

Random Walks for Image Segmentation

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Content of the Paper

- An user interactive K-way Image Segmentation Algorithm
- Users define K seeds belonging to K different objects
- Apply an algorithm similar to random walk
- Solve the corresponding Dirichlet problem
- Some theoretical and behavioral properties of the algorithm

Random walk or Drunkard's Walk

A mathematical formalisation of a trajectory that consists of taking successive random steps

Penny Matching Game

- Say John and Peter have 5 pennies
 - On each match Peter **lose or gain a penny with probability 0.5**
 - When Peter have all the five or zero penny the game ends
 - Find **p(x) (probability of Peter's win) with x initial pennies**
-
- $p(0) = 0$
 - $p(N) = 1$
 - $p(x) = \frac{p(x-1) + p(x+1)}{2}$

$$\text{How } p(x) = \frac{p(x-1)+p(x+1)}{2} \ ?$$

E: Probability that Peter wins if it starts with x pennies

F: The first step is to loose a penny

G: The first step is to gain a penny

$$P(E) = P(E/F)P(F) + P(E/G)P(G)$$

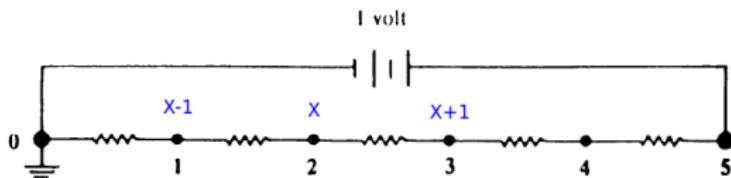
$$\text{Now, } P(E) = p(x) \text{ and } P(F) = P(G) = \frac{1}{2}$$

$$P(E/F) = p(x-1) \text{ and } P(E/G) = p(x+1)$$

So,

$$p(x) = \frac{p(x-1)+p(x+1)}{2}$$

Electric Circuit Analogue



By applying Ohm's law and Kirchoff's law: $v(x) = \frac{v(x-1) + v(x+1)}{2}$

If resistors between x and $x + 1$ is R_x :

$$v(x) = \frac{\frac{1}{R_{x-1}}}{\frac{1}{R_{x-1}} + \frac{1}{R_x}} v(x-1) + \frac{\frac{1}{R_x}}{\frac{1}{R_{x-1}} + \frac{1}{R_x}} v(x+1)$$

Harmonic Function

Is a function that satisfies Laplace's equation: $\nabla^2 u = 0$

In 1-dimensional:

Let, $S = \{0, 1, 2, \dots, N\}$, $D = \{1, 2, \dots, N - 1\}$, $B = \{0, N\}$

A function $f(x)$ defined on S is harmonic if: $f(x) = \frac{f(x-1) + f(x+1)}{2}$

Dirichlet Problem: Find a harmonic function given it's boundary values.

Uniqueness Principle: There can't be two different harmonic functions having same boundary values.

Qualities of Interactive Segmentation

- Fast Computation
- Fast Editing
- Limited interaction
- Intuitive segmentation

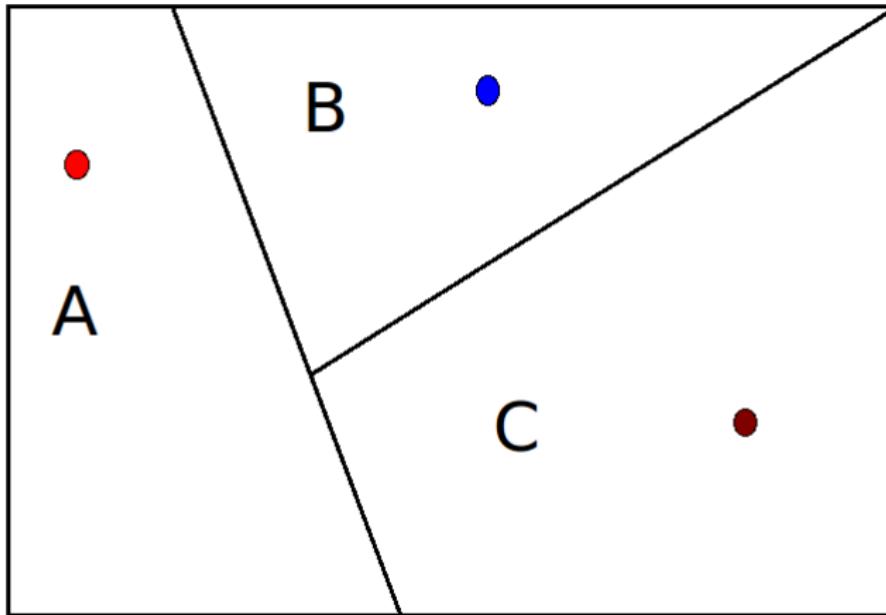
This algorithm has:

- All four of these properties
- **noise robustness**
- **avoids small cut**

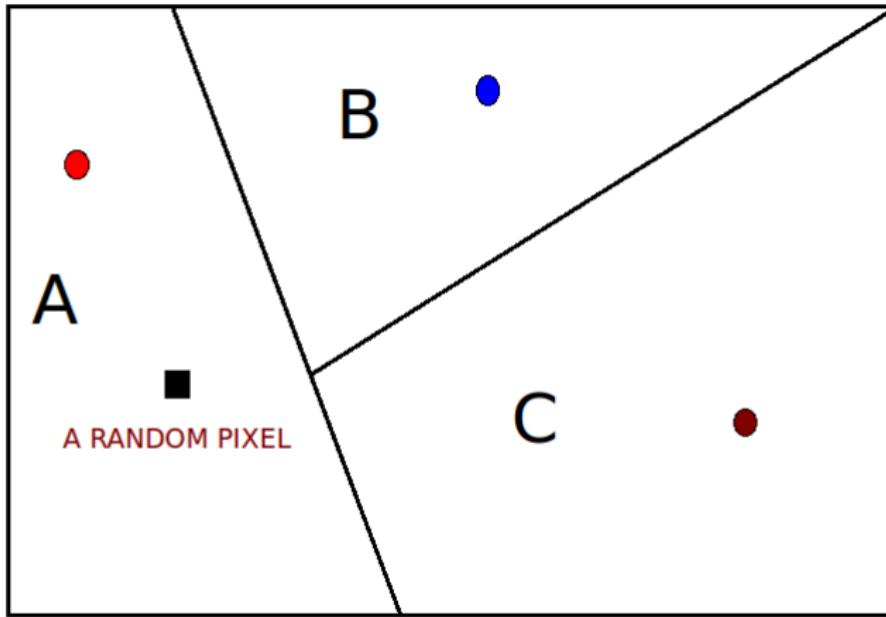
Graph on Image

- We can define a **connected** and **undirected** graph on the image
 $G = (V, E)$
- A vertex is v_i , an edge between v_i and v_j is e_{ij}
- Degree of a vertex v_i is $d_i = \sum w(e_{ij})$
- $w_{ij} = \exp(-\beta(\text{intensity}(g_i) - \text{intensity}(g_j)))$

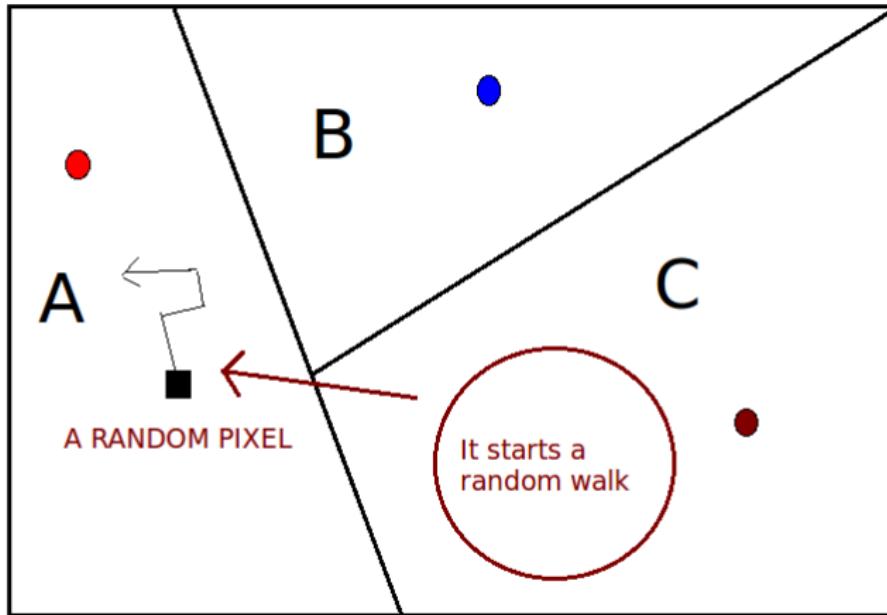
Random walk on a finite graph is a path of length k , which starts at some node O and continue selecting the next node randomly from it's neighbours, until it has $k-1$ transitions.



