

(3)

FMA —

Q. Determining # distinct elements in stream.

I/p stream $\Rightarrow 1, 3, 2, 1, 2, 3, 4, 3, 1, 2, 3, 1$.
hash fnr. $h(x) = (6x+1) \bmod 5$

x	$h(x) = (6x+1) \bmod 5$	$(h(x))_2$	# Trailing 0's
1	$(6(1)+1) \bmod 5 = 7 \bmod 5 = 2$	010	1
3	4	100	2
2	3	011	0
1	2	010	1
2	3	011	0
3	4	100	2
4	0	000	0
3	4	100	2
1	2	010	1
2	3	011	0
3	4	100	2
1	2	010	1

$$\therefore \text{Max # Trailing 0's} = \underline{2} = \sigma$$

\therefore Distinct Value $R = 2^{\sigma} - 2^0 = 4$
 \therefore There are σ distinct elements.

0 0 0	0
0 0 1	1
0 1 0	2
0 1 1	3
1 0 0	4
1 0 1	5
1 1 0	6
1 1 1	7
2 2 2	0

④ Bloom Filter - membership test using prob.
 Eg. checking availability of a user, "it is taken".
 Search it from a list of registered user names.
 FP.

Empty Bloom filter \rightarrow 1 0 1 0 0 0 0 0 0 0
 empty array of 0's of length 8.

Q. 'at throw' to be inserted
 'catch' to be searched. window size = 10

	1	0	0	0	0	0	0	0	0
	a	t	h	e	w	n	d	o	u
1	2	0	8	1	8	1	5	2	3
$\div 10$									
1	0	8	8	5	0	0	3	0	0

1	1	0	1	0	0	0	0	1	0
0	1	2	3	4	5	6	7	8	9

Catch search:

C a t + c b

3	1	2	0	3	8
---	---	---	---	---	---

 $\div 10 \quad \div 10 \quad \div 10 \quad \div 10 \quad \div 10$

search for indices 3, 1, 2, 3, 8 if they're set to 1

they indeed are
 \therefore FP

spell checks.

Q. Bloom filter.

$$m = 5$$

0	1	2	3	4
0	0	0	0	0

$$h_1(x) = x \bmod 5$$

$$h_2(x) = (2x + 3) \bmod 5$$

hash
2 hashfn help avoid
overlap = fp in hsgl
↓ # hashfn = 5

1. Insert 9

$$(9)_2$$

0	1	2	3	4
0	0	0	0	1

2. Insert 11

$$(11)_2$$

1	1	0	0	1
0	1	2	3	4

3. check 15

$$15$$

1	1	0	0	1
0	1	2	3	4

$$h_1(x)$$

$$9 \bmod 5 = 4$$

$$11 \bmod 5 = 1$$

$$15 \bmod 5 = 0$$

$$h_2(x)$$

$$[2(9) + 3] \bmod 5$$

$$21 \bmod 5 = 1$$

$$[2(11) + 3] \bmod 5$$

$$25 \bmod 5 = 0$$

$$[2(15) + 3] \bmod 5 = 3$$

not set

Surely Not present.

(5)

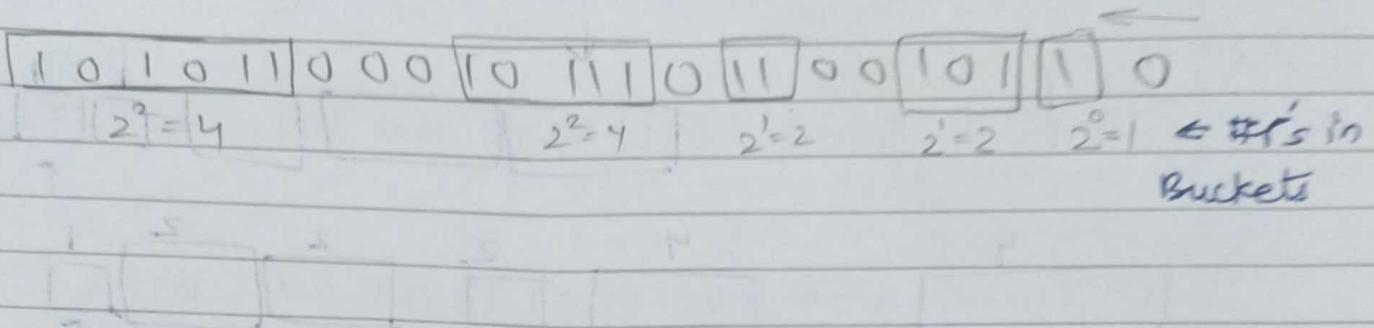
DGM Algo →

Error < 50%

estimate #: 1's

Q.

