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PART-C

To combine the advantages of matrices and node link diagrams, we use "Matrix Explorer".

The main goal is to visualize the network.

It consists of the following steps:

1. Initiation of Exploration
2. Explore interactively and iteratively
3. Find a consensus in data
4. Present findings

Initiation of Exploration:

- * Matrix representation is readily easy to create and has faster initiation for exploration because the rendering time is low.
- * In matrix representation, the black colored cells imply that there is communication b/w the row and column.

	A	B	C	D	E
A					
B					
C					
D					
E					

Matrix

* In traditional forced layout, the graph groups the connections b/w the nodes cross, there is overlapping in the nodes and it is not visually readable.

* Dense networks are present and peripheral nodes might have less connections.

- * In matrices, this readability isn't an issue
- * If dense networks, then many black cells are present. Gray cells represent that there are no connection between the nodes.
- * If properly ordered it could help in exploring many insightful details.
- * A particular row or column in black cells implies there is a link from an institution / organization to another institution.
- * Iterative Exploration: ~~obtaining a picture of the exploration process in itself~~
- * ~~The exploration process in itself is iterative.~~ To make it iterative, we manipulate the visualizations.
- * Both node-link and matrix representation is allowed.

~~It includes the following for interaction:~~

- * ~~Iteratively find the visual characteristics~~
- * ~~Grouping by color, label, hierarchy is used~~
- * ~~Iterative layout and ordering~~
- * ~~Manually specifying where the node position must be placed for~~
- * ~~Automated layout and ordering~~
- * ~~Using computer algorithms to specify layout and ordering~~

- * computer assisted layout
- * subset of the network
- * Filtering: Filtering or a text filter
- * Combining: Joining similar sets +
- * Overview + Focus: ~~Re~~
- * Focus Content: Use binoculars for reading
- * Tree map: Used for hierarchical structures
- * Filtering is used to find consensus in data

Find consensus in data

- * Different visualizations
- * When reading, I have different perspectives / clusters
- * Solution is to use colors that are present
- * If a node's members are different, then differentiate them by

- * Computer assisted layout and ordering: only a subset of the network uses layout and ordering algorithms for specific types of nodes and edges. It is often combined with a clustering step.
 - * Filtering: Filtering or removal of elements based on value typed in a text field or from a list.
 - * Combining: Joining / aggregating the elements that are similar.
 - * Overview + Focus: Reading matrices, treemaps, filters etc and used for navigation and zooming.
 - ↳ Focus Content: Use birds eye for overview of network and magnify details.
 - ↳ Treemap: Used for macrostructure of network diagram.
 - ↳ Filtering is used to find communities within the network.
 - Find consensus in Data:
 - * Find consensus in Data:
 - * Different visualizations provide with ~~set~~ insights in data.
 - * When reading, each time matrices provide a different perspective / cluster.
 - * Solution is to iteratively do reading, and check clusters against all representations.

Finding Correlation in Data:

- Finding consensus in Data -

 - * Different visualizations provide with ~~best~~ insights in data.
 - * When reading, each time matrices provide a different perspective / cluster
 - * Solution is to iteratively do reading, and check clusters that are present in almost all representations.
 - * If a node's membership is doubtful, then could differentiate them by using lighter colours.

Present Findings: ~~Matrix representation is helpful if we want overview of the entire network or keep it simple when we have to present to a wider audience.~~

- * Matrix representation is helpful if we want overview of the entire network or keep it simple when we have to present to a wider audience.
- * Node-link is useful when we have to work with multiple views of node-link representations.

