

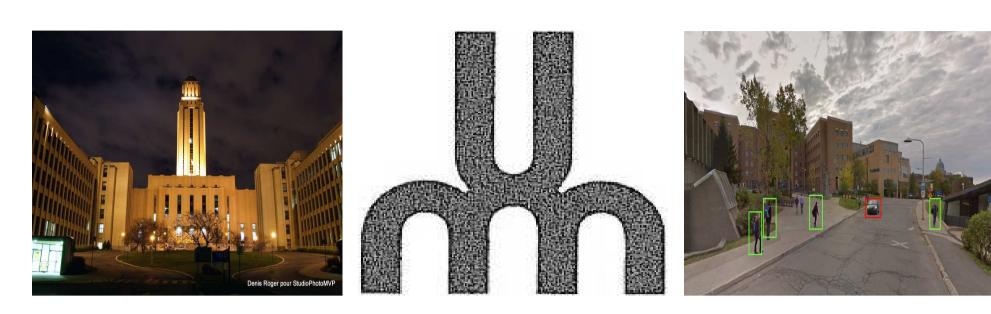
GCE-BASED MODEL FOR THE FUSION OF MULTIPLES COLOR IMAGE SEGMENTATIONS

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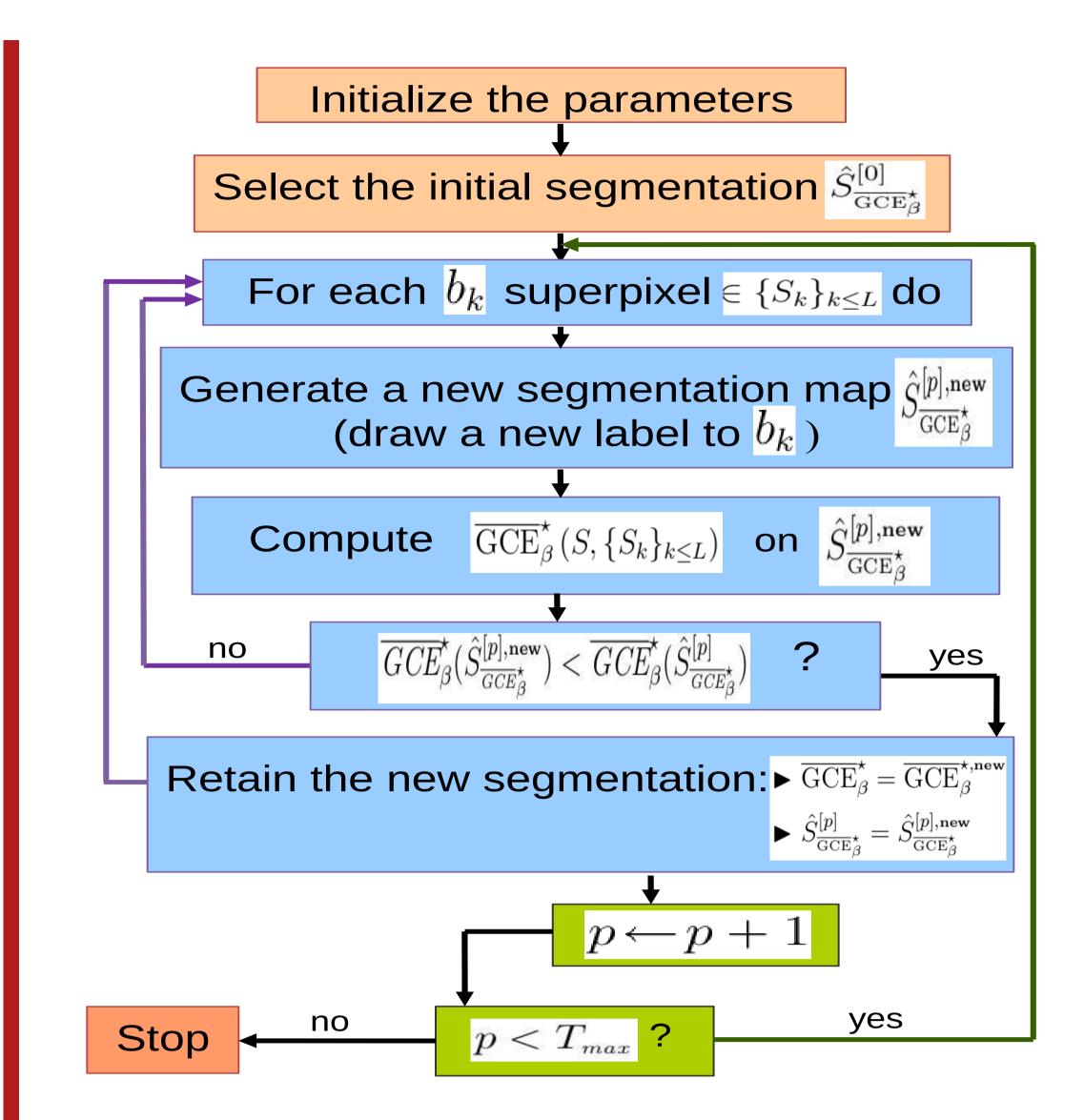


