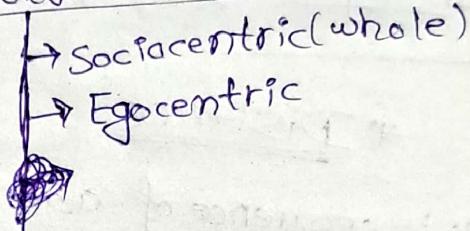


Social Network

- A social network is social structure of people, related to each other through a common relation or interest.
- SNA is study of social networks to understand their structure & behaviour.

People - nodes
relations - edges

Social Network Analysis

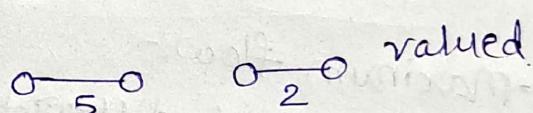
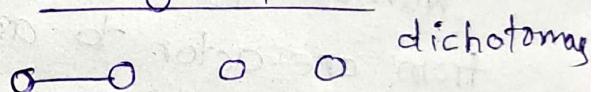


Types of social relations

- friendship
- acquaintance
- kinship
- advice
- hindrance
- sex.

class code :-

Strength of Tie



3 c + fmde

o actor

o o dyad.



triad.

friendship

Kinship

Relations.

o o adjacent node to

o ← o incident node to

o ← o

o → o

o ← o

o → o

3 isomorphism classes

→ null dyad.

→ mutual dyad.

→ asymmetric.

converse of graph:

Reverse direction of all arcs.

Density :- Number of ties that are present
amount of ties that can be present

out-degree :- sum of connections from an actor to others.

In-degree :- sum of connections to an actor.

* Distance

- walk :- sequence of actors & relations that begins & ends with actors.

- Geodesic distance:-
Number of relationships in the shortest possible walk
from one actor to another.

- maximum flow:-
amount of different actors in neighbourhood of a source that lead to pathways to Target.

• Degree:-
sum of connections from or to an actor.

• closeness centrality:-
- Dist. of one actor to all others.

• Between centrality:-
no. that represents how frequently an actor is between other actors geodesic path.

centrality Measures :-

Who is the most prominent?

Who knows the most actors?

* Who controls knowledge flow?

* Who has shortest distance?

closeness

Who has more power?

Between.

Degree



What are LinkedIn connections?

1st - directly connected

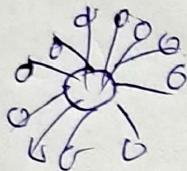
2nd - people connected to 1st degree

3rd - people connected to 2nd degree.

Why 6 degrees of separation?

6 degrees of
Freedom.
separation

Each friend connected to 44 people



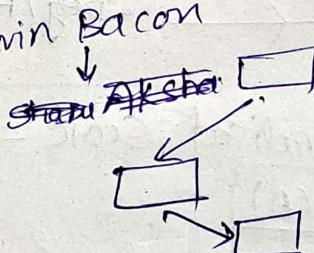
$$(44)^6 \approx 7.26 \text{ Billion people}$$

small world theorem.

You can get job connections easily through random acquaintances than through your close friends! - Veritasium.

strength of weak Tie

The oral of Bacon! \rightarrow Kevin Bacon



Bacon 13.

Erdős Number project

• Adjacency matrix
Lone mode

• Affiliation matrix
Two mode.

Adjacency matrix vs Attribute Matrix

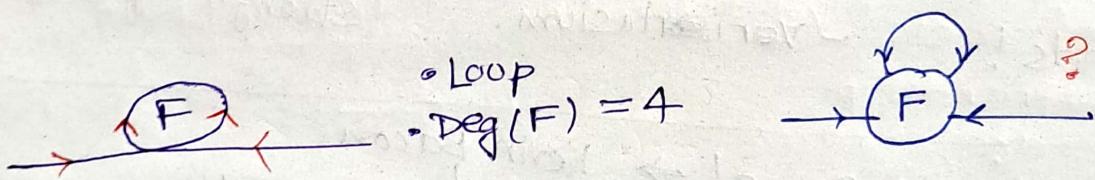
order :- no. of nodes in graph.

cardinality :- no. of edges.
size :- max. no. of edges , ie $\frac{n(n-1)}{2}$ for undirected.
 $n(n-1)$ for directed.

degree centrality

for directed node \rightarrow in-degree
 \rightarrow out-degree.

