



Consensus-Based Image Segmentation via Topological Persistence

Qian Ge and Edgar Lobaton

Department of Electrical and Computer Engineering, North Carolina State University

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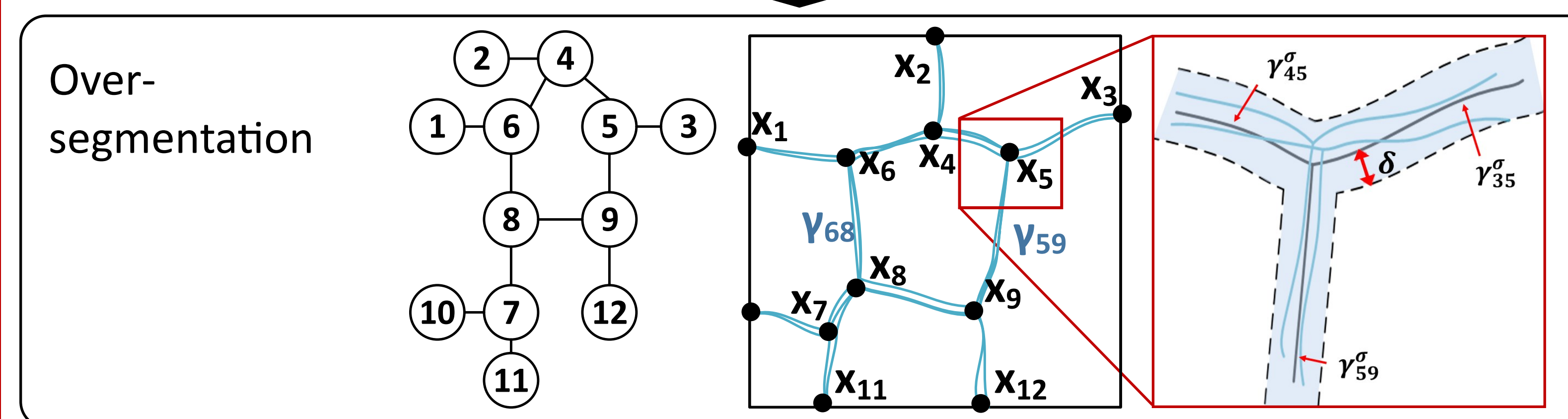
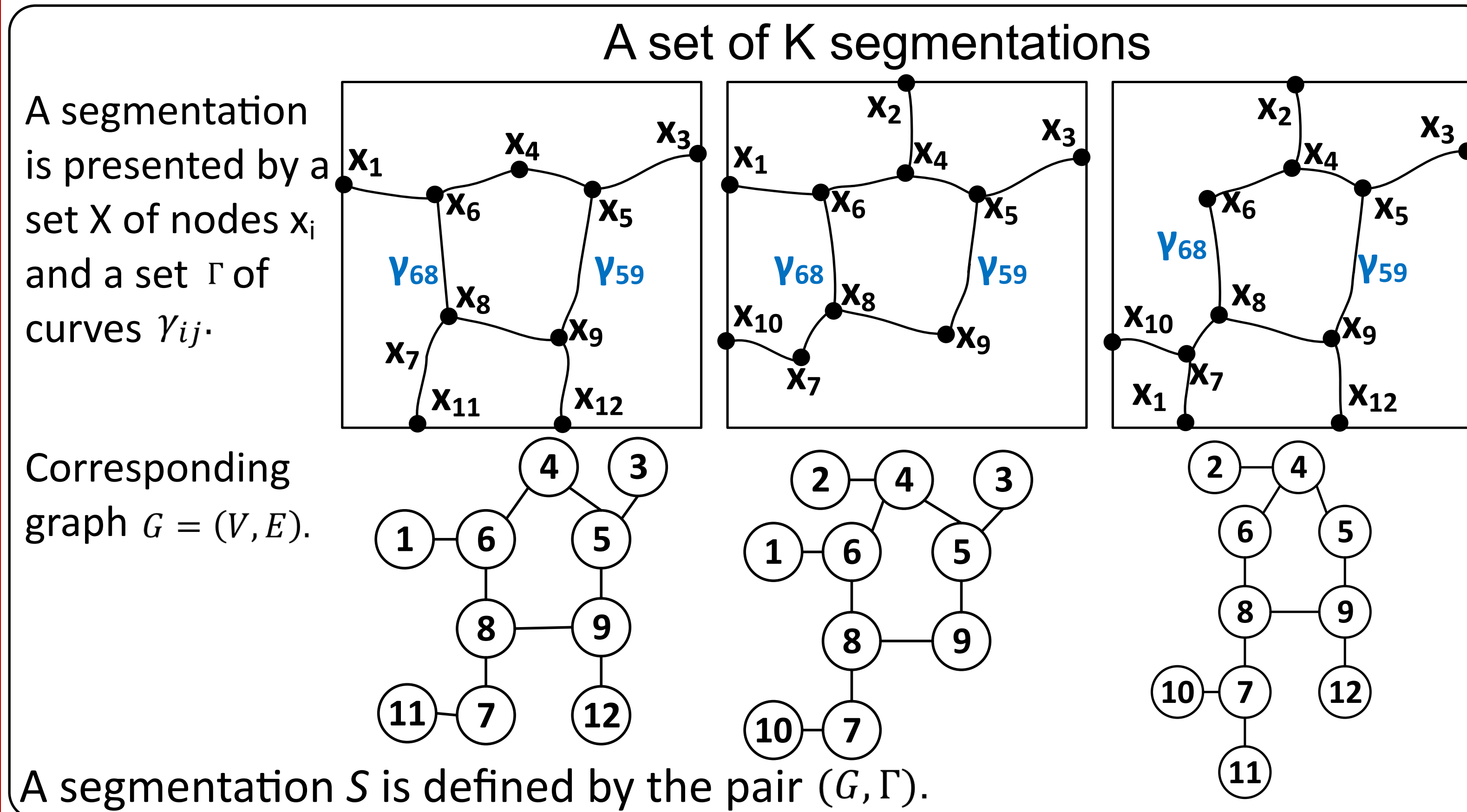
CVPR 2016

