How to Learn and Evaluate an Interactive Computer Vision

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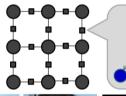


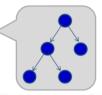
Interactive Image and Data manipulation





Applied Optimization, Models, and Learning





Inverse rendering from moving images







3D Scene Understanding



Benchmarking and Label collection



BioImaging

... looking for PostDocs and PhD students

Inverse Rendering & Scene Understanding

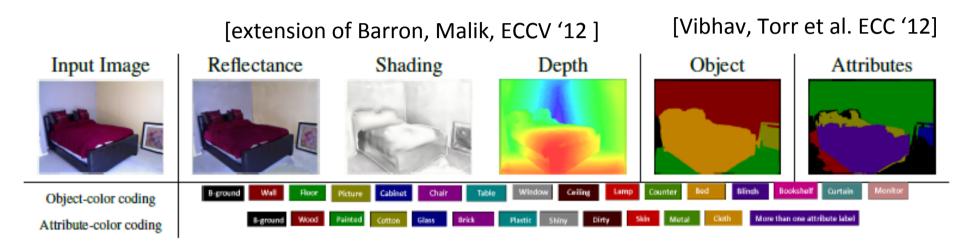


Figure 1: Given an image, our algorithm jointly estimates the intrinsic properties such as reflectance, shading and depth maps, along with the estimation of the per-pixel object and attribute labels.

How to Learn and Evaluate an Interactive Computer Vision

Joint work with:
Pushmeet Kohli, Hannes Nickisch,
Christoph Rhemann

Learning and Evaluation of Interactive Segmentation Systems

Interactive Image Segmentation



Image (X)



Demo



Key Questions

1. How to evaluate different segmentation systems?

2. How to learn the parameters of a given segmentation system?

... How GrabCut got into Office 2010+

[Markov Random Fields for Computer Vision, Blake, Kohli and Rother]



Database of 200 Images



Image



Ground Truth



Rectangle Labelling



Lasso Labelling





... How GrabCut got into Office 2010+

[Markov Random Fields for Computer Vision, Blake, Kohli and Rother]

Product team got big spread sheet with error rates

Product team: "What do these numbers mean?"

Me: "Explain Error rate"

Product team: "Hm"

Me: "Ok we do something better"



