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

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## Section 1

## Connectivity

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# 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.

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#### 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

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

*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  $j \in T$ .

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# Menger's theorem

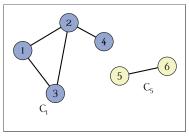
#### Theorem (Edge-connectivity version)

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  $\{C_i\}$  form a *partition* of the nodes

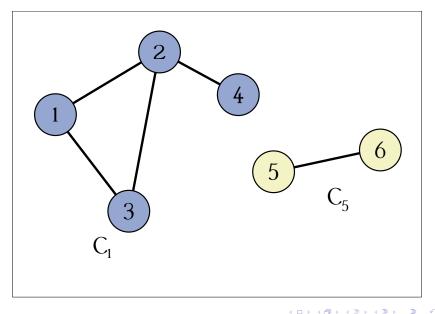
#### Definition (Partition)

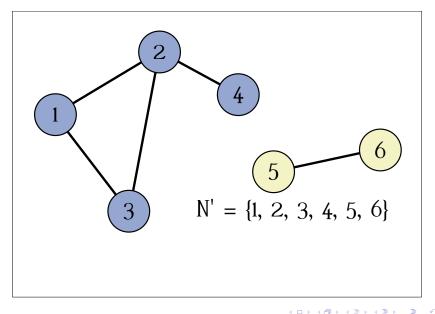
A *partition* is a set of covering and disjoint subsets, i.e.,  $\{C_i\}_{i=1}^n$  is a partition of *C* iff

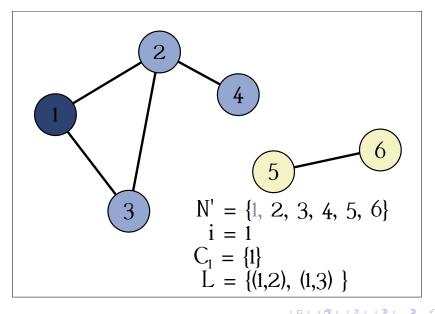
$$\bigcup_{i=1}^{i} C_i = C \quad \text{and} \quad C_i \cap C_j = \phi, \ \forall i \neq j.$$

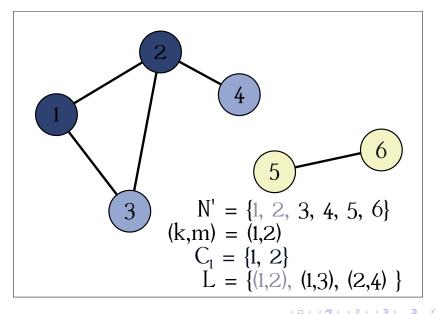
# Connected Components Algorithm

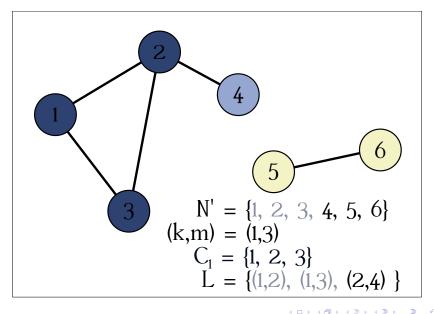
```
Data: A Graph G = (N, E)
   Result: A set of connected components \{C_i\}
 1 Initialise N' = N:
   while (N' \neq \phi) do
 2
       Choose a node i \in N' and delete it from N':
 3
       Set C_i = \{i\} and L = \{(i, j) | (i, j) \in E\};
 4
       while (L \neq \phi) do
 5
           Choose a link (k, m) \in L;
 6
           if m \notin C_i then
 7
               add m to C_i:
 8
               delete m from N':
 9
               add all links (m, l) \in E to L;
10
           end
11
           delete (k, m) from L;
12
       end
13
14 end
```

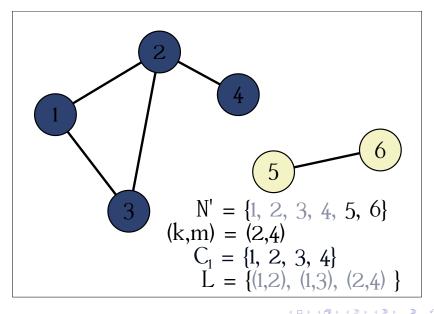


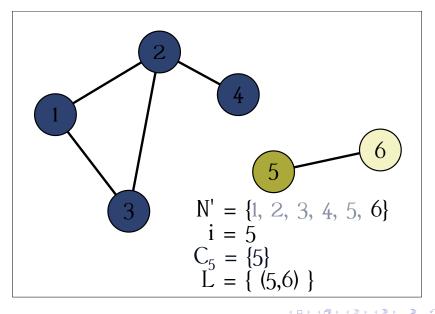


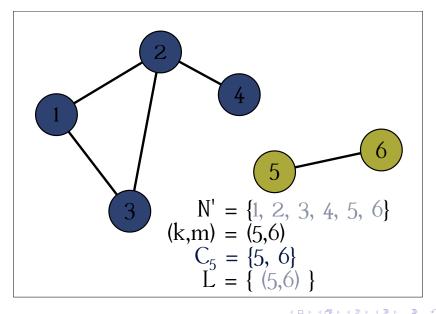












Key requirements for critical infrastructure networks (*e.g.*, Internet, Water, Power,  $\dots$ )