• Aims to represent the image content into different regions of coherent properties with homogeneous characteristics:



where

 $S^{t} = \{C_{1}^{t}, C_{2}^{t}, \ldots, C_{R^{t}}^{t}\}$: Segmentation $S^{g} = \{C_{1}^{g}, C_{2}^{g}, \ldots, C_{R^{g}}^{g}\}$: Segmentation R^{t} : Number of region in S^{t} $R^{\mathfrak{g}}$: Number of region in $S^{\mathfrak{g}}$ p_i : Pixel $C_{< p_i >}^{t}$: Segments in S^{t} including p_i $C_{< p_i>}^{\mathfrak{g}}$: Segments in $S^{\mathfrak{g}}$ including p_i Set differencing operator *n*: Number of pixel within the image |C|: Cardinality of the set of pixels C



texture color movement

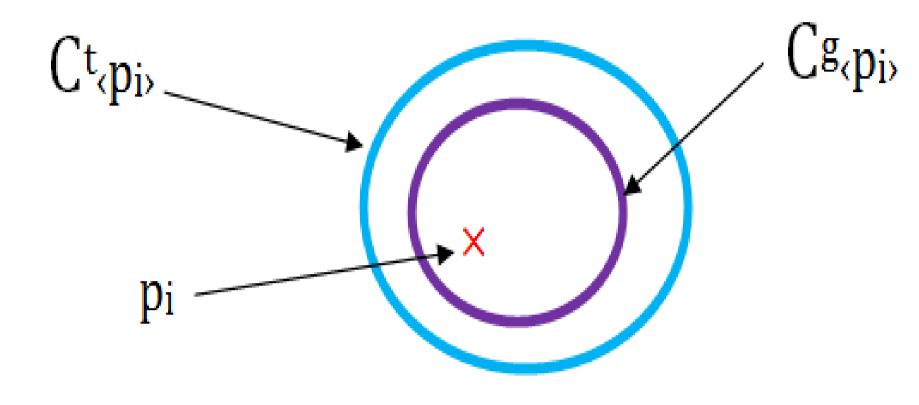
Problem: III-posed problem

- Selected criterion for grouping pixels depends on the goal of the segmentation
- A unique algorithm cannot perform adequately for all applications

