14.15**NETWORKS** PROBLEM SET 3

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1. Problem 1

1.1. **Problem 1.a. Problem:** Let A_1 denote the event that node 1 has at least $l \in \mathbb{Z}^+$ neighbors. Do we

observe a phase transition for this event? If so, finnd the threshold function and explain your reasoning. **Solution:** Yes, we do observe a phase transition when $t(n) = \frac{l}{n-1}$. To show that there exists a phase transition, lut us examine the case when $\frac{p(n)}{t(n)} = \frac{p(n)(n-1)}{l} \to 0$. In this case we can use Markov's inequality. We define X as a random variable of the number of neighbors of node 1 and we note that $P(A_1) = P(X \ge l)$. Thus, we want to obtain the expected value of X. However, we know that a node has an expected (n-1)p(n)neighbors, because each neighbor has a p(n) probability of being connected by an edge, and there are n-1neighbors. Thus, we see that $E[A_1] = (n-1)p(n)$. Markov's inequality allows us to obtain the following bound:

(1)
$$P(A_1) = P(X \ge l) \le \frac{E[A_1]}{l} = \frac{p(n)(n-1)}{l}$$

However, since $\frac{p(n)}{t(n)} = \frac{p(n)(n-1)}{l} \to 0$, we see that $P(A_1) \to 0$ as well. Thus, we have shown that the first part of the phase transition, namely that under the threshold, A_1 does not occur. Now we shall show that above the threshold, the event occurs almost surely. We want to see what happens when $\frac{p(n)}{t(n)} = \frac{p(n)(n-1)}{l} \rightarrow$ ∞ . We shall use Chebyshev's inequality:

(2)
$$P(|X - E[X]| \ge |E[X] - l|) \le \frac{\operatorname{Var}(X)}{(E[x] - l)^2}$$

(3)
$$= \frac{p(n-1)}{((n-1)p-l)^2}$$

$$= \frac{p(n-1)}{(n-1)^2 p l^2 + l^2 - 2(n-1)pl}$$

(5)
$$= \frac{1}{(n-1)l^2 + \frac{l^2}{(n-1)p} - 2l}$$

Where we know that Var(X) = E[X] = (n-1)p because X can be approximated as a poisson random variable (see Newman p. 402) for large n. Moreover, we know that l is a constant so that as $\frac{p(n)(n-1)}{l} \to \infty$ we have $\frac{l^2}{(n-1)p} \to 0$. Therefore, we also see that $P(|X-E[X]| \ge |E[X]-l|) \le \frac{1}{(n-1)l^2-2l}$ as $\frac{p(n-1)}{l} \to \infty$. Since l is a constant we see that $n \to \infty$ implies that this probability goes to zero. In other words, the probability that X deviates by more than E[X] - l from its expected value goes to zero. This shows that $P(A_1) \to 1$ when $E[X] \ge l$. \square

1.2. **Problem 1.b. Problem:** Let B denote the event that a cycle with k edges (for a fixed k) emerges in the graph. Do we observe a phase transition of this event? If so, find the threshold function and explain your reasoning.

Solution: Yes a threshold function does exist for this event. Take $t(n) = \frac{1}{n}$ as the threshold function. We will first show that as $\frac{p}{t(n)} = pn \to 0$, then $P(B) \to 0$. We first want to find the expected number of cycles of length k in the graph. Let X denote the number of cycles of length k. We know that there are $\binom{n}{k}$ ways to select k nodes, and that for each of these subsets of nodes, there are (k-1)!/2 ways to create a cycle. This follows because we set the first node of the cycle, then there are k-1 ways of picking the second node, k-2 ways of picking the third node, etc. We divide by two because we could go either backwards or

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forwards in this cycle (clockwise or counterclockwise). Each of these cycles has probability p^k of emerging. Therefore, we have:

(6)
$$E[X] = \binom{n}{k} \frac{(k-1)!}{2} p^k$$

(7)
$$= \frac{n!}{k!(n-k)!} \frac{(k-1)!}{2} p^k$$

$$= \frac{n!}{2k(n-k)!}p^k$$

However, we know that $\frac{n!}{(n-k)!} = n(n-1)\dots(n-k+1) \le n^k$. This means that we have $E[X] \le \frac{(np)^k}{2k}$. Thus, as $np \to 0$, we see that $E[X] \to 0$, which implies that $P(X \ge 1) \le \frac{E[X]}{1} = E[X] \to 0$. Since $P(X \ge 1) = P(B)$, we see that $P(B) \to 0$ as $\frac{p}{t(n)} \to 0$. Now to show the second half of the phase transition, we need to show that as $\frac{p}{t(n)} \to \infty$ we have $P(B) \to 1$.