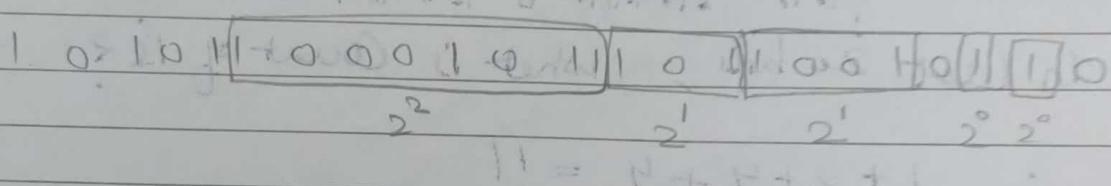
101011000101110110010110



Rightmost Side is Start

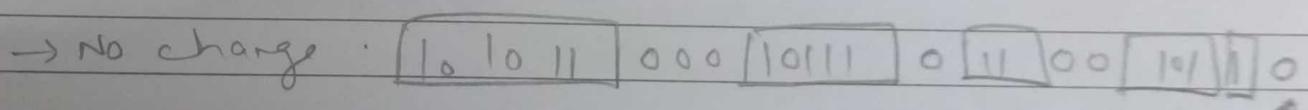
Rightmost bit of Bucket = 1

0's can be in Bucketless

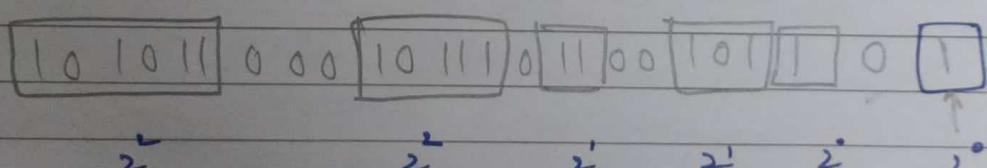
Right to Left size of Bucket = $2^0, 2^0, 2^1, 2^1, 2^2, 2^2$ 

u can repeat a Bucket size twice

i) New Bit '0' arrives (from right to left)



ii) New Bit '1' arrives

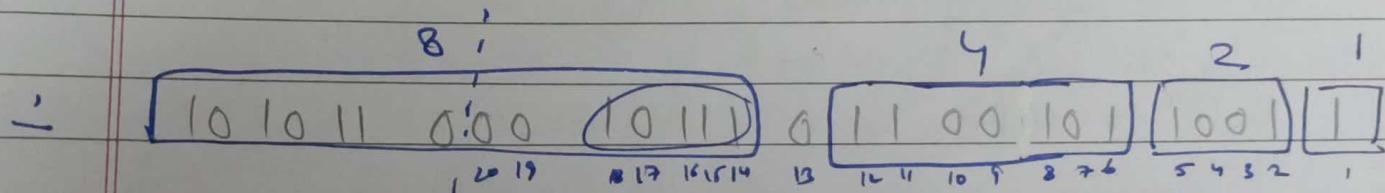
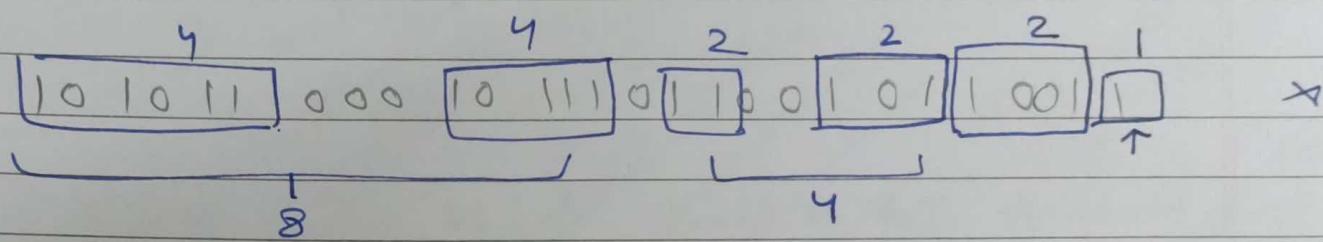
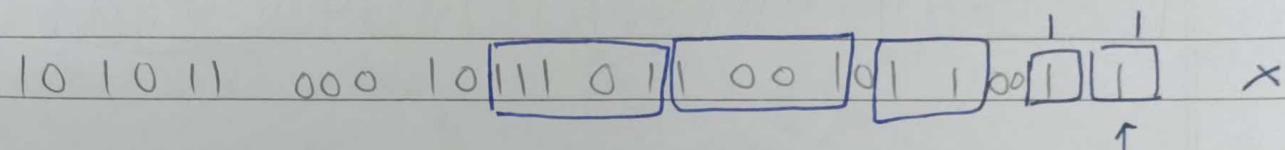
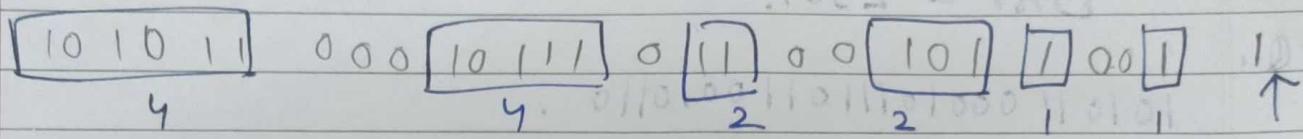


still ok.

20, 30 = 20 1st window } u need whole window in Mn
3 30 - 20 = 10 2nd window }

iii)

One more 1 arrives.



In last 20 bit, what is Count of 1's?