Reliability

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Key requirements for critical infrastructure networks (*e.g.*, Internet, Water, Power,  $\dots$ )

- Reliability
- Reliability

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- Reliability
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Key requirements for critical infrastructure networks (*e.g.*, Internet, Water, Power,  $\dots$ )

- Reliability
- Reliability
- Reliability
- Cost

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Key requirements for critical infrastructure networks (*e.g.*, Internet, Water, Power,  $\dots$ )

- Reliability
- Reliability
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- Performance

Key requirements for critical infrastructure networks (*e.g.*, Internet, Water, Power,  $\dots$ )

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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

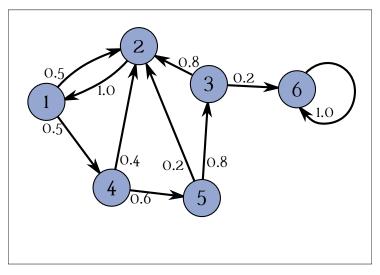
Markov chains probability transition matrix

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

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



The transition matrix is just a weighted adjacency matrix.

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

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- 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

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## Depth-First Search

Visit a neighbour's children before you visit the next neighbour

```
1 Function 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 | DFS(G, j);

6 | end

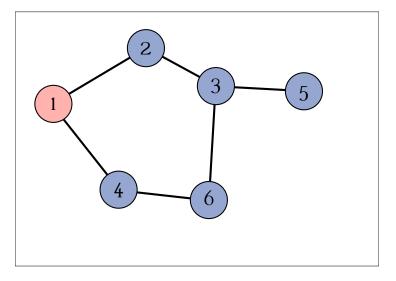
7 end
```

- 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 ...