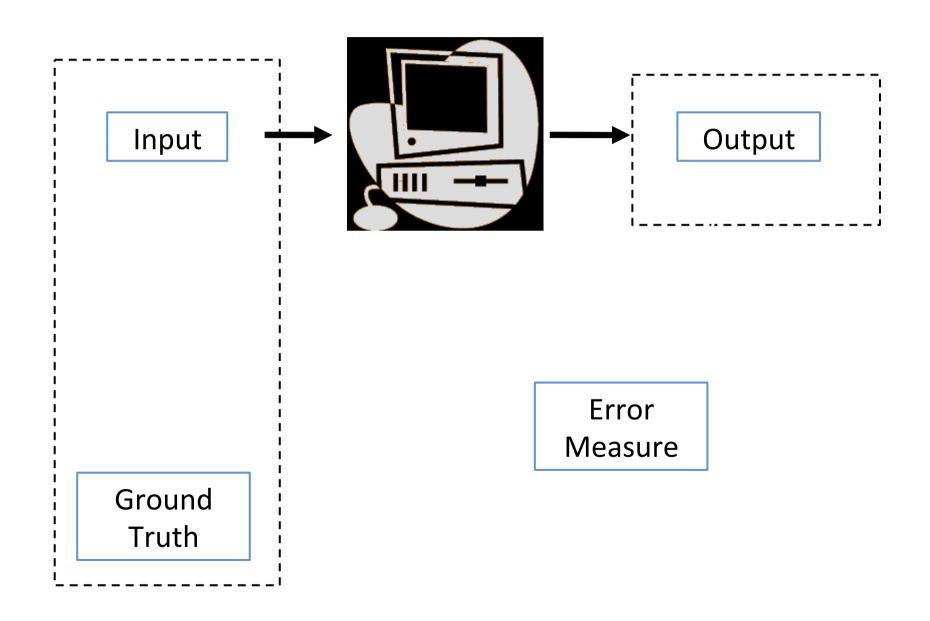


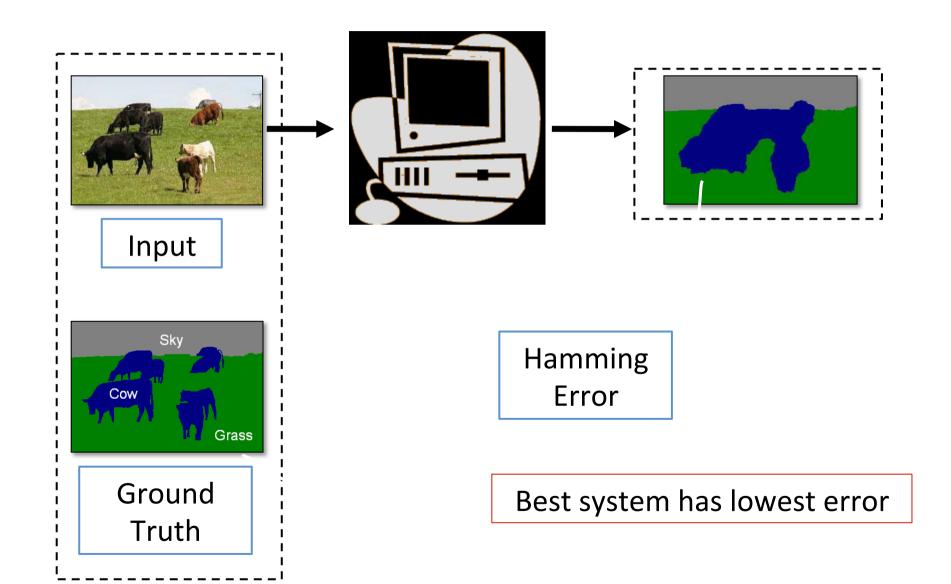






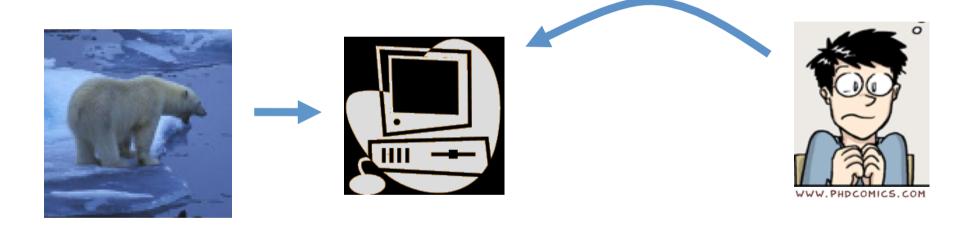
5% error





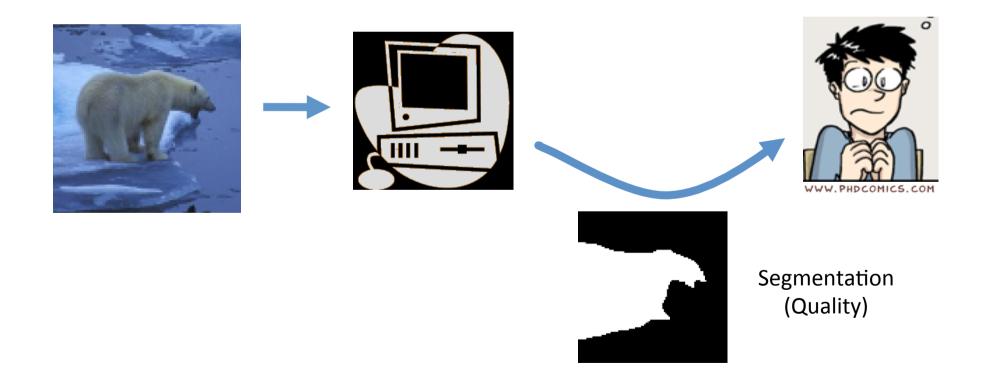


User Scribbles (Effort)



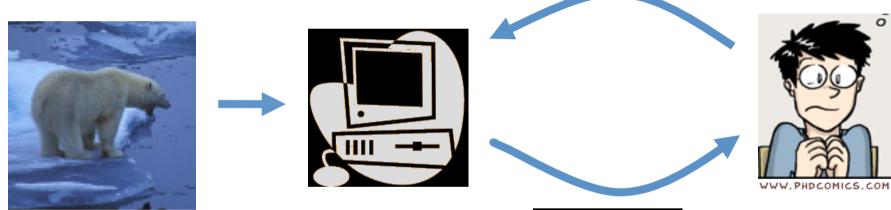


User Scribbles (Effort)





