boy (6/2/13)



(c) Grandpa (4/2/11)



(d) Twins (4/4/12)

