

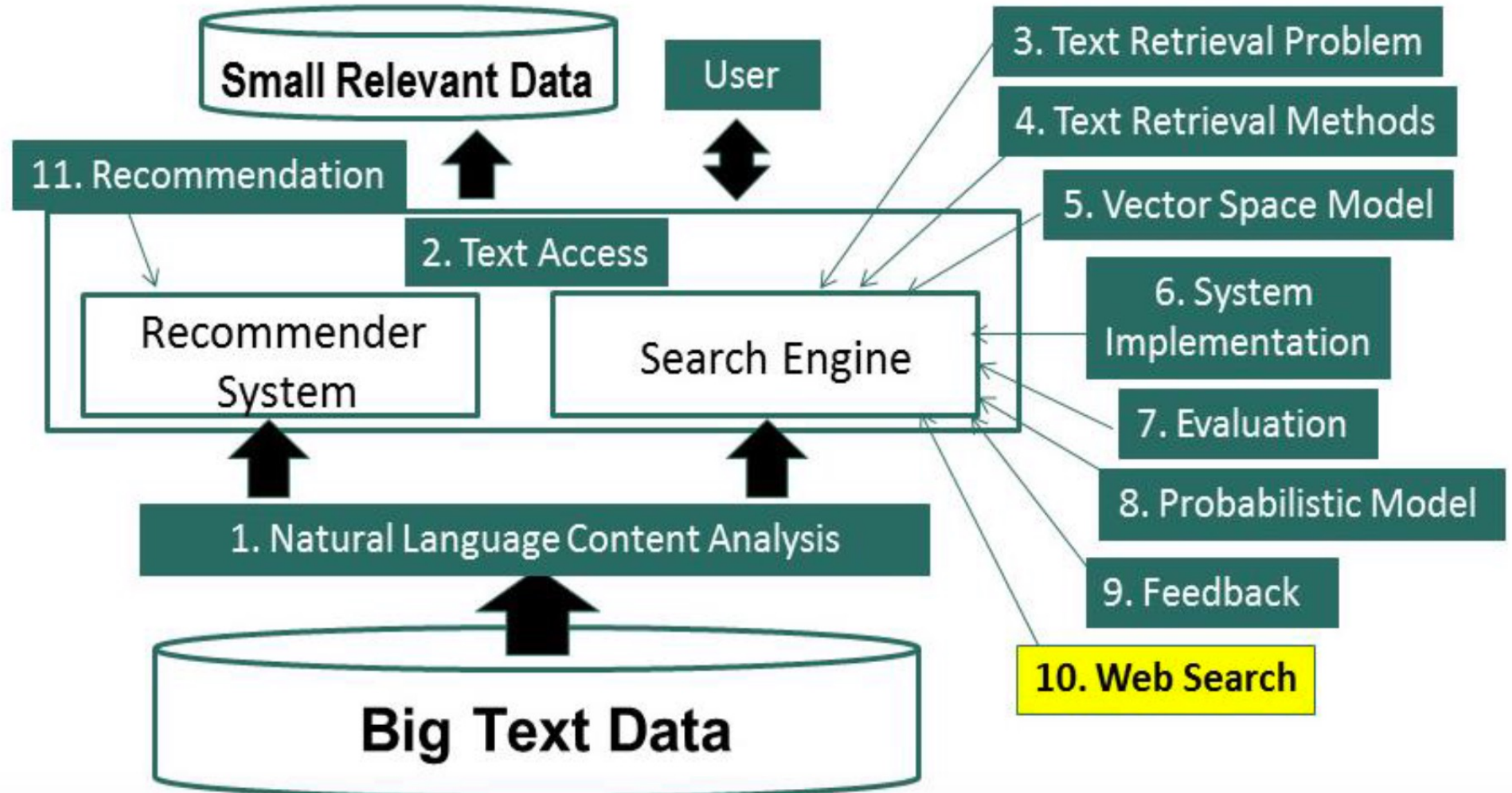


Information Retrieval

Web Search: Page Rank

Dr. Iqra Safder

Course Schedule



Page Rank Algorithm

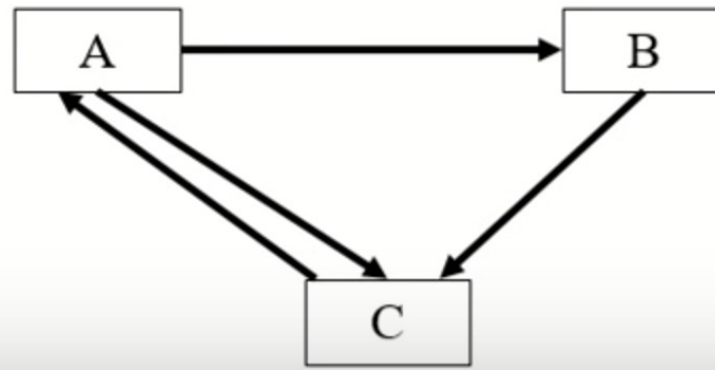
- **P**ageRank is a “vote”, by all the other pages on the Web, about how important a page is.
- **A** link to a page counts as a vote of support
- **T**he original PageRank algorithm was designed by Lawrence Page and Sergey Brin.