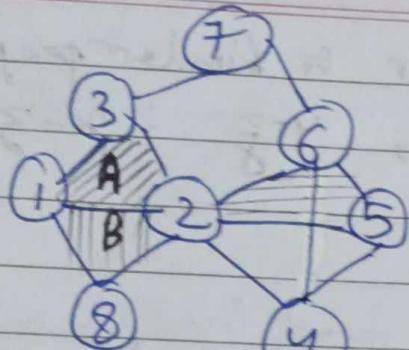
$$\therefore 1 + 2 + 4 + 4 = \underline{11}.$$

No common nodes
↓
no community

classmate

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Q.



$$\{1, 2, 8\} \neq \{1, 2, 3\}$$

cliques

drawings →

1, 2, 3

1, 2, 8

2, 6, 5

2, 4, 5

4, 5, 6

2, 4, 6

2, 4, 5, 6

K=3

K=4

are there 2
adj nodes?

A > (A, B) 1, 2

B > 2, 5 (C, D)

C > 4, 6 (E, F)

D > 4, 6 (E, F)

1, 2, 3, 8

2, 6, 5, 4

4, 5, 6, 2

Q.

1, 2, 3

1, 3, 4

4, 5, 6

5, 6, 7

5, 6, 8

5, 7, 8

6, 7, 8

> (3)

5, 6

> 7, 8

1, 2, 3, 4

4, 5, 6, 7, 8

communities

now, see if
 B-C
 B-D
 C-D

ABD not triangle
 ACD triangle

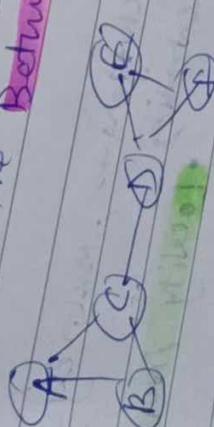
write all edges down & sort their start & end in G.
 write all edges down & sort their start & end in G.
 are they connected via E.
 are they connected via F.

$$\delta(x,y) = \infty \quad \text{if } x \neq y$$

$$\delta(x,y) = 0 \quad \text{if } x = y$$

$$\delta(x,y) = \infty \quad \text{if } x \neq y \text{ and } y \text{ is not present}$$

- Calculate the Betweenness Centrality for node C



Betweenness Centrality for node C

(x,y)	$\delta(x,y)$	via C	via E	via F	via G	Total
(A,B)	1	1	0	0	0	1
(A,C)	1	1	0	0	0	1
(A,D)	1	1	0	0	0	1
(B,C)	1	1	0	0	0	1
(B,D)	1	1	0	0	0	1
(C,D)	1	1	0	0	0	1
(E,F)	1	0	1	0	0	1
(E,G)	1	0	1	0	0	1
(F,G)	1	0	0	1	0	1
Total		5	3	2	1	10

(x,y)	$\delta(x,y)$	via C	via E	via F	via G	Total
(A,B)	1	1	0	0	0	1
(A,C)	1	1	0	0	0	1
(A,D)	1	1	0	0	0	1
(B,C)	1	1	0	0	0	1
(B,D)	1	1	0	0	0	1
(C,D)	1	1	0	0	0	1
(E,F)	1	0	1	0	0	1
(E,G)	1	0	1	0	0	1
(F,G)	1	0	0	1	0	1
Total		5	3	2	1	10

WAT Measure of Similarity

Context
 - keywords
 in veget. doc

total
 rep

Context

= keywords
 in veget. doc

Date
 Page

ASSNATE

Date
 Page

ASSNATE

Date
 Page

- Calculate the Betweenness Centrality for All nodes

B.C. for All nodes

Date
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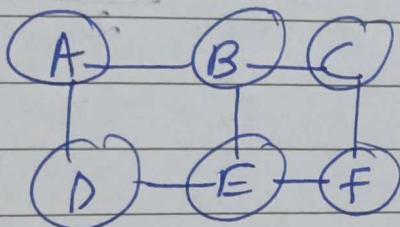
Girvan Newman algo. — BFS, hierarchical clustering

Divisive

Edge
Betweenness

- Calculate the Betweenness centrality for edges.
- Solve using GNA.

→



$$\begin{matrix} A & + & B & + & D \\ B & | & C, E \\ D & | & E \\ E & | & F \end{matrix}$$

$$\begin{matrix} 1 + \left(\frac{1+2}{3}\right)/2 & 1 + \frac{1}{3} \\ 1 + \frac{1}{3} & 1 + \left(\frac{1+2}{3}\right)/2 \\ 1 + \left(\frac{1+2}{3}\right)/2 & 1 + \frac{1}{3} \\ 1 + \left(\frac{1+2}{3}\right)/2 & 1 + \frac{1}{3} \\ (from B=1, D=1) & \\ (from C=1, E=2) & \end{matrix}$$

$$\begin{matrix} \frac{1}{3} & F_3 \\ C_1 & E_2 \\ (1 + \frac{1}{3}) & (1 + \frac{2}{3}) = 0.83 \\ BI & DI \\ (1 + \frac{1}{3}) + 0.83 & 0.83 + 1 \\ A & \end{matrix}$$