Average degree $\Rightarrow \frac{\text{Deg}(A) + \text{Deg}(B) + \dots + \text{Deg}(N)}{n}$



Degree centralised score = $\frac{\text{Deg}(A)}{\max(\text{Deg of all other nodes})}$

$$= \frac{\text{Deg}(A)}{n-1}$$

standardised score or normalised form.

closeness centrality:- $\Rightarrow \arg \left(\min \left(\text{dist with all nodes} \right) \right)$

\Rightarrow inverse of sum of distance between a node & other nodes in a network.

closeness score $\Rightarrow \frac{1}{\text{total dist}}$

Standardised score $\Rightarrow \frac{1}{\text{total dist}} \times (n-1)$

most closest node \Rightarrow highest standardised score

Betweenness centrality :-

$$g(v) = \sum_{s \neq v \neq t} \frac{\sigma_{st}(v)}{\sigma_{st}}$$

$$\text{normal}(g(v)) = \frac{g(v) - \min(g)}{\max(g) - \min(g)}$$

which results in $\max \text{normal} \Rightarrow 1$

σ_{st} is total no. of shortest path from $s \rightarrow t$.
 $\sigma_{st}(v)$ is no. of paths passing through v . (v is not s or t).

~~This is not~~

For friendship network, which is most popular person?

Q] For friendship network, which is most popular person?
degree centrality (but there is never fixed answer)

ans:-

Q] Find section in an information flow network that can most efficiently ~~info~~ info flow info?

ans:- closeness centrality.

Q] Which section controls, information flow?

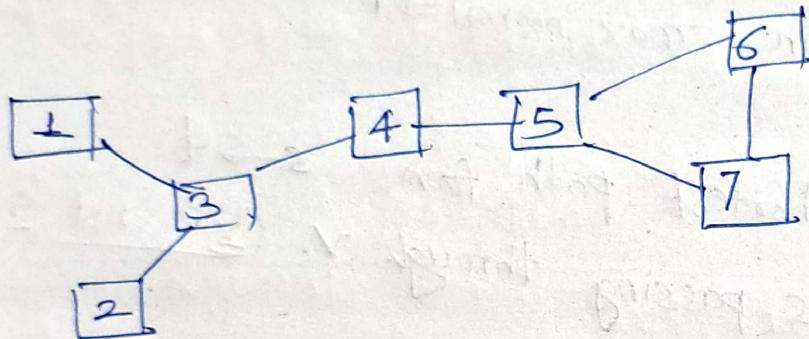
ans:- betweenness centrality

THANK YOU!

- * do all maths yourself.
- * GEPHI network Lab practical.
- * Install NetworkX.
- * Next to Next class :- Quiz.

Data Management :-

- Adjacency Matrix.
 - Asymmetric, binary
 - Asymmetric weighted.
- Affiliation Matrix.



consider only paths involving 4 in them.
As they will have contribution in betweenness.

$$(1-3) \times$$

$$(1-5) \checkmark$$

$$1-3-4-5$$

Node 3

1-2	1
1-4	1
1-5	1
1-6	1
1-7	1

for directed divide by
 $(n-1)(n-2)$

for undirected divide by

$$\frac{(n-1)(n-2)}{2} = \frac{(7-1)(7-2)}{2} = 15$$

2-4	1
2-5	1
2-6	1
2-7	1

9

Hence: - $\frac{9}{15}$ (Here undirected).