Motivation

- It is difficult for a single algorithm with the same parameters to segment all the images successfully due to variations between images.
- The desired boundaries are detected more consistently than other boundaries.

Methodology

Segmentation Model



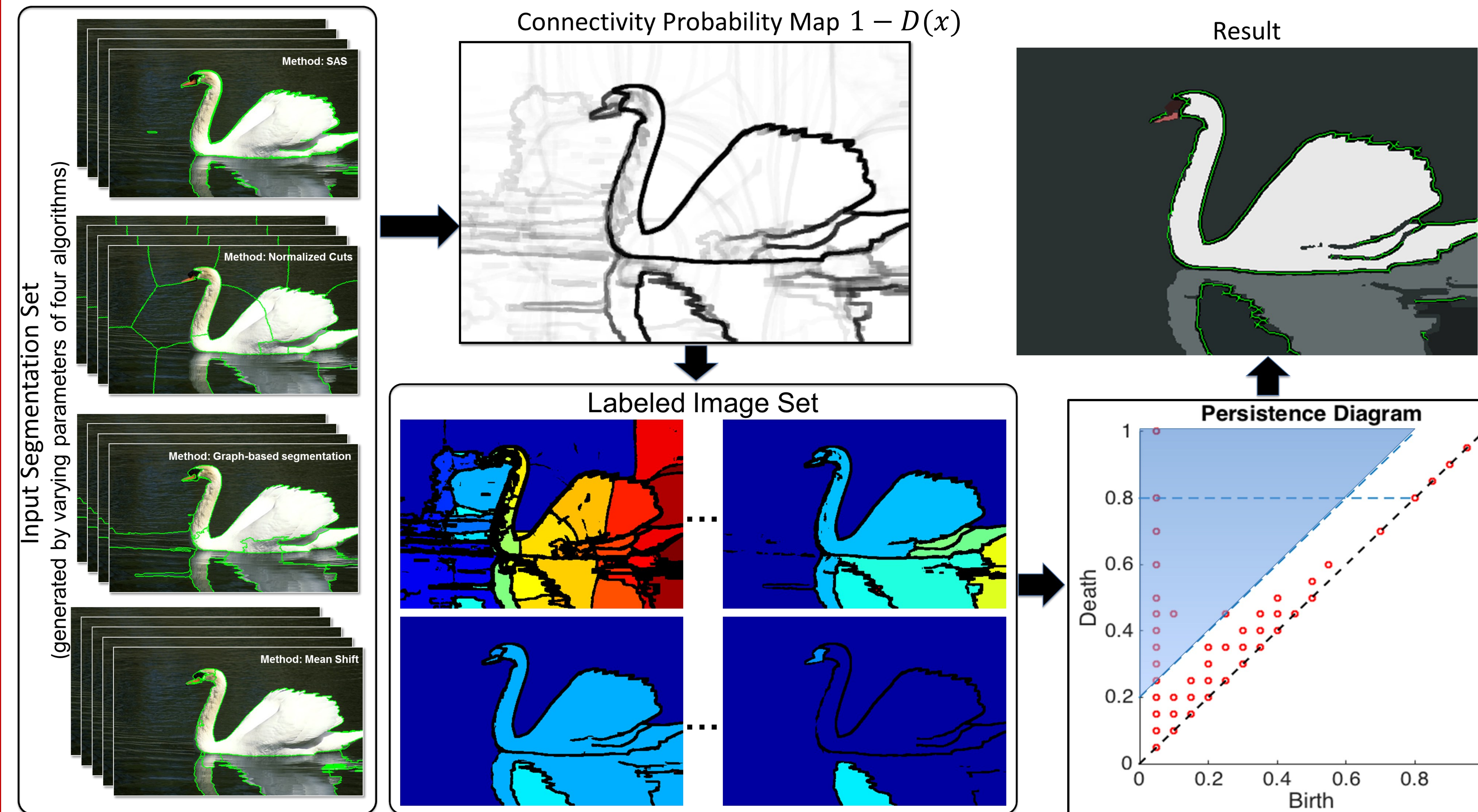
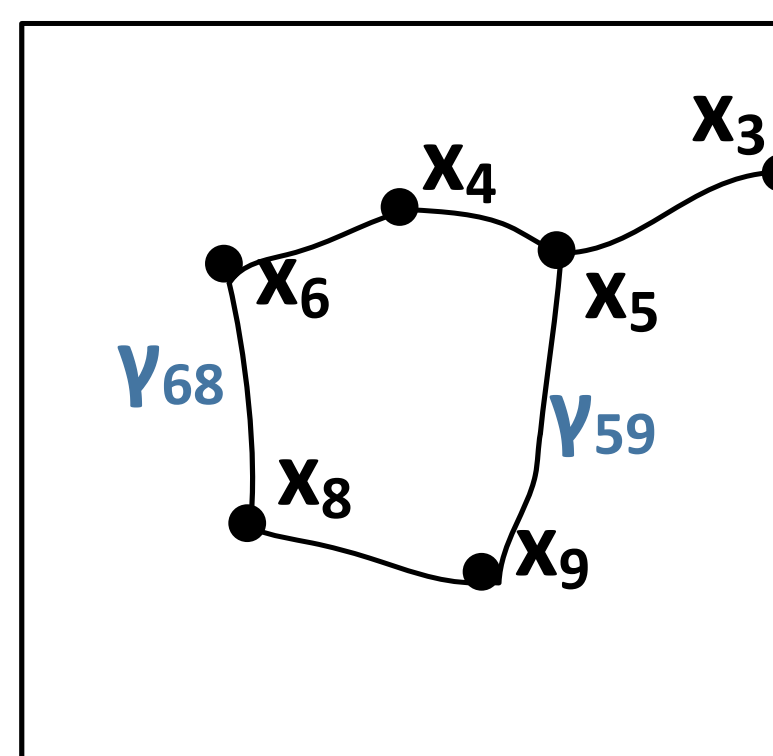
- p_{ij} is defined as the probability of the curve γ_{ij} being present in a segmentation.
- Assume for all the edges with $p_{ij} > \sigma$, there exists a constant δ and a set Γ_σ of curves γ_{ij}^σ that $|\gamma_{ij} - \gamma_{ij}^\sigma|_\infty \leq \delta$.
- Our objective is to estimate $S_\sigma = (G_\sigma, \Gamma_\sigma)$.

