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Reference

Saptarshi Ghosh, Bimal Viswanath, Farshad Kooti, Naveen Kumar Sharma, Korlam Gautam, Fabricio Benevenuto, Niloy Ganguly, and Krishna P. Gummadi. 2012. *Understanding and Combating Link Farming in the Twitter Social Network*. Proceedings of the 21st International World Wide Web Conference (WWW), Lyon, France.

Link Farming

Link Farming in Web

Websites exchange reciprocal links with other sites to improve ranking by search engines

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A form of spam

Heavily penalized by search engines

Twitter as a Web within the Web

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- Search engines rank users by follower-rank, Pagerank to decide whose tweet to return as search results
- High indegree (no. of followers) is seen as a metric of influence

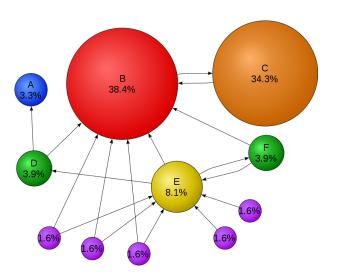
PageRank (Page and Brin, 1998)

Basic Intuition

A webpage is important if other important pages point to it

- PageRank is a "vote" by all other webpages about the importance of a page.
- $v(i) = \sum_{j \to i} \frac{v(j)}{deg^{out}(j)}$
- v is the stationary distribution of the Markov chain
- "The \$25,000,000,000 Eigenvector: The Linear Algebra Behind Google"

PageRank Example (Source: Wikipedia)

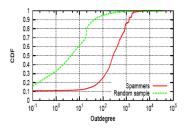


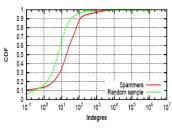
Spammers in contrast to a random user

What do you expect about the indegree and outdegree distribution

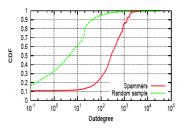
Think about a random user and a spammer

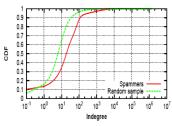
Link Farming in Twitter: Motivation for the study





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Link Farming in Twitter

Spammers follow other users and attempt to get them to follow back

Motivation is similar

Higher indegree will give better ranks in search results

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Who engages in link farming?

- Web spammers
- Twitter spammers,

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Additional factors in Twitter

'Following back' considered as a social ettiquette

Link farmers in Twitter

Idea: start with spammers

Study how spammers acquire social links

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Study how spammers acquire social links

Large amounts of spam in Twitter

- Spam-URLs get much higher clickthrough rates than spam-URLs in email
- Spammers are successfully acquiring social links and social influence

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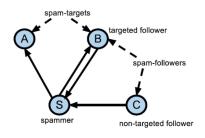
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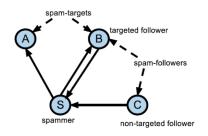
Attempt to crawl user's profile page - if the user is suspended, crawl would lead to http://twitter.com/suspended

- 379,340 accounts suspended during Aug 2009 Feb 2011
- Suspension either due to spam-activity or long inactivity
- 41,352 suspended accounts posted at least one blacklisted URL shortened by bit.ly or tinyurl

Terminology for spammers' links

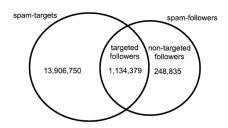


Terminology for spammers' links



- Spam-targets: users followed by spammers
- Spam-followers: users who follow spammers
- Targeted followers: spam-target as well as spam-follower
- Non-targeted followers: follow spammers without being targeted

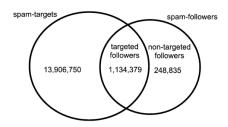
Link farming by spammers



Spammers farm links at large scale

Over 13 million users (27% of total) targeted by 41,352 spammers (0.08% of total)

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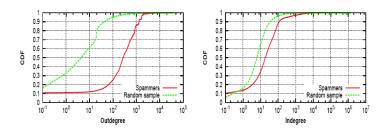
Spammers farm links at large scale

Over 13 million users (27% of total) targeted by 41,352 spammers (0.08% of total)

1.3 million spam-followers

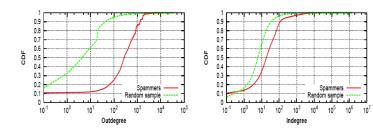
82% are targeted \rightarrow spammers get most links by reciprocation

Link farming makes spammers influential



- Spammers get more followers than an average Twitter user
- Some spammers acquire very high Pageranks :

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Who are the spam-followers?