How Page Rank is Calculated

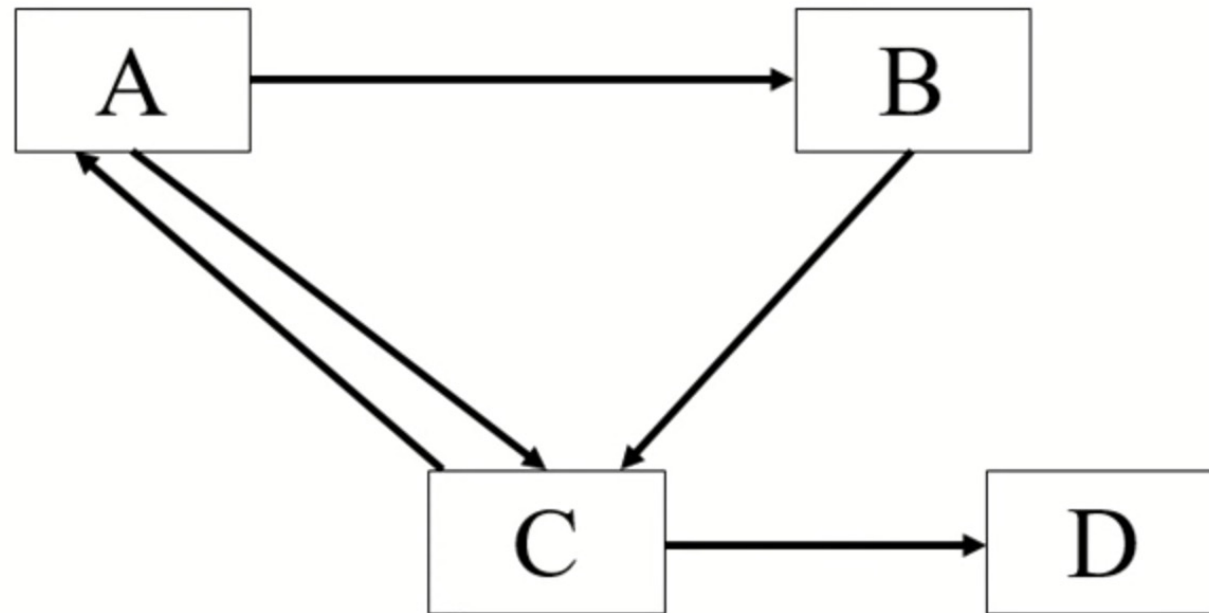
- **T**he rank of a document is given by the rank of those documents which link to it.
- **T**he PR of each page depends on the PR of the pages pointing to it.
- **B**ut we won't know what PR those pages have until the pages pointing to them have their PR calculated and so on.