a. Depth first search from a node

```
import networkx as nx
import matplotlib.pyplot as plt
```

```
G = nx.generators.small_bachman_kite_graph(7)
```

a. Display adjacency list

```
print(G.adj)
```

b. Print neighbors of each node

```
for node in G.nodes():
    print("The neighbours of", node, "are")
```

```
    print(G[node])
    print(G.neighbors(node))
```

c. Depth first search with node 0 as starting point

```
import networkx.algorithms.traversal
```

```
def dS-find(G, start, visited=[], edges={}):
    visited.append(start)
```

```
    print(start)
```

```
    for neighbour in G[start]:
```

```
        if neighbour not in visited:
            edges.append((start, neighbour))
            visited.append(neighbour)
            dS-find(G, neighbour, visited, edges)
```

edges.append((c, e))

dS-find(G, n)

return edges

dfs(G, 0)

d. Shortest path from n

print(nx.algorithms.shortest_path(G, source=n, target=t))

e. Average shortest path

print(nx.algorithms.average_shortest_path_length(G))

f. Depth first search

print(DFS(G, start))

print(BFS(G, start))

`edges.append((c.start, neighbor))`

8-TAAQ

`dfs-find(G, neighbor, visited, edges)`

`return edges`

`dfs(G, 0)`

d. Shortest path from node 0 to 7.

`print(nx.algorithms.shortest_path(G, 0, 7))`

e. Average shortest path

`print(nx.algorithms.average_shortest_path(G))`

f. Depth first search (Note: this uses built-in methods like `list` and `len`)

`def dfs(G, start):`

`stack = [start]`

`while stack:`

`node = stack.pop()`

`if node not in visited:`

`visited.add(node)`

`for neighbor in G.neighbors(node):`

`if neighbor not in visited:`

`stack.append(neighbor)`

`return visited`

`print(dfs(G, 0))`

`print(len(dfs(G, 0)))`

`print(dfs(G, 0) == nx.bfs_tree(G, 0).nodes())`

`print(dfs(G, 0) == nx.dfs_tree(G, 0).nodes())`

`print(dfs(G, 0) == nx.dfs_preorder_nodes(G, 0).nodes())`

`print(dfs(G, 0) == nx.dfs_postorder_nodes(G, 0).nodes())`

`print(dfs(G, 0) == nx.dfs_labeled_nodes(G, 0).nodes())`

`print(dfs(G, 0) == nx.dfs_breadth_first(G, 0).nodes())`

`print(dfs(G, 0) == nx.bfs_labeled_nodes(G, 0).nodes())`

`print(dfs(G, 0) == nx.bfs_width(G, 0).nodes())`

`print(dfs(G, 0) == nx.bfs_level_order(G, 0).nodes())`

`print(dfs(G, 0) == nx.bfs_level_order_reversed(G, 0).nodes())`

`print(dfs(G, 0) == nx.bfs_level_order_descending(G, 0).nodes())`

`print(dfs(G, 0) == nx.bfs_level_order_descending_reversed(G, 0).nodes())`

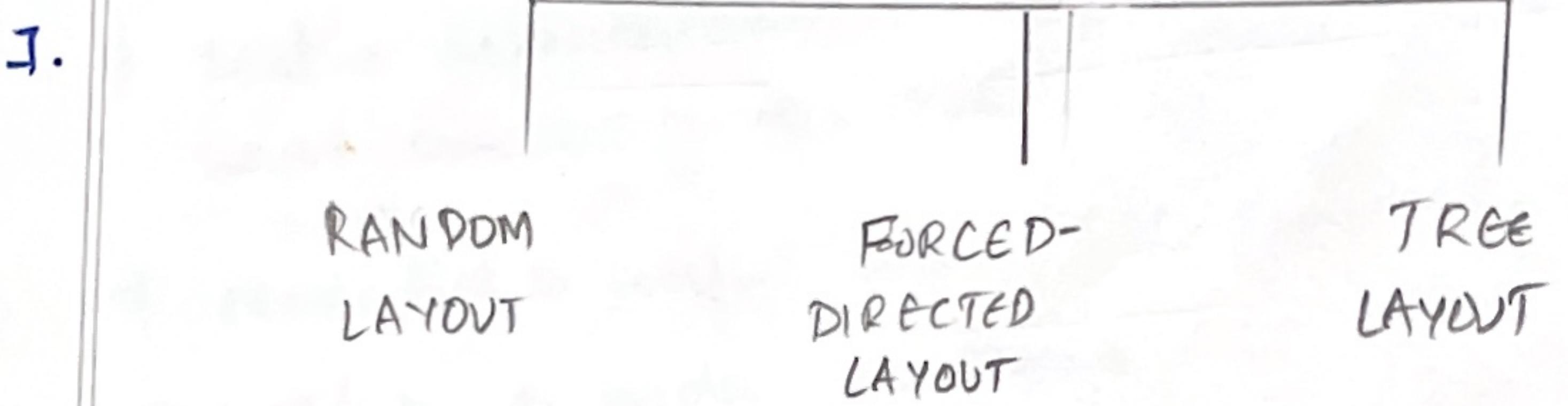
`print(dfs(G, 0) == nx.bfs_level_order_descending_descending(G, 0).nodes())`