User Scribbles (Effort)



Interactive Algorithms

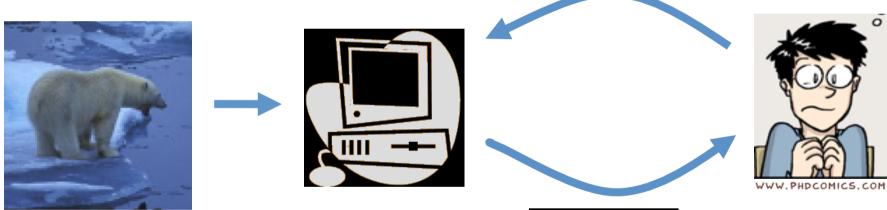
Best system requires lowest amount of effort to reach a reasonable answer



Segmentation (Quality)



User Scribbles (Effort)



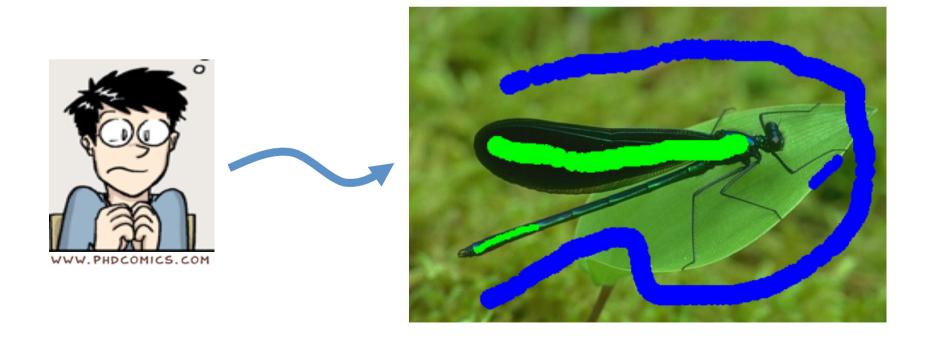
Interactive Algorithms
Best system requires lowest
amount of effort to reach a
reasonable answer



Segmentation (Quality)

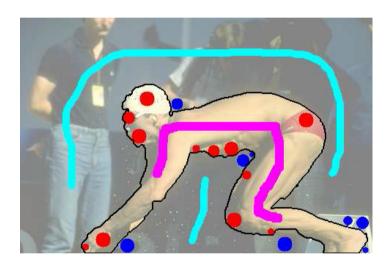
How to measure "effort" and "reasonable"?

- Number of brush strokes
- Stroke complexity
- Time taken

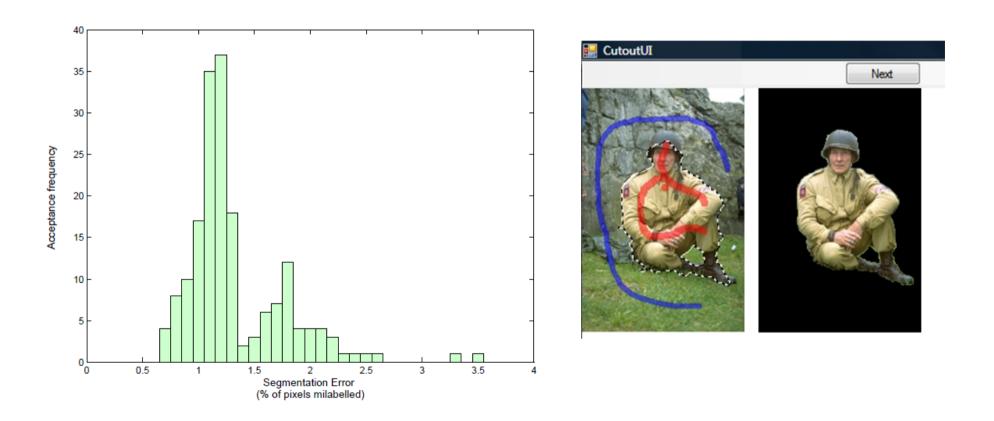


Circular brush





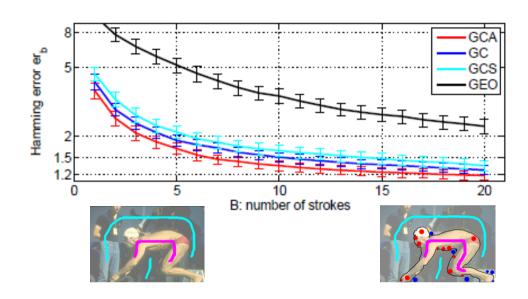
How to quantify reasonable accuracy?