Non-targeted spam-followers

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- Mostly sybils / hired helps of spammers
- Most have now been suspended by Twitter (9,725 among top 10,000, having links to spammers)

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Non-targeted spam-followers

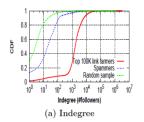
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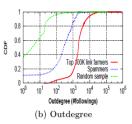
Targeted spam-followers

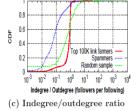
- Ranked on the basis of number of links to spammers
- 60% of the follow-links acquired by spammers come from the top 100,000 targeted followers

- Analyzed the status of the top 100,000 link farmers (July, 2011)
- 76% still exist and have not been suspended by Twitter
- 235 verified as real, well-known users
- much higher indegree as well as outdegree compared to spammers
- Most of their tweets contain valid URLs

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Top 5 link farmers according to	
#links to spammers	Pagerank
Larry Wentz: Internet, Affil-	Barack Obama: campaign
iate Marketing	staff
Judy Rey Wasserman:	Britney Spears: It's Britney
Artist, founder	
Chris Latko: Interested in	NPR Politics: Political cover-
tech. Will follow back	age and conversation
Paul Merriwether: helping	UK Prime Minister: PM's of-
others, let's talk soon	fice
Aaron Lee: Social Media	JetBlue Airways: Follow us
Manager	and let us help

17/1

Highly influential users

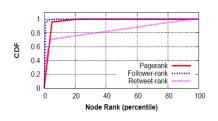
Rank within top 5% as per Pagerank, follower-rank, retweet-rank

Highly influential users

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Mostly social marketers, enterpreneurs, ...

- Want to promote some online business/website
- Heavily interconnect with each other - density 0.018 (10⁻⁷ for the whole graph)
- Aim: to acquire social capital





Combating the problem

Not practical for Twitter to suspend / blacklist top link-farmers

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Solution

- Strategy to disincentivize users from following / reciprocating to unknown people
- Penalize users for following spammers

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Collusionrank: inverse of pagerank

- Negatively bias a small set of known spammers
- Propagate negative scores from spammers to spam-followers

Collusionrank

Algorithm 1 Collusionrank

Input: network, G; set of known spammers, S; decay factor for biased Pagerank, α

Output: Collusionrank scores, c

initialize score vector d for all nodes n in G

$$d(n) \leftarrow \left\{ \begin{array}{ll} \frac{-1}{|S|} & \text{if } n \in S \\ 0 & \text{otherwise} \end{array} \right.$$

/* compute Collusionrank scores */ $c \leftarrow d$

while c not converged do

for all nodes n in G do

transfer all nodes
$$n$$
 in G do
$$tmp \leftarrow \sum_{nbr \in followings(n)} \frac{c(nbr)}{|followers(nbr)|}$$

$$c(n) \leftarrow \alpha * tmp + (1 - \alpha) * d(n)$$

end for

insert leaked scores uniformly across all nodes such that $\sum_{n} c(n) = -1$

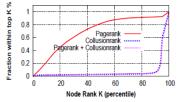
return c

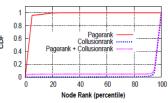
Pagerank+Collusionrank

- Computed Collusionrank considering 600 known spammers
- Rank users by Pagerank + Collusionrank

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 - → Effectively filters out spammers and link-farmers (top spam-followers) from top ranks





(a) Rankings of all 41,352 spam- (b) Rankings of Top 100,000 capimers

talists