## Directed Probabilistic Watershed



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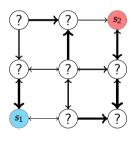
HCI/IWR at Heidelberg University



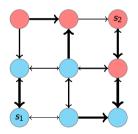




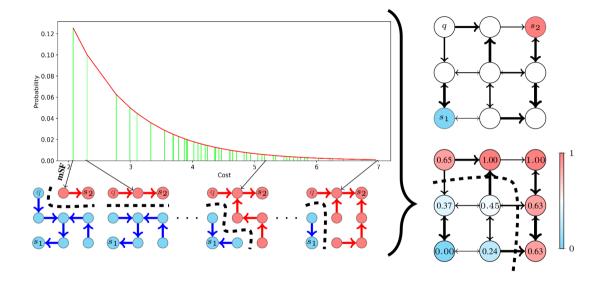
# Transductive semi-supervised learning algorithm on **directed** graphs



⇒ Directed Probabilistic Watershed



- Web graphs
- Citation graphs



### Probabilistic Watershed

Undirected graphs

Fita Sanmartín et al. (2019)

Forests

#### Directed Probabilistic Watershed

Directed graphs

In-forests rooted at the seeds

Probabilistic Watershed: Sampling all spanning forests for seeded segmentation and semi-supervised learning, Fita Sanmartín et al. (2019)

#### Probabilistic Watershed

Undirected graphs

Fita Sanmartín et al. (2019)

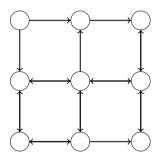
Directed Probabilistic Watershed

Directed graphs

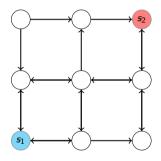
Forests  $\Longrightarrow$ 

In-forests rooted at the seeds

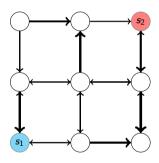
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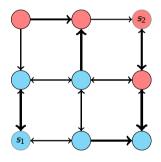
Directed Graph



- Directed Graph
- Seeds (labeled nodes)

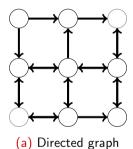


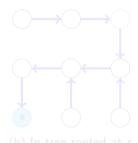
- Directed Graph
- Seeds (labeled nodes)
- Edge-Costs,  $c_e$  ( $\sim$ affinity between nodes)



- Directed Graph
- Seeds (labeled nodes)
- Edge-Costs,  $c_e$  ( $\sim$ affinity between nodes)
- Classification

#### Definition in-tree & in-forest

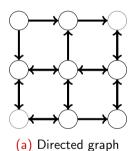


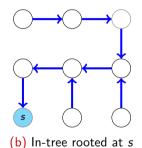


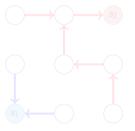


(c) In-forest rooted at  $s_1$  and  $s_2$ 

#### Definition in-tree & in-forest

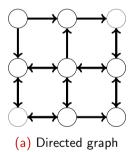


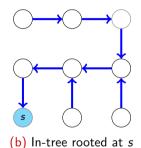


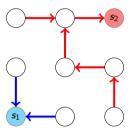


(c) In-forest rooted at  $s_1$  and  $s_2$ 

#### Definition in-tree & in-forest



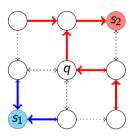




(c) In-forest rooted at  $s_1$  and  $s_2$ 

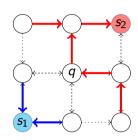
### Main question

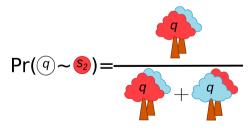
What is the probability of sampling an in-forest such that a node of interest, *q*, belongs to a tree rooted at a certain seed?



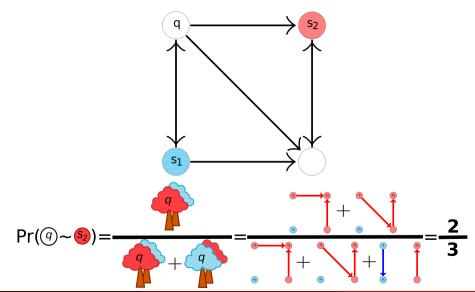
### Main question

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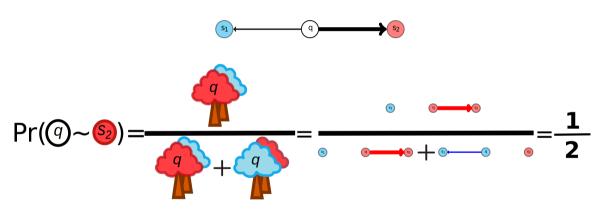


### Directed Probabilistic Watershed Probabilities



Enrique Fita Sanmartín

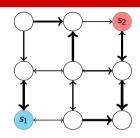
#### Directed Probabilistic Watershed Probabilities