Algo is for Detection & Analysis of Community Structure

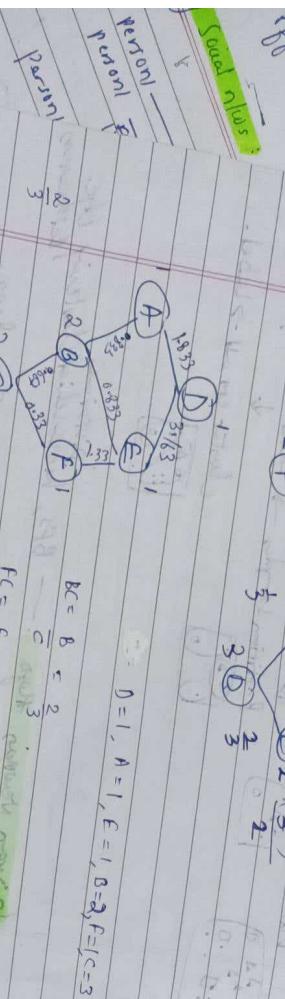
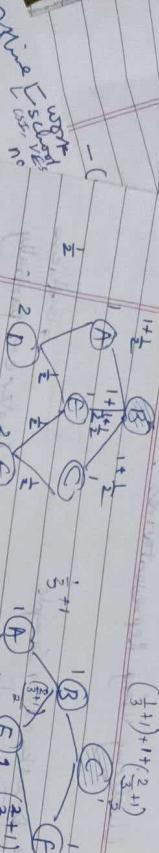
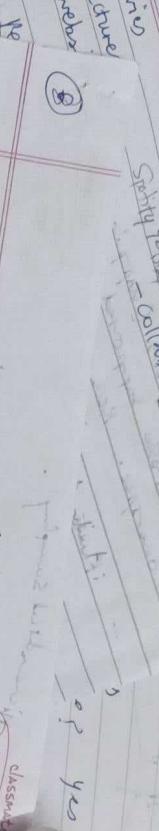
— depends on iterative elimination of edges w/ highest values.

Recommendations = Search query
System based (A) previous work
Society based (C) literature people
Society based (B) friends Community

loc queries

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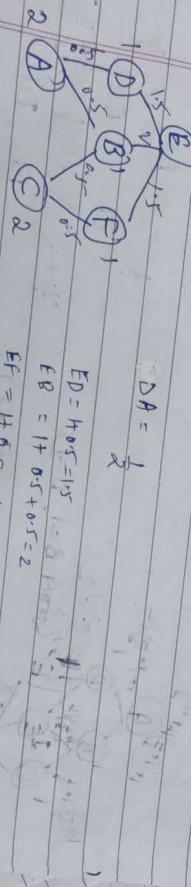
Everything
queried for
not stammered
and (containing)



$$PF = 0.33 + E - 1.33 = 0.83$$

$$DA = 0.833 + D = 1.833$$

$$DE = D + 1.33 + 0.833 = 3.063$$

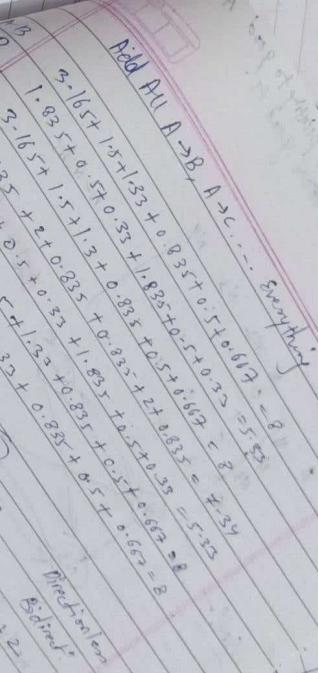


$$EF = 1 + 0.5 + 0.5 = 2$$

$$DA = \frac{1}{2} = 0.33$$

$$ED = 1 + 0.33 = 1.33$$

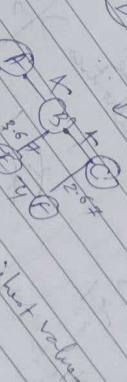
$$CB = \frac{(1+0.67)}{2} = 0.835$$



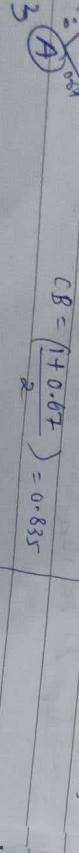
Algorithm for fast

queries

revert edges to highest value
stop: will lead to
brother communication
node community



return edges to highest value
stop: will lead to
brother communication
node community



