

### 1.5.6 → Random graphs

→ overview → many of the previous tools can be applied to random graphs.  
→ need to understand graphs first

→ notes → A graph consists of a set of nodes & a set of arcs connecting the nodes.

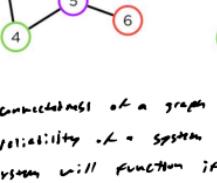
Let  $n$  be the set of nodes &  $A$  be the set of arcs.

A graph has a maximum of  $\binom{n}{2}$  arcs (i.e.  $\binom{n}{2}$  different pairs of nodes).

A graph can be subdivided into subgraphs, called components, where each component consists of non-overlapping connected nodes.

→ A graph is connected if there is only one component. In other words, given any two nodes, there must be a path from one node to the other for the graph to be connected.

→ Examples: list all subgraphs & components of this graph



→ 8 subgraphs:  $\{\{1,2,3,4,5,6\}, \{7\}, \{8\}\}$

5 arcs:  $\{\{1,2\}, \{2,4\}, \{3,4\}, \{4,5\}, \{5,6\}\}$

4 components:  $\{\{1,2,3,4,5,6\}, \{7\}, \{8\}, \{6\}\}$

$\Rightarrow$  not connected

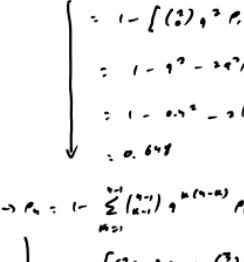
→ Random graphs

→ now consider a graph w/  $n$  nodes, where any two nodes (node & nodes) are not connected with certainty, but w/ probability of  $p_{ij}$ . This graph is known as a random graph. Let  $X_{ij} = 1$  if arc that represents the existence of an arc between node  $i$  & node  $j$ .

$$X_{ij} = \begin{cases} 1 & \text{if } p_{ij} \text{ is true} \\ 0 & \text{otherwise} \end{cases}$$

$$\Rightarrow P(X_{ij}=1) = p_{ij} + P(X_{ij}=0) = 1 - p_{ij}$$

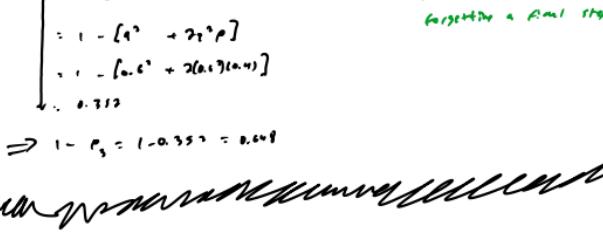
→ suppose we are interested in the probability that a random graph is connected. note that a graph doesn't need to have every node connected to every other node to be considered connected.



→ the connectedness of a graph is similar to the reliability of a system. the reliability of a system depends on the reliability of its components. a system will function if it only has at least one minimal path set functioning.

→ like a system, a graph has minimal path/cut sets. A random graph is connected as long as all sets of at least one minimal path set exist. A random graph w/  $n$  nodes has  $\binom{n-1}{n-2}$  minimal path sets &  $2^{n-1}-1$  maximal cut sets.

w/ regard random graph w/ 3 nodes has  $2^{3-1}-1=3$  minimal path sets



→ let's assume each arc is the fail (success). Let  $P_0$  = P[random graph w/  $n$  nodes is connected]. Then

$$P_0 = 1 - \sum_{k=1}^{\binom{n}{2}} \binom{\binom{n}{2}}{k} p_k^{k(1-p)} (1-p)^{\binom{n}{2}-k}$$

where  $p_k = 1 - p + k - 1$ . Note that this is a recursive formula (the value of  $p_k$  depends on the values of all  $p_i$  from  $i=1, \dots, k-1$ ).

This formula gets complicated quickly

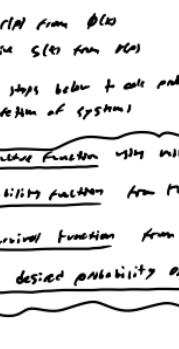
→ to avoid the exact calculation, we just we bound

$$(1-p)^{n-1} = \left(\frac{1}{2}\right)^{n-1} \leq 1 - p \leq (1-p)^{n-1}$$

→ for large  $n$ , it can be shown

$$p \approx (1-p)^{n-1}$$

→ random graph w/  $\binom{n}{2}$  = 28 arcs (on average)



→ calculating bounds is just that & that ...

→ Example: given info on the graph

- 4 nodes
- prob of arc between any two nodes is  $p$
- Every arc is  $\perp\!\!\!\perp$

→ find the prob the graph is connected

→ want  $P_0 \Rightarrow$  know  $P_1 = 1$

$$P_1 = p = 0.6$$

→ first find  $P_2$

$$P_2 = 1 - \sum_{k=1}^{\binom{4}{2}} \binom{\binom{4}{2}}{k} p_k^{k(1-p)} (1-p)^{\binom{4}{2}-k}$$

$$= 1 - [(2)^2 + 3 \cdot 1^2] p^2 = 1 - 2^2 p^2$$

$$= 1 - 0.4^2 = 0.84$$

$$\rightarrow P_3 = 1 - \sum_{k=1}^{\binom{4}{3}} \binom{\binom{4}{3}}{k} p_k^{k(1-p)} (1-p)^{\binom{4}{3}-k}$$

$$= 1 - [(1)^3 + 3 \cdot 2^2 p^3 + 3 \cdot 1^2 p^3]$$

$$= 1 - 0.4^3 + 3 \cdot 0.6^2 \cdot 0.6 + 3 \cdot 0.6^3 \cdot 0.6 = 0.745$$

$$= 0.745$$

→ first part graph is not connected

$$\rightarrow P_2 = 1 - p = 0.4$$

$$\rightarrow P_3 = 1 - \sum_{k=1}^{\binom{4}{2}} \binom{\binom{4}{2}}{k} p_k^{k(1-p)} (1-p)^{\binom{4}{2}-k}$$

$$= 1 - [(2)^2 + 3 \cdot 1^2] p^2 = 1 - 2^2 p^2$$

$$= 1 - 0.4^2 = 0.84$$

$$\rightarrow 1 - P_3 = 1 - 0.84 = 0.16$$

→ Assignment

→ a) Given info about a parallel system w/ two components

→ minimal path sets =  $\{\{1,2\}, \{1,3\}, \{2,3\}\}$

→ components are  $\perp\!\!\!\perp$

→  $\text{MTTF}_1 = 10$  months,  $\text{MTTF}_2 = 10$  months

→  $P(T_1 > t) = 1 - e^{-t/\text{MTTF}_1} = 1 - e^{-t/10}$

→  $P(T_2 > t) = 1 - e^{-t/\text{MTTF}_2} = 1 - e^{-t/10}$

→  $P(T_1 > t) \cdot P(T_2 > t) = e^{-t/10} \cdot e^{-t/10} = e^{-2t/10}$

→  $P(T_1 > t) + P(T_2 > t) = 1 - e^{-2t/10}$

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→  $P(T_1 > t) +$