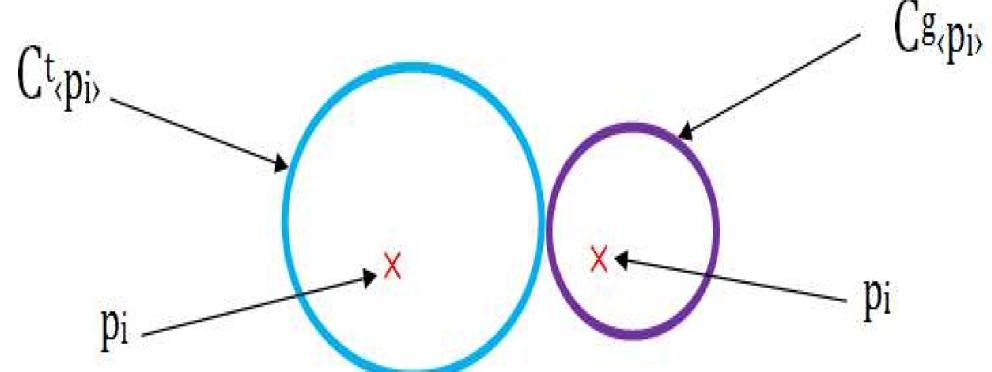
Key Idea: Combining multiple, quickly estimated segmentation maps of the same image to obtain a final refined segmentation



● GCE= 0 Ľ Perfect match between the two segmentations to be compared

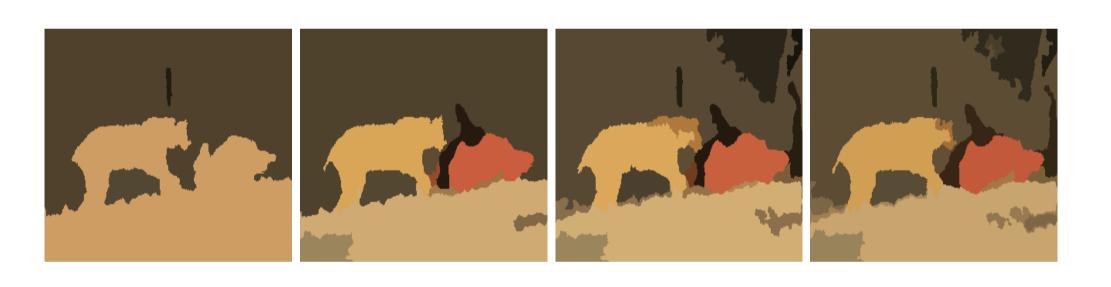


• GCE= 1 \square Maximum difference between the two segmentations

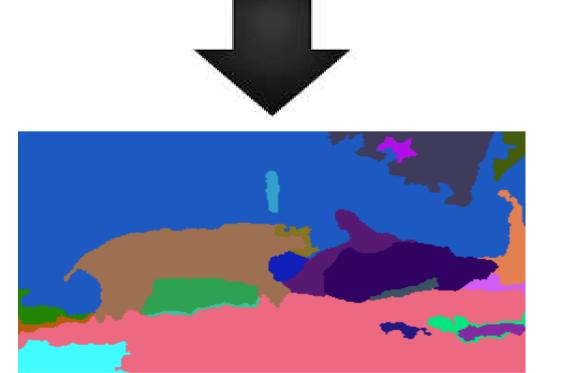


Results

Example of segmentation solutions generated for different values of $\overline{\mathcal{R}}$







Segmentation criterion: Global Consistency **Error (GCE)**

GCE-Based Fusion Model

$$\hat{S}_{\overline{\mathtt{GCE}}_{\beta}^{\star}} = \arg\min_{S\in\mathcal{S}_n} \Big\{ \overline{\mathtt{GCE}}^{\star} \big(S, \{S_k\}_{k\leq L} \big) + \beta E_{\mathtt{Reg}}(S) \Big\} (3)$$

$$\overline{\mathsf{GCE}^{\star}}(S^{\mathsf{t}}, \{S_k^g\}_{k \le L}) = \frac{1}{L} \sum_{k=1}^{L} \mathsf{GCE}^{\star}(S^{\mathsf{t}}, S_k^g)$$