`print(dfs(G, 0) == nx.bfs_level_order_descending_descending_reversed(G, 0).nodes())`

PART-B:

((random, world)) keeps nodes

(nug, between random, id) binds it's

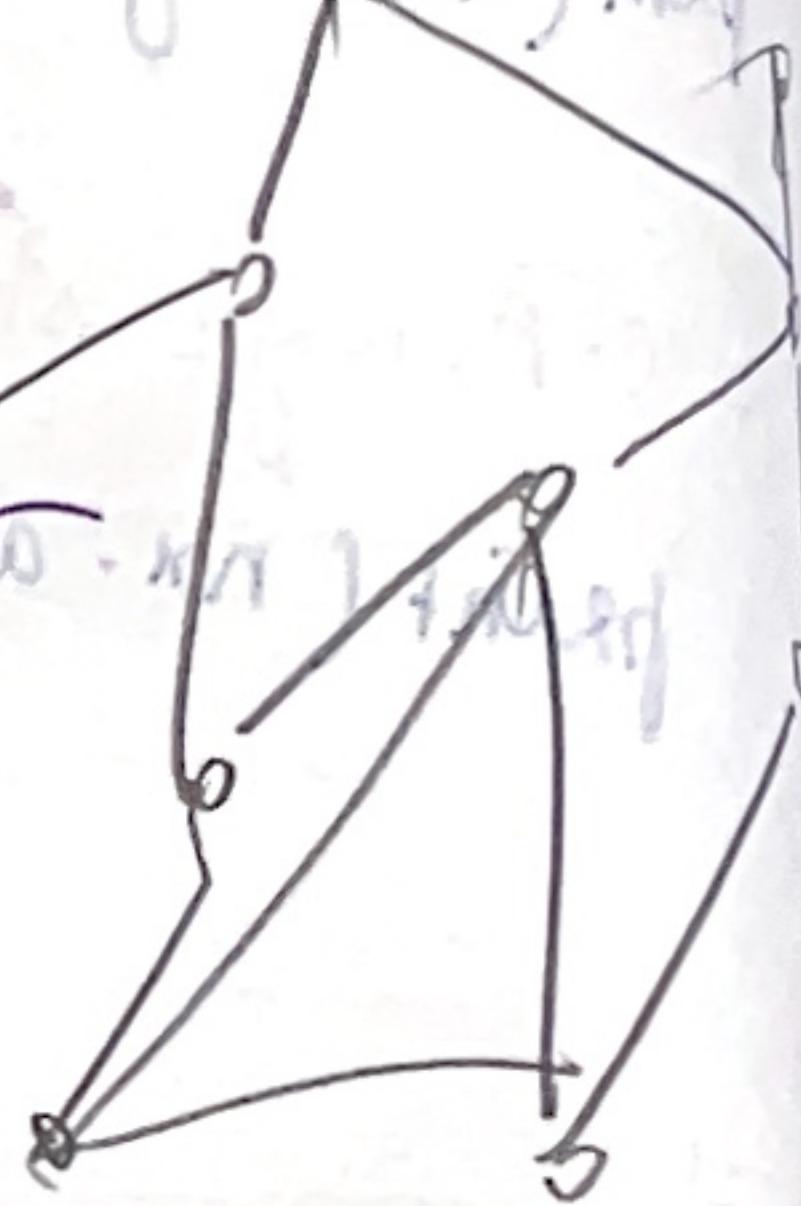


up to nodes

(0, n) is it

Random Layout:

- * Nodes are present in random geographical positions.
- * Links are also in randomised manner.
- * Cannot be used for more than 1000 nodes
- * The time for drawing the network is linear with number of nodes $\approx O(n)$
- * Can be used for large networks



Forced-directed Layout

- * Nodes are like repelling objects and the edges are like springs

- * They follow Hooke's and Coulomb's law

- * Initially it starts as a random layout



- * The nodes then
- * This process is known as
- * It makes
- * It takes no

TR

is

- * It consists of nodes connected to other connected to other connected to other nodes.

- * There is a structure
- * Further divided into

extremes like

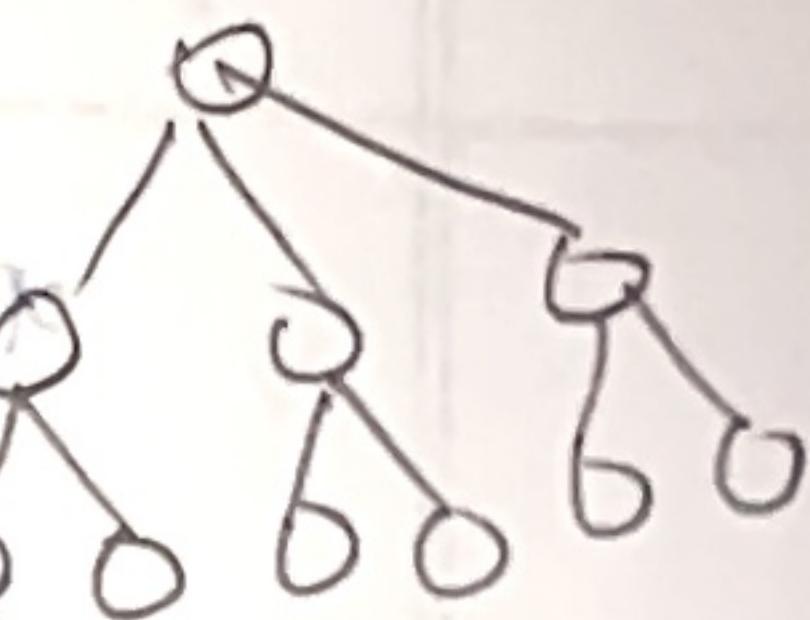
- * It uses forces + consideration
- * Both local and global

User centric visualization
functions
+ Heuristics

- * The nodes then position themselves in various iteration
- * This process is continued until the force of attraction between nodes remain unchanged
- * It makes better use of the space layout
- * It takes $O(N \log N)$ or $O(E)$ time and is therefore not suitable for more than 100 nodes.

