

## Finding the source

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Warmup: simple random walk on  $\mathbb{Z}$ .

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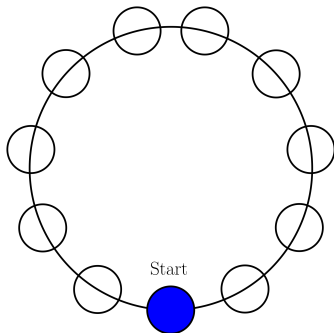


Which was the most likely starting point?

**A:** They're all equally likely!

Re-index SRW by record times, compute explicitly.

OR: last vertex visited by SRW on the ring is uniform.



Random growth process on a connected graph  $G = (V, E)$

- The source: start with  $A_0 = \{v^*\}$
- Given  $A_t$ , randomly generate  $A_{t+1} \supset A_t$
- Spread along edges of  $A$ , speed at most 1:  $A_{t+1} \subset A_t \cup \partial A_t$

e.g. SI model: spread along each  $e \in \partial_E A_t$  with probability  $p$

Given a 'snapshot,'  $A_t$  at some (large) time  $t$ , try to guess  $v^*$

$A_t$  = set of infected sites at time  $t$ , started from  $A_0 = \{v^*\}$

## (Maximum) likelihood

For any set  $A \subset G$ ,  $v \in A$ ,

$$L(v|A) = \mathbb{P}(A_t = A | v^* = v).$$

Maximum likelihood estimator:

$$\hat{v}_{ML} = \arg \max_{v \in A_t} L(v|A_t).$$

Often ML is hard to compute, can work with other estimators.

## Detection probability

The observer correctly identifies the source with probability

$$\mathbb{P}(\hat{v}_{MLE}(A_t) = v^*)$$

Motivation: protecting privacy of metadata

Goals for the message spreading algorithm:

- *Spreading*: spread to many sites
- *Obfuscation*: minimize the detection probability for patient zero
- *Multiple observations*: obfuscate even if observer has  $> 1$  independent observations
- *Local spreading* (new): spread to all sites near patient zero

Previous results: SI model

SI: edges pass information at rate 1 all independently

### Theorem (Shah, Zaman, '10)

Consider the SI spreading model on the  $d$ -regular tree for  $d \geq 3$ . The detection probability is bounded away from 0 as  $t \rightarrow \infty$ .

ML is described by 'rumor centrality':

$$R(v) = \prod_{w \in A_t} |T_w^v|.$$

$T_w^v$  = subtree rooted at  $w$  'away' from  $v$

Fast spread and local spread, but no obfuscation.

Similar results for SI model on GW-trees; multiple observations on  $T_d$



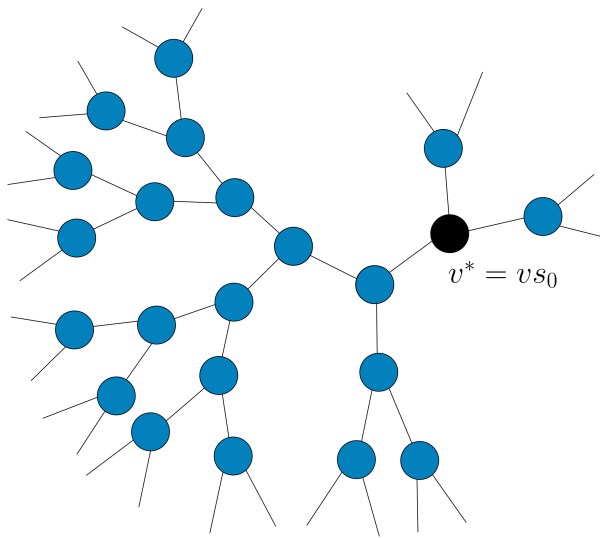
Adaptive diffusions: designed to hide the source

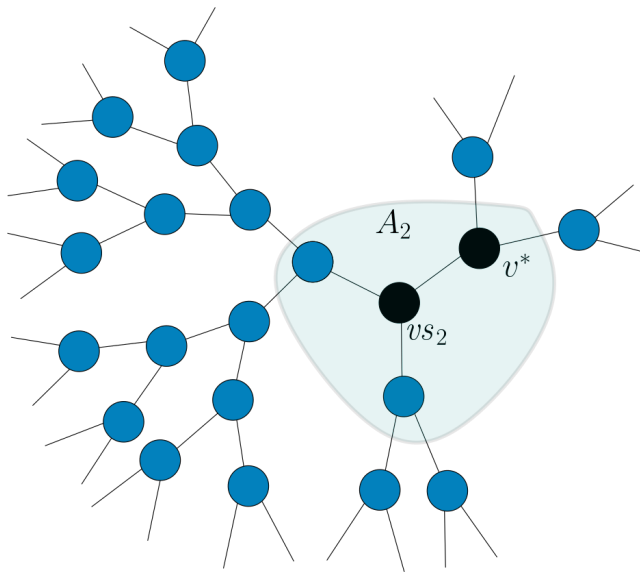
- Fix transition probabilities  $h(t, x)$  for a random walk  $H(t)$  on  $\mathbb{Z}^{\geq 0}$ :

