

TrustRank



Trust Rank

- Observation
 - Good pages tend to link good pages.
 - Human is the best spam detector
- Algorithm
 - Select a small subset of pages and let a human classify them
 - Propagate goodness of pages

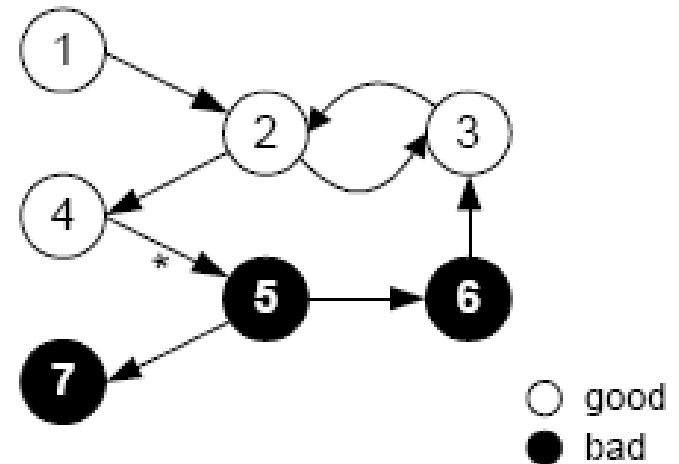
Propagation

- Trust function T
 - $T(p)$ returns the probability that p is a good page
- Initial values
 - $T(p) = 1$, if p was found to be a good page
 - $T(p) = 0$, if p was found to be a spam page
- Iterations:
 - propagate Trust following out-links
 - only a fixed number of iteration M .

Propagation (2)

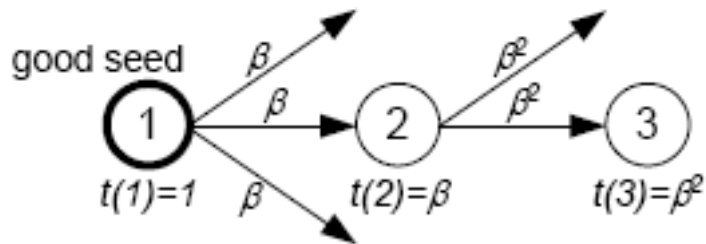
- Problem with propagation

- Pages reachable from good seeds might not be good
- the further away we are from good seed pages, the less certain we are that a page is good.



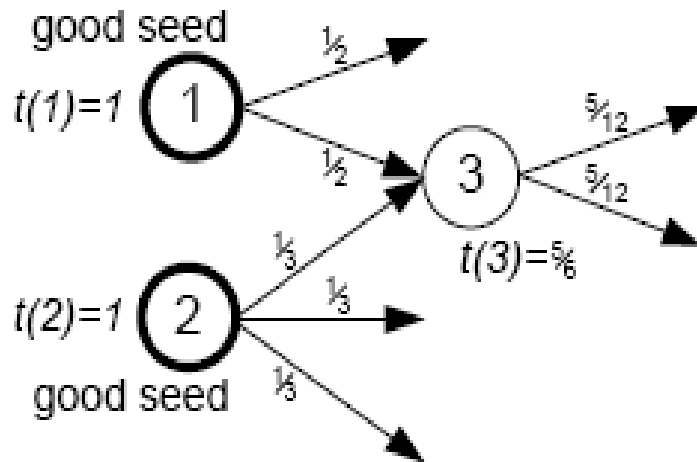
- **solution:** reduce trust as we move further away from the good seed pages (trust attenuation).

Trust attenuation – dampening



- Propagate a dampened trust score $\beta < 1$ at first step
- At n -th step propagate a trust of β^n
- How to deal with multiple in-links? (max, mean, etc.)

Trust attenuation – splitting



- Parent trust value is splitted among child nodes
- Observation: the more the links the less the care in choosing them
- Mix damp and split? β^n (splitted trust)

Selection – Inverse PageRank

- The seed set S should:
 - be as small as possible
 - cover a large part of the Web
- Covering is related to out-links in the very same way PageRank is related to in-link
 - ***Inverse PageRank !***
- Perform PageRank on a graph with inverted links
 - $G' = (V, E')$ where $(p, q) \in E' \Leftrightarrow (q, p) \in E$.
- Alternatively, using *High PageRank* showed slightly worse performance

Algorithm

1. Select seeds (s) and order by preference
2. Invoke oracle (human) on the first L seeds,
3. Initialize and normalize oracle response d
4. Compute TrustRank score (as in PageRank formula):

$$t^* = \beta \cdot T \cdot t^* + (1 - \beta) \cdot d$$

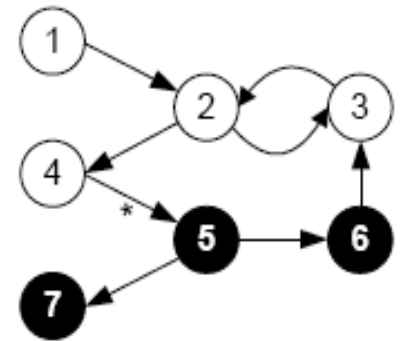
T is the adjacency matrix of the Web Graph.