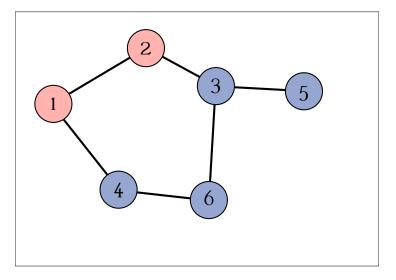
## Breadth-First Search

Visit all neighbours before you visit their children

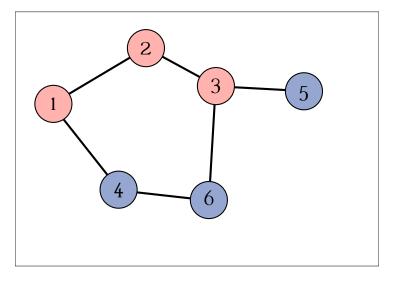
```
1 Function BFS (G, i):
   Input: A Graph G = (N, E), and start node i \in N
2 label i as explored;
3 create queue Q;
4 put i on Q;
  while Q not empty do
5
      take i off the front of Q:
 6
      forall k \in neighbourhood{i} do
 7
          if k is unexplored then
 8
              label k as explored;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



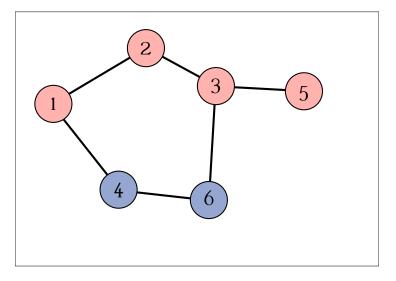
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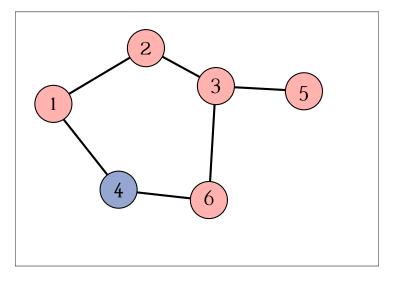


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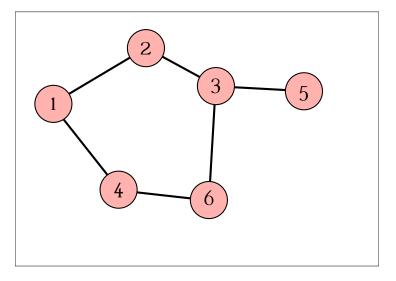
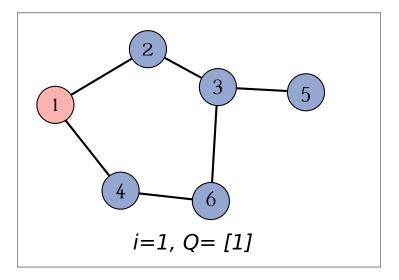


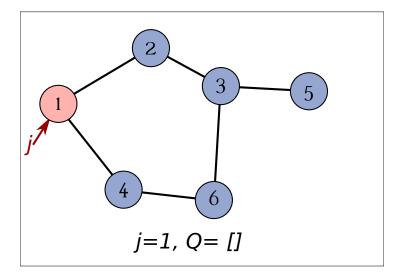
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## Breadth-First Search Example



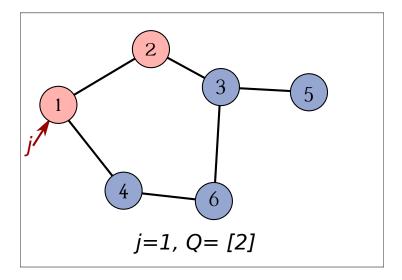
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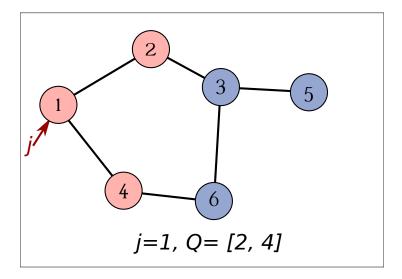


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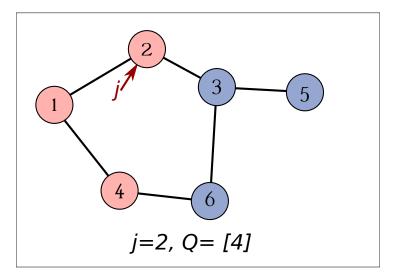


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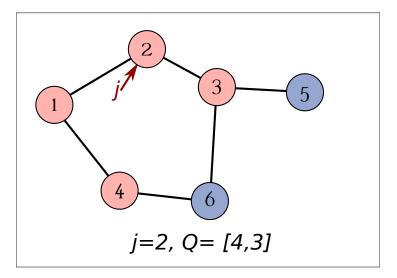
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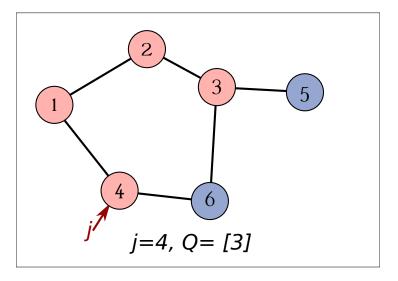
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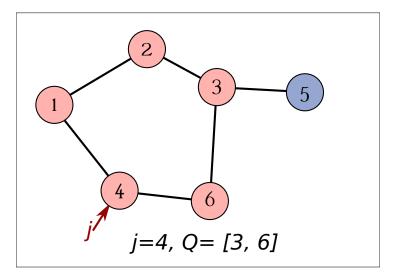
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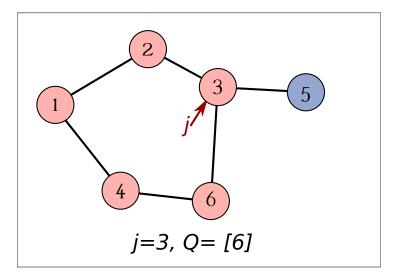
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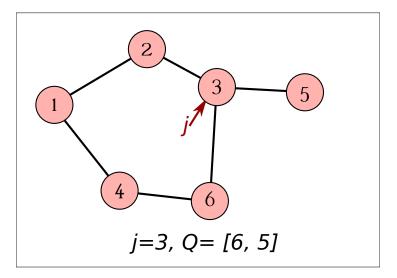


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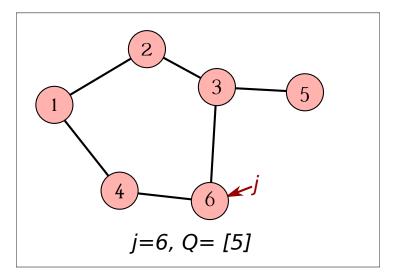


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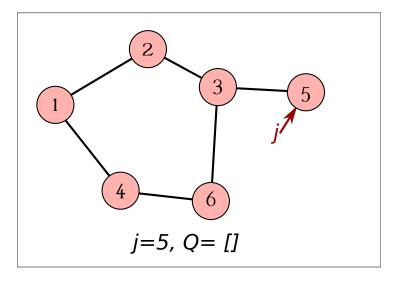
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# Further reading I

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