Inbound, Outbound and Dangling Links

- Inbound link for a web page always increases that page's PageRank.
- An important aspect of outbound links is the lack of them on web pages.
- When a web page has no outbound links, its PageRank cannot be distributed to other pages.
- Lawrence Page and Sergey Brin characterise links to those pages as dangling links.



Dangling Links



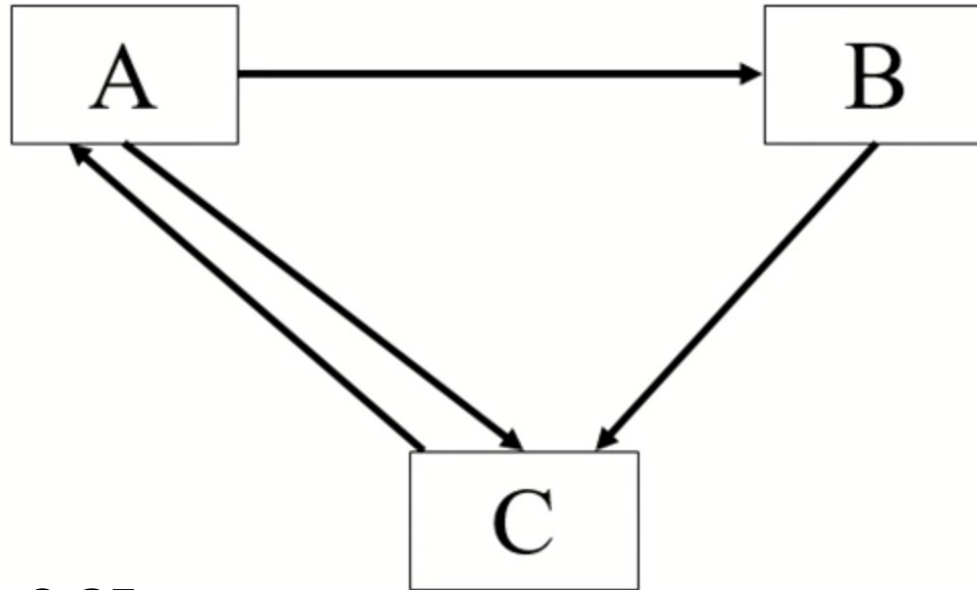
Page Rank Initialization

- Initially Page Rank (PR) for all the web pages = 1

The PageRank theory holds that an imaginary surfer who is randomly clicking on links will eventually stop clicking.

The probability, at any step, that the person will continue is a damping factor d . Various studies have tested different damping factors, but it is generally assumed that the damping factor will be set around 0.85

Damping Factor = $d = 0.85$



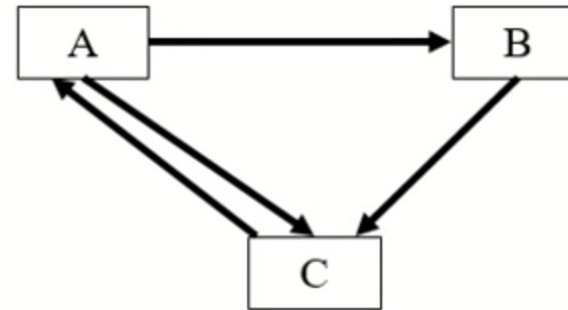
$$PR(A) = (1-d) + d(PR(T_i)/C(T_i) + \dots + PR(T_n)/C(T_n))$$

Page Rank of A & B

- Initially Page Rank (PR) for all the web pages = 1

$$PR(A) = (1-d) + d(PR(T_i)/C(T_i) + \dots + PR(T_n)/C(T_n))$$

$$\begin{aligned} PR(A) &= (1-d) + d [PR(C) / C(C)] \\ &= (1-0.85) + 0.85 [1/1] \\ &= 0.15 + 0.85 [1] \\ &= 0.15 + 0.85 \\ &= 1 \end{aligned}$$