An Image

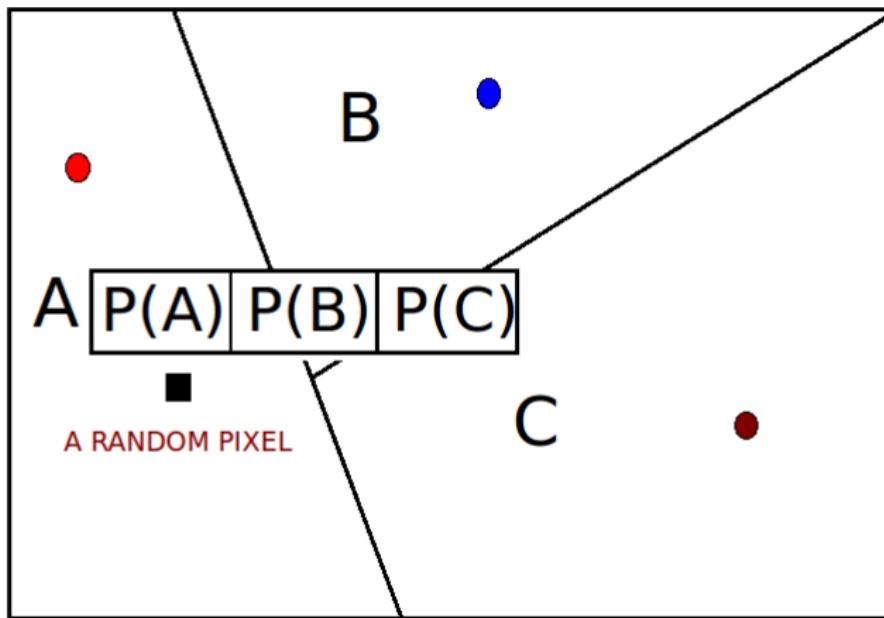


An Image



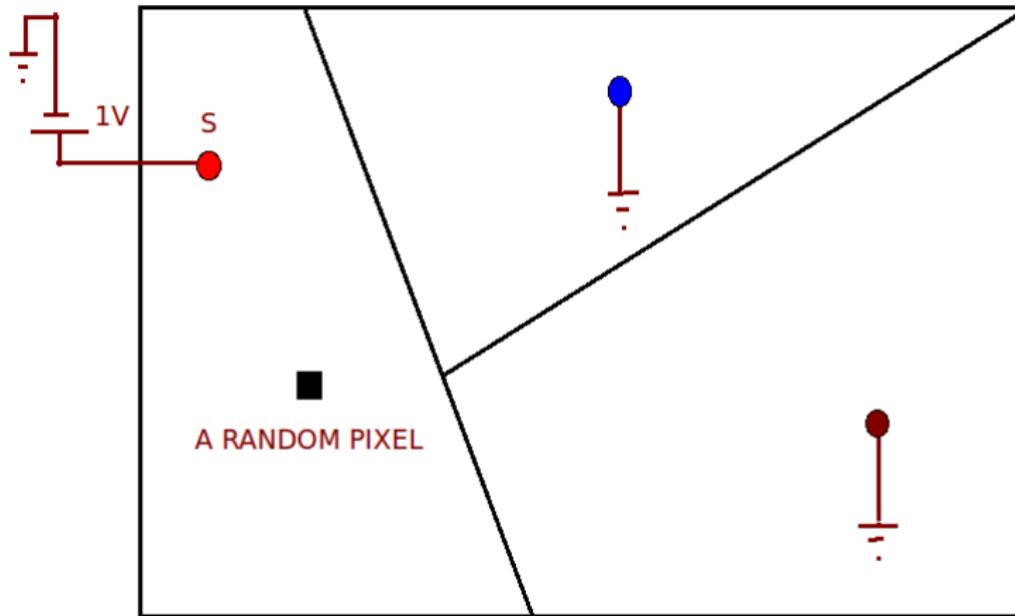
An Image

$$P(A) + P(B) + P(C) = 1$$



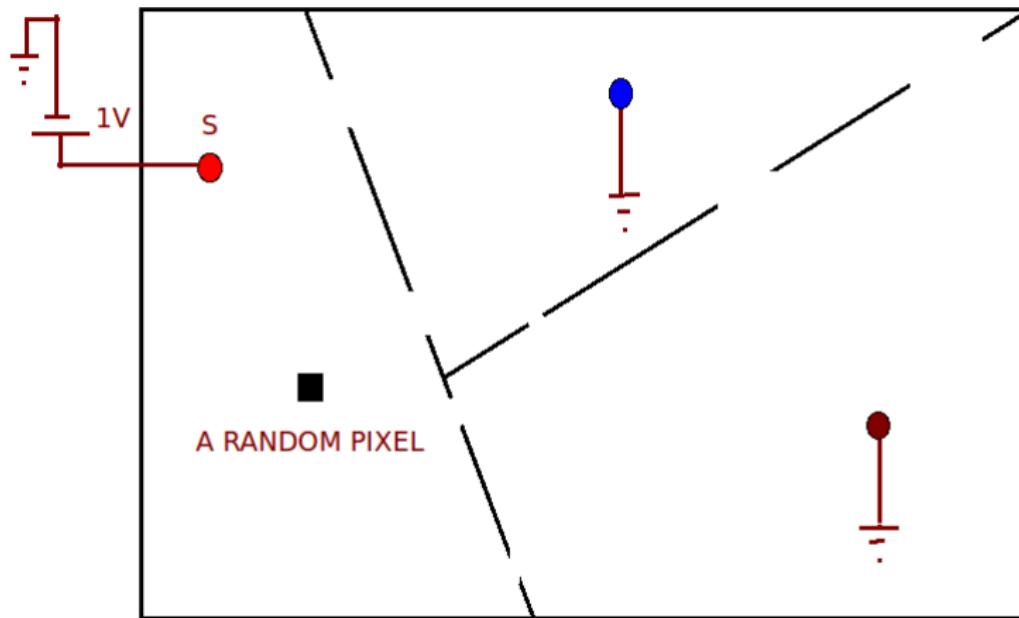
An Image

ELECTRIC ANALOGUE AGAIN !!!



An Image

What if boundaries of image are not well defined ?



An Image

Some Definitions

Dirichlet Integral:

$$D[u] = \frac{1}{2} \int_{\Omega} |\nabla u|^2 d\Omega$$

Combinatorial Laplacian Matrix:

$$L_{ij} =$$

$$\begin{cases} d_i, & \text{if } i=j \\ -w_{ij}, & \text{if } v_i \text{ and } v_j \text{ are adjacent nodes} \\ 0, & \text{otherwise} \end{cases}$$

$m \times n$ edge node Incidence Matrix:

$$A_{e_{ij}v_k} =$$

$$\begin{cases} +1 & \text{if } i=k \\ -1 & \text{if } j=k \\ 0, & \text{otherwise} \end{cases}$$

Some more definitions

Constitutive Matrix (C)

Diagonal matrix with weights of each edge along the diagonal.