Node 5

1-6	1
1-7	1
2-6	1
2-7	1
3-6	1
3-7	1
4-6	1
4-7	1
6-7	0
$\boxed{\frac{8}{15}}$	

node	score
1	0
2	0
3	$\frac{9}{15}$
4	$\frac{9}{15}$
5	$\frac{8}{15}$
6	0
7	0

Another formula

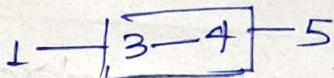
$$\text{between}(S|t) = \frac{\text{no. of paths}}{\text{no. of nodes.}}$$

ex) $(1,5) \Rightarrow \frac{1}{2}$
 between.

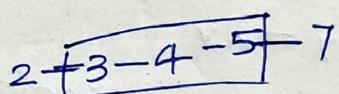
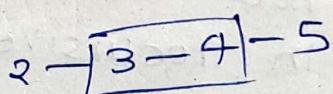
$$(3-5) \Rightarrow \frac{1}{1}$$

$$(2-5) \Rightarrow \frac{1}{2}$$

$$(2-7) \Rightarrow \frac{1}{3}$$



3-4-5.



Graph Measures :-

Eccentricity ($\epsilon(v)$)

$\epsilon(v)$ of vertex 'v' in connected graph G is max. graph distance between 'v' and any other vertex u of G .

$$\epsilon(v) = \max_{u \in V} d(u, v)$$

For disconnected point $\epsilon(v) = \infty$.

• Diameter :- max eccentricity

• Radius :- min eccentricity

• Central Point :- $\epsilon(v) = \text{Radius}$

Eigenvector centrality (You are powerful if friends are powerful)

Influence of node in network:-

Importance of node depends on, importance of its neighbours in recursive manner.

$$Av = \lambda v$$

Select an eigen vector associated with largest eigen value

Eigen Vector

$$A - \lambda I | I | = 0$$

- A is some matrix.
- size of matrix, tells no. of Eigen values

e.g. 3×3 , then 3 eigen values

Greatest Eigen value is Eigenvector centrality

~~compute~~ A is adjacency matrix here.

Katz centrality

- almost same as Eigen vector centrality.
- quality (who) and quantity (how many) matters
- Also considers Attenuation in influence.

Bonacich centrality

Two parametric centrality measure $c(\alpha, \beta)$.

α is normalisation factor

β can be +ve or -ve.

connected to powerful $\xrightarrow{\text{connected to weak node.}}$

Eigen vector centrality works well for strongly connected network.
but not for weakly connected network.

Eigenvector Centrality Example 1 - Exam Question

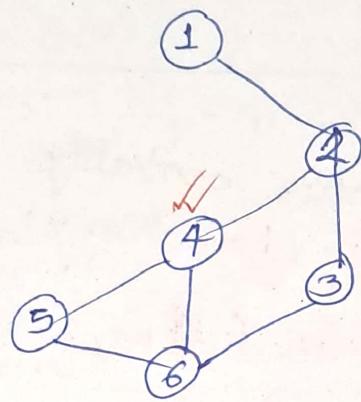
Solve $A - \lambda I$.

$$\lambda = 2, 1, 2.54, 1.5 \\ \text{max.}$$

$$Av = 2.54v$$

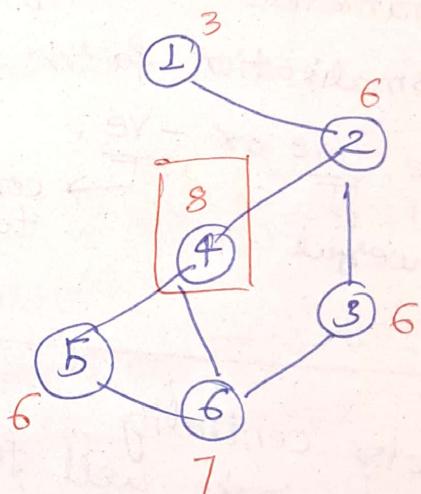
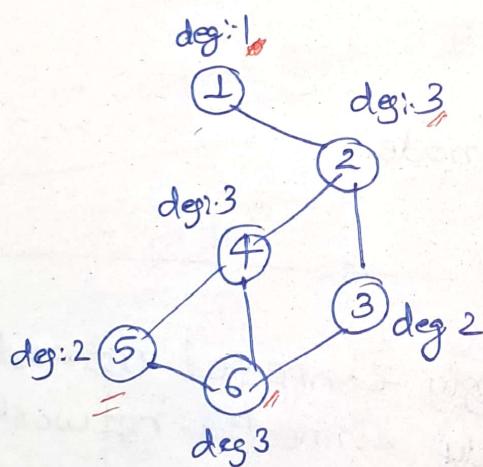
choose eigen vector v , with highest eigen value.