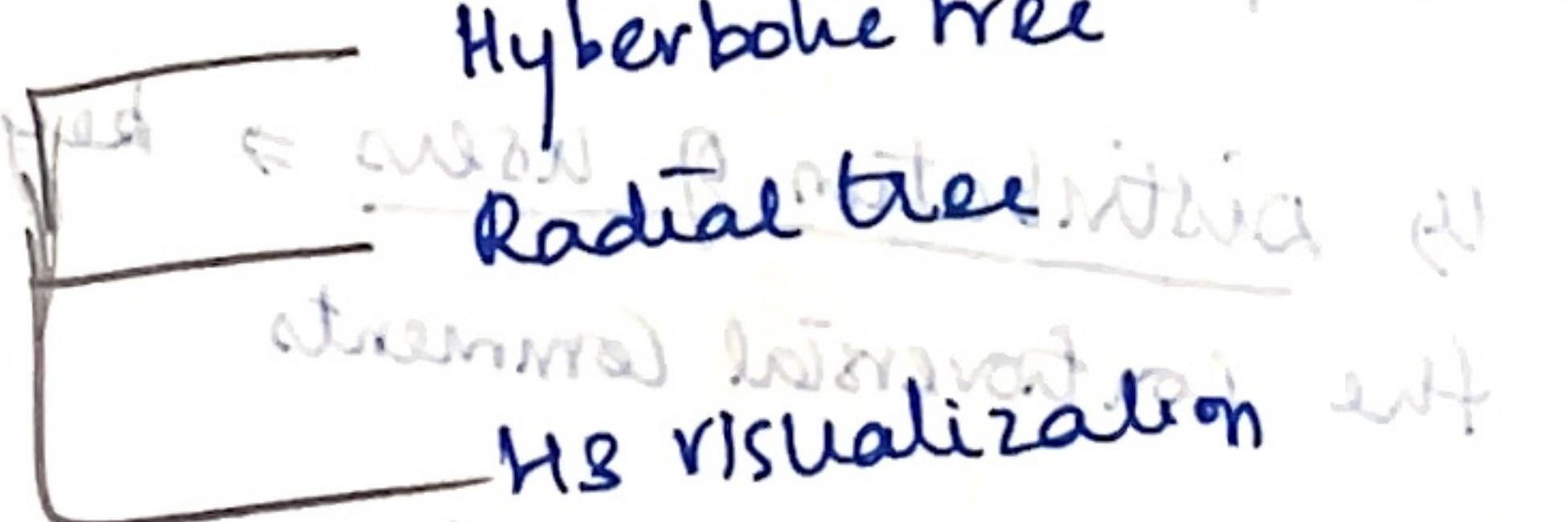
Tree Layout:

- * A tree layout is more grasps to human's eye than a normal graph layout



- * It consists of a root node, its child node and the nodes connected to it as children nodes and the nodes connected to the children node as grand children nodes.

- * There is a structure and a context in the layout
- * Further divided into



- * It uses focus + context + animation consideration
- * Both local and global view in 2D and 3D is possible

- 5/1
- visualization
functionalities:
8. User centric visualization
- * Here the characteristics of a person is presented and also the subjects and relationships of interest are presented
 - * The user centric functionalities of visualization could include
 - groups users with similar features
 - key actors with high impact
 - people with high social relationships or active social interactions.
 - * The user centric approach allows people to access and discover social networks of their choice.
9. Content centric visualization:
- * Various different types of contents are provided to the people who analyse social network
 - ↳ distribution of users → key user opinions and the controversial comments
 - ↳ users who big high impact to social communities, especially in the near future
 - ↳ Relationships between different content groups

Hybrid Visualization

- * Usually social network is really diverse and cannot be done with only few approaches
- * There is a need for hybrid approach that combines different aspects of attributes like Content and people's ways
- * This is especially of importance in daily and email services