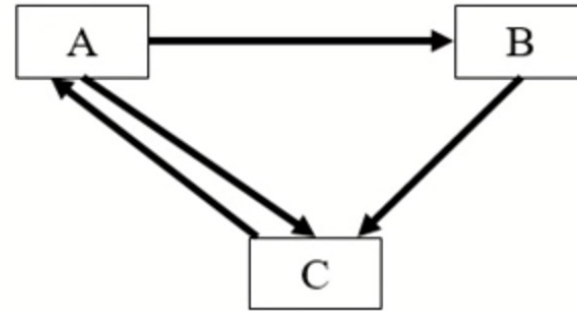


$$\begin{aligned} PR(B) &= (1-d) + d [PR(A) / C(A)] \\ &= (1-0.85) + 0.85 [(1) / 2] \\ &= 0.15 + 0.85 [0.5] \\ &= 0.15 + 0.425 \\ &= 0.575 \end{aligned}$$

Page Rank of C

- Initially Page Rank (PR) for all the web pages = 1

$$PR(A) = (1-d) + d(PR(T_i)/C(T_i) + \dots + PR(T_n)/C(T_n))$$

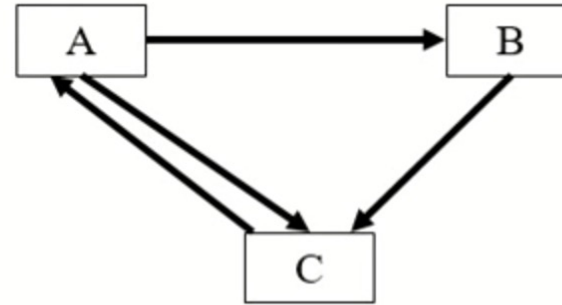


$$\begin{aligned} PR(C) &= (1-d) + d [PR(A) / C(A) + PR(B) / C(B)] \\ &= (1-0.85) + 0.85 [(1/2) + (0.575 / 1)] \\ &= 0.15 + 0.85 [0.5 + 0.575] \\ &= 0.15 + 0.85 [1.075] \\ &= 0.15 + 0.91375 \\ &= 1.06375 \end{aligned}$$

Iterations

- Initially Page Rank (PR) for all the web pages = 1

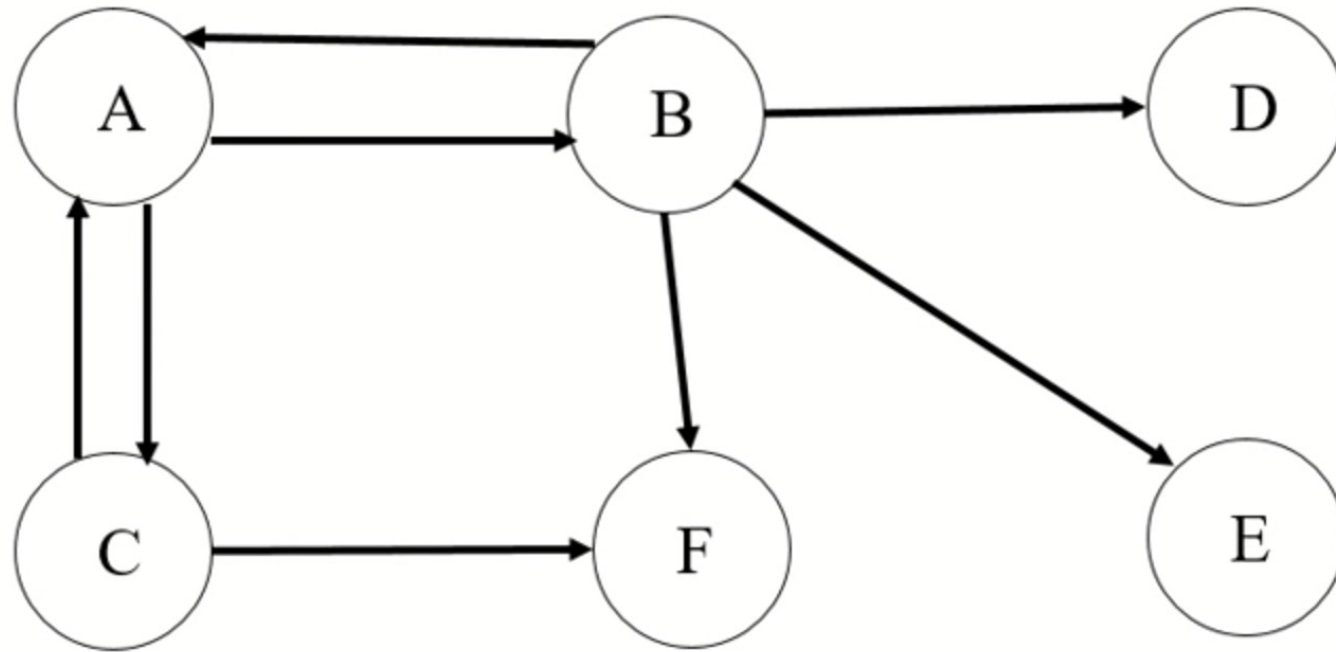
$$PR(A) = (1-d) + d(PR(T_i)/C(T_i) + \dots + PR(T_n)/C(T_n))$$