$$v = \begin{bmatrix} 0.31 \\ 0.79 \\ 0.69 \\ 1.00 \\ 0.78 \\ 0.97 \end{bmatrix}$$



(use: Matrixcalc.org
Matrix calculator)

Trick: add degree of your friends.



new power = sum of power of neighbours

Intuitive Approach (Power iteration method)

net power \leftarrow sum of power of neighbours

	0	1	2	3	4	normalise (divide by max, here 52)
1	1	1	3	6	17	0.32
2	1	1	3	6	17	0.73
3	1	2	6	13	37	0.71
4	1	3	8	19	52	1.00
5	1	2	6	15	39	0.75
6	1	3	7	20	47	0.90

similar
as
eigen vector

$$v = \begin{bmatrix} 0.31 \\ 0.79 \\ 0.69 \\ 1.00 \\ 0.78 \\ 0.97 \end{bmatrix}$$

why stop at round 4? (no ~~centered~~ search at Home).
 because at round 5, ~~maybe~~ instead of node 4,
 node 6 will

	5
1	38
2	106
3	85
4	124
5	99
6	128

Stop?

$$X_t = A^T X_{t-1}$$

Stop when new-old < threshold.
 or do this until highest eigen value.

Why iterations?
 because w.r.t time, friends and connections will change

Lets check

	0	1	2	3	4
1	1	0.33	0.375	0.3	0.28
2	1	1	0.75	0.85	0.73
3	1	0.66	0.75	0.65	0.71
4	1	1	1	0.95	1
5	1	0.66	0.75	0.25	0.75
6	1	1	0.875	1	0.90

reaching eigen
value.

stopping criteria

$$1) L_2 \text{ norm } \sqrt{\sum x^2}$$

$$(t_t - t_{t-1}) \leq \delta$$

2) Fluctuation / oscillation.

$$3) |X_t - X_{t-1}| < \delta$$

4) Rayleigh Quotient

H.W

Read Tutorial on PCA by Li Smith.

