Metadata-Conscious Anonymous Messaging

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Anonymity matters



Jason Rezaian's Year of Imprisonment in Iran

Wednesday marks the one-year anniversary of the *Washington Post* reporter's detention in the Islamic Republic.

Russian Activists and Journalists Attacked at Chechen Border

Saudi Man Gets 10 Years, 2,000 Lashes Over

Atheist Tweets

By THE ASSOCIATED PRESS • RIYADH, Saudi Arabia — Feb 27, 2016, 8:26 AM ET

Politics | Fri Nov 23, 2007 4:54pm EST

Syria blocks Facebook in Internet crackdown

Adaptive Diffusion

Adaptive diffusion breaks symmetry to provide strong anonymity. Intuition: carefully adapt the information flow rate and direction

Adaptive diffusion protocol:

The source node chooses a neighbor at random, and passes the message to it with m = 1 & color = orange
An orange node

. passes the message to a randomly chosen neighbor with an incremented m & color = orange,

2. and then to all other neighbors with a **decremented m & color = purple**.

3. A purple node with m > 1 passes the message to all its neighbors with a **decremented m & color = purple** 4. A purple node with m = 0 stops spreading





How can we empower people to speak without fear of social or political retribution?

The problem

- Design a distributed messaging algorithm that:
- a) Prevents a powerful adversary from identifying the true message source,
- b) Spreads content quickly over contact graphs.

The adversaries



Regular Trees: snapshot adversary



Colluding "spies" share message contents, metadata, and underlying graph structure to infer message authorship.

Adversary knows the underlying graph and can con collect a snapshot containing information about who got the message.

David

Information flow in social networks



This spreading model is known as the **diffusion model**. Messages flow in all directions at the same rate. With high probability, diffusion places the **true source in the center** of the graph. This helps adversaries infer the source. [Shah & Zaman 2011, Pinto et. al. 2012]

Snapshot Adversary

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low likelihood

Theorem: On regular trees, adaptive diffusion hides the source **in all the leaf nodes** and spreads the message **exponentially quickly**. Therefore, the probability of rumor source detection is inversely proportional to the number of noes with the message.

Regular Trees: spy-based adversary





Intuition: Break the symmetry around the center. Spread in a way such that the source is always on the boundary of the information spread.

T = 4

Theorem: On regular trees, the probability of detection is always greater than or equal to p. Moreover, the probability of detection under adaptive diffusion is p + o(p). The limit of the probability of detection, as the degree of the tree goes to infinity, is equal to p.

Adaptive Diffusion on Real Graphs