Iteration	A	B	C
0	1	1	1
1	1	0.575	1.06375
2	1.0541875	0.5980296875	1.06354922

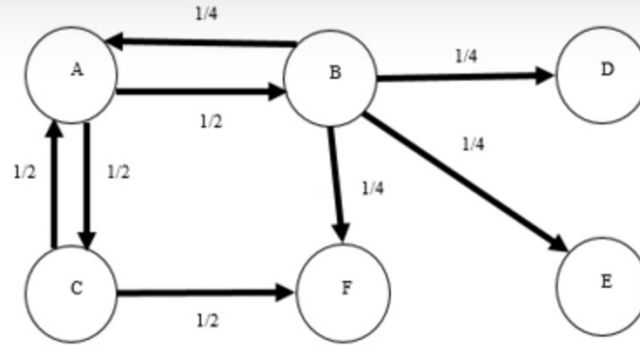
Page Rank using Matrix

- Initially Page Rank (PR) for all the web pages = 1



Matrix Representation

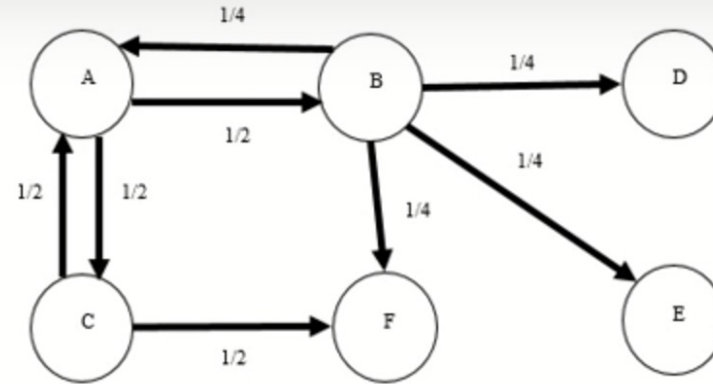
- Teleport Factor = 0.8



	A	B	C	D	E	F
A	0	1/2	1/2	0	0	0
B	1/4	0	0	1/4	1/4	1/4
C	1/2	0	0	0	0	1/2
D	0	0	0	0	0	0
E	0	0	0	0	0	0
F	0	0	0	0	0	0

Matrix Transpose

- Teleport Factor = 0.8



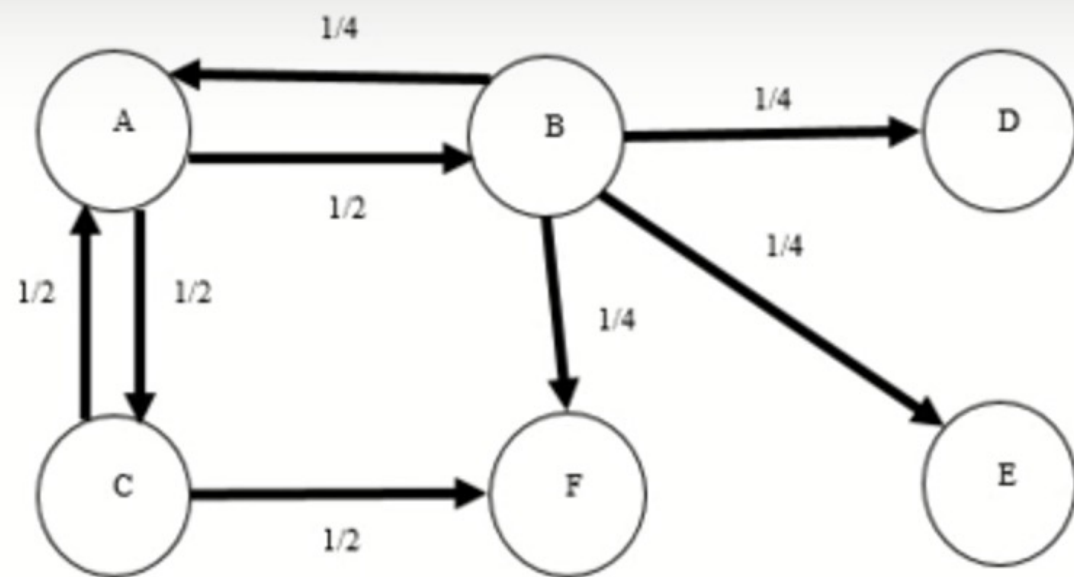
M^T

	A	B	C	D	E	F
A	0	1/4	1/2	0	0	0
