# Complex-Network Modelling and Inference <br> Lecture 4: Graph connectivity and traversal 

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## Section 1

## Connectivity

## Connectivity

## Definition

Two nodes are connected if a path exists between them.

## Definition

A graph is connected if all pairs of nodes are connected.

## Definition

A strongly connected digraph is connected in the sense above, whereas a weakly connected digraph is connected if we include all reverse links.

## Definition

A graph is $k$-edge connected if the graph remains connected after the removal of any set of $k-1$ edges, and $k$-node connected if the graph remains connected after the removal of any set of $k-1$ nodes.

## Definition (Cut)

A cut is a partition of the nodes of a graph into two subsets $C=(S, T)$.

## Definition (Cut-set)

The cut-set of a cut $C=(S, T)$ is the set of edges

$$
\{(u, v) \in E \mid u \in S, v \in T\}
$$

i.e., the edges that cross the cut.

## Definition (Edge Cut)

An (minimum) edge cut is the minimum number of edges whose removal disconnects two nodes $i$ and $j$, i.e., a minimal cut-set with $i \in S$ and $j \in T$.

## Menger's theorem

Theorem (Edge-connectivity version)
For an undirected graph $G$, the size of the minimum edge cut for an arbitrary pair of nodes $i \neq j$ is equal to the maximum number of edge-disjoint paths from $i$ to $j$.

- Edge-disjoint means they share no common edges
- There is also a node connectivity version
- It also works for digraphs and infinite graphs
- The theorem is generalised in many optimisation algorithms: e.g., maximum flow algorithms.


## Connected Components

- A connected component is a maximal connected subgraph

- The set of connected components $\left\{C_{i}\right\}$ form a partition of the nodes


## Definition (Partition)

A partition is a set of covering and disjoint subsets, i.e, $\left\{C_{i}\right\}_{i=1}^{n}$ is a partition of $C$ iff

$$
\bigcup^{n} C_{i}=C \quad \text { and } \quad C_{i} \cap C_{j}=\phi, \quad \forall i \neq j
$$

## Connected Components Algorithm

Data: A Graph $G=(N, E)$
Result: A set of connected components $\left\{C_{i}\right\}$
1 Initialise $N^{\prime}=N$;
2 while $\left(N^{\prime} \neq \phi\right)$ do
3 Choose a node $i \in N^{\prime}$ and delete it from $N^{\prime}$;
$4 \quad$ Set $C_{i}=\{i\}$ and $L=\{(i, j) \mid(i, j) \in E\}$;
$5 \quad$ while $(L \neq \phi)$ do
6
Choose a link $(k, m) \in L$;
if $m \notin C_{i}$ then
add $m$ to $C_{i}$;
delete $m$ from $N^{\prime}$; add all links $(m, I) \in E$ to $L$;
end
delete $(k, m)$ from $L$;
end
14 end

## Connected Components Example



## Connected Components Example



## Connected Components Example



## Connected Components Example



## Connected Components Example



## Connected Components Example



## Connected Components Example



## Connected Components Example



## Application 1

Key requirements for critical infrastructure networks (e.g., Internet, Water, Power, ...)

- Reliability


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## Application 1

The simplest definition of "reliability" used in networks is some variant of $k$-connectedness

- The particular variant depends on the failure modes of the network
- do the nodes fail, or the edges (or both)?
- Leads to network designs with redundancy
- not necessarily $k$-fold redundancy
- Is this a good enough definition of reliability


## Application 2

Markov chains probability transition matrix

$$
P=\left(\begin{array}{llllll}
0.0 & 1.0 & 0.0 & 0.0 & 0.0 & 0.0 \\
0.5 & 0.0 & 0.8 & 0.4 & 0.2 & 0.0 \\
0.0 & 0.0 & 0.0 & 0.0 & 0.8 & 0.0 \\
0.5 & 0.0 & 0.0 & 0.0 & 0.0 & 0.0 \\
0.0 & 0.0 & 0.0 & 0.6 & 0.0 & 0.0 \\
0.0 & 0.0 & 0.2 & 0.0 & 0.0 & 1.0
\end{array}\right)
$$

## Application 2

Markov chains are described by a directed graph with self-loops, e.g.,


The transition matrix is just a weighted adjacency matrix.

## Application 2

First step of studying a Markov chain is to check its properties

## Definition

State $i$ is accessible from state $j$ if it is possible to get from $i$ to $j$.
Accessible $=$ a path from $i$ to $j$ exists

## Definition

A Markov chain is irreducible if it is possible to get to any state from any state.

Irreducibility $=$ strong connectivity of the graph

## Definition

A communicating class is a maximal set of mutually accessible states.
Communicating class $=$ connected component

## Section 2

## Graph Traversal

- Connected components as described above has a little vagueness
- when I say "choose" how do you choose?
- It's an example of graph traversal
- where we want to visit each node of a graph (at least once)
- ordered nodes by connectivity
- Traversals used for lots of algorithms
- could be to search for an element
- or to calculate a value for each node

Maybe you don't have the whole graph stored in memory, but have to read bits, e.g., traversing Facebook graph

- There are two main strategies
- Depth-First Search
- Breadth-First Search
- For the sake of simplicity, we will assume graphs are connected
- Easiest to understand in neighbour-list representation


## Depth-First Search

Visit a neighbour's children before you visit the next neighbour
1 Function $\operatorname{DFS}(G, i)$;
Input: A Graph $G=(N, E)$, and start node $i \in N$
2 label $i$ as explored;
3 forall $j \in$ neighbourhood $\{i\}$ do
4 if $j$ is unexplored then
5
6 $\operatorname{DFS}(G, j)$;
end
7 end

- We could make this faster by avoiding edges going backwards.
- At the moment the algorithm doesn't do anything
- a search also checks something about the node, and returns the first one that checks out
- but we might also do some sort of update
- or use to find a connected component ...


## Breadth-First Search

Visit all neighbours before you visit their children

```
    1 Function BFS(G,i);
    Input: A Graph G = (N,E), and start node i N N
2 label i as explored;
3 create queue Q;
4 put i on Q;
5 while Q not empty do
6 take j off the front of Q;
7 forall }k\in\mathrm{ neighbourhood{j} do
8 if k is unexplored then
9
10
11
12
    end
13 end
```


## Depth-First Search Example



## Depth-First Search Example



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## Depth-First Search Example



## Depth-First Search Example



## Depth-First Search Example



## Breadth-First Search Example



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## Further reading I

