

情報検索システム特論

Advanced Information Retrieval Systems

第11回 Lecture #11

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Alternative Retrieval Models (cont'd)

refer the material for Lecture #09-10 pp.54-80

Page Rank Technology

Page Rank

- ▶ Web Information Retrieval
 - ▶ Crawling
 - ▶ Indexing
 - ▶ Retrieving --- Scoring
- ▶ Page Rank is trademark of Google

Motivation

- ▶ Want to retrieve well-written Web pages
→ Giving high score for well-written pages
- ▶ Recognizing “well-written” - How?
 - ▶ Reading the page as a human does
→ natural language processing, knowledge processing

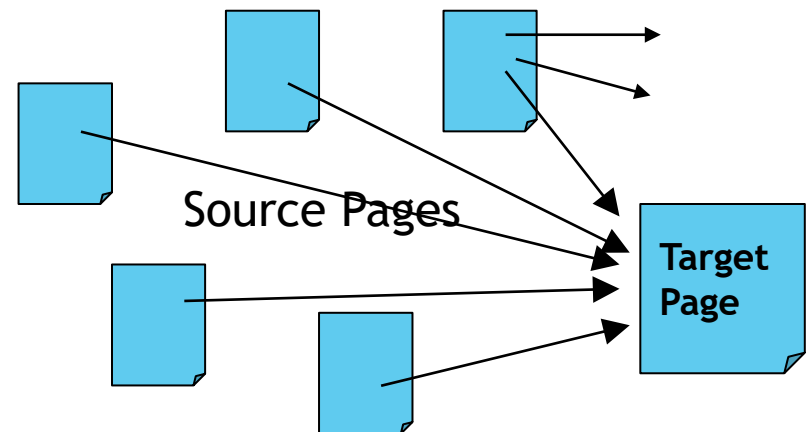


Too complicated

- ▶ Using another information, which is special in WWW

The Idea

- ▶ A page which linked by a lot of well-written pages is also a well-written page
- ▶ Criteria of “well-written” for the web page
 - ▶ If a source page is “well-written”, the target page would be “well-written”
 - ▶ If a source page has less outgoing links, the target page would have better quality
 - ▶ The target page are linked by more source pages which are “well-written” and have less links, the target page would be very good page



Principle

- ▶ A Link from Page i to Page $j \rightarrow$ Page i votes Page j
- ▶ Score of Page j is determined based on
 - ▶ The number of votes
 - ▶ Score of Page i
 - ▶ The number of outgoing links in Page i (less is better)

Computation

- ▶ A Link from Page i to Page j

$$a_{i,j} = \begin{cases} 1, & \text{there is a link from Page } i \text{ to Page } j \\ 0, & \text{there is no link from Page } i \text{ to Page } j \end{cases}$$

- ▶ Create a matrix, which $a_{i,j}$ is an element at row i and column $j \rightarrow$ adjacency matrix

Transpose the matrix

Normalize for each column

Divide each element by the number of total links

Transition probability matrix

Computation (cont'd)