Matrix based

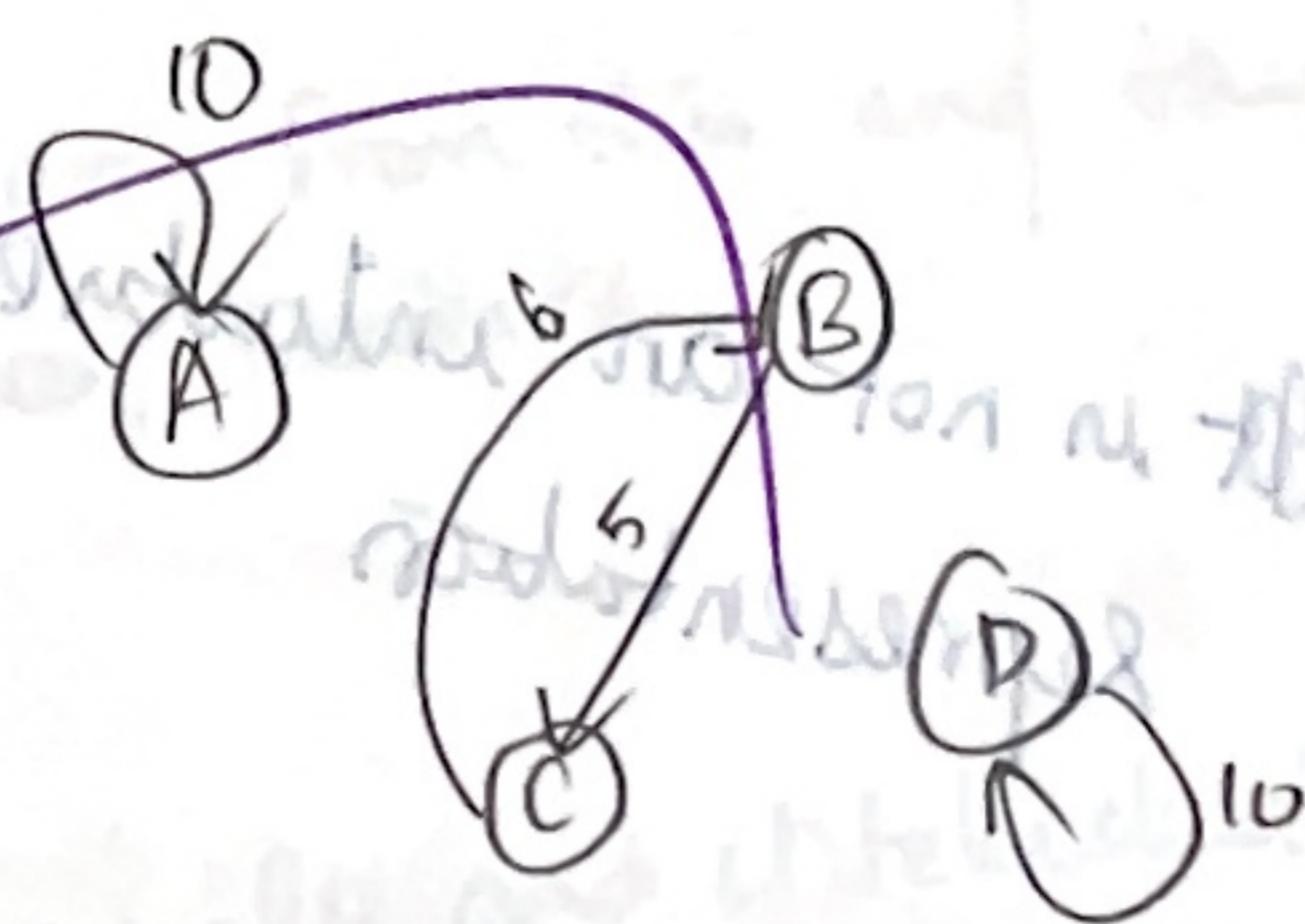
- * Nodes are represented in rows and columns
- * The values between in the cell represent connection between the nodes.

	A	B	C	D
A	10	0	0	0
B	0	5	0	0
C	0	6	0	0
D	0	0	0	10

The weight are specified in the matrix

Node Link Diagram

- * Nodes are represented as circles labeled with their names
- * Edges are lines joining the nodes and the weights are present on the lines



- * Matrix representation is suitable for large networks
 - * Used for dense networks
 - * Has low initiation time for exploration
 - * cannot be used for path finding
 - * No problem of ~~edge~~ node overlapping and cross links
 - * It is not an intuitive representation
- * Node link representation is suitable for small network or for global & compact representation of networks
- * Generally suitable for sparse networks
- * Has high initiation time for exploration
- * can be used for path finding
- * Node overlapping and cross links make it difficult to understand the representation (it) it becomes unreadable
- * It is an intuitive representation that can be given for a wide audience.