Theorem. The set $C = \{x | D_\delta(x) \geq \sigma\}$ satisfies:

$$im(S_\sigma) \subset C \text{ and } im(S_\sigma)^c \ominus B_{2\delta} \subset C^c,$$

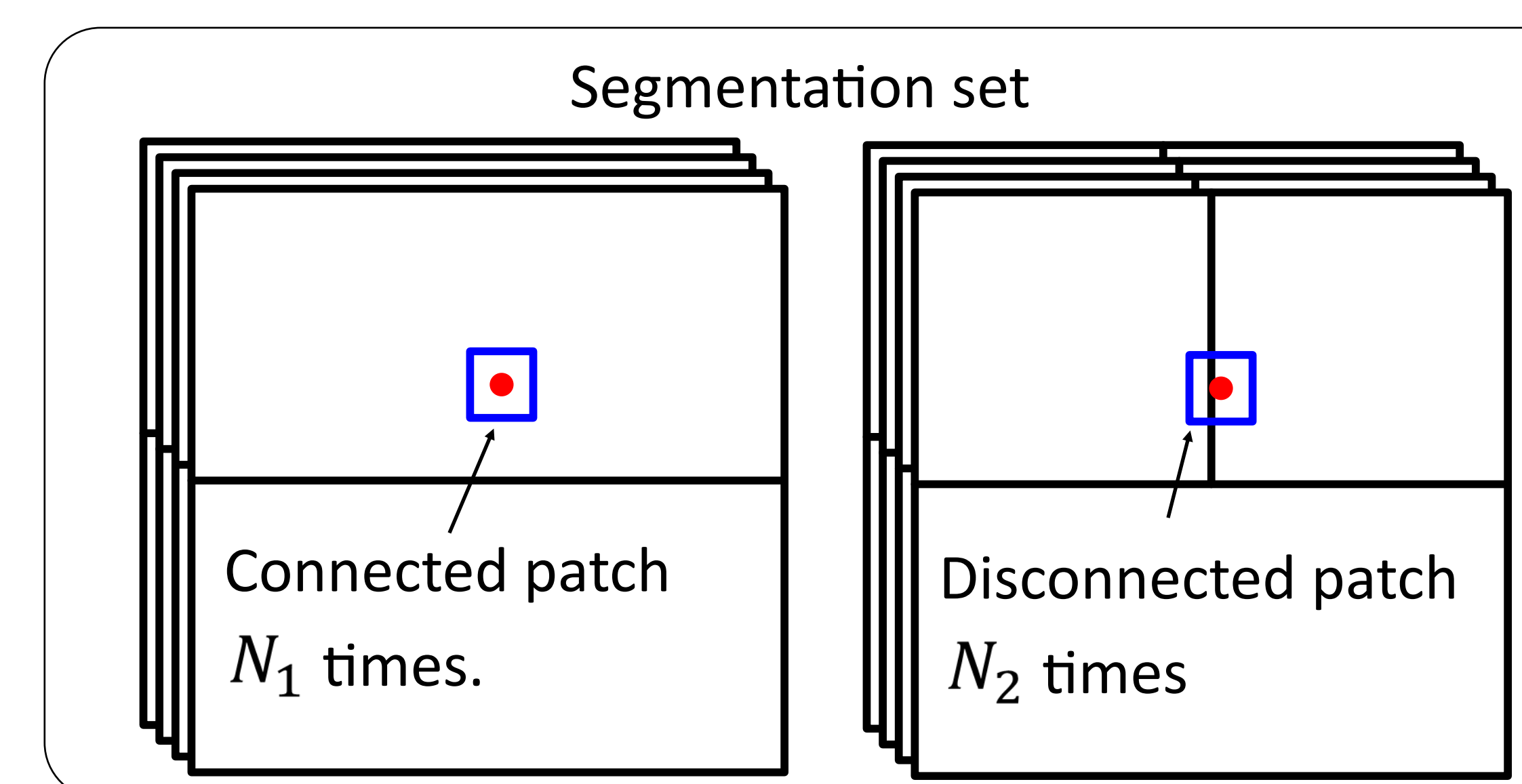
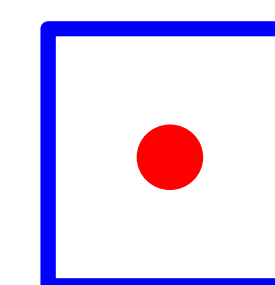
where $im(S_\sigma) = \cup_{(i,j) \in E_\sigma} im(\gamma_{ij}^\sigma)$, $im(\gamma_{ij}^\sigma) = \{x = \gamma_{ij}^\sigma(s) \in \Omega | s \in [0,1]\}$,