β is the dampening factor. (usually .85)

Algorithm - example

- $\mathbf{s} = [0.08, 0.13, 0.08, 0.10, 0.09, 0.06, 0.02]$
- Ordering = [2, 4, 5, 1, 3, 6, 7]
- $L=3 \{2, 4, 5\} \mathbf{d}=[0, 0.5, 0, 0.5, 0, 0, 0]$
- $\beta=0.85 \ M=20$
- $\mathbf{t}^* = [0, 0.18, 0.12, 0.15, 0.13, 0.05, 0.05]$

- NB. max=0.18
- Issues with page 1 and 5



Evaluation metrics

- Pairwise orderness
 - fraction of pairs without mistakes
- Precision
 - fraction of good pages among those with trust above threshold
- Recall

$$\text{pairord}(\mathbb{T}, \mathbb{O}, \mathcal{P}) = \frac{|\mathcal{P}| - \sum_{(p,q) \in \mathcal{P}} I(\mathbb{T}, \mathbb{O}, p, q)}{|\mathcal{P}|}$$

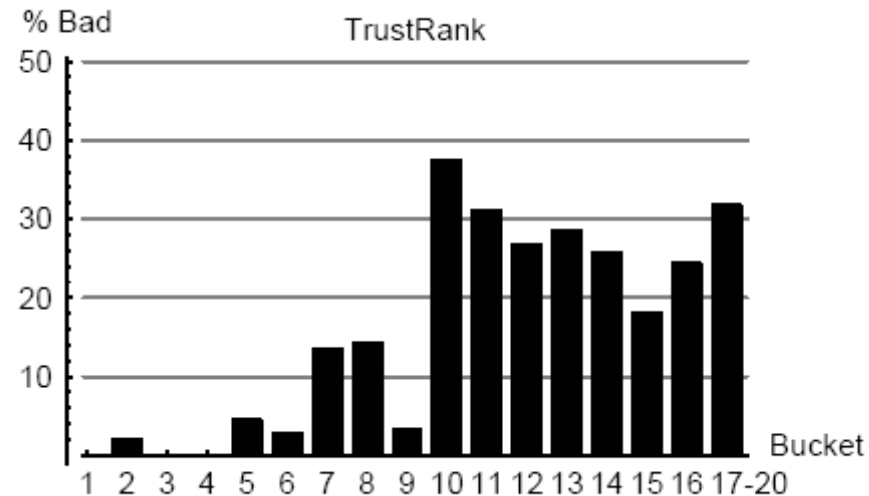
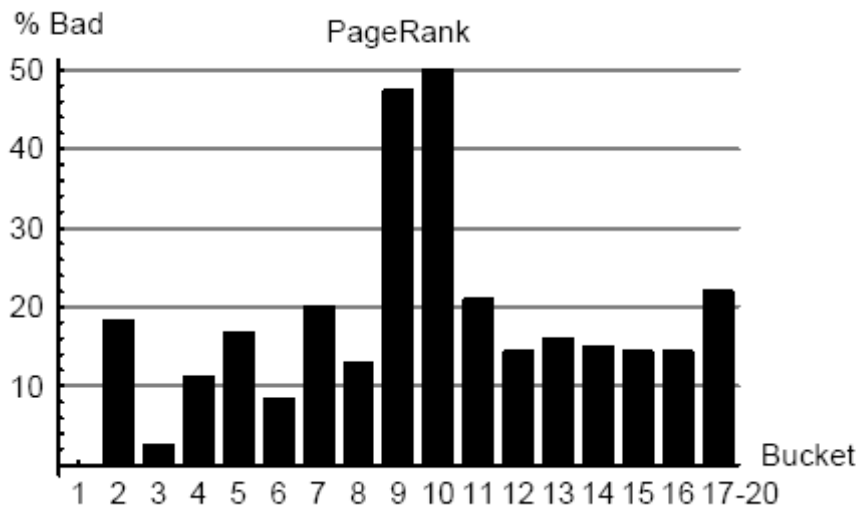
$$\text{prec}(\mathbb{T}, \mathbb{O}) = \frac{|\{p \in \mathcal{X} | \mathbb{T}(p) > \delta \text{ and } \mathbb{O}(p) = 1\}|}{|\{q \in \mathcal{X} | \mathbb{T}(q) > \delta\}|}$$

$$\text{rec}(\mathbb{T}, \mathbb{O}) = \frac{|\{p \in \mathcal{X} | \mathbb{T}(p) > \delta \text{ and } \mathbb{O}(p) = 1\}|}{|\{q \in \mathcal{X} | \mathbb{O}(q) = 1\}|}$$

Results – evaluation data

- August 2003 dataset
- Approximation to websites instead of page
- 31 million websites
- 1 third (13 million) were unreferenced
- 178 seeds were choosed among those the oracle evaluated as good seeds
- 748 sample sites used to evaluate TrustRank

