



# Probabilistic Watershed: Sampling all spanning forests for seeded segmentation and semi-supervised learning

Enrique Fita Sanmartín, Sebastian Damrich, Fred A. Hamprecht

HCI/IWR at Heidelberg University

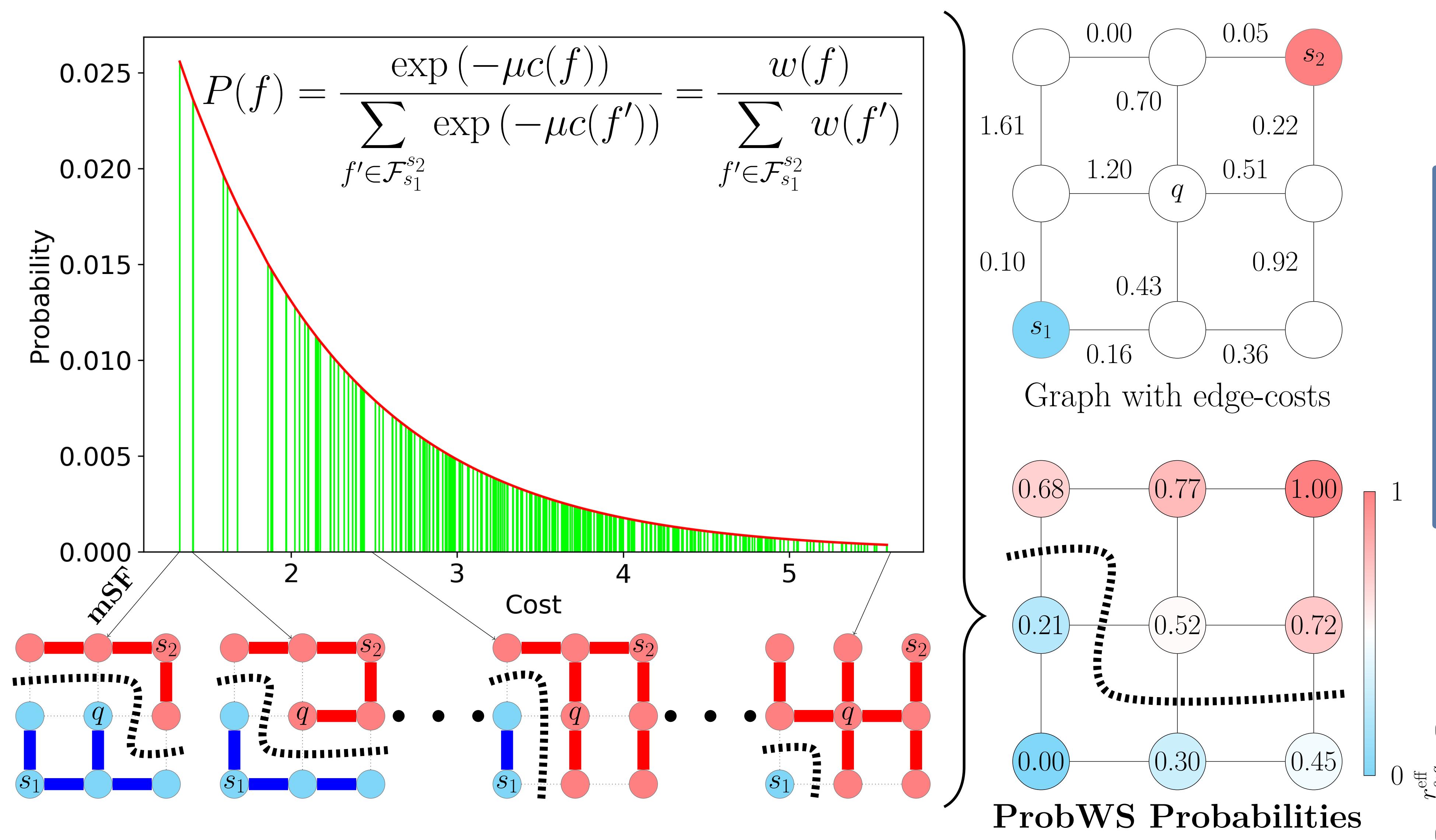
## Approach

### Question

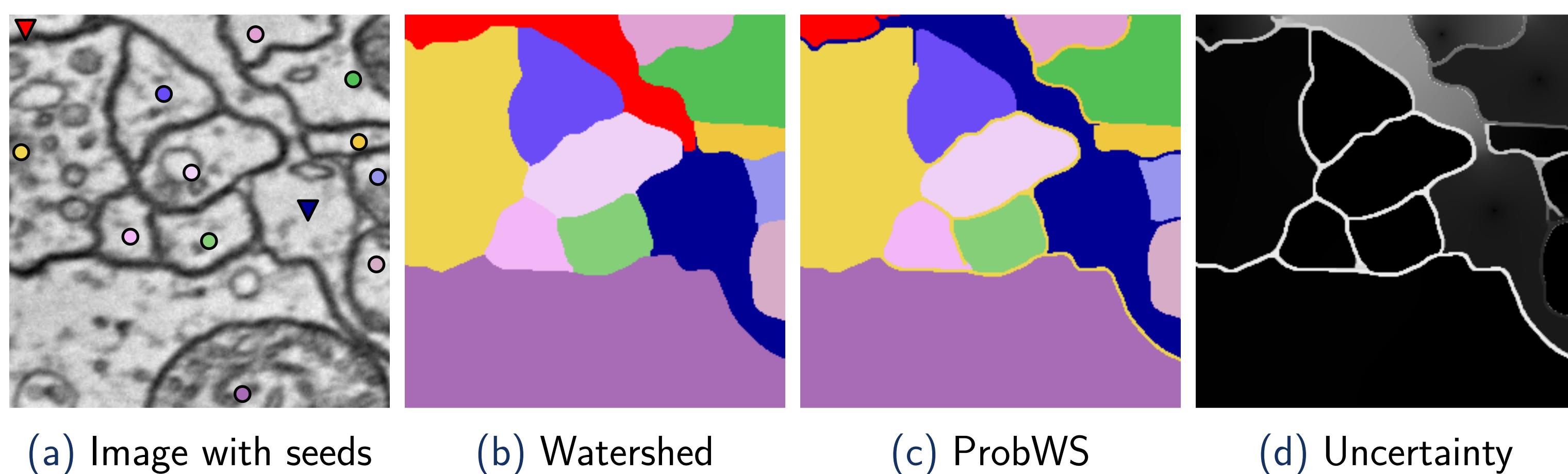
What is the probability of sampling a Gibbs-distributed forest such that a node of interest is connected to a certain seed? What is the relation between the probability and the cost of a forest?

- Seeded **Watershed** computes a minimum spanning cost forest (**mSF**) → "winner-takes-all".
- The **Probabilistic Watershed (ProbWS)** takes all spanning forests into account according to their cost by means of a Gibbs distribution over the spanning forests.

$$P(q \sim s_1) := \sum_{f \in \mathcal{F}_{s_1,q}^{s_2}} P(f) = \frac{\sum_{f \in \mathcal{F}_{s_1,q}^{s_2}} w(f)}{\sum_{f' \in \mathcal{F}_{s_1}^{s_2}} w(f')} = \frac{w(\mathcal{F}_{s_1,q}^{s_2})}{w(\mathcal{F}_{s_1}^{s_2})}.$$



- Applications: Seeded segmentation.



## Computation $w(\mathcal{F}_{s_1}^{s_2})$ , $w(\mathcal{F}_{s_2,q}^{s_1})$ and $w(\mathcal{F}_{s_1,q}^{s_2})$

- $r_{uv}^{\text{eff}} :=$  Effective resistance distance between the nodes  $u$  and  $v$ .
- As a consequence of the **Matrix Tree Theorem** [3] we are able to find a **closed-form computation of the weight of the set of forests**:

$$w(\mathcal{F}_{s_1}^{s_2}) = w(\mathcal{T}) r_{s_1 s_2}^{\text{eff}} \propto r_{s_1 s_2}^{\text{eff}}.$$

- We define the following linear system:

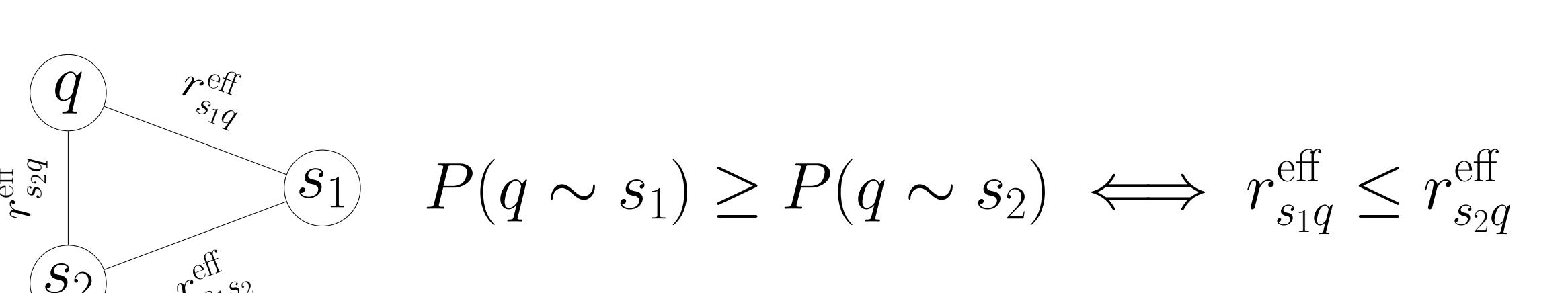
$$\begin{aligned} f \in \mathcal{F}_{s_2,q}^{s_1} & \quad f \in \mathcal{F}_{s_1,q}^{s_2} & w(\mathcal{F}_{s_1,q}^{s_2}) + w(\mathcal{F}_{s_1,s_2}^q) &= w(\mathcal{F}_{s_2}^q) \\ & \quad \times \quad | \quad | & w(\mathcal{F}_{s_1,s_2}^q) + w(\mathcal{F}_{s_2,q}^{s_1}) &= w(\mathcal{F}_{s_1}^{s_2}) \\ & \quad \times \quad | \quad | & w(\mathcal{F}_{s_2,q}^{s_1}) + w(\mathcal{F}_{s_1,q}^{s_2}) &= w(\mathcal{F}_{s_2}^{s_1}). \end{aligned}$$

$f \in \mathcal{F}_{s_1}^{s_2} = \mathcal{F}_{s_1,q}^{s_2} \cup \mathcal{F}_{s_2,q}^{s_1}$

## Closed-form ProbWS

$$\begin{aligned} P(q \sim s_1) &= \frac{w(\mathcal{F}_{s_2}^q) + w(\mathcal{F}_{s_1}^{s_2}) - w(\mathcal{F}_{s_1}^q)}{2w(\mathcal{F}_{s_1}^{s_2})} \\ &= \frac{r_{s_2 q}^{\text{eff}} + r_{s_2 s_1}^{\text{eff}} - r_{s_1 q}^{\text{eff}}}{2r_{s_2 s_1}^{\text{eff}}}. \end{aligned}$$

- ProbWS is proportional to the effective resistance's triangle inequality gap.



## Random Walker = ProbWS

The probability,  $x_q^{s_2}$ , that a Random Walker starting at node  $q$  reaches first  $s_2$  before reaching  $s_1$  ([2]) is equal to the probability defined by the ProbWS.