To show this we note that $P(X \leq 0) = P(E[X] - X \geq E[X]) \leq \frac{\operatorname{Var}(X)}{E[X]^2}$ by Chebyshev. We can also bound E[X] from below by using the fact that $\frac{n!}{(n-k)!} = n(n-1)\dots(n-k+1) \geq (n-k)^k$. This shows that $E[X] \geq \frac{(n-k)^k p^k}{2k}$. Since k is a constant we know that $nk - kp \to \infty$ when $np \to \infty$. Thus, we see that $(n-k) \to \infty$, which implies that $E[X] \geq \infty$ when $np \to \infty$.

Now we use the fact that X can be approximated as a poission distribution to note the fact that Var(X) = E[X]. This means that $P(X \leq 0) \leq \frac{1}{E[X]} \to 0$. This implies that $P(X \geq 0) \to \infty$, which means that $P(B) \to \infty$, just as we wanted. \square

2. Problem 2

2.1. Problem 2.a. Problem: Show that the mean degree of a vertex in this network is 2c.

Solution: We know that the expected number of connected trios is the total possible number of triples of nodes times the probability each triple becomes a triangle. This is $\binom{n}{3}\frac{c}{\binom{n-1}{2}}=\frac{n!}{3!(n-3)!}\frac{c}{(n-1)!}(n-1-2)!2!=\frac{2!nc}{3!}=\frac{nc}{3!}$. Since we know that the total number of edges is just 3 times the number of triangles we have nc total expected edges. Moreover, the expected total degree is 2 times the total number of expected edges (since each edge has two endpoints). This means we expect 2nc total degree in the graph, and since there are n nodes, each node has an expected degree of $\frac{2nc}{n}=2c$. \square

2.2. Problem 2.b. Problem: Show that the degree distribution is

(9)
$$p_k = \left\{ \begin{array}{ll} e^{-c} c^{k/2} / (k/2)! & \text{if } k \text{ is even} \\ 0 & \text{if } k \text{ is odd} \end{array} \right\}$$

Solution: We can assume that the degree distribution is a poisson random variable because of the fact that each triangle is selected randomly with a particular probability. This is a binomial distribution, which converges to a poission distribution as n grows large. Therefore, we only need to find the expected degree of a node. The probability that a node has degree k is really equal to the probability that it is inside of k/2 triangles (because each triangle provide two edges). The expected value is given by the number of triangles that a node can connect to times the probability of occurring. This is just $\binom{n-1}{2} \frac{c}{\binom{n-1}{2}}$.

Therefore, we see that $\lambda=c$ in this poisson distribution and that there are m=k/2 different triangle possibilities. We therefore have the probability distribution of a possion random variable $\frac{\lambda^m e^{-\lambda}}{m!}$ which we can substitute $\lambda=c$ and m=k/2 to obtain $\frac{c^{k/2}e^{-c}}{(k/2)!}$. This is the degree distribution for when k is even, because k cannot be odd. This is because whenever a new triangle is added, 2 more edges are added to each node, and thus, one cannot have an odd degree. Therefore, we have shown that the degree distribution is given by:

(10)
$$p_k = \left\{ \begin{array}{ll} e^{-c} c^{k/2} / (k/2)! & \text{if } k \text{ is even} \\ 0 & \text{if } k \text{ is odd} \end{array} \right\}$$

2.3. **Problem 2.c. Problem:** Show that the clustering coefficient is $C = \frac{1}{2c+1}$. Solution: \square

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2.4. Problem 2.d. Problem: Show that when there is a giant component in the network, its expected size S, as a fraction of the network size, satisfies $S = 1 - e^{-cS(2-S)}$.

Solution: Suppose that u_i is the probability that node i does not belong to the giant component. For i to not belong in the giant component we must either have 1) that i is not connected to some triangle j or 2) that i is connected to the triangle j but j itself is not part of the giant component. The first occurs with probability 1-p, while the second occurs with probability pu_i^2 , since both nodes of the connecting triangle cannot belong to the giant component. This means the probability that i is not connected to the giant component through traingle j is $1 - p + pu_i^2$.

Since each node is independent and triangles are randomly selected, we can say $u_i = u_j = u$. Since there are $\binom{n-1}{2}$ triangles for which we must do this analysis for, we see the following:

(11)
$$u = (1 - p + pu^2)^{\binom{n-1}{2}}$$

(12)
$$= \left(1 - \frac{c}{\binom{n-1}{2}} + \frac{c}{\binom{n-1}{2}}u^2\right)^{\binom{n-1}{2}}$$

Taking logs of both sides we see that we can simplify the expression:

(13)
$$\ln u = \binom{n-1}{2} \ln \left(1 - \frac{c}{\binom{n-1}{2}} + \frac{c}{\binom{n-1}{2}} u^2 \right)$$

$$= {n-1 \choose 2} \frac{cu^2 - u}{{n-1 \choose 2}}$$

$$(15) = cu^2 - u$$

Where we have used the fact that ln(1+x) = x for small x. Exponentiating both sides of the resulting expression gives us that $u = e^{cu^2 - u}$. However, we know that S = 1 - u because u is the probability of any particular node not belong to the giant component, and S is the probability of belonging to the giant component. Therefore, we see that $1 - S = e^{-c+c(1-S)^2}$ which simplifies to $S = 1 - e^{-cS(2-s)}$, which proves our result. \square

2.5. **Problem 2.e. Problem:** What is the value of the clustering coefficient when the giant component fills half of the network?