$D_\delta(x) := P[im(S) \cap B_\delta(x) \neq \emptyset]$, $B_{2\delta}$ is the ball of radius 2δ and \ominus is the morphological erosion operator.



Boundary Characterization

$D(x)$ is estimated by an $n \times n$ patch centering at x .



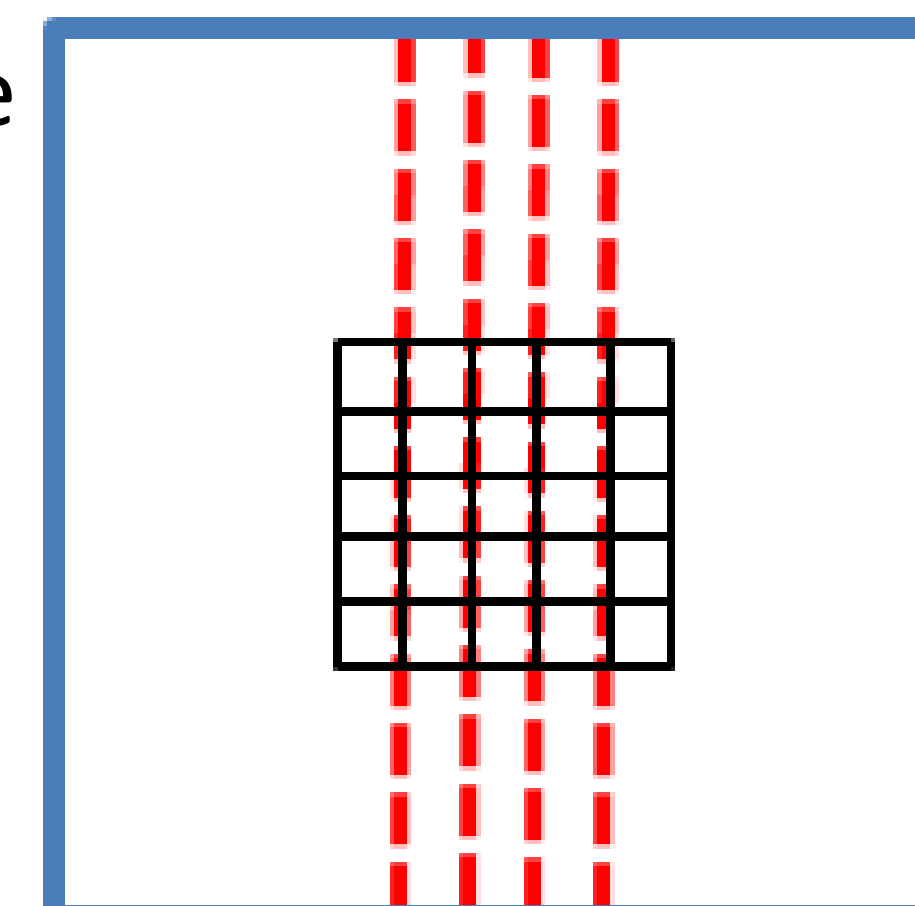
$$D^*(x) = \frac{N_2}{N_1 + N_2}$$

Connectivity probability map is obtained by estimating $D(x)$ for all x in the image.