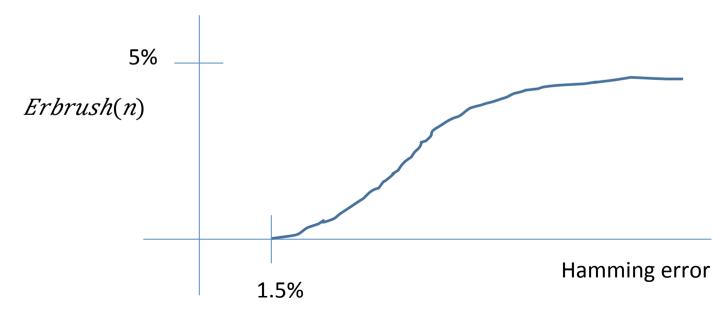


We truncate the hamming error with 1.5%

Error measure

 $Er = \sum_{n=1}^{\infty} \int B Erbrush(n)$





User Study or Crowdsourcing

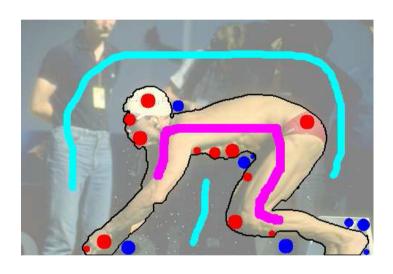
- Measure performance using actual users
 - Few users[Li et al. 2004, Bai & Sapiro 2007]
 - Crowd-sourcing with different incentives
 [MTurk, Community-based (Labelme), ESP game]

Problems:

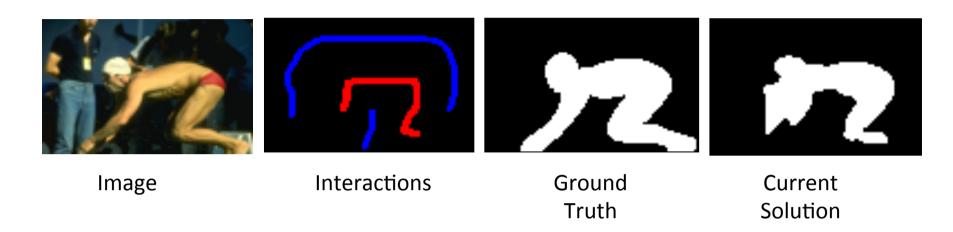
- Expensive
- Slow (participants should be allowed learning time)
- Unsuitable for parameter learning
 (where you might need to evaluate many systems)

Robot User

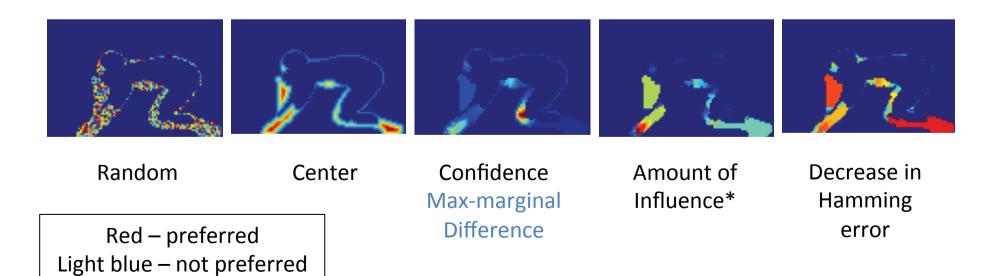




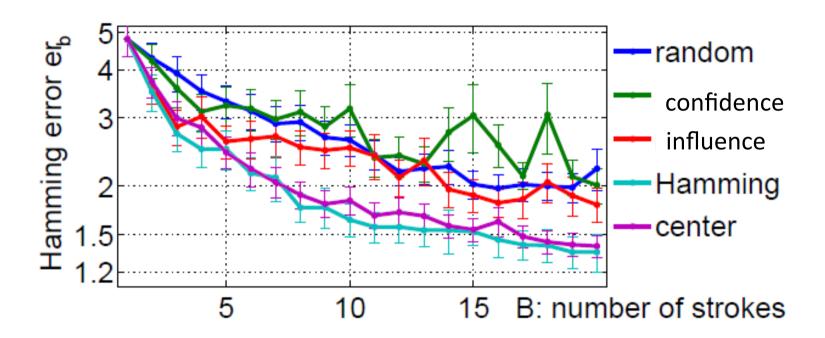
Where to place the brush strokes?