- ▶ Compute the eigenvector for maximum eigenvalue

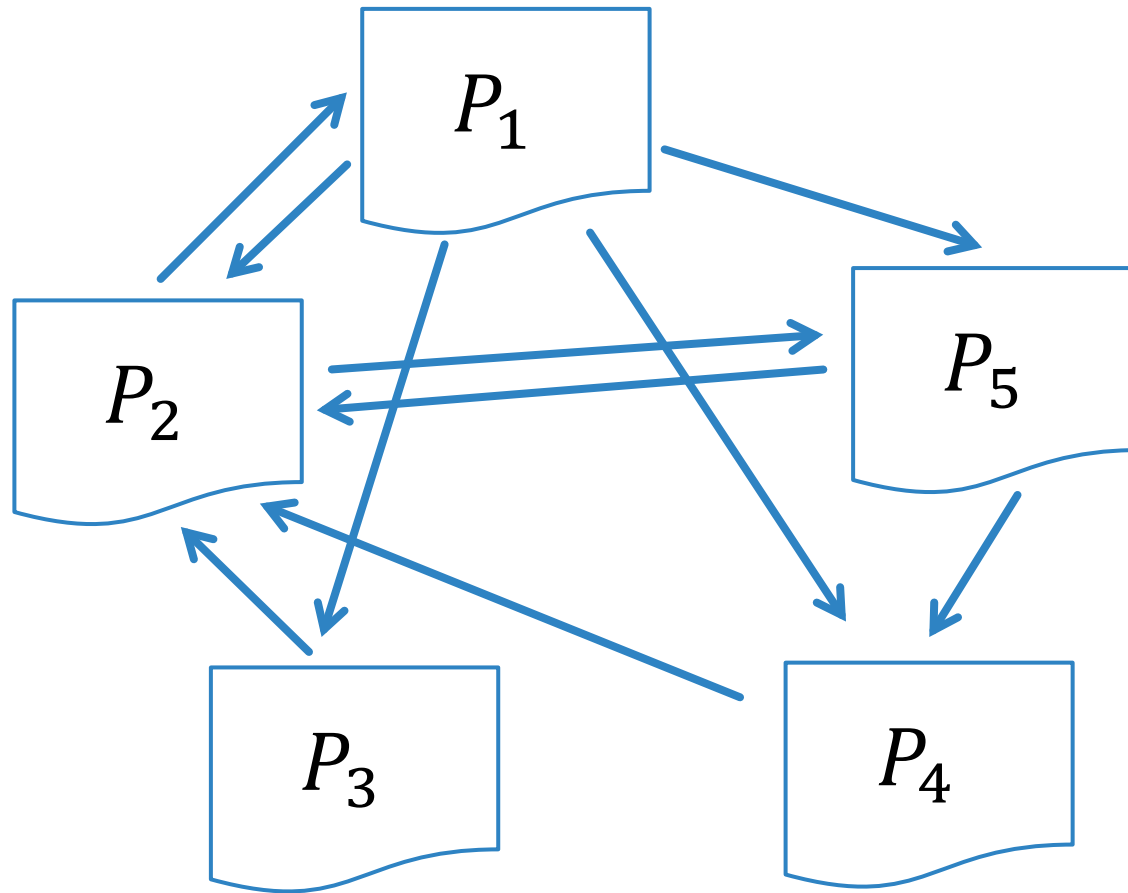
$$\mathbf{Ax} = \lambda \mathbf{x}$$

eigenvalue

eigenvector

- ▶ Normalize the eigenvector
- ▶ i -th element is the score for Page i

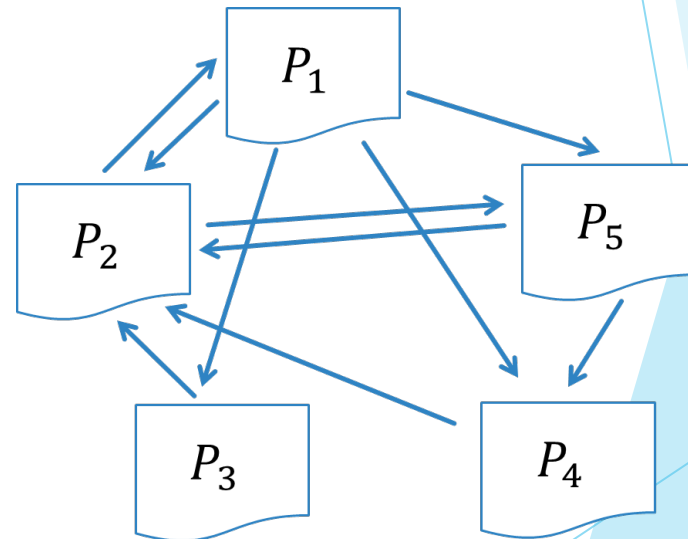
PageRank Example



PageRank Example (cont'd)

- adjacency matrix (隣接行列)

	P_1	P_2	P_3	P_4	P_5
P_1	0	1	1	1	1
P_2	1	0	0	0	1
P_3	0	1	0	0	0
P_4	0	1	0	0	0
P_5	0	1	0	1	0



PageRank Example (cont'd)

- ▶ transposed adjacency matrix (隣接行列の転置)

- ▶ adjacency matrix

$$\begin{pmatrix} 0 & 1 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 \end{pmatrix}$$

- ▶ transposed adjacency matrix

$$\begin{pmatrix} 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 & 0 \end{pmatrix}$$

PageRank Example (cont'd)

- ▶ transition probability matrix

- ▶ transposed adjacency matrix

$$\begin{pmatrix} 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 & 0 \end{pmatrix}$$

- ▶ transition probability matrix

$$\begin{pmatrix} 0 & 1/2 & 0 & 0 & 0 \\ 1/4 & 0 & 1 & 1 & 1/2 \\ 1/4 & 0 & 0 & 0 & 0 \\ 1/4 & 0 & 0 & 0 & 1/2 \\ 1/4 & 1/2 & 0 & 0 & 0 \end{pmatrix}$$