$$\text{cov}(x, x) = \text{var}(x).$$

$$\text{cov}(x, y) = \text{cov}(y, x)$$

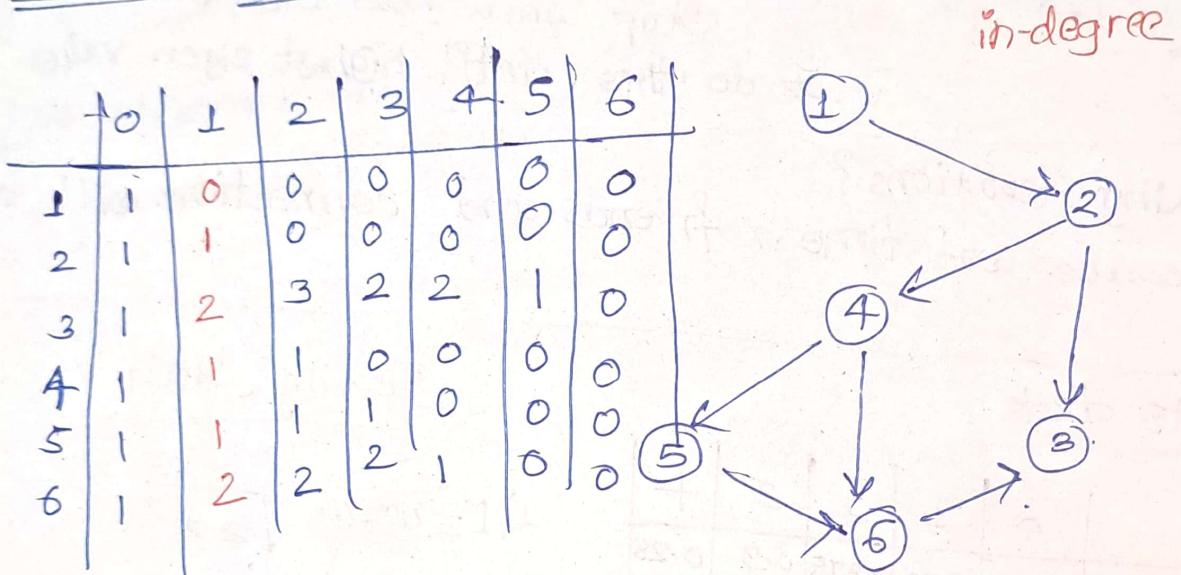
For directed
→ indirected

In links

$$A \xrightarrow{\text{in links}} AX_t = X_{t+1}$$

$$A^T X_t = X_{t+1}$$

calculate Eigen vector centrality for directed graph:-



* acyclic digraph.

• all power leaks out.

citation networks are acyclic, so eigen vector centrality is useless.

Adaptation for directed networks:-

$$X = \underbrace{\alpha AX \cdot \mathbf{B}}_{\text{power from neighbours}} + \underbrace{\mathbf{B}^\perp}_{\text{intrinsic power}}$$

power from neighbours

when this converges, we get:-

$$X = \mathbf{B}(1 - \alpha A^T)^{-1} \mathbf{I}_1$$

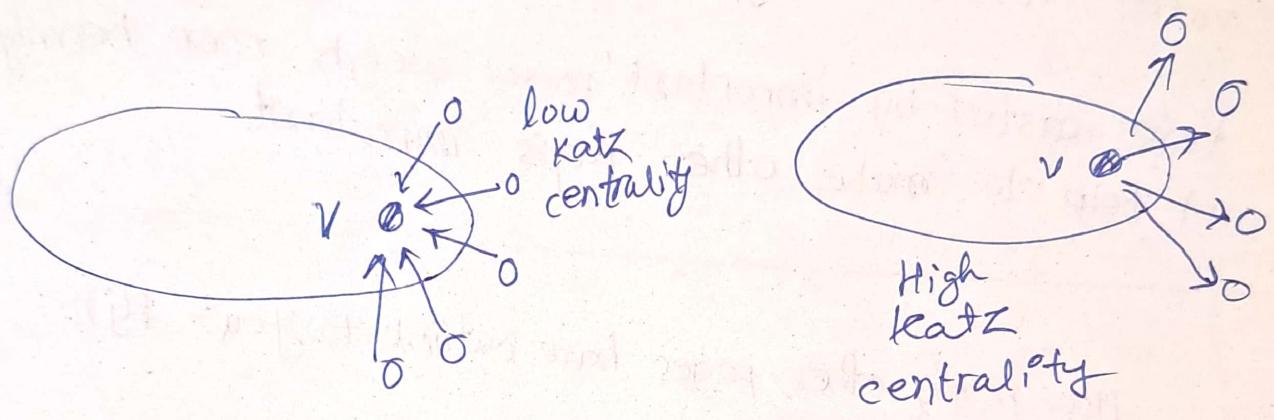
$$X = (1 - \alpha A^T)^{-1} \mathbf{I}_1$$

Ratz centrality
 $\beta = 1$

- If $\alpha=0$, $x=I$, centralities = 1
- As $\alpha \uparrow$, centralities increase
- At $\frac{1}{\lambda_1}$, (λ_1 is largest eigen value) things diverge.

Hence

$$\left[0 \leq \alpha \leq \frac{1}{\lambda_1} \right]$$



PageRank is most powerful measure.

Next class - 1, 2, 3 chapter.

Eigen, Katz, bonachich.

Hard copy Quiz.

Note:- We stop when Eigen vector values converge.

PAGE RANK

For web link analysis models.

PageRank interprets a hyperlink from page x to page y as a vote, by page x , for page y .

However, Page Rank looks at more than sheer number of votes; it also analyses the page that casts the vote.