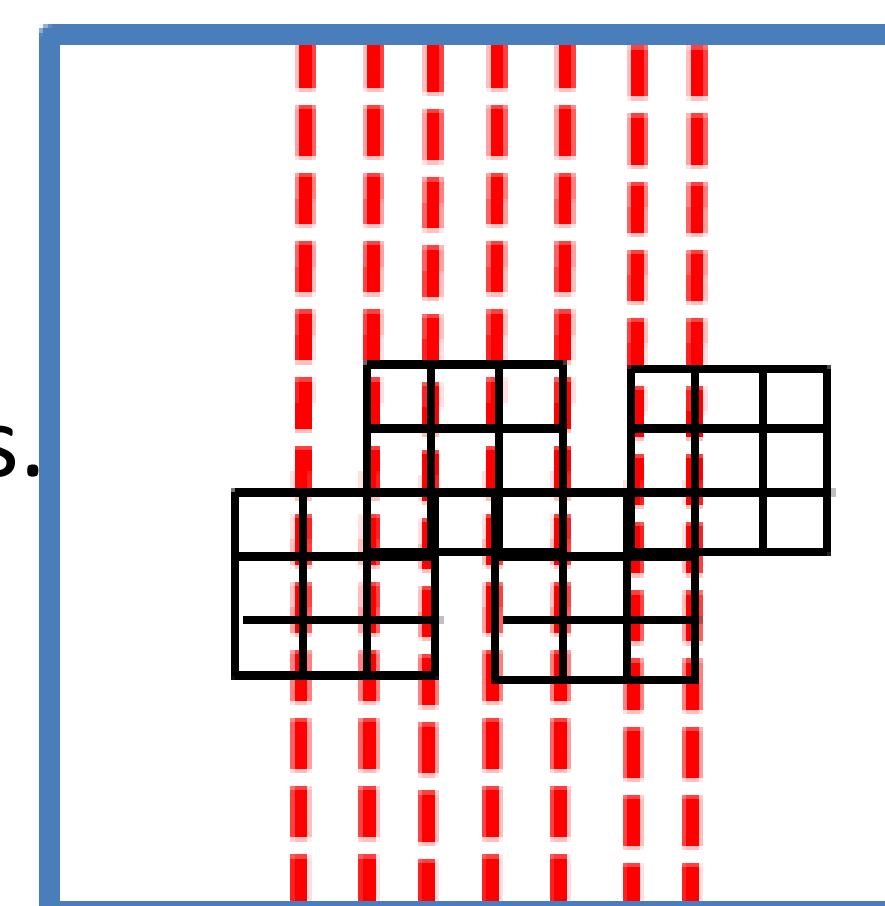
Choice of Parameter n

Let 2σ be the range of perturbation of an edge.

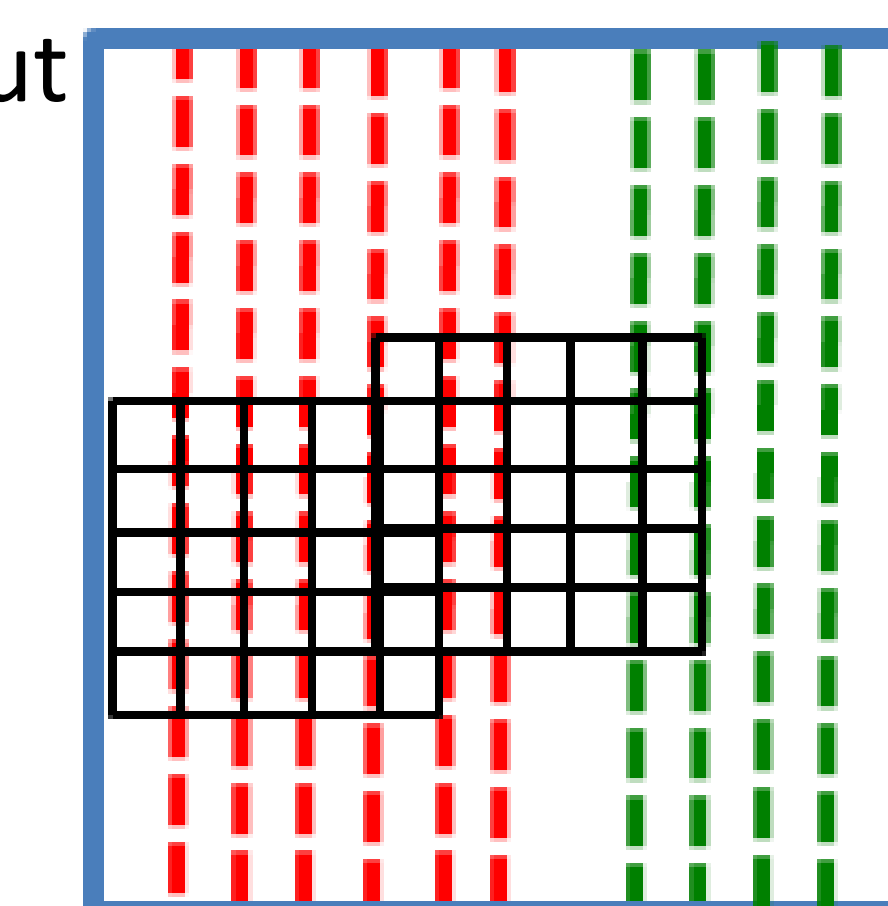
Case 1
 $n - 2 \geq 2\delta, D_n^*(x) = p_{ij}$.
The range can be covered by a single patch.