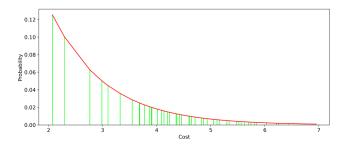


### Gibbs distribution

#### Gibbs distribution

$$\Pr(F) \propto w(F) := \exp\left(-\mu \underbrace{c(F)}_{e \in F}\right)$$

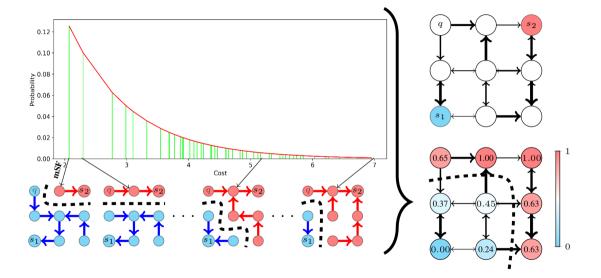




### Directed Probabilistic Watershed Probabilities

$$Pr(q \sim s_2) = \frac{W(q)}{W(q)} + W(q)$$

#### Directed Probabilistic Watershed Probabilities



### Computation Probabilities Directed Probabilistic Watershed

Number in-forests increases exponentially with the number of nodes and edges



Naive approach infeasible

An elementary proof of a matrix tree theorem for directed graphs, Leenheer (2019)

### Computation Probabilities Directed Probabilistic Watershed

Number in-forests increases **exponentially** with the number of nodes and edges

 $\Rightarrow$  Naive approach infeasible

Directed Matrix Tree Theorem

 $\Rightarrow$  Effic

Efficient computation probabilities

#### Linear System

$$L_U^{\top} x_U^{s_2} = -[B_1^{\top}]_{s_2}$$

An elementary proof of a matrix tree theorem for directed graphs, Leenheer (2019)

### Comparison linear systems

#### Probabilistic Watershed

$$L_U x_U^{s_2} = -[B_1^{\top}]_{s_2}$$

#### Directed Probabilistic Watershed

$$L_U^{\top} x_U^{s_2} = -[B_1^{\top}]_{s_2}$$

Probabilistic Watershed: Sampling all spanning forests for seeded segmentation and semi-supervised learning, Fita Sanmartín et al. (2019)

### Equivalence Random Walker

#### Probabilistic Watershed

Random Walker (Seed absorption probabibility) Grady (2006)

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Probabilisic Watershed
Fita (2019)

#### Directed Probabilistic Watershed

Directed Random Walker (Seed absorption probabibility)

 $\sim$ 

Directed Probabilisic Watershed

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Random walks for image segmentation, Grady (2006)

### Minimum entropy case

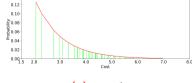
#### Gibbs distribution

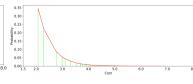
$$\Pr(F) \propto \exp\left(-\mu c(F)\right) = \exp\left(-\mu \sum_{e \in F} c_e\right)$$

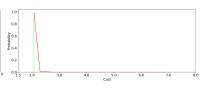
$$\mu \to \infty \Longrightarrow$$

Minimum entropy

Count minimum cost in-forests (in-mSF)







(a)  $\mu = 1$ 

(b)  $\mu = 2$ 

(c)  $\mu = 20$ 

### Minimum entropy case

#### Probabilistic Watershed

Power Watershed

Couprie et al. (2011)

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Minimum entropy Probabilisic Watershed Fita (2019)

#### Directed Probabilistic Watershed

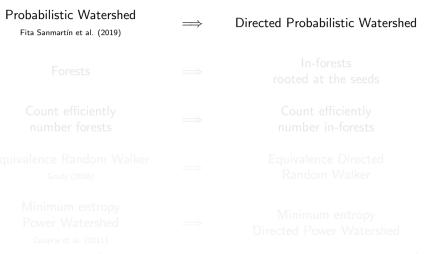
Directed Power Watershed

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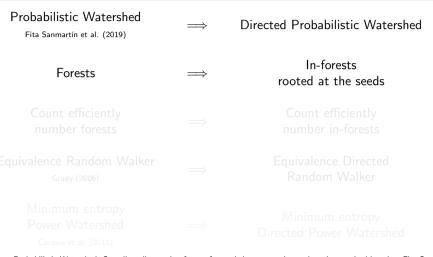
Minimum entropy
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Power Watershed: A Unifying Graph-Based Optimization Framework, Couprie at al. (2011)

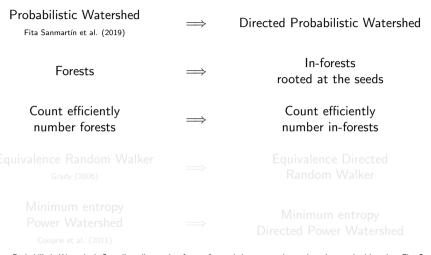


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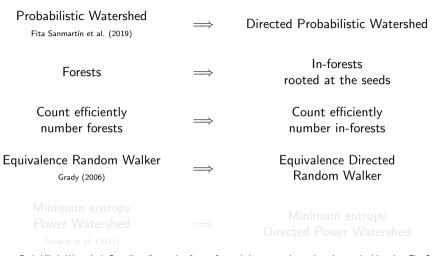
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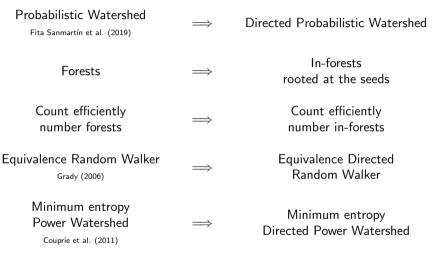


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Thank you for your attention