

# Efficient and Decentralized PageRank Approximation in a P2P Network

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September 13, 2006



# Outline

- 1 Introduction
- 2 Related Work
- 3 The JXP Algorithm
- 4 Mathematical Analysis
- 5 Experimental Results
- 6 Conclusions and Ongoing Work



# Introduction

## Computational Model

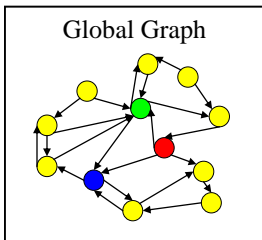
Every peer crawls Web fragments at its discretion and has its own local & personalized search engine



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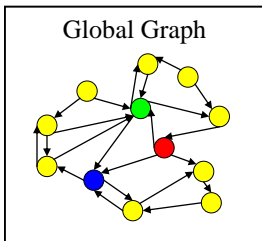
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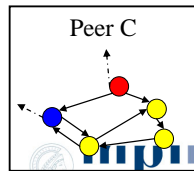
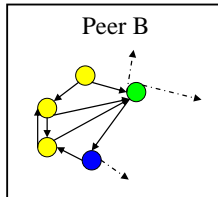
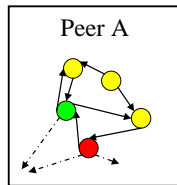
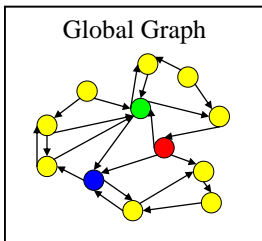
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## Goal

Compute “global” authority scores of pages in the network.



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## Problems

- Peers have only local (incomplete) information
- Pages might link to or be linked by pages at other peers
- No control over overlaps between local graphs





# PageRank

## PageRank [Brin and Page, WWW'98]

- Importance of a page depends on the importance of the pages that point to it
- Stationary distribution of a Markov chain that describes a random walk over the graph
- Can be computed using the power iteration method

## PageRank Formulation

$$PR(q) = \epsilon \times \sum_{p|p \rightarrow q} PR(p)/out(p) + (1 - \epsilon) \times 1/N$$



# Related Work

## Efficient PR

- Graph Aggregation [Broder et al., WWW'04]
- Iterative Aggregation [Langville & Meyer, WWW'04]

## Decentralized PR

- *Local PageRank & ServerRank* [Wang & DeWitt, VLDB'04]
- *BlockRank* [Kamvar et al., Stanford Tech. Report'03]

## Markov Chains Aggregation/Disaggregation Techniques

- Kemeny & Snell [1963]
- Stewart [1994]
- Meyer [2000]

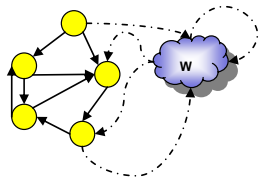
# JXP Algorithm

## JXP Algorithm

- Decentralized algorithm for computing authority scores of pages in a P2P Network, with arbitrary overlapping
- Runs locally at every peer
- No coordinator, asynchronous
- Combines local PageRank computations + Meetings between peers
- JXP scores converge to the true global PageRank scores

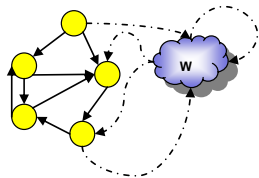


# World Node



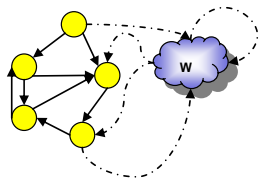
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- Special node added to each local graph



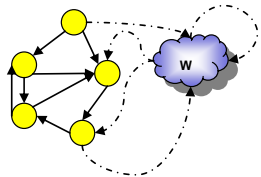
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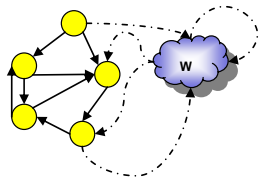
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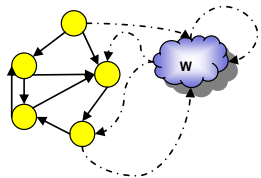
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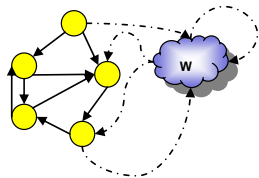
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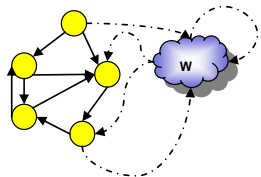