B	1/2	0	0	0	0	0
C	1/2	0	0	0	0	0
D	0	1/4	0	0	0	0
E	0	1/4	0	0	0	0
F	0	1/4	1/2	0	0	0

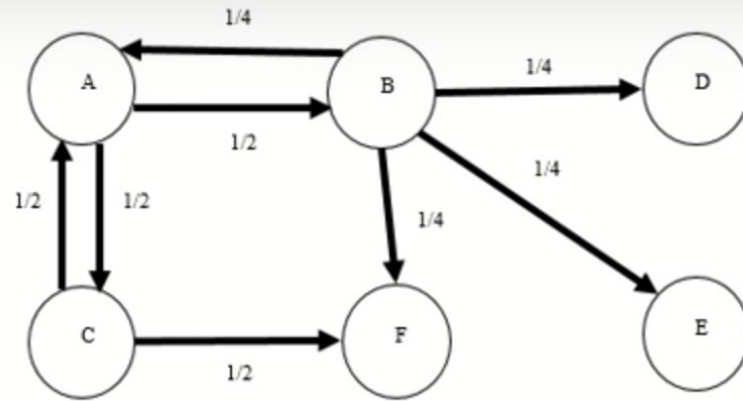
- Teleport Factor = 0.8

$M^1 =$

$$\begin{pmatrix}
 0 & 1/4 & 1/2 & 0 & 0 & 0 \\
 1/2 & 0 & 0 & 0 & 0 & 0 \\
 1/2 & 0 & 0 & 0 & 0 & 0 \\
 0 & 1/4 & 0 & 0 & 0 & 0 \\
 0 & 1/4 & 0 & 0 & 0 & 0 \\
 0 & 1/4 & 1/2 & 0 & 0 & 0
 \end{pmatrix}
 *
 \begin{pmatrix}
 0.8 \\
 0.8 \\
 0.8 \\
 0.8 \\
 0.8 \\
 0.8
 \end{pmatrix}
 =
 \begin{pmatrix}
 0.6 \\
 0.4 \\
 0.4 \\
 0.2 \\
 0.2 \\
 0.6
 \end{pmatrix}$$