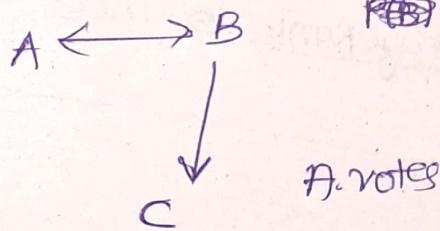
Notes casted by "important" pages weigh more heavily and help to make other pages important.

How many other pages have outlink to you :- $P(j)$

How many votes they give :- O_j

B. votes = 2.

$$P(A) = \frac{P(B)}{2} = 1.5.$$



$$P(B) = \frac{P(A)}{1} = 1.5$$

A. votes = 1

Let's assume $P(B) = 3$.
Then calculate $P(A)$ & $P(C)$.

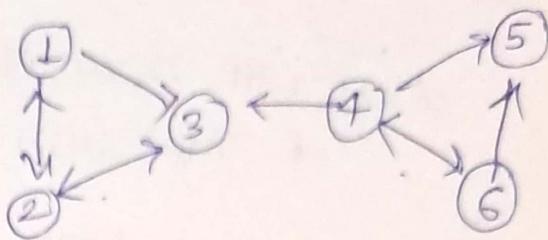
If your network has low rank pages,
Your page Rank will also decrease."

Solving Page Rank

$$P_{i+1} = A^T P_i$$

Stopping criteria $\Rightarrow P_{i+1} - P_i < 0.1$.

$$\bar{A} = \begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} & 0 & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 & 0 & 0 \\ 0 & \frac{1}{2} & 0 & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{3} & 0 & \frac{1}{3} & \frac{1}{3} \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} & 0 \end{bmatrix}$$



Let $P_0 = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \end{pmatrix}$

$$A^T = \begin{bmatrix} 0 & \frac{1}{2} & 0 & 0 & 0 & 0 \\ \frac{1}{2} & 0 & 1 & 0 & 0 & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 & \frac{1}{3} & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & \frac{1}{2} \\ 0 & 0 & 0 & \frac{1}{3} & 0 & \frac{1}{2} \\ 0 & 0 & 0 & \frac{1}{3} & 0 & 0 \end{bmatrix}$$

$$P_1 = A^T P_0$$

$$P_2 = A^T P_1$$

\vdots

$$P_g = A^T P_8$$

$$P_8 = \begin{bmatrix} 4.518 \\ 8.898 \\ 6.784 \\ 0.210 \\ 0.315 \\ 0.182 \end{bmatrix} \quad P_g = \begin{bmatrix} 4.501 \\ 9.096 \\ 6.830 \\ 0.173 \\ 0.213 \\ 0.122 \end{bmatrix}$$

$P_g - P_8 < 0.1$, so stop.

$\therefore \text{Rank} :- 2, 3, 1, 5, 4, 6$

Irrespective of P_0 , same Rank will be achieved.

How compare importance of pages.

start with normalisation.
if Total sum = 1

∴ 6 pages, Each Rank is $\frac{1}{6}$.

Total sum = 1.

∴ Updated page rank :-

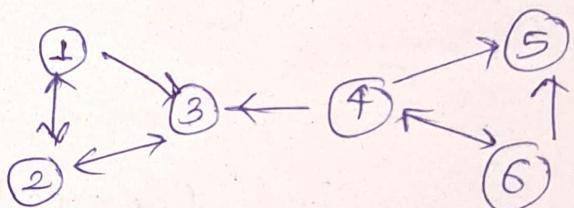
$$p(i) = (1-d) + d \sum_{(j, i \in E)} \frac{p(j)}{o_j}$$

Also $d = 0.8$
is recommended.

$$= (1-d) + d(A^T P)$$

Ex- Let $P_0 = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \end{bmatrix}$

$$d = 0.8$$



~~Page 2~~ ~~Iteration 1~~

$$p(1) = (1 - 0.8) + 0.8 \left(\frac{2}{2}\right) = 1$$

$$p(3) = (1 - 0.8) + 0.8 \left(\frac{1}{2} + \frac{3}{2} + \frac{4}{3}\right) = 2.88.$$

$$p(5) = (1 - 0.8) + 0.8 \left(\frac{1}{2} + \frac{3}{2} + \frac{4}{3}\right) = 2.88.$$

Likewise

Ranking : $[5, 2, 3, 4, 1, 6]$
Iteration 1)

Iteration 2) $\rightarrow [2, 3, 1, 5, 4, 6]$.

which is answer.

