



Rumor Source Detection: A Probabilistic Perspective

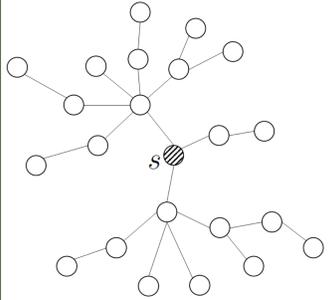
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Rumor Source Detection Problem

At time 0, start spreading.



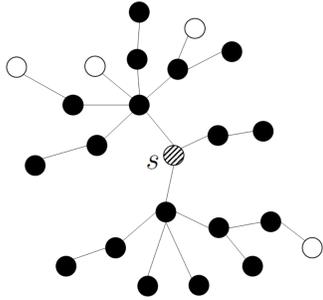
SI Model Spreading

Two types of nodes:

Susceptible and Infected.

Takes time $\tau_{ij} \sim P_{ij}$ for i to infect j .
 τ_{ij} 's are independent.

At unknown time t , observe G_I .



► Goal: Find s from G_I .

► It is natural to estimate s through the MLE:

$$\hat{s}_{MLE} = \arg \max_{v \in \mathcal{V}} L(v|G_I), \quad L(v|G_I) \triangleq \Pr(\exists t, \mathcal{G}_I^s(t) = G_I | s = v).$$

Challenge and Previous Works

✓ Challenge:

- MLE's computation time is **exponential** in # of infected nodes [1].
- Previous works resort to approximations.

✓ Previous Approximation Schemes:

► [1] Rumor Center \hat{s}_R :

$R(v, G_I)$: # of ways infection starting from v leading to G_I .

$$\hat{s}_R = \arg \max_{v \in \mathcal{V}_I} R(v, G_I).$$

► [2] Jordan Center \hat{s}_J :

$d(u, v)$: minimum # of hops from node u to node v .

$$\hat{s}_J = \arg \min_{u \in \mathcal{V}_I} \max_{v \in \mathcal{V}_I} d(u, v).$$

► [3] Dynamic Message Passing \hat{s}_D :

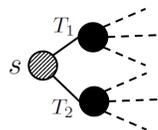
$\tilde{P}_v(i, t)$: the estimate of the probability that when node v is the source, node i is infected at time t .

$$\hat{s}_D = \arg \max_{v \in \mathcal{V}_I} \max_{t \in \mathbb{R}_+} \prod_{i \in \mathcal{V}_I} \tilde{P}_v(i, t) \prod_{j \notin \mathcal{V}_I} (1 - \tilde{P}_v(i, t)).$$

Key Ideas

► If G_I is composed of two trees, \mathcal{T}_1 and \mathcal{T}_2 , then the joint likelihood is factorized:

$$\mathbb{P}(G_I | s, t) = \mathbb{P}(\mathcal{T}_1 | s, t) \mathbb{P}(\mathcal{T}_2 | s, t).$$



- Recursively factorizing leads to the Belief Propagation Algorithm.
- In Contrast, the likelihood of the node is not factorizable because

$$\mathbb{P}(G_I | s) = \int_t \mathbb{P}(\mathcal{T}_1 | s, t) \mathbb{P}(\mathcal{T}_2 | s, t) p(t) dt \neq \mathbb{P}(\mathcal{T}_1 | s) \mathbb{P}(\mathcal{T}_2 | s).$$

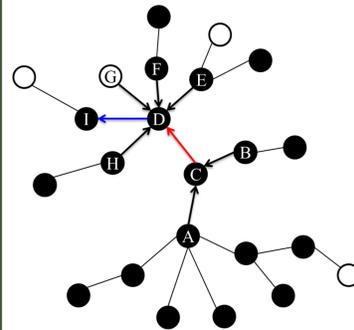
► Thus, we aim to find the *Joint Maximum Likelihood Estimator*:

$$(\hat{s}_{JMLE}, \hat{t}_{JMLE}) = \arg \max_{(v, t) \in \mathcal{V} \times \mathbb{R}_+} L(v, t | G_I), \quad L(v, t | G_I) \triangleq \Pr(\mathcal{G}_I^s(t) = G_I | s = v, t).$$

Belief Propagation Algorithm

► $m_{i \rightarrow j}(t)$: probability that if j is the source, j causes the infection in the subtree rooted at i .