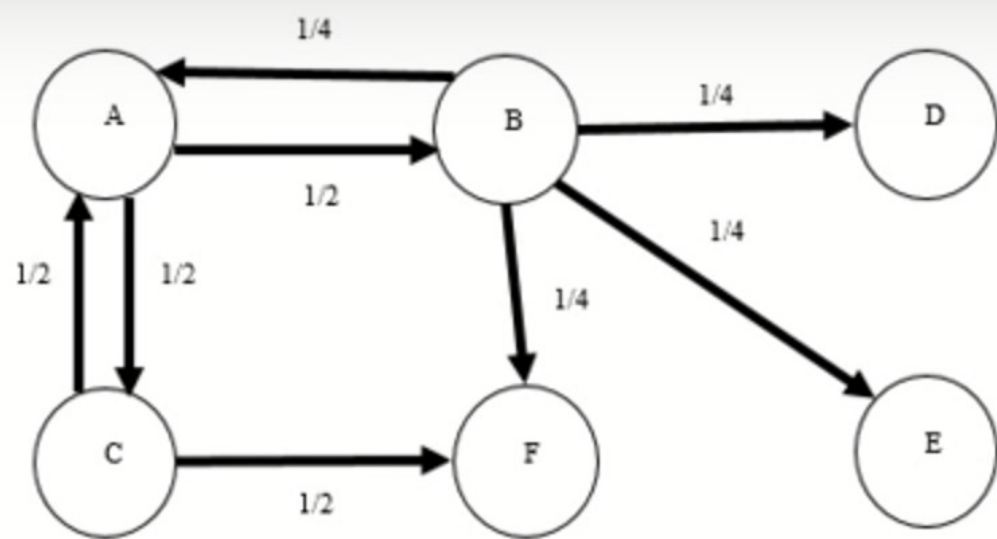


Inbound, Outbound and Dangling Links



$M^2 =$

$$\begin{pmatrix}
 0 & 1/4 & 1/2 & 0 & 0 & 0 \\
 1/2 & 0 & 0 & 0 & 0 & 0 \\
 1/2 & 0 & 0 & 0 & 0 & 0 \\
 0 & 1/4 & 0 & 0 & 0 & 0 \\
 0 & 1/4 & 0 & 0 & 0 & 0 \\
 0 & 1/4 & 1/2 & 0 & 0 & 0
 \end{pmatrix}
 *
 \begin{pmatrix}
 0.6 \\
 0.4 \\
 0.4 \\
 0.2 \\
 0.2 \\
 0.6
 \end{pmatrix}
 =
 \begin{pmatrix}
 0.3 \\
 0.3 \\
 0.3 \\
 0.1 \\
 0.1 \\
 0.3
 \end{pmatrix}$$