$$DA = \frac{1}{2} = 0.33$$

$$ED = 1 + 0.33 = 1.33$$

$$CB = \frac{(1+0.67)}{2} = 0.835$$

Q. Hits. Calculate Hub & Auth. score

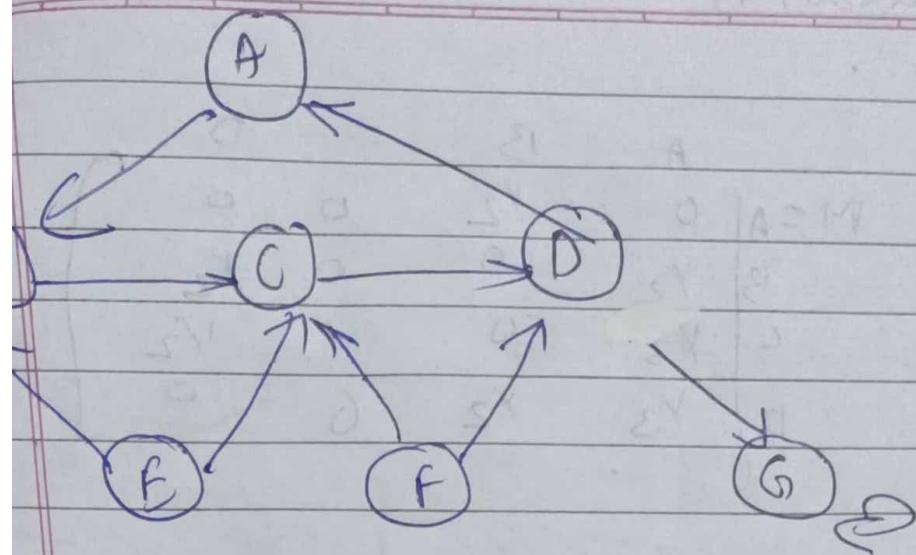
	1	2	3
1	0	0	1
2	0	0	1
3	0	0	0

Initial Hub weight vector = $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = 4$

$$\text{Auth. score} = V = A^T \cdot U$$

$$= \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix}$$

$$\text{Hub score updated} = U = A \cdot V = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix}$$



Identify Spider traps & dead ends

	A	B	C	D	E	F	G
A	0	1	0	0	0	0	0
B	0	0	1	0	0	0	0
C	0	0	0	1	0	0	0
D	1	0	0	0	0	0	1
E	0	1	1	0	0	0	0
F	0	0	1	1	0	0	0
G	0	0	0	0	0	0	1

No dead ends \therefore no full row = 0 (no node with #outlinks = 0).

G is a spider trap

\because G is a Dead End \therefore It has a selfloop except

next Come back from G.

\therefore it ends up getting all the PR

(or Honey pot : where da search, ends up that page itself)

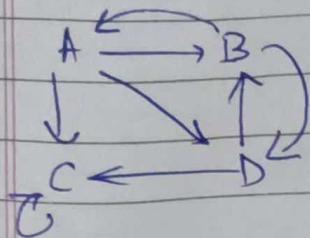
Solⁿ: Teleportation / Taxation

Solⁿ: to link spam : Topic sensitive PR

$$V_{t+1} = M \cdot V_t$$

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Q. Teleport? / Iteration
assume $\beta = 0.8$.



$$M = A^T \begin{bmatrix} 0 & Y_2 & 0 & 0 \\ Y_2 & 0 & 0 & Y_2 \\ X_3 & 0 & 1 & Y_2 \\ Y_3 & Y_2 & 0 & 0 \end{bmatrix}$$

C, A Spider Trap

$$[BM]v + (1-B)c$$

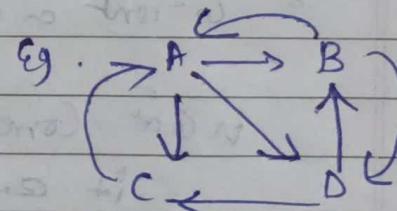
$$= \begin{bmatrix} 0 & 2/5 & 0 & 0 \\ 4/15 & 0 & 0 & 2/5 \\ 4/15 & 0 & 4/15 & 2/5 \\ 4/15 & 2/5 & 0 & 0 \end{bmatrix} \begin{bmatrix} v_A \\ v_B \\ v_C \\ v_D \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} + \frac{0.2}{4} \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$v' = \begin{bmatrix} 9/60 \\ 13/60 \\ 25/60 \\ 13/60 \end{bmatrix} \neq \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$$

if normal PR method followed

so on for many iterations

Q. Topic Specific PR \rightarrow



Given: Teleport set $\{B, D\}$

$$\beta = 0.8$$

$$(1-\beta) \frac{e}{n} = \frac{0.2}{2} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 1/10 \\ 0 \end{bmatrix}$$

rest same

(n=2 only B & D)

Simple PR

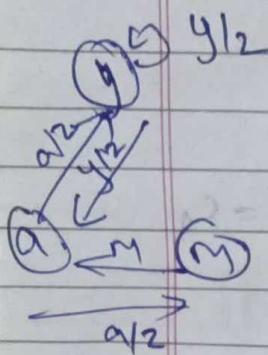
Links = votes

Recursive formulation

flow Model

nodes	in	out
y	y, a	a, y
a	y, m	y, m
m	a	a

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$$\text{PR of } a \Rightarrow \frac{y}{2} + \frac{m}{1} \quad (\text{from})$$

$$\text{PR of } y \Rightarrow \frac{y}{2} + \frac{a}{2}$$

$$\text{PR of } m \Rightarrow \frac{a}{2}$$

$$\begin{bmatrix} y \\ y \\ a \\ m \end{bmatrix} = \begin{bmatrix} y & a & m \\ y & 0 & 0 \\ a & 0 & 1 \\ m & 0 & 0 \end{bmatrix} = M$$

power iteration 1.