Solution: We choose S = 1/2 and we solve for c. We have:

$$\frac{1}{2} = 1 - e^{\frac{1}{2}c(\frac{1}{2} - 2)}$$

(16)
$$\frac{1}{2} = 1 - e^{\frac{1}{2}c(\frac{1}{2}-2)}$$
(17)
$$\frac{1}{2} = 1 - e^{-\frac{3}{4}c}$$
(18)
$$e^{-\frac{3}{4}c} = \frac{1}{2}$$

(18)
$$e^{-\frac{3}{4}c} = \frac{1}{2}$$

$$-\frac{3}{4} = -\ln(2)$$

$$(20) c = \frac{4}{3}\ln(2)$$

Now, we substitute into our equation for the clustering coefficient of $C = \frac{1}{2c+1}$ so we obtain $C = \frac{1}{8\ln(2)/3+1}$.

3. Problem 3

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4. Problem 4

Problem: Consider a graph G with n nodes generated according to the configuration model with a particular degree distribution P(d). Show that the overall clustering coefficient is given by $Cl(G) = \frac{\langle d \rangle}{n} \left[\frac{\langle d^2 \rangle - \langle d \rangle}{\langle d \rangle^2} \right]^2$.

Solution: We note that the clustering coefficient Cl is the average probability that two neighbors of a vertex are neighbor of each other as wel. Thus, we are looking for the probability that if i and j are connected, given that they are connected to v. First, we know that the probability of an edge between nodes i and j is given by $\frac{k_i k_j}{2m}$, were k_i is the degree of node i. This is because there are k_j edges coming into node j, so there is a $k_j/2m$ probability that any edge goes to j. Since there are k_i edges leaving node i, there are k_i times the probability of any given edge going to j. Therefore, we see that the probability of an edge existing between nodes i and j is $\frac{k_i k_j}{2m}$.

Next, we notice that the excess degree distribution for node i can be given by $q_k = \frac{k+1}{k} p_{k+1}$. This can be seen from the derivation in Newman (Equation 13.46). Thus, the probability that any node i, which is already connected to some node v, has degree of k is given by q_k . Thus, we can write the clustering coefficient as the sum over all possible degrees k_i, k_j for nodes i and j of the probability of obtaining k_i and k_j times the probability of an edge between nodes i and j. This is given by:

$$Cl(G) = \sum_{k_i=0}^{\infty} \sum_{k_i=0}^{\infty} q_{k_i} q_{k_j} \frac{k_i k_j}{2m}$$

$$= \frac{1}{2m} \left(\sum_{k=0}^{\infty} k q_k \right)^2$$

(23)
$$= \frac{1}{2m} \left(\sum_{k=0}^{\infty} (k+1)p_{k+1} \right)^2$$

This was obtaining by noting that $q_k = \frac{k+1}{k} p_{k+1}$. Next, we notice that if we multiply by $\frac{\langle k \rangle^2}{\langle k \rangle^2} = 1$ and make a change of variable for k = k + 1, we obtain the following expression:

(24)
$$Cl(G) = \frac{1}{2m\langle k \rangle^2} \left(\sum_{k=0}^{\infty} k(k+1) p_{k+1} \right)^2$$

$$= \frac{1}{2m\langle k \rangle^2} \left(\sum_{k=0}^{\infty} (k-1)kp_k \right)^2$$

Since 2m is the total number of edges in the graph, we know that $2m = n\langle k \rangle$ so we can substitute in the denominator of the fraction and expand $(k-1)kp_k = k^2p_k - kp_k$ to obtain:

(26)
$$Cl(G) = \frac{1}{n\langle k \rangle^3} \left(\sum_{k=0}^{\infty} k^2 p_k - k p_k \right)^2$$

$$= \frac{1}{n\langle k \rangle^3} \left(\langle k^2 \rangle - \langle k \rangle \right)^2$$

Now, set d=k and we see that $Cl(G)=\frac{\langle d \rangle}{n}\left[\frac{\langle d^2 \rangle - \langle d \rangle}{\langle d \rangle^2}\right]^2$, which is what we wanted. \Box **Problem:** Solution: \Box