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- Self-loop link to represent transitions among external pages



# The Algorithm

## Initialization step

- Local graph is extended by adding the world node;
- PageRank is computed in the extended graph → JXP scores



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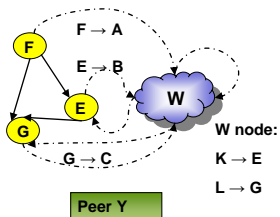
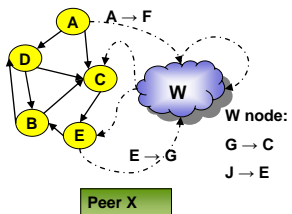
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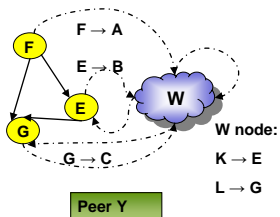
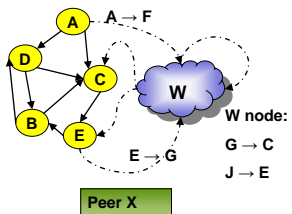
## Main Algorithm (for every $P_i$ in the network)

- Select  $P_j$  to meet
- Update world node
  - Add edges for pages in  $P_j$  that point to pages in  $P_i$
  - If an edge already exists at the world node, the score of the source page is updated by taking the highest of both scores
- Compute PageRank → JXP scores

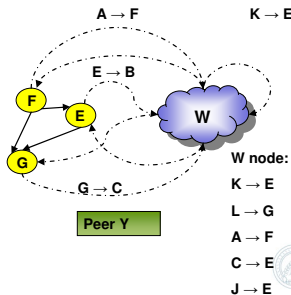
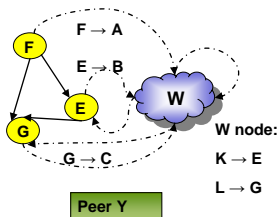
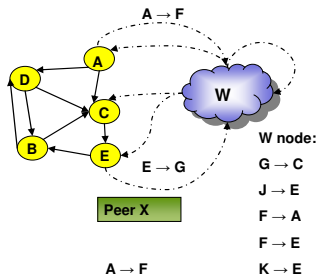
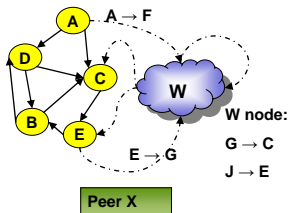
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# Peer Selection Strategy

## Motivation

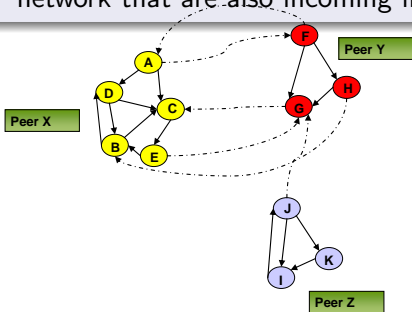
- Peers' contribution for the convergence are different
- Finding peers with high contribution would speed up convergence
- “Quality indicator”: Number of outgoing links of a peer in the network that are also incoming links in the local graph



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# Peer Selection Strategy

## Good strategy

Find promising peers without increasing much bandwidth consumption

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## Statistical synopses

Approximation technique for comparing data of different peers without explicitly transferring their contents.

- Compact representation of sets
- Can be used to estimate cardinality of the intersection between two sets
- JXP uses Min-Wise Independent Permutations (MIPs) [Broder et al., 1997]

# Pre-meetings Strategy

- Each peer  $P_i$  computes  $local(P_i)$  and  $successors(P_i)$  MIPs vectors (256-integer vectors)
- When  $P_i$  meets  $P_j$ 
  - Uses MIPs vectors to estimate percentage of local pages pointed by pages in  $P_j$
  - If percentage above threshold,  $P_i$  caches  $P_j$ 's ID
  - Uses MIPs again to estimate overlap between the two local graphs
  - If there is high overlap, peers exchange their list of cached ID's and store them in a temporary list
  - **Idea:** Peers on the temporary list are potential candidates for the next meeting