PageRank Example (cont'd)

- ▶ Eigenvector for the maximum eigenvalue

$$(8 \quad 16 \quad 2 \quad 7 \quad 10)$$

- ▶ Normalized eigenvector

$$\left(\frac{8}{43} \quad \frac{16}{43} \quad \frac{2}{43} \quad \frac{7}{43} \quad \frac{10}{43} \right)$$

$$\approx (0.186 \quad 0.372 \quad 0.047 \quad 0.163 \quad 0.233)$$

P_1

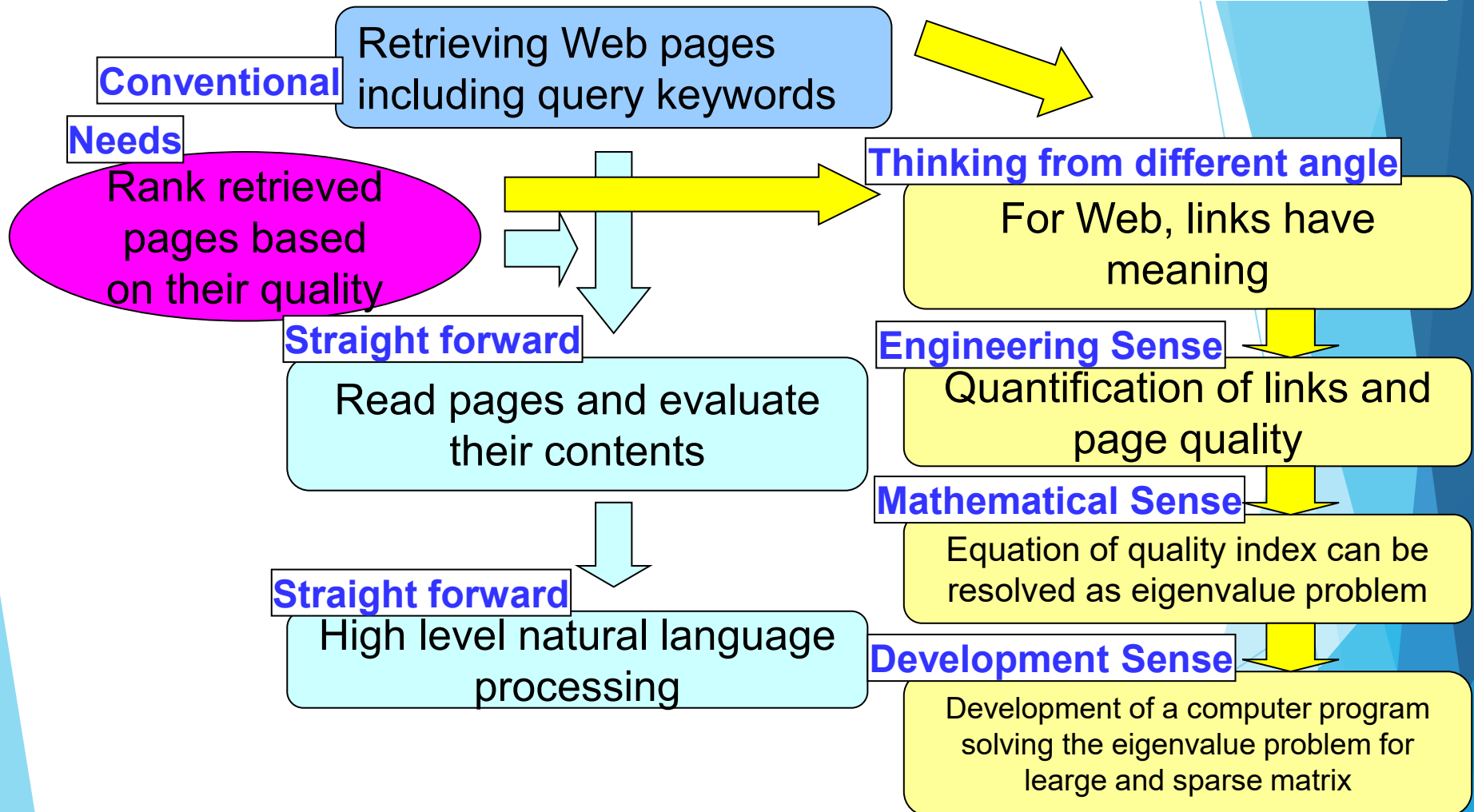
P_2

P_3

P_4

P_5

PageRank Technology : Structure of the Idea



That's it today

Assignment #2 in next class (July 12th)