Case 2
 $n - 1 < 2\delta, p_{ij}/m \leq D_n^*(x) < p_{ij}$, where $m = [(2\delta + 1)/(n - 1)]$.
The range can be covered by m patches.

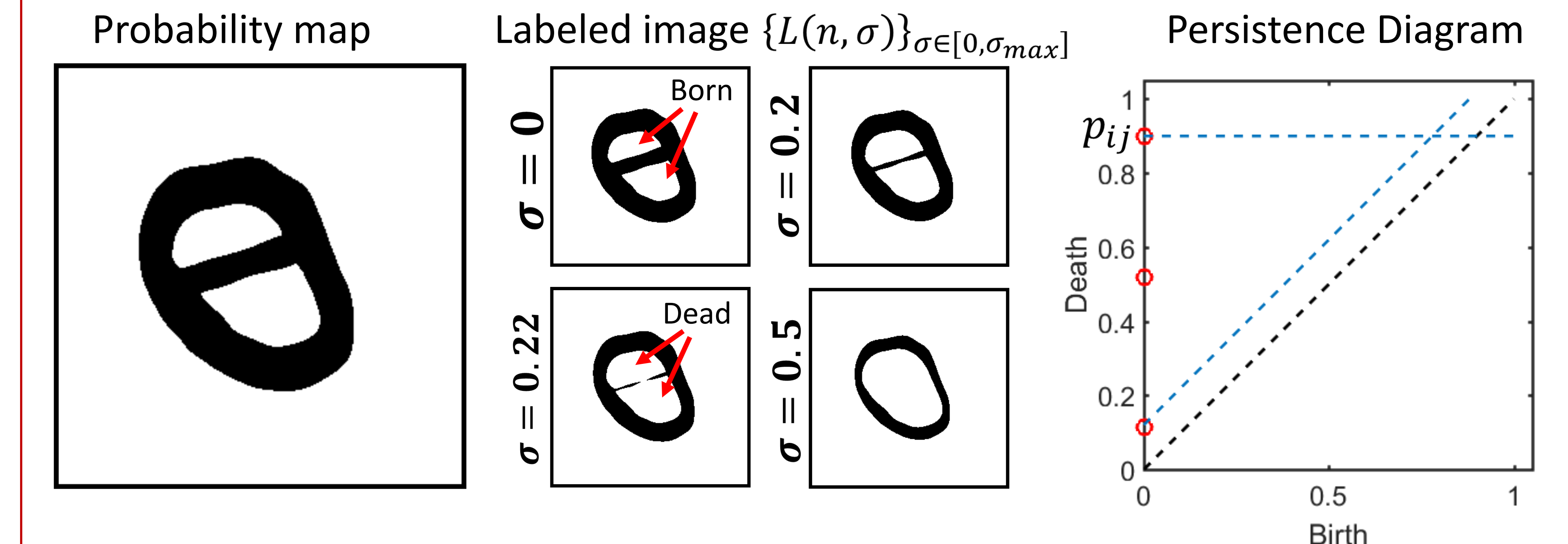


Case 3
Perturbation range of two edges $n-1$ pixels away from each other can be identified without the influence of other edges.