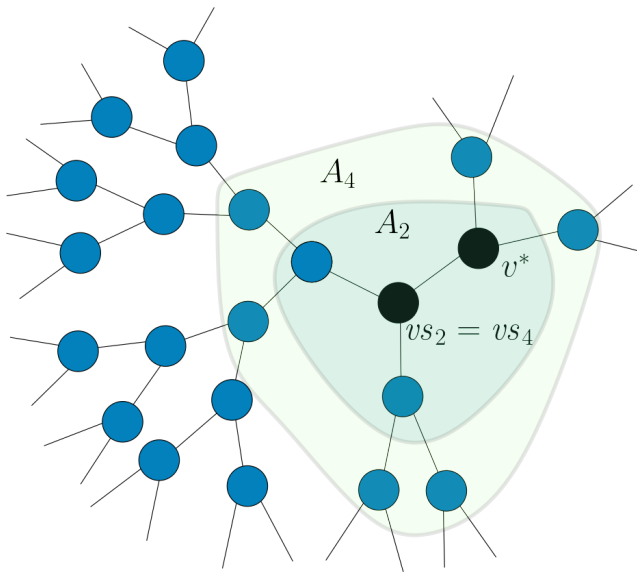
$$h(t, x) = \mathbb{P}(H(t+1) - H(t) = 1 | H(t) = x)$$

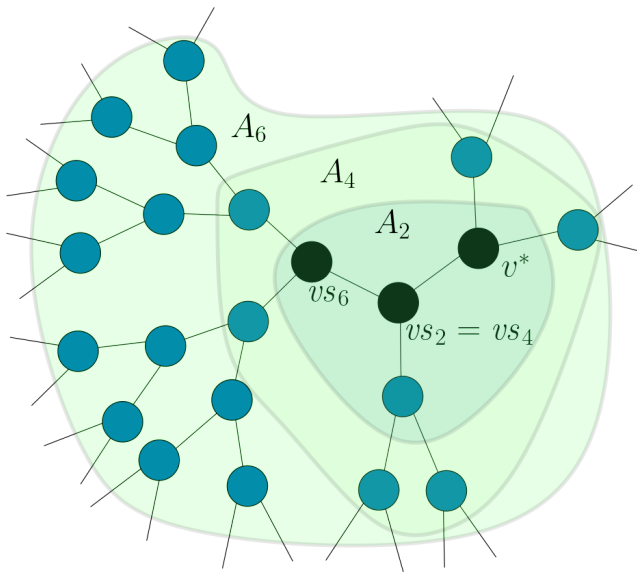
$$H(t) - H(t) \in \{0, 1\} \text{ for all } t$$

- Evolve a single particle  $VS_t$  on  $G$  with  $VS_0 = v^*$  and  $VS_t$  at depth  $H(t)$  for all  $t$  (choose uniform child when stepping)
- For  $t = 2, 4, 6, 8, \dots$ ,  $A_t =$  ball of radius  $t/2$  in  $G$  centered at  $VS_t$









## Spreading

For adaptive diffusion,

$$|A_t| = N_t = \frac{1}{d-2}(d-1)^{t/2}.$$

deterministically at even times  $t$ . (Order-optimal spreading)

## Detection

$$\mathbb{P}(\hat{v}_{MLE} = v^*) = \begin{cases} \Theta(N_t^{-1}) & \text{no detection} \\ \Theta(N_t^{-\gamma}) & \text{polynomial detection} \\ \Theta(1) & \text{perfect detection} \end{cases}$$

SI: good spread and local spread, perfect detection. [Shah, Zaman '10]

### Adaptive diffusion (Fanti, Kairouz, Oh, Viswanath '15)

Let  $G = T_d = d$ -regular tree. There exists an adaptive diffusion algorithm that achieves **no detection**:

$$\mathbb{P}(\hat{v}_{ML} = v^*) = \Theta(N_t^{-1})$$

*Pf sketch:* Choose transition probabilities for the virtual source so that it is uniformly distributed over a ball

Local spreading?

### Definition

The *local spread*  $L(t)$  is the radius of the largest ball centered at  $v^*$  and contained in  $A_t$ .

The adaptive diffusion algorithm that cannot be detected has constant order local spread,  $L(t) = \Theta(1)$  – no local spread!