► $f_{ij}(t)$: PDF of the time it takes for node i to infect node j .



$$m_{C \rightarrow D} = (m_{A \rightarrow C} m_{B \rightarrow C}) * f_{DC}$$

$$m_{D \rightarrow I} = \left(\prod_{k \in \{C, E, F, G, H\}} m_{k \rightarrow D} \right) * f_{DI}$$

* denotes convolution.

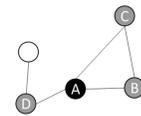
$$L(j, t | G_I) = \prod_{i \in \text{Nei}(j)} m_{i \rightarrow j}$$

- By Fast Fourier Transform, a convolution takes time $O(L \log L)$.
- Each edge needs to propagate two times (forward and backward).
- Since a tree has $n - 1$ edges, the time complexity is $O(nL \log L)$.

Extension to General Graph (GGT Heuristic)

Generate a weighted spanning tree. Then apply the BP algo.

1. Given G_I and the rates on edges, Assume A is the source. $\mathbf{I} = \{A\}$



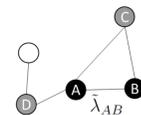
2. Choose the next infected node as $\arg \max_{v \in \mathcal{V} \setminus \mathbf{I}} \sum_{j \in \text{InNei}(v)} \lambda_{jv}$

Assume B is chosen. $\mathbf{I} = \{A, B\}$

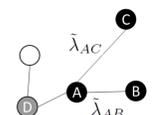
3. Update B's parent and infection rate.

$$\text{Parent}_B = \arg \min_{i \in \mathbf{I} \setminus \{B\}} \mathbb{E}[\tilde{T}_i] = A$$

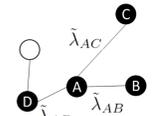
$$\tilde{\lambda}_{AB} = 1 / \mathbb{E}[\tilde{T}_B - \tilde{T}_A | \tilde{T}_B > \tilde{T}_A]$$



4. Choose the next infected node again. Assume C is chosen. $\mathbf{I} = \{A, B, C\}$
Update C's parent and infection rate.



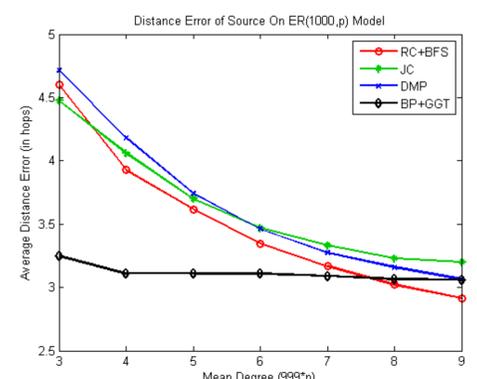
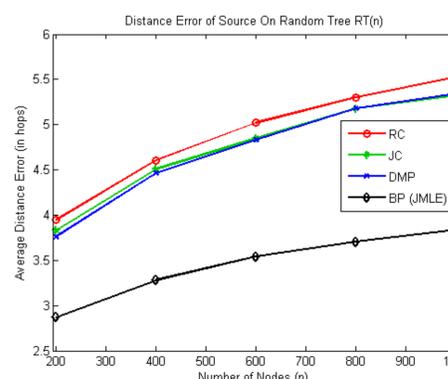
5. Do again for D.



6. Compute $L(A, t | \text{GGT}_A)$ by BP algo. GGT_A is the weighted tree above.

Note: \tilde{T}_i 's are computed and modeled by Gamma distribution.

Experimental Results



- All simulation is run with $\tau_{ij} \stackrel{\text{i.i.d.}}{\sim} \text{Exponential}(1)$.
- The more similar a graph is to a tree, the more powerful the BP algorithm will be.

Reference:

- [1] D. Shah and T. Zaman, "Rumors in a network: Who's the culprit?" *IEEE Transactions on Information Theory*, vol. 57, no. 8, pp. 5163–5181, Aug 2011.
- [2] W. Luo, W. P. Tay, M. Leng, and M. K. Guevara, "On the universality of the jordan center for estimating the rumor source in a social network," in *2015 IEEE International Conference on Digital Signal Processing (DSP)*, July 2015, pp. 760–764.
- [3] A. Y. Lokhov, M. M'ezard, H. Ohta, and L. Zdeborov'a, "Inferring the origin of an epidemic with a dynamic message-passing algorithm," *Phys. Rev. E*, vol. 90, p. 012801, Jul 2014.