Next improvement (improved normalisation)

$$P = (1-d) \frac{1}{n} + d \left(\frac{PR(i)}{\alpha(i)} + \frac{PR(j)}{\alpha(j)} + \dots + \frac{PR(n)}{\alpha(n)} \right)$$

cutpoint - delete node to increase components

bridge - delete edge, to increase components

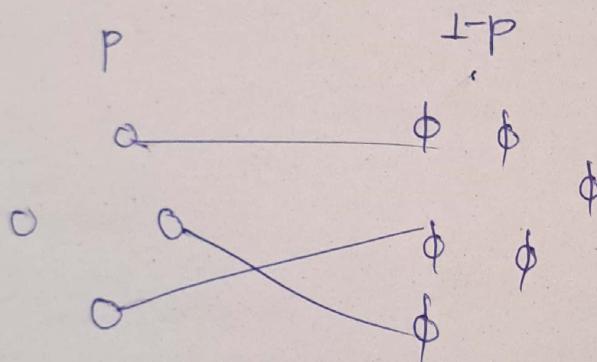
Homophily :- Birds of feather flock together.

Netlogoweb \rightarrow social science \rightarrow segmentation.

consider random network $G(V, E^r)$
where each node is assigned ~~not~~ male: p
female: $1-p$.

Let $G = (V, E)$ be random sample of R with p fraction of male and $1-p$ fraction of female.

- consider any edge $(i, j) \in E^r$ of random network R .



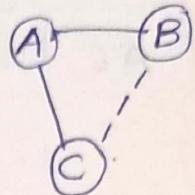
$$E(X^2) - (E(X))^2 = p(1-p) + (1-p)p = 2p(1-p).$$

Computer networks :- Hoax calls, mails (phishing links)

uptill Homophily mid-sem

Social vs selection Influence

- * Triadic closure.
- * Focal closure
- * membership closure



* community Homophily :-

Asian origin vertices:- 5, 7, 8, 10, 11, 12.

Caucasian origin vertices :- 1, 2, 3, 4, 6, 13, 14.

