

**ECS 253 / MAE 253, Network Theory and Applications**  
Spring 2023  
**Advanced Problem Set # 4, Due May 31**  
**Topic: Configuration model random graphs**

**Note 1:** Some of the problems herein require you to implement computer code. The purpose is to teach you generic skills about how to build a random graph and how to run simulations on a graph. For these problems, you can either use your favorite graph library (e.g., networkX, igraph, ...) or do everything from scratch (e.g., by implementing your own adjacency list as a vector of vector). Although we encourage you to use a graph library, you should only call “basic graph operations” from such libraries. Indeed, calling a “high level” function such as `nx.configuration_model` defeats the purpose of learning how to build your own random graph. The following operations are “basic graph operations”: creating a network with  $N$  nodes and no edges; adding/deleting a node; adding/deleting an edge; requesting who are the neighbors of a node; storing information in a node/edge (i.e., node/edge properties). Use common sense.

**Note 2:** This document uses the first  $N$  natural numbers (i.e.,  $1, 2, 3, \dots, N - 1, N$ ) to refer to each node of a graph. Depending of the programming language you are using, indexing the elements of a vector may start at 0 or 1. Hence, you are free to use the first  $N$  non-negative integers (i.e.,  $0, 1, 2, \dots, N - 2, N - 1$ ) in your computer code if you want to.

## 1 Building a configuration model random graph

Let  $\mathbf{k} = (k_1, k_2, k_3, \dots, k_N)$  be a vector of  $N$  non-negative integers such that  $\sum_{i=1}^N k_i$  is an even number. An instance of the configuration model random graph with degree sequence  $\mathbf{k}$  can be obtained as follows.

- Make sure that  $\sum_{i=1}^N k_i$  is an even number (give an error if not).
- Create an undirected graph containing  $N$  nodes and no edges.
- Create a vector (or list or multiset...) such that, for each  $1 \leq i \leq N$ , it contains  $k_i$  copies of  $i$ . This object will hereafter be known as the “stub list”. *Example: The vector  $(1, 2, 3, 4, 5, 5, 6, 6, 6, 7, 7, 7)$  is a valid stub list in the case  $\mathbf{k} = (k_1, k_2, k_3, k_4, k_5, k_6, k_7) = (1, 1, 1, 1, 2, 3, 3)$ .*
- Sample uniformly at random one element from the stub list; call that element  $i$  and remove it from the stub list. Sample uniformly at random another element from the (now shorter) stub list; call that element  $j$  and remove it from the stub list. Add an edge between node  $i$  and node  $j$  in the network. Repeat this step as long as the stub list is not empty.

The resulting network will be a random graph with degree sequence  $\mathbf{k}$ . For the purpose of this problem, we do not worry about repeated edges (i.e., more than one links between two nodes) and self loops (i.e., a node with an edge to itself).

- (a) Write a computer implementation of this algorithm.
- (b) Using the degree sequence  $\mathbf{k} = (1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 3, 3)$ , generate a configuration model random graphs and display the result as a figure.
- (c) Let  $\mathbf{n}$  be a vector such that the  $k$ th component represents the number of nodes of degree  $k$ . For example,  $\mathbf{n} = (0, 8, 4, 2)$  corresponds to the  $\mathbf{k} = (1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 3, 3)$ . Write a function that receives  $\mathbf{n}$  as an input and returns  $\mathbf{k}$ .
- (d) Let  $\mathbf{p}$  be a vector such that the  $k$ th component denotes the probability of having a node of degree  $k$ . Write a function that receives  $\mathbf{p}$  and returns  $\mathbf{k}$ .

## 2 Percolation and spreading

Consider the three functions described below.

**percolation:**

- Receive as input a graph  $G$  and a real number  $p$  such that  $0 \leq p \leq 1$ .
- Make a graph with no edges and as many nodes as  $G$  has; call this graph without edges  $G'$ .
- Iterate over all the edges of  $G$ . Suppose the current edge is between nodes  $u$  and  $v$ . Get a random number uniformly distributed in the interval  $[0, 1)$ . If that number is lower than  $p$ , add in  $G'$  an edge between nodes  $u$  and  $v$ .
- Return the graph  $G'$ .

Hence, each edge of  $G$  has probability  $p$  to be present in  $G'$ .

**spreading:**

- Receive as input a graph  $G$ , a node index  $v$ , and a real number  $T$  such that  $0 \leq T \leq 1$ .
- “Mark” all nodes as “unreached” by one of the following two methods:
  - Create a vector of bool called `is_reached` containing as many entries as there are nodes in  $G$ . Initialize all its entries to “False”; **OR**
  - Give to each node in  $G$  the bool “node property” `is_reached`. Initialize them all to “False”.
- Create an empty vector (or other appropriate container) of indices; call it `unresolved`, and place  $v$  in it.

- Set an integer variable `number_reached` with value 1.
- Repeat the following as long as `unresolved` is not empty. Get in  $u$  the value of an element of `unresolved`, and remove said element from `unresolved`. If  $u$  is marked as unreached (i.e., if `is_reached` is “False” for that node), do the following:
  - Increment `number_reached` by one.
  - Mark  $u$  as reached (i.e., set `is_reached` to “True” for that node).
  - For each neighbor  $w$  of  $u$ , generate a random number uniformly distributed in the interval  $[0, 1)$ . If the number is lower than  $T$ , add  $w$  to `unresolved`.
- Return `number_reached`.

`component_size`:

- Receive as input a graph  $G$  and a node index  $v$ .
  - Call your `spreading` function for the graph  $G$ , the node index  $v$ , and using  $T = 1$ .
  - Return the `number_reached` returned by the aforementioned function.
- (a) The fifth bullet of `spreading` does not specify *which* element of `unresolved` should be removed to become  $u$ . Does the outcome `number_reached` depend on this choice? Why?
- (b) Explain why the value returned by `component_size` corresponds to the size of the component to which  $u$  belong.
- (c) Let  $G'$  be a graph returned by `percolation` (with parameters  $G$  and  $p$ ). Show (by hand) that calling `spreading` with the parameters  $G', v$  and  $T$  is statistically equivalent to calling `spreading` with the parameters  $G, v$  and  $pT$ . From this result, deduce a relationship between `spreading` and the size of the component to which  $v$  belong in a graph returned by `percolation`.
- (d) Implement these three functions.
- (e) We shall now use the code you have written up to this point to understand percolation/spreading on a random network with a given degree sequence. For a given graph, you will simulate spreading with a given probability of transmission ( $T$ ) and compute the probability distribution for the total number of nodes reached starting from a node at random. The process is described below. Write code to do the following:
- Receive  $T, \mathbf{n}$  as inputs.
  - Initialize a vector `res` to store the result. If the number of nodes in the network is  $N$  this vector has size  $N$ . Since spreading process could reach at most  $N$  nodes (and always starts with one random node).
  - Repeat the following steps `number_simulations` times:

- Create a configuration model random graph with  $\mathbf{n}$  as the input.
- Do  $\mathcal{A}$  or  $\mathcal{B}$  (See below). Store the result in  $S$
- Increment the  $S$ th entry of the vector  $\mathbf{res}$  by 1.
- Normalize the vector  $\mathbf{res}$  such that component  $i$  gives the probability of the process spreading to  $i$  nodes starting from a random node. This is the result we are interested in.

$\mathcal{A}$ : Call `spreading`.

$\mathcal{B}$ : Call `percolation` and then `component_size`.

(f) Run the above process for  $\mathbf{n} = (0, 8, 4, 2)$  and  $T = 0.4$ . Report the result.