Interaction chosen by taking the maximum in the preference map



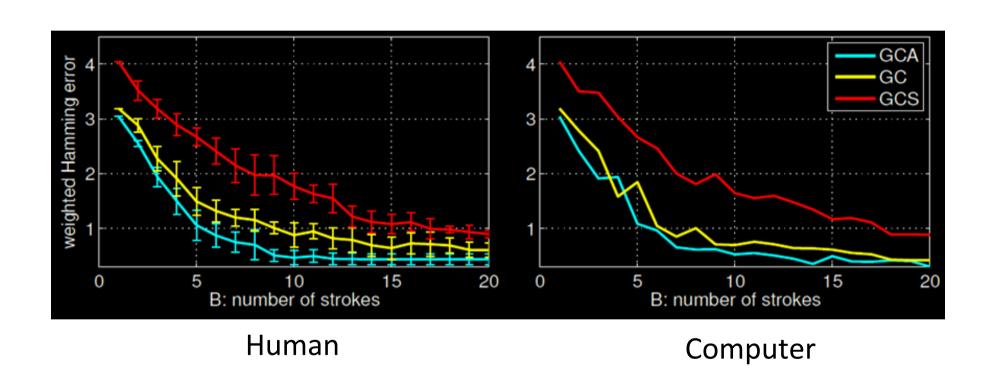
Where to place brush strokes



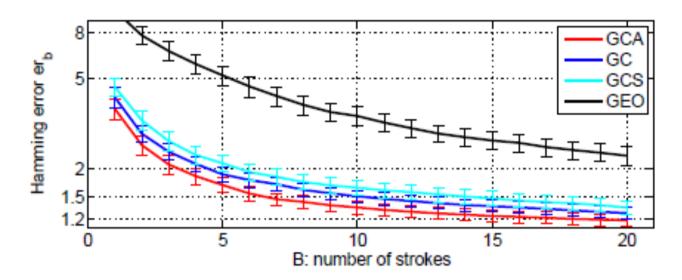
Number of Interactions

Side comment: using the uncertainty of the system is worse than random!

Is it a good model for Users?



Comparison

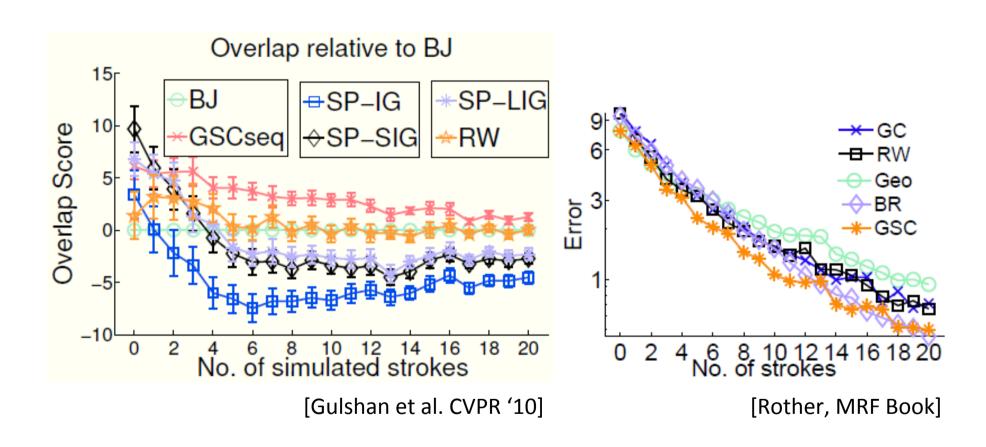


- Grabcut

[Boykov Jolly 2001] [Rother et al SIGGRAPH 2004]

- GCS(Simplified fixed colour models)
- GCA(Advanced enforces connectivity of segmentation)
- GEO
 (Geodesic distance based methods) [Bai Sapiro, ICCV 07]

Used to compare systems



What is ignored?