Topological Persistence

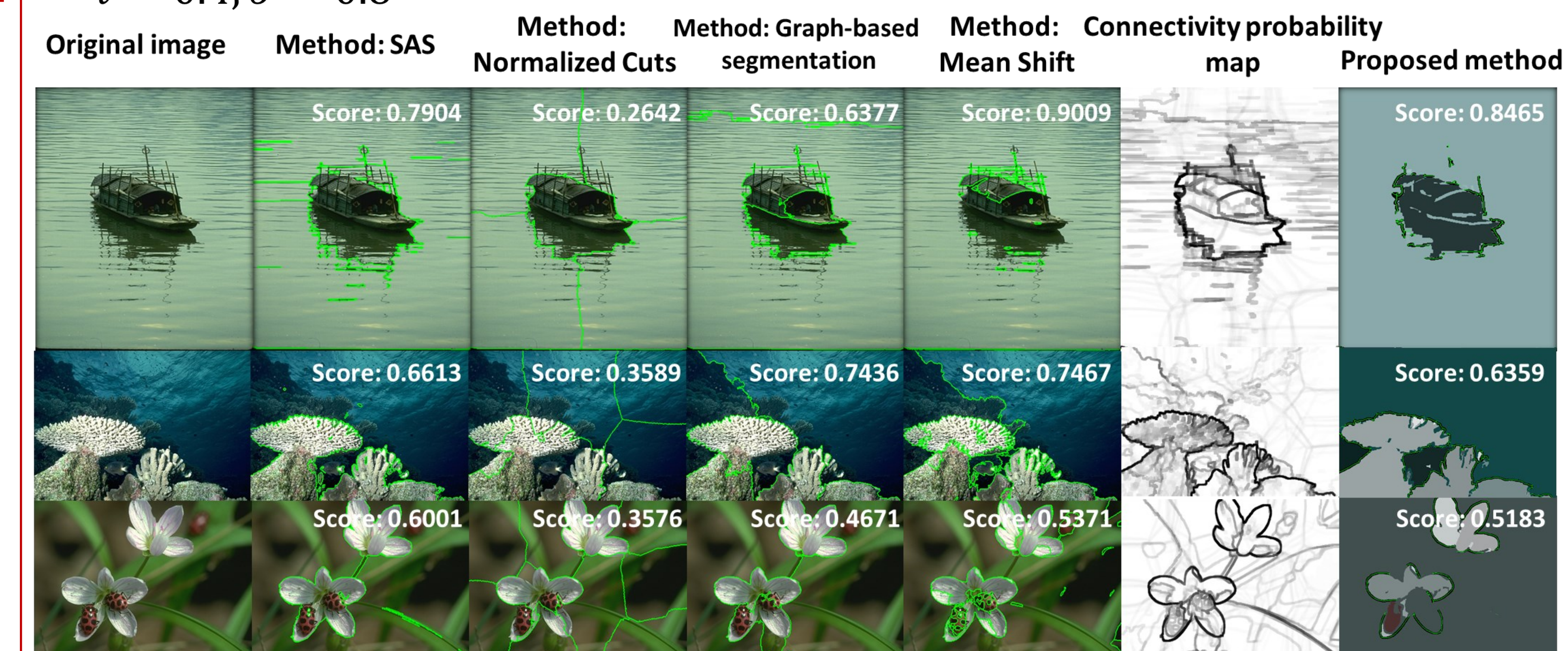
- Extract robust regions by using topological persistence



- Apply a persistence threshold τ to avoid the connected components generated by noise. Set to make $\sigma_{max} = p_{ij}$ sure capture the curves present with probability greater than p_{ij} within a range $2\delta \leq n - 2$.

Results

- $\tau = 0.4, \sigma = 0.8$



Future Work

- Consider multiple neighborhoods of size $n \times n$ (treating n as another parameter).
- Find ways to select an appropriate range of parameters for the base segmentation algorithms.
- Obtain a good estimate for the persistence threshold via training using user-input.
- Identify better metric to qualify the quality of a segmentation.