$$\therefore v_1 = M \cdot v, \quad v = \begin{bmatrix} y_3 \\ y_3 \\ y_3 \end{bmatrix}$$

$$= \begin{bmatrix} y \\ y \\ y \end{bmatrix} \begin{bmatrix} y_3 \\ 3/6 \\ y_6 \end{bmatrix}$$

$$(2\text{nd iter.}) \quad v_2 = \begin{bmatrix} 5/12 \\ 2/13 \\ 3/12 \end{bmatrix}$$

$$9/24$$

$$11/24$$

$$1/6$$

$$v_2 = M \cdot v_1$$

$$v_3 = M \cdot v_2$$

$$= M \cdot (M \cdot v_1) = M^2 \cdot v_1$$

now check for highest Again.

	old A ⁱⁿ	old H ^{out}	New A	New H	
N1	1/9	3/9	2/9	8/9	outdeg(N1) = N2, N3, N4 C Auth.
N2	2/9	3/9	6/9	7/9	$= \frac{2}{9} + \frac{4}{9} + \frac{2}{9} = \frac{8}{9}$
N3	4/9	2/9	$\frac{3+3}{9} + \frac{2+1}{9} = 1$	6/9	
N4	2/9	1/9	$\frac{3+2}{9} = \frac{5}{9}$	4/9	
(9)			EA	EH	

$$\text{Indegree}(N1) = \text{only } N2 \Rightarrow \text{outdegree}(N2) = \frac{\text{old } H}{9} N2$$

$$\text{Indegree}(N2) = N1 \& N3 = \sum \text{outd. } N1 \& N2 = \frac{3+3=6}{9} = \frac{6}{9}$$

do on.

as $K \uparrow$, H & A values do converge

calculate in & outdegrees

↓
Normalize ems

New A for node N1

= [Integ. of N1]'s outdegree
--- (adj column)

New H for node N1

= [outdeg of N1]'s indegree
--- 1st col.

0A_{in} 0M_{out}

Exponentially Decaying Windows

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Sliding window [time based] {Recent / After active
size based}

but to avoid making this distinct, & j assign weights
↑ weight to new.

weight | weight Decays over time Exponentially
time

Q. Fifa, ip, Fifa, ip, ip, ip, Fifa — Exponential
Decaying Windows Numerical
if u see, # Fifa = 3
ip = 4

Sol. — Current wood = Fifa . (Add 0 if wood = ip)
Fifa = 1 * (1 - 0.1) = 0.9

$$ip = 0.9 * (1 - 0.1) + 0 = 0.81$$

$$fifa = 0.81 * (1 - 0.1) + 1 = 1.729$$

$$ip = 1.729 * (1 - 0.1) + 0 = 1.5561$$

$$ip = 1.5561 * (1 - 0.1) + 0 = 1.4085$$

$$ip = 1.4085 * (1 - 0.1) + 0 = 1.2605$$

$$fifa = 1.2605 * (1 - 0.1) + 1 = 2.135$$

now current wood = ip (Add 1 if wood = ip)
Fifa = 0 * (1 - 0.1) + 0 = 0

$$ip = 0 * (1 - 0.1) + 1 = 1$$

$$fifa = 1 * (1 - 0.1) + 0 = 0.9$$

$$ip = 0.9 * (1 - 0.1) + 1 = 1.81$$

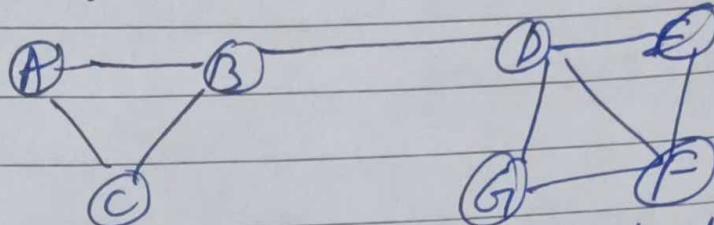
$$ip = 1.81 * (1 - 0.1) + 1 = 2.7919$$

$$ip = 2.7919 * (1 - 0.1) + 1 = 3.764$$

$$fifa = 3.764 * (1 - 0.1) + 0 = 3.3264$$

Score of Fifa < Score of ip
 $3.3264 \therefore ip$ is more TRENDING than Fifa

Locality property: friendship paradox



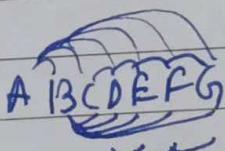
ticks over
edges that
do exist fr.

does this graph exhibit locality property?

# Vertices	neighbour/ Adj vertices	Combin ations	# ✓ / Total Combos
A	B, C	BC	1/1 = 1
B	A, C, D	AC, AD, CD	1/3 = 0.33
C	A, B	AB	✓/1 = 1
D	B, E, G, F	BE, BG, BF, EG, EF, GF	2/6 = 0.167
E	D, F	DF	1/1 = 1
F	G, D, E	GD, DE, GE	2/3 = 0.667
G	D, F	DF	✓/1 = 1

$(9/16)$
 $= 0.563$

Nodes = 7



Edges = 9

$$\# \text{possible edges} = \frac{n(n-1)}{2} = \frac{7 \times 6}{2} = 21$$

$$\frac{\text{actual # edges}}{\text{total possible # edges}} = \frac{9}{21} = 0.429.$$