As an operator:

A : **Gradient** Operator

A^T : Combinatorial **Divergence** Operator

So, Combinatorial **Laplacian**: $L = A^T A$ and

Combinatorial **Laplace-Beltrami** Operator: $L = A^T C A$

Dirichlet Again

$$D[x] = \frac{1}{2}(Ax)^T C(AX) = \frac{1}{2}x^T Lx = \frac{1}{2} \sum_{e_{ij} \in E} w_{ij} (x_i - x_j)^2$$

V_M : Marked Nodes

V_U : Unmarked Nodes

$$\begin{aligned} D[x_U] &= \frac{1}{2} [x_M^T x_U^T] \begin{bmatrix} L_M & B \\ B^T & L_U \end{bmatrix} \begin{bmatrix} x_M \\ x_U \end{bmatrix} = \\ &\quad \frac{1}{2} (x_M^T L_M x_M + 2x_U^T B^T x_M + x_U^T L_U x_U) \end{aligned}$$

Differentiating $D[x_U]$ with respect to x_U and finding the critical points yield:

$$L_U x_U = -B^T x_M$$

x_i^s : Probability/potential assumed at node v_i for each label s

$$Q(v_j) = s, \text{ for } v_j \in V_M, s \in \mathbb{Z}, 0 \leq s \leq K$$

Define a column vector m_s of length $|V_M|$ for a label s and $v_j \in V_M$

$$m_j^s =$$

$$\begin{cases} 1 & \text{if } Q(v_j) = s \\ 0 & \text{if } Q(v_j) \neq s \end{cases}$$

So for label s , the solution for Combinatorial Dirichlet problem is found by solving

$$L_U x^s = -B^T m^s \text{ for one label or,}$$

$$L_U X = -B^T M$$

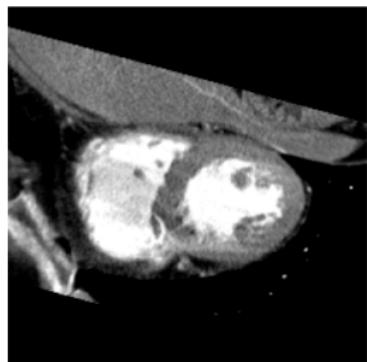
for all labels, where X and M have the columns x^s and m^s respectively.

We also have the constraint:

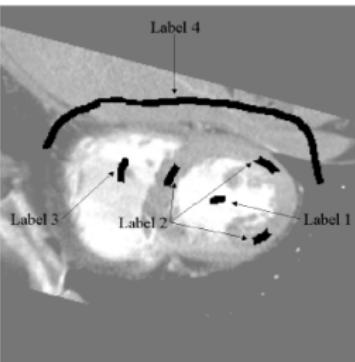
$$\sum_s x_i^s = 1, \text{ For all } v_i \in V,$$

So we have $K-1$ sparse linear equations to solve !!!

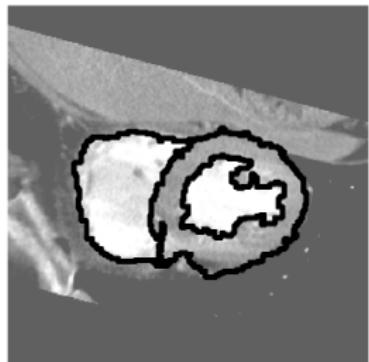
Experimental Results



(a) Original



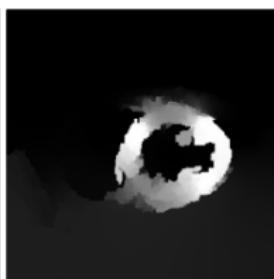
(b) Seeds indicating four objects



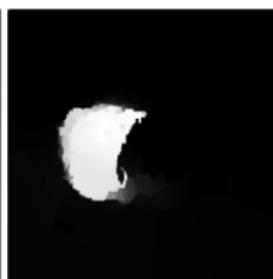
(c) Resulting segmentation



(d) Label 1 probabilities



(e) Label 2 probabilities



(f) Label 3 probabilities



(g) Label 4 probabilities

Comments and Conclusion

- A segmentation algorithm with user interaction
- Random walk has been used very nicely
- There are some nice discussions on theoretical and behavioral properties of the algorithm
- Not sure why there is no experiment with natural images
- The algorithm works well with weak boundaries and for noisy images