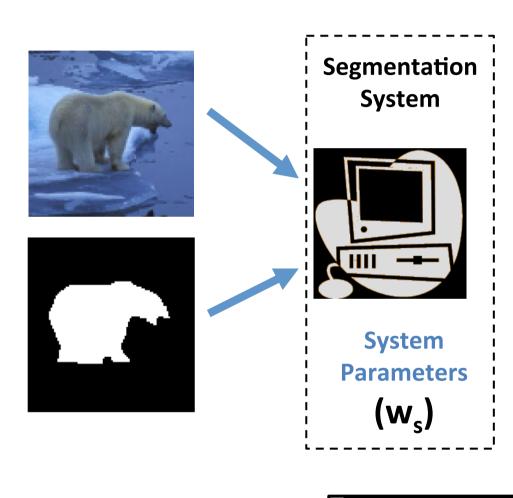
- Speed of system? (faster better)
- User adjust to the system

Key Questions

1. How to evaluate different segmentation systems?

2. How to learn the parameters of a given segmentation system?

The Learning Task



[Generative Learning]
Maximum Likelihood

[Kumar et al 2005] [Pletscher, Nowozin, Kohli, Rother, DAGM 2011]

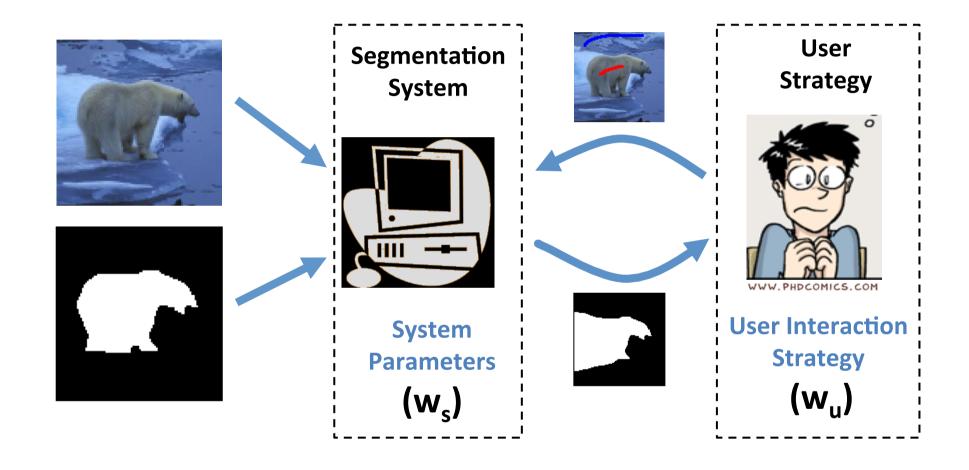
$$\mathbf{w}^* = \operatorname{arg\,max}_{\mathbf{w}} \left[\prod_{k=1}^K \mathbb{P}_{\mathbf{w}}(\mathbf{x}^k, \mathbf{y}^k) \right]$$
 $\mathbb{P}_{\mathbf{w}}(\mathbf{x}, \mathbf{y}) = \frac{1}{Z} \exp\left(-E_{\mathbf{w}}(\mathbf{x}, \mathbf{y})\right)$

[Discriminative Learning] Max-margin based methods

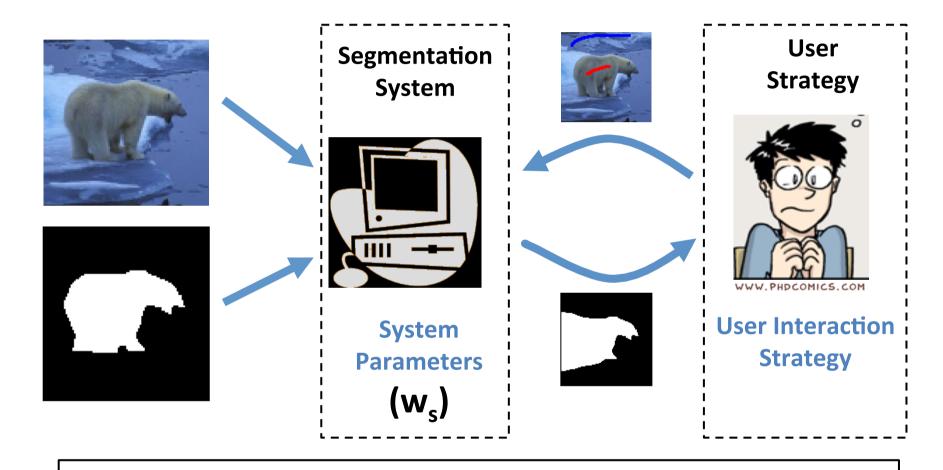
[Tsochantaridis et al 2001] [Taskar et al 04,05] [Szummer et al 08]