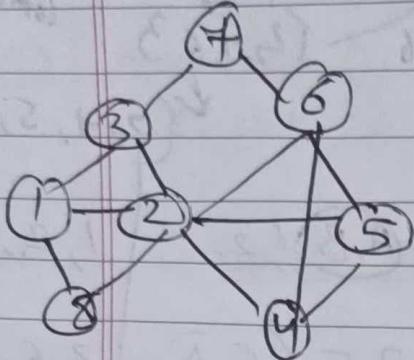
∴ Conclude
locality prop

For 2 edges, $\frac{9-2}{21-2} = 0.368$ prob. \rightarrow Exist here
 $\therefore 0.563 > 0.368$

Sol. : Teleportation / Taxation

So 1ⁿ to link span : Topic sensitive PR

0. Triangle Colabut:



vertex	degree	edge
1	3	(1,2)(1,3)(1,8)
2	6	(2,6)(2,5)(2,4)(2,3)(2,7)(2,8)
3	3	(3,1)(3,2)(3,7)
4	3	(4,2)(4,5)(4,6)
5	3	(5,2)(5,6)(5,4)
6	4	(6,2)(6,5)(6,7)(6,8)
7	2	(7,1)(7,3)
8	2	(8,1)(8,2)

key	count: 2 if Nodir.	values
(1,2)(2,1)		w
(1,3)(3,1)		
(1,8)(8,1)		
(2,6)(6,2)	x	
(2,5)(5,2)		
(2,4)(4,2)		
(2,3)(3,2)		
(3,7)(7,3)		
(4,5)(5,4)	y	
(4,6)(6,4)		
(5,6)(6,5)		
(6,7)(7,6)	z	

We can identify if (6,2) exists or no
either bidir or uni

Arrange the table in ascending
order of degrees.



10.2.1 Parameters Used in Graph (Social Network)

- Q. Explain the graph parameters listed below :
- (i) Degree (ii) Geodesic distance (iii) Density.
- Q. How degree, closeness, between's centrality is measured?

Every node is distinct in a network and it is part of graph by set of links. Some general parameter consider for any social network as graph are :

- Degree** : Number of adjacent nodes (considering both out degree and in-degree). Degree of node n_i denoted by $d(n_i)$.
- Geodesic Distance** : Actual distance between two node n_i and n_j , expressed by $d(i, j)$.
- Density** : It gives correctness of a graph, it is useful to count closeness of network.
- Centrality** : It tells about degree centrality i.e. nodes appearance in the centre of network centrality has types.

Example :

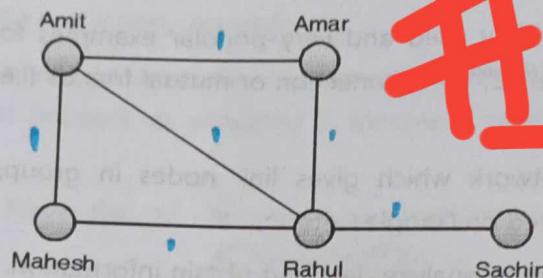


Fig. 10.2.1

Degree of each nodes are as follows :

Nodes	Degree
Amit	3
Amar	2
Mahesh	2
Rahul	4
Sachin	1

Density of undirected graph is 0.6.

Geodesic Distances between two nodes is as follows :

	Amit	Amar	Mahesh	Rahul	Sachin
Amit	-	1	1	1	2
Amar	1	-	2	1	2
Mahesh	1	2	-	1	2
Rahul	1	1	1	-	1
Sachin	2	2	2	1	-

#edges

$2 * \# \text{node}$

= 6

~~245~~

Explain how dead ends are handled in Page Rank.

In a given part of web, if we encountered with a page which doesn't have links which are going out from that page or component.

This will affect the transition matrix directly as, the column containing entry for dead end page will lead to sum = 0 instead of 1.

The property of having sum = 1 for most of the columns in a given transition matrix is known as 'Stochasticity' and if there are dead ends then some of the columns have '0' entries.

Consider the Fig. 8.1.10.

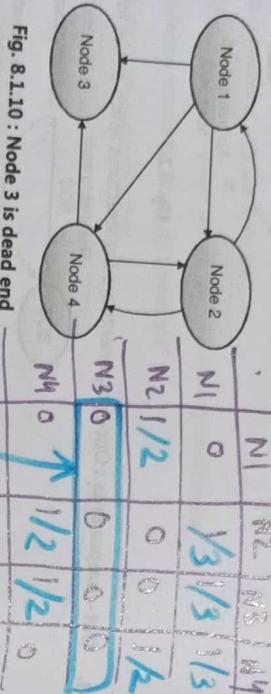


Fig. 8.1.10 : Node 3 is dead end

referring Fig. 8.1.10 we came to know that, Node 3 is dead end and while searching if we countered on the Node 3, i.e. at dead end then web surfing will stuck at that page or node as there

is a single out link from Node 3. Hence, transition matrix for Fig. 8.1.10 is,

$$M = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1/2 & 0 & 0 \\ 1/3 & 0 & 0 & 1/2 \\ 0 & 0 & 1/2 & 0 \end{bmatrix}$$

are the ways to deal with dead ends :

ach to deal with dead ends we can delete that node by removing their incoming links.

antage of this approach is it will introduce more dead ends which has to be solved with the same

ch in recursive manner.

we delete the node but total page rank for a given graph to or web will be kept as it is the ssors for the calculation of page rank.

ally we can consider the successor nodes and have a division operation.

procedure, some nodes might be there which are not available in graph G but they are done predecessors calculations.

e iterations all nodes has their page rank the order in which page ranks are calculated is posite to node deletion order.