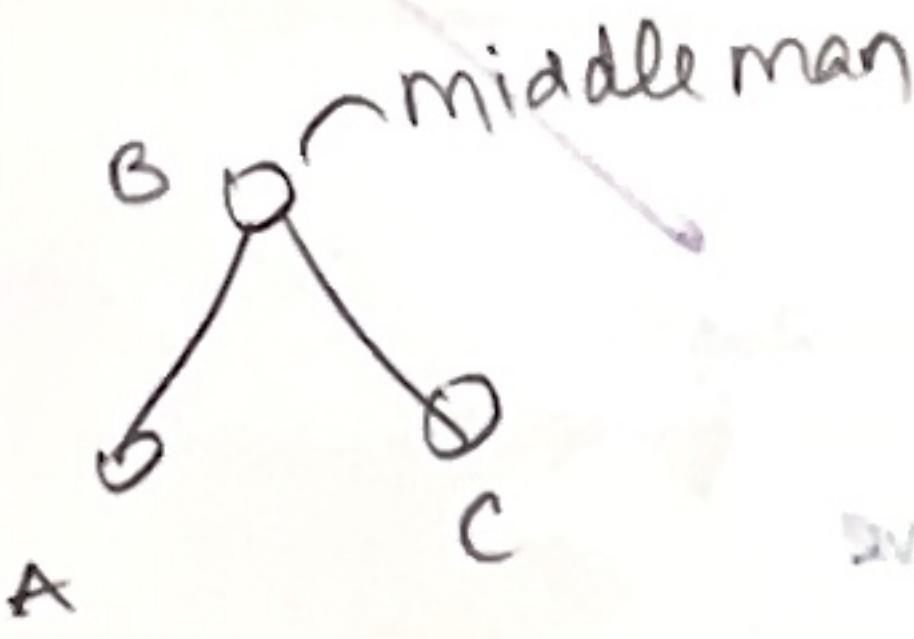
PART-A:

1. A triad is a set of 3 nodes that could be interlinked in any ways. If directed there are 16 different ways of arranging 3 nodes.

Triadic analysis is to find how many times a node has repeated one of the 16 patterns in its network.

e.g. The hijacker pilot of 9/11 attack had 300 such connections.

2. A forbidden triad is the one in which 2 ends don't communicate directly with each other and requires a middleman.



If A and C have to communicate, it has to go through B -> middleman

Significance of forbidden triads in network

* This middleman could gain from this and they can take forms as broker for houses, brokers, entrepreneurs etc.

* Bridges gap between socially and intellectually different communities e.g. Musicians, scientists, musicians and students.

3. Visualization is an important concept in SNA.

: A-T9A9

* It helps in linking human vision to

computer

* It helps in pattern finding between

* It helps to get meaningful insights out of data

* It helps in finding structural information

4. For matrix based visualization, the following are best suited :-

* Large Networks that require low computation time.

* Networks with node overlap

* Networks with cross links

* Directional data

* High Density networks

* Network that do not have path finding applications

5. The 3 popular centrality measures are:

By Degree, Betweenness, Closeness

Degree Centrality * It is the number of edges a node is incident on.

* For directed graph in-degree and out-degree centrality is present

* For undirected, it is degree centrality

Betweenness Centrality: * It is a measure of the extent to which a node is present between other nodes.

- * The extent to which it acts as a bridge node for any as the shortest path.

* High betweenness \Rightarrow crucial for information passing

Closeness Centrality:

* It is a measure of how near other nodes are to the node.

* Calculated w.r.t mean shortest path

* If the closeness centrality is high, then it is used for distribution of information and is a very crucial node.

Advantages of Node Link Diagrams:

* It can be presented to a wider audience as the representation is intuitive.

* Can be used for also small / sparse networks

* Can be used for compact representation of network

* Can be used for path finding applications.