# Pre-meetings Strategy

## Pre-meetings phase

- $P_j$  contacts peers on the temporary list and ask for their MIPs vectors
- Assign scores to each peer
- For next (real) meeting,  $P_i$  chooses  $P_k$  where
  - $P_k$  is best scored peer in temporary list, with prob.  $\alpha$
  - $P_k$  is one of the already cached peers, with prob.  $\beta$
  - $P_k$  is a random peer, with prob.  $(1 - \alpha - \beta)$



# Mathematical Analysis

## Assumptions

Global transition matrix  $\mathbf{C}_{N \times N}$  and global stationary distribution vector  $\boldsymbol{\pi}$

Local transition matrix and local stationary distr. (JXP scores)

$$\mathbf{P} = \left( \begin{array}{ccc|c} p_{11} & \cdots & p_{1n} & p_{1w} \\ \vdots & \cdots & \vdots & \vdots \\ p_{n1} & \cdots & p_{nn} & p_{nw} \\ \hline p_{w1} & \cdots & p_{wn} & p_{ww} \end{array} \right)$$

$$\boldsymbol{\alpha} = \left( \alpha_1 \quad \cdots \quad \alpha_n \mid \alpha_w \right)^T$$

$$p_{ij} = \begin{cases} \frac{1}{\text{out}(i)} & \text{if } \exists i \rightarrow j \\ 0 & \text{otherwise} \end{cases}$$

$$p_{iw} = \sum_{\substack{i \rightarrow r \\ r \notin G}} \frac{1}{\text{out}(i)}$$

for every  $i, j$ ,  $1 \leq i, j \leq n$ . ( $G$  is the set of local pages)

# Mathematical Analysis

## World Node transitions prob.

$$p_{wi}^t = \sum_{\substack{r \rightarrow i \\ r \in W^t}} \frac{\alpha(r)^t}{\text{out}(r)} \cdot \frac{1}{\alpha_w^{t-1}} \quad p_{ww}^t = 1 - \sum_{i=1}^n p_{wi}^t$$

$W^t$ : Set of pages represented at the World Node during meeting  $t$

## Random Jumps

$$\mathbf{P}' = \epsilon \mathbf{P} + (1 - \epsilon) \frac{1}{N} ( 1 \quad \dots \quad 1 \mid (N - n) )$$





# Mathematical Analysis

## Meeting Step

Considering one link addition/update at a time

$$\mathbf{P}^t = \mathbf{P}^{t-1} + \mathbf{E} \quad \mathbf{E} = \left( \begin{array}{cccc|cc} 0 & & \dots & & 0 & 0 \\ \vdots & & & & \vdots & \vdots \\ 0 & & \dots & & 0 & 0 \\ \hline 0 & \dots & 0 & \delta & 0 & \dots & 0 & -\delta \end{array} \right)$$



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## Theorem 1

*The JXP score of the world node, at every peer in the network, is monotonically non-increasing.*

Proof: Based on the study of the sensitivity of Markov Chains [Cho & Meyer, 1999].

# Mathematical Analysis

## Theorem 2

*The sum of scores over all pages in a local graph, at every peer in the network, is monotonically non-decreasing.*



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## Theorem 3

*Consider the true stationary probabilities (PR scores) of pages  $i \in G$  and the World Node  $w$ ,  $\pi_i$  and  $\pi_w$ , and their JXP scores after  $t$  meetings  $\alpha_i^t$  and  $\alpha_w^t$ . The following holds throughout all JXP meetings:*

$$0 < \alpha_i^t \leq \pi_i \text{ for } i \in G \text{ and } \pi_w \leq \alpha_w^t < 1.$$



# Mathematical Analysis

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## Theorem 4

*In a fair series of JXP meetings, the JXP scores of all nodes converge to the true global PR scores.*

# Setup

## Amazon collection

- 55,196 pages
- 237,160 links
- 10 categories (e.g. Computers, Sports, Travel, etc)

## Web collection

- 103,591 pages
- 1,633,276 links
- 10 categories (e.g. Movies, Music, Politics, etc)



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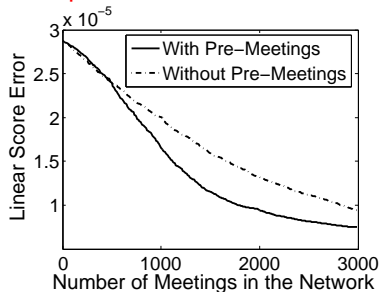
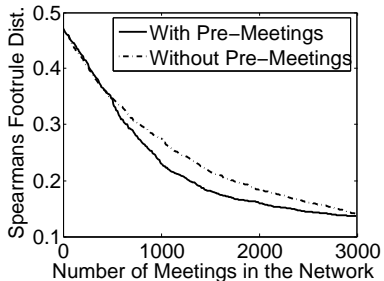
## Evaluation Measures