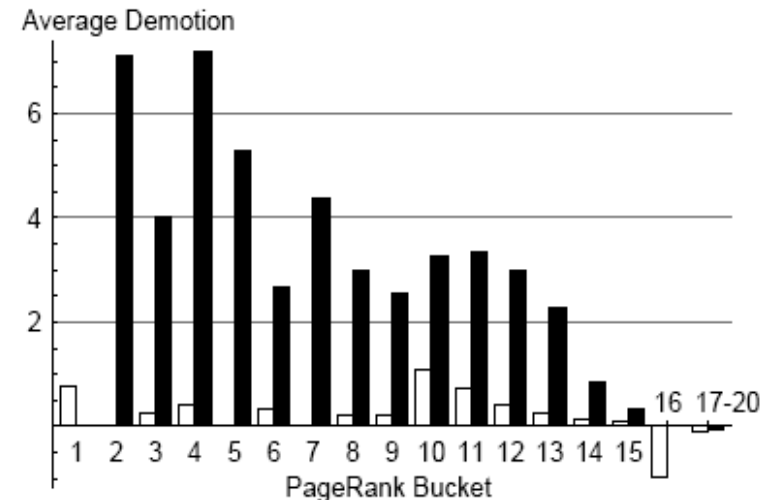
Results – compare with PageRank



- Almost no spam in the first 5 buckets of TrustRank

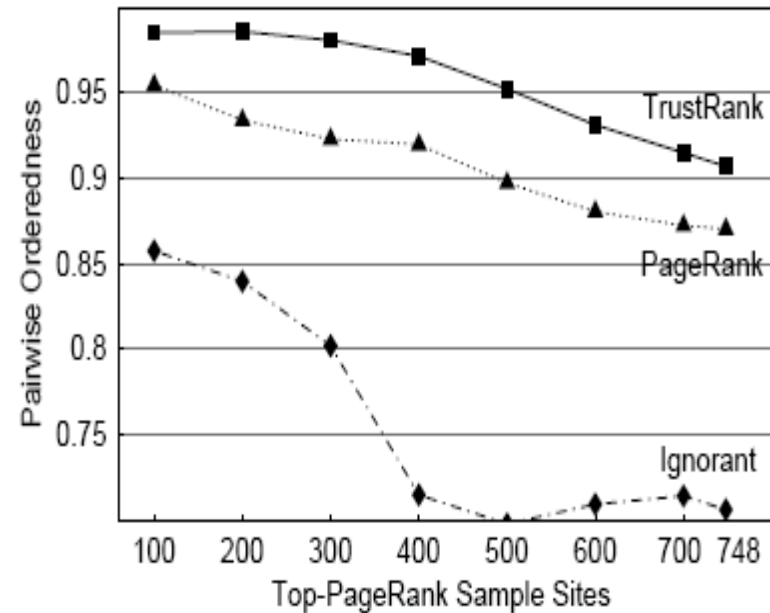
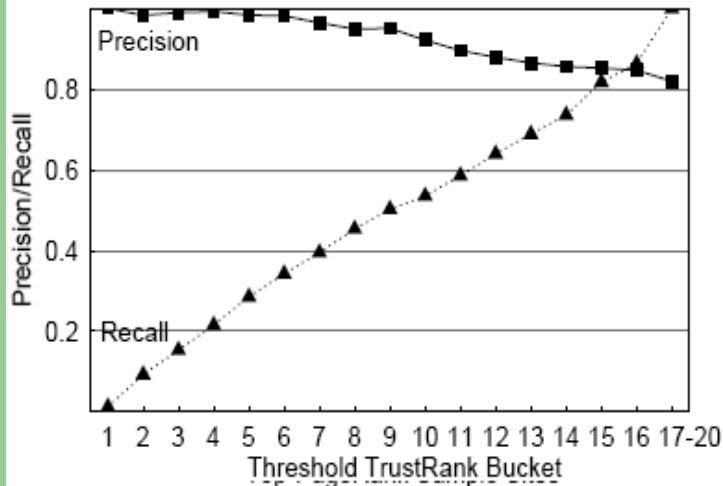
Results – compare with PageRank

- The vertical axis shows the number of buckets by which sites from a specific PageRank bucket got demoted in TrustRank on average.
- White bars represent the reputable sites, while black ones denote spam.
- Example: spam sites in PageRank bucket 2 got demoted seven buckets on average (around bucket 9)
- Promotion example: in PageRank bucket 16, good sites appear on average one bucket higher in the TrustRank ordering.



Results – evaluation metrics

- Pairwise orderness in TrustRank, PageRank and the ignorant trust function.



- Precision and recall. Threshold chosen according to buckets.

Further refinements

- further explore the interplay between dampening and splitting for trust propagation.
- iterative process: after the oracle has evaluated some pages, we could reconsider what pages it should evaluate next, based on the previous outcome.

fine.



PageRank

- *PageRank in one equation:*
 - $PR(p) = \alpha M + (1 - \alpha) V_p$
 - M is the adjacency matrix of the Web Graph.
 - α is the *damping* factor. (usually .85)
 - in case of fairness $V_p = 1/N$ (N = # of pages in the Web).
 - V is the personalization vector.