- Suppose we have graph containing nodes and these nodes are arranged in following manner as shown in Fig. 8.1.11.

MU : Dec. 16, 5 Marks

	Node 1	Node 2	Node 3	Node 4	Node 5
Node 1	1	0	1	0	0
Node 2	0	1	0	0	0
Node 3	1/2	0	0	0	0
Node 4	0	0	0	1	0
Node 5	0	0	0	0	1

Fig. 8.1.11

- If we observe the Fig. 8.1.11 to calculate the page rank. We find that, Node 5 is the dead end as it doesn't have any forward links i.e. the links going out from Node 5.
- So hence, to avoid the dead ends, delete the Node 5 and its corresponding arc coming from Node 3.

	Node 1	Node 2	Node 3	Node 4	Node 5
Node 1	1	0	1	0	0
Node 2	0	1	0	0	0
Node 3	1/2	0	0	0	0
Node 4	0	0	0	1	0
Node 5	0	0	0	0	1

Fig. 8.1.12

- By observe the Fig. 8.1.12 we came to know that now 'node 3' is 'dead end'
- Now as we are avoiding the dead ends. Hence delete 'Node 3' and it is respective in coming edges.

Now, Graph G becomes,

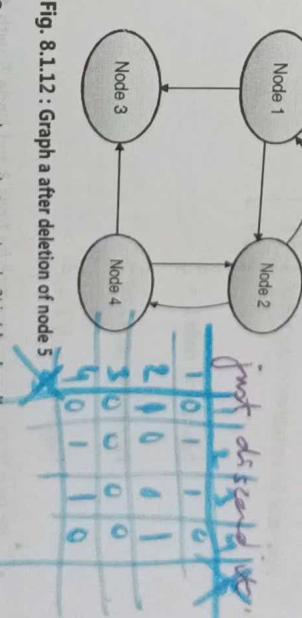


Fig. 8.1.12 : Graph a after deletion of node 3

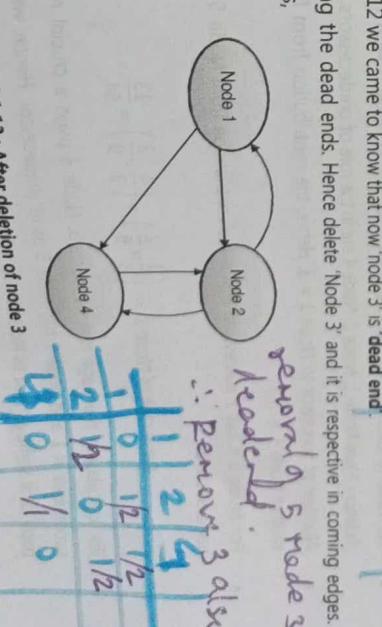


Fig. 8.1.13 : After deletion of node 3

Now every node : Node 1, Node 2, and Node 3 has edges coming out i.e. forward links. Hence, is no dead end in graph 'G'.



- The transition matrix for above graph will be,

$$M = \begin{bmatrix} 0 & 1/2 & 0 \\ 1/2 & 0 & 1 \\ 1/2 & 1/2 & 0 \end{bmatrix}$$

We can have component vector representation for above matrix as follows :

$$\begin{bmatrix} 1/3 \\ 1/3 \\ 1/3 \end{bmatrix} - (1) \text{ Iteration 1}$$

$$\begin{bmatrix} 1/6 \\ 3/6 \\ 2/6 \end{bmatrix} - (2) \text{ Iteration 2}$$

$$\begin{bmatrix} 3/12 \\ 5/12 \\ 4/12 \end{bmatrix} - (3) \text{ Iteration 3}$$

Final value for component vector will be,

$$\begin{bmatrix} 2/9 \\ 4/9 \\ 3/9 \end{bmatrix}$$

$$\text{Page rank for node 1} = \frac{2}{9}$$

$$\text{Page rank for node 2} = \frac{4}{9}$$

$$\text{Page rank for node 4} = \frac{3}{9}$$

P.R. for node 3 = ?

5 = ?

- We have to calculate the page rank for Node 3 and Node 5 with the exact opposite order of node deletion. Here Node 1, Node 2, Node 4 are in the role of predecessors.
- Number of successor to Node 1 = 3, Hence, the contribution from Node 1 for calculating the page rank of Node 3 is $1/3$
- For Node 5 it has 2 successors. Hence the contribution from Node 5 for calculating the page rank of node 3 is $1/2$.

$$\text{Page rank of Node 3} = \left(\frac{1}{3} \times \frac{2}{9} \right) + \left(\frac{1}{2} \times \frac{3}{9} \right) = \frac{13}{54}$$

- For calculating the page rank of Node 5, Node 3 plays a crucial role. As Node 3 has, number of successors = 1 and node 1 has node 3 as its predecessor. Hence, we can conclude that Node 5 was page rank same as that of Node 3.
- As the aggregate of their page rank is greater than 1, so it doesn't indicate the distribution for a given user who is surfing through that web page. Still it highlights the importance of web page relatively.