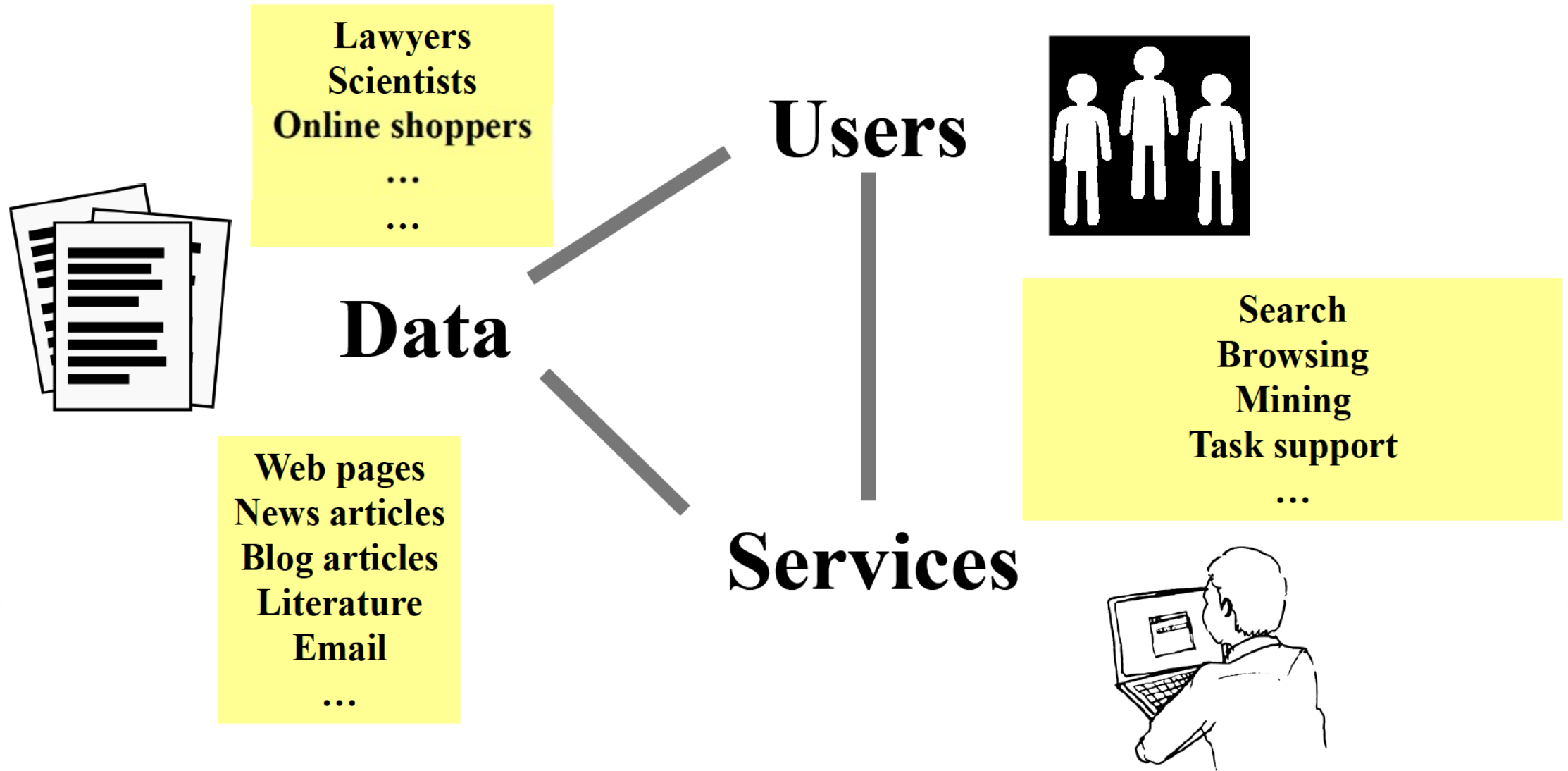
$M^3 =$

$$\begin{pmatrix}
 0 & 1/4 & 1/2 & 0 & 0 & 0 \\
 1/2 & 0 & 0 & 0 & 0 & 0 \\
 1/2 & 0 & 0 & 0 & 0 & 0 \\
 0 & 1/4 & 0 & 0 & 0 & 0 \\
 0 & 1/4 & 0 & 0 & 0 & 0 \\
 0 & 1/4 & 1/2 & 0 & 0 & 0
 \end{pmatrix}
 *
 \begin{pmatrix}
 0.3 \\
 0.3 \\
 0.3 \\
 0.1 \\
 0.1 \\
 0.3
 \end{pmatrix}
 =
 \begin{pmatrix}
 0.225 \\
 0.15 \\
 0.15 \\
 0.075 \\
 0.075 \\
 0.225
 \end{pmatrix}$$

Next Generation Search Engines

- More specialized/customized (vertical search engines)
 - Special group of users (community engines, e.g., CiteSeer)
 - Personalized (better understanding of users)
 - Special genre/domain (better understanding of documents)
- Learning over time (evolving)
- Integration of search, navigation, and recommendation/filtering (full-fledged information management)
- Beyond search to support tasks (e.g., shopping)
- Many opportunities for innovations!

The Data-User-Service (DUS) Triangle



Future Intelligent Information Systems