- “Global” JXP ranking vs. Global PageRank ranking
- Spearman’s Footrule Distance at top-k
- Linear Score Error at top-k



# Experimental Results

## Amazon Collection, top-10000

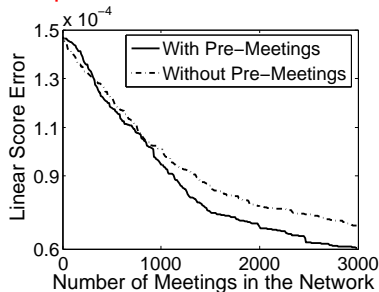
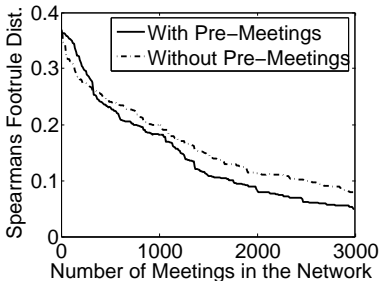


For a footrule distance of 0.2 number of meetings was reduced  
from **1,770** to **1,250**



# Experimental Results

## Web Collection, top-1000



For a footrule distance of 0.1 number of meetings was reduced  
from **2,480** to **1,650**

# Bandwidth Consumption

## Web Collection

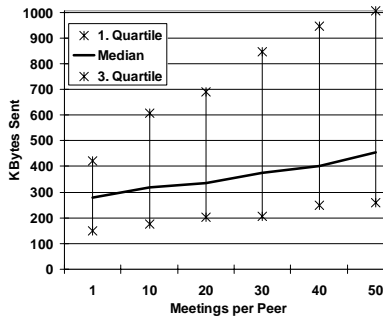
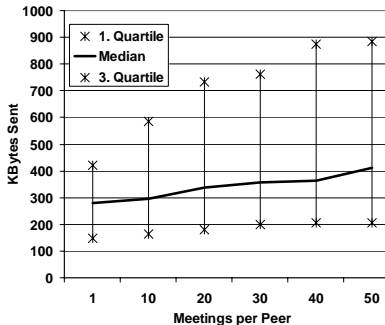


Figure: Without pre-meetings

Figure: With pre-meetings

Message size (in KBytes) for the Web crawl setup

# JXP in P2P Search

JXP integrated into our P2P search engine Minerva.  
(Minerva Project Website: <http://www.minerva-project.org>)

## Setup

- Bigger subset of Web (250,760 docs & 3,123,993 links)
- 40 peers, high overlap
- 15 queries <sup>a</sup>, using the Minerva query routing mechanism
- Results were ranked in two ways:
  - tf\*idf only
  - weighted sum of tf\*idf and JXP scores
- Precision at top-10 measured (based on manually assessments)

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<sup>a</sup>taken from Borodin et al., ACM TOIT, 2005

# Results

Query	tf*idf	(0.6 tf*idf + 0.4 JXP)
affirmative action	40%	40%
amusement parks	60%	60%
armstrong	20%	<b>80%</b>
basketball	20%	<b>60%</b>
blues	20%	20%
copyright	<b>30%</b>	20%
cheese	40%	<b>60%</b>
iraq war	<b>50%</b>	30%
jordan	40%	40%
moon landing	<b>90%</b>	70%
movies	30%	<b>100%</b>
roswell	30%	<b>70%</b>
search engines	20%	<b>60%</b>
shakespeare	60%	<b>80%</b>
table tennis	50%	<b>70%</b>
<b>Average</b>	40%	<b>57%</b>



# Conclusions and Ongoing Work

## Conclusions

- JXP algorithm for dynamically computing authority scores of pages distributed in a P2P network
- Fully decentralized (no coordinator), asynchronous
- Combines local PageRank computation with meetings between peers
- JXP scores are proved to converge to global PageRank scores

## Ongoing Work

- Integrate JXP into the query routing mechanism [P2PIR'06]
- JXP in dynamic networks
- Adapt JXP to work in the presence of malicious peers