$$x_q^{s_2} = P(q \sim s_2).$$

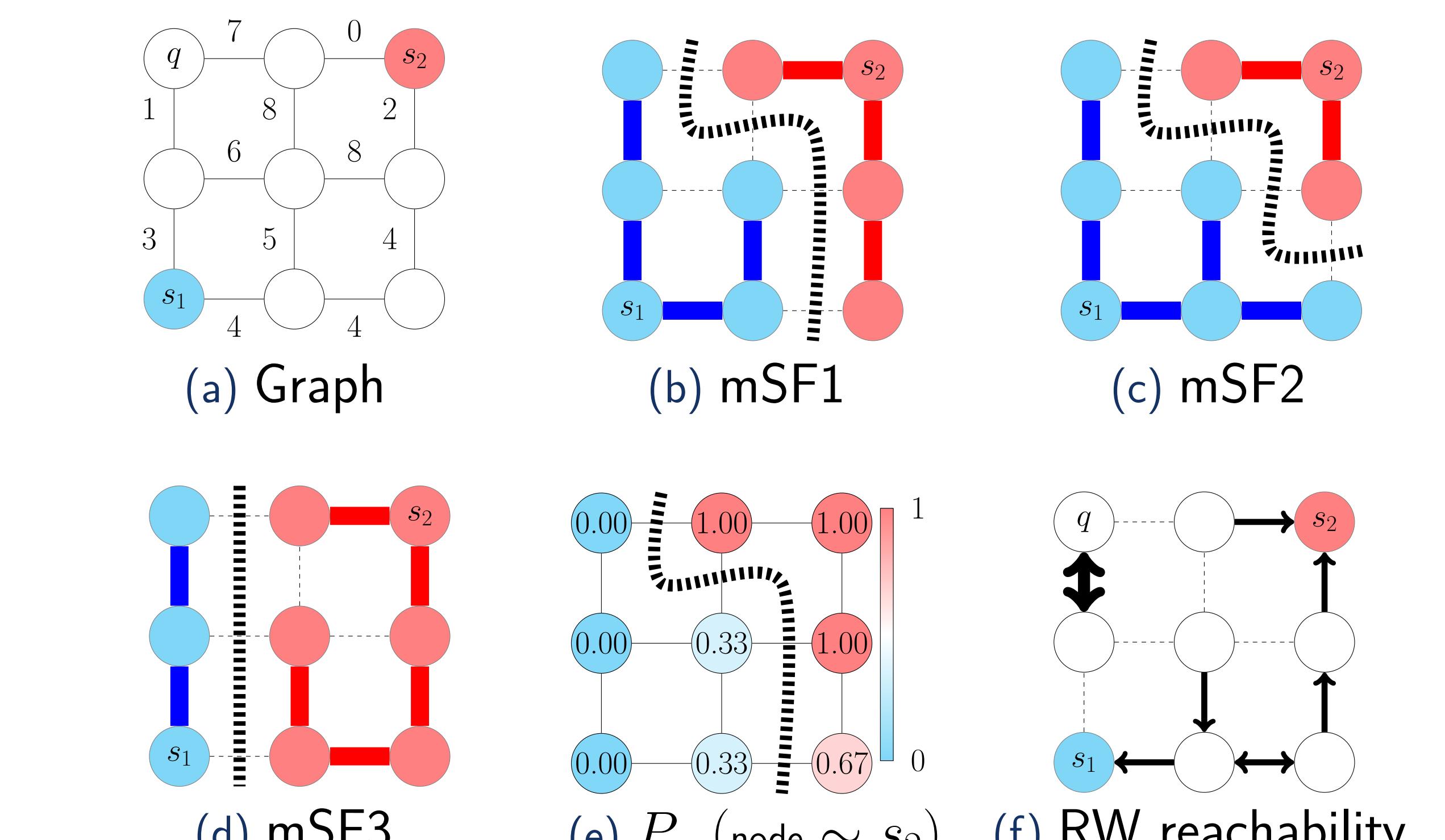
## Power Watershed

Power Watershed [1] was introduced as a limit of the Random Walker.

### Power Watershed counts mSF

Given  $S = \{s_1, s_2\}$  a set of seeds, let us denote the potential of node  $q$  being assigned to seed  $s_1$  by the Power Watershed[1] with  $\beta = 2$  as  $x_q^{\text{PW}_1}$ . Let further  $c_{\min}$  be  $\min_{f \in \mathcal{F}_{s_1}^{s_2}} c(f)$ . Then

$$x_q^{\text{PW}_1} = P_\infty(q \sim s_1) := \frac{|\{f \in \mathcal{F}_{s_1,q}^{s_2} : c(f) = c_{\min}\}|}{|\{f \in \mathcal{F}_{s_1}^{s_2} : c(f) = c_{\min}\}|}.$$

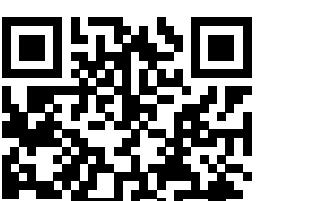


## Contributions

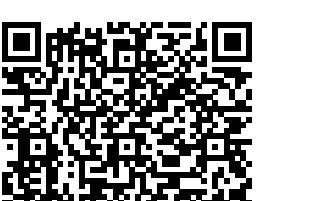
- Analytical calculus of the Probabilistic Watershed probability, i.e. the probability that a node is assigned to a particular seed in an ensemble of Gibbs distributed spanning forests.
- Probabilistic Watershed (ProbWS) = Random Walker (RW).
- Explicit relation between effective resistance and ProbWS/RW.
- Power Watershed counts minimum cost spanning forests.

## References

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