(4)

and

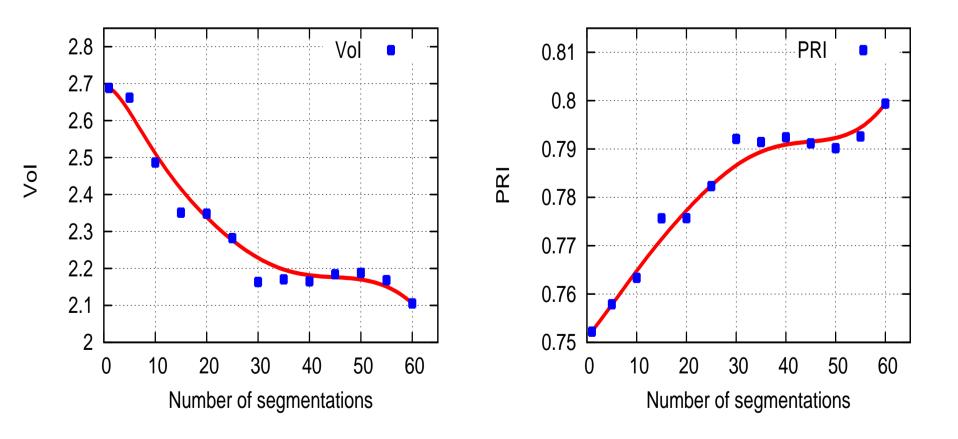
$$E_{\mathsf{Reg}}\left(S = \{C_k\}_{k \le R}\right) = \left|-\sum_{k=1}^{R} \frac{|C_k|}{n} \log \frac{|C_k|}{n} - \overline{\mathcal{R}}\right|$$
(5)

(2)

Performances on BSDS300

Algorithms	$\text{Vol}\downarrow$	$\textbf{BDE}\downarrow$	$\textbf{GCE} \downarrow$	PRI ↑
Humans	1.10	4.99	80.0	0.87
GCEBFM	2.10	8.73	0.19	0.80
PRIF [2]	1.97	8.45	0.21	0.80
FCR [3]	2.30	8.99	0.21	0.79

Progression of the PRI and the Vol according to the number of segmentations (L)



Example of segmentations result

• Measures the extent to which one segmentation map can be viewed as a refinement of another segmentation

 $\mathsf{GCE}^{\star}(S^{\scriptscriptstyle \mathrm{t}},S^{\scriptscriptstyle \mathrm{g}}) =$ $\frac{1}{2n} \left\{ \sum_{i=1}^{n} \mathsf{LRE}(S^{\mathsf{t}}, S^{\mathsf{g}}, p_i) + \sum_{i=1}^{n} \mathsf{LRE}(S^{\mathsf{g}}, S^{\mathsf{t}}, p_i) \right\} (1)$

 $\mathsf{LRE}(S^{\mathsf{t}}, S^{\mathsf{g}}, p_i) = \frac{|C^{\mathsf{t}}_{< p_i >} \backslash C^{\mathsf{g}}_{< p_i >}|}{|C^{\mathsf{t}}_{< p_i >}|}$

with

β : Regularization parameter

 $|C_k|$: Number of pixel in the k-th region C_k of the segmentation map S

 $\overline{\mathcal{R}}$: Internal parameter defines the mean entropy of the a priori defined acceptable segmentation solutions

Optimization

 Based on the ensemble of regions provided by each individual segmentations to be combined

• Algorithm: Iterative Conditional Modes (ICM)[1]



References

[1] J. Besag. On the statistical analysis of dirty pictures. Journal of the Royal Statistical Society., B-48:259-302, 1986.

[2] M. Mignotte. A label field fusion Bayesian model and its penalized maximum Rand estimator for image segmentation. IEEE Trans. Image Process., 19(6):1610-1624, 2010.

[3] M. Mignotte. Segmentation by fusion of histogram-based K-means clusters in different color spaces. IEEE Trans. Image Process., 17:780-787, 2008.