## Spreading/detection trade-off [Racz, R. '18]

Consider any adaptive diffusion with **polynomial detection** of order  $\gamma \in (0, 1)$ , i.e.

$$\mathbb{P}(\hat{v}_{ML} = v^*) = O(N_t^{-\gamma}).$$

Then the average **local spreading** is bounded from above:

$$\mathbb{E}[L_t] \leq \frac{1}{2}(1 - \gamma)t + O(\log t).$$

Obfuscation (non-detection) and local spreading are **inversely linked** in this case.

The trade-off is essentially tight:

### Spreading/obfuscation trade-off [Racz, R. '18]

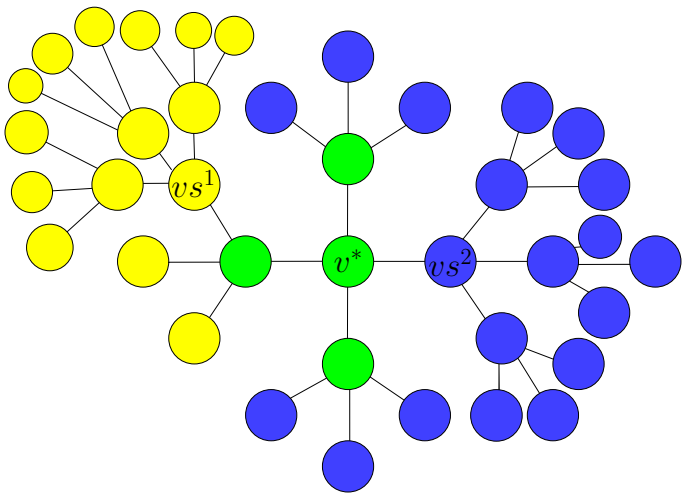
For every  $\gamma \in (0, 1)$ , there exists an adaptive diffusion with both **polynomial detection** of order  $\gamma$ ,

$$\mathbb{P}(\hat{v}_{ML} = v^*) = O(N_t^{-\gamma}),$$

and **order optimal local spreading**

$$\mathbb{E}[R_t] \geq (1 - \gamma) \frac{t}{2}.$$

Suppose the observer has access to  $k > 1$  independent snapshots  $\{A_t^i\}_{i=1}^k$  of the diffusion started from the same source  $v^*$ .



## Two independent observations (Racz, R. '18)

Suppose the observer has two iid adaptive diffusion snapshots  $A_t^1$  and  $A_t^2$  started from the same source  $v^*$ . For any  $t$ ,

$$\mathbb{P}(\hat{v}_{ML} = v^*) \geq \frac{d-1}{d} \cdot \frac{2}{t}.$$

Moreover, there exists a protocol such that for any  $t$ ,

$$\mathbb{P}(\hat{v}_{ML} = v^*) \leq \frac{d-1}{d} \cdot \frac{7}{t}.$$

Nearly perfect detection now!

It gets worse:

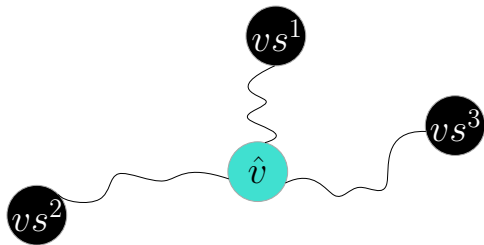
### Three or more independent observations (Racz, R. '18)

Suppose the observer has  $k \geq 3$  iid snapshots  $A_t^i$ ,  $i \in [k]$  started from the same source  $v^*$ . For any  $t$ ,

$$\mathbb{P}(\hat{v}_{ML} = v^*) \geq 1 - d \exp\left(-\frac{(d-2)^2}{2d^2}k\right).$$

Perfect detection!

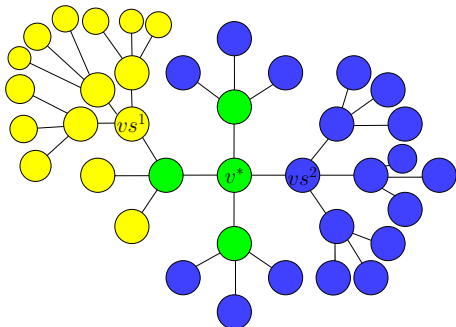
*Proof:* Pick any three virtual sources and draw the paths between them.



When the three virtual sources lie in different sub-trees away from the root, there will be a unique intersection point  $\hat{v}$ .

Necessary condition to hide the source under multiple observations

Simple estimator: guess a green vertex



## Question

Does there exist a spreading algorithm that achieves **order-optimal spreading** and at most **polynomial detection** given  $\geq 2$  observations?

Should look at algorithms that have **order-optimal local spreading**:

$$\mathbb{P}(\hat{v}_{MLE} = v^*) \geq \mathbb{E} \left[ \left| \bigcap_{i=1}^k G_t^i \right|^{-1} \right],$$

RHS is large if local spread is typically small.

GW-trees? Real-world networks?