$$\mathbf{w}^* = \mathop{\mathrm{arg\,max}}_{\mathbf{w}} \left[\mathop{\mathrm{min}}_{\mathbf{y} \in \mathcal{Y} \setminus \mathbf{y}^k} \left\{ E_{\mathbf{w}}(\mathbf{x}^k, \mathbf{y}) - E_{\mathbf{w}}(\mathbf{x}^k, \mathbf{y}^k) \right\} \right]$$

The Learning Task



The Learning Task



Few parameters



Learn by grid-search over the evaluation score on test data

System

$$E(\mathbf{y}) = \sum_{p \in \mathcal{V}} E_p(y_p) + \sum_{(p,q) \in \mathcal{E}} E_{pq}(y_p, y_q)$$

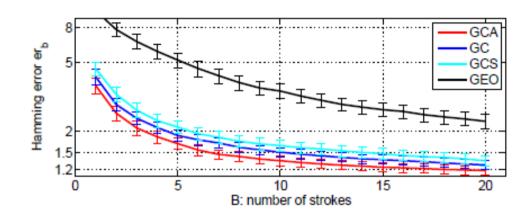
$$E_{pq}(y_p, y_q) = \frac{|y_q - y_p|}{\text{dist } (p, q)} \left(w_i + w_c \exp \left[-\beta ||x_p - x_q||^2 \right] \right)$$

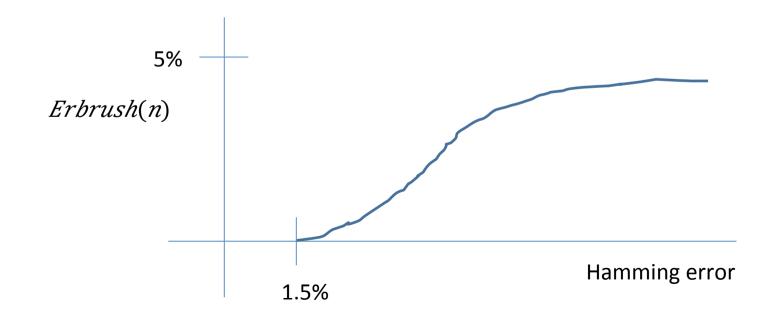
$$\beta = 0.5 \cdot w_{\beta} / \left\langle \left\| x_p - x_q \right\|^2 \right\rangle$$

3 free parameters

Training objective

 $Er = \sum_{n=1}^{n} 1 \uparrow B \text{ } Erbrush(n)$



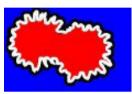


Extreme scenarios

Training Time





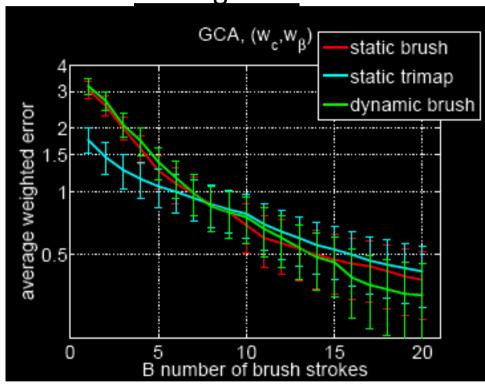


Static brush

Static trimap

Trimap	w _c
Static trimap	0.07 +/- 0.09
Static brush	0.22 +/- 0.52

Testing Time



Other Application Scenarios

Interactive systems is more than just a work-around to not yet automatic systems

- Computer vision:
 - segmentation, optical flow, shape, etc.
 - Mechantical turk for gathering ground truth
- Biology
- Physical simulations
- Human-Computer-Interaction
- Interactive visualization

Open Questions

- Speed of system? (faster better)
- User adjust to the system
- Other solutions ...