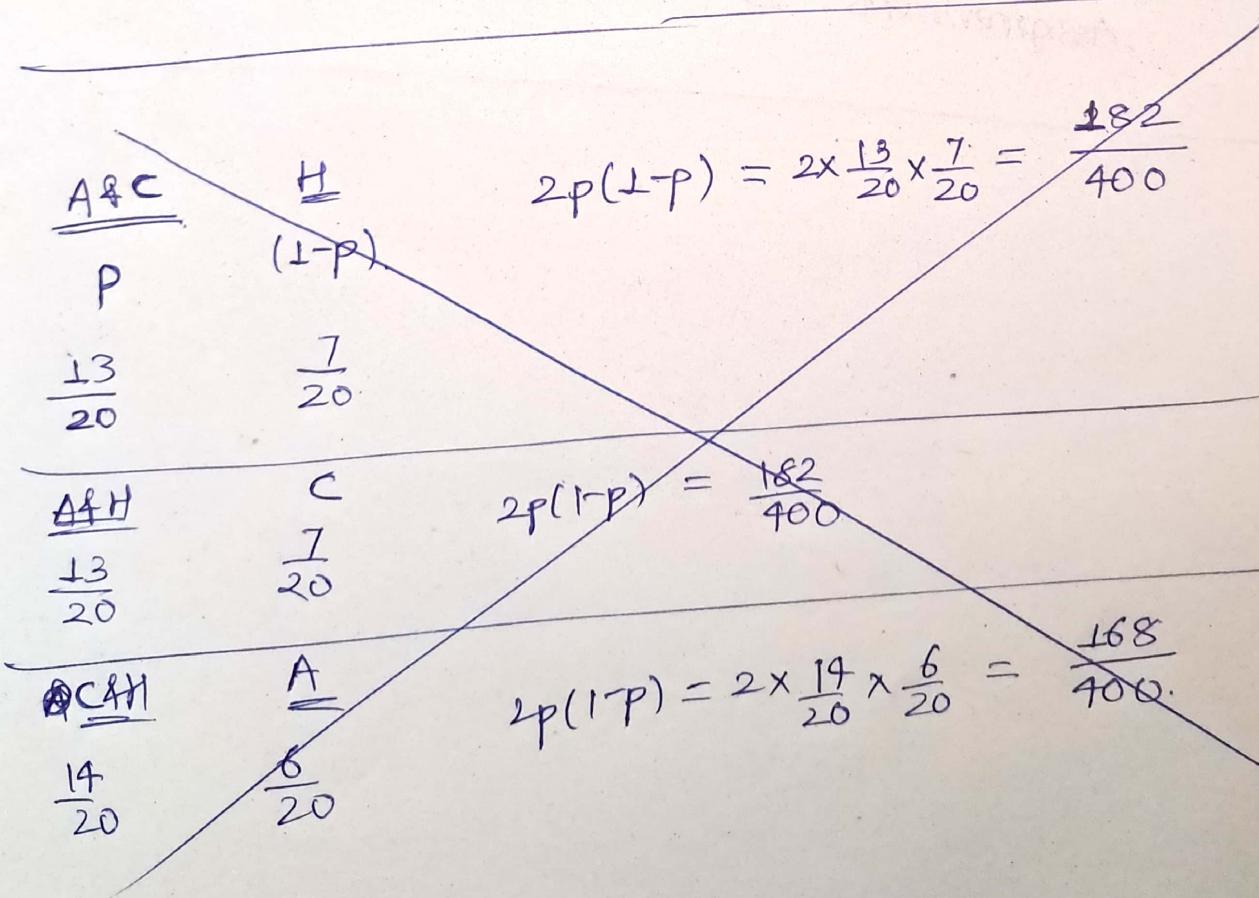
Hispanic origin vertices:- 9, 15, 16, 17, 18, 19, 20.

$$A : \frac{6}{20} = 0.3$$

$$C : \frac{7}{20} = 0.35$$

$$H : \frac{7}{20} = 0.35.$$

$$\# \text{Links} = \frac{\text{sum of degrees}}{2} = 47$$

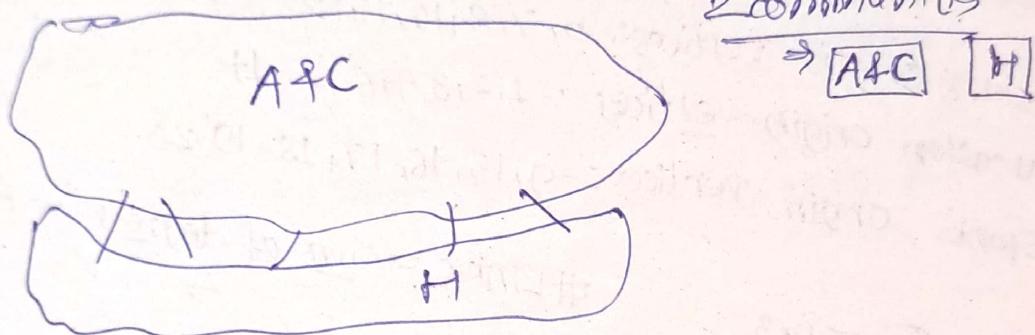


consider any 2 groups.

$$\text{Fraction of A-C Links} = \frac{12}{47} = 0.26$$

$$> 2 * A * C = 2 * 0.3 * 0.5 = [0.21]$$

Fraction Final community detected



Mid-sem exam syllabus over

~~Aze ration~~

Homophily

Assignments :- 35 marks