## CSL851: Algorithmic Graph Theory

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## Lecture 5: September 24

Lecturer: Anand Srivastav

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### 5.1 Threshold for connectedness in Random Graphs

Theorem 5.1 Let $\alpha=\alpha(n)$ be a function with $\alpha(n) \rightarrow \infty$ as $n \rightarrow \infty$. Then $S(n):=\frac{\ln (n)-\alpha(n)}{n}$ is a lower threshold function and $t(n):=\frac{\ln (n)+\alpha(n)}{n}$ is an upper threshold function for connectivity of $G(n, p)$.

## Proof:

1. Lower Threshold Function: If $p \leq s$ then by Theorem $4.3 \mathrm{G}(\mathrm{n}, \mathrm{p})$ has isolated vertices approximately almost surely. Therefore, $\mathrm{G}(\mathrm{n}, \mathrm{p})$ graph are not connected.
2. Upper Threshold Function: We assume $\alpha(n) \leq \ln (n)$ for all n. For each $k \leq n$, let $X_{k}$ be the random variable which counts the number of connected components of size k .

Let $Y=\sum_{k \leq n} X_{k}$ and let $X:=\sum_{k \leq\left\lfloor\frac{n}{2}\right\rfloor} X_{k}$. Note here Y counts all the connected components in $\mathrm{G}(\mathrm{n}, \mathrm{p})$ whereas X counts all the connected components of size at most $\left\lfloor\frac{n}{2}\right\rfloor$.
We thus have $Y \geq 1$ implies $X \geq 1$ because there is some smaller connected component inside the bigger component.
Therefore( Using Markov inequality ), $P(Y \geq 1) \leq P(X \geq 1) \leq \frac{E(X)}{1}=E(X)$

Claim 5.2 $E(X) \rightarrow 0$ as $n \rightarrow \infty$

Proof: Consider $S \subseteq V$,
$\mathrm{P}(\mathrm{S}$ forms a maximal connected component in $\mathrm{G}(\mathrm{n}, \mathrm{p})) \leq \mathrm{P}\left(\right.$ no vertex in S is connected to $\left.S^{c}\right)$
$\mathrm{P}\left(\right.$ no vertex in S is connected to $\left.S^{c}\right)=(1-p)^{|S|(n-|S|)} \leq e^{-p|S|(n-|S|)} \leq e^{-t|S|(n-|S|)}$
Note the last inquality holds if $p \geq t$ (and we can this Equation 1).
Now we are going to analyze the expectation of each $X_{k} \forall k \leq\left\lfloor\frac{n}{2}\right\rfloor$

$$
E\left(X_{k}\right) \leq e^{-t k(n-k)}\binom{n}{k}
$$

Now we need to give a good estimation of the Binomial Coefficient. Using Stirling's formula :

$$
E\left(X_{k}\right) \leq\left(\frac{n e}{k}\right)^{k} e^{-t k(n-k)}
$$

Next inserting the formula for t we obtain :

$$
E\left(X_{k}\right) \leq e^{-\alpha(n)}\left(\frac{e^{1-\left(1-\frac{1}{n}\right) \alpha(n)}+\frac{k}{n}(\ln (n)+\alpha(n))}{k}\right)^{k}=e^{-\alpha(n)}\left(B_{k}\right)^{k}
$$

We can the above Equation 2 and now turn our attention to $B_{k}$.
Case 1: When $k \leq\left\lfloor n^{\frac{3}{4}}\right\rfloor$ then :

$$
\frac{k}{n}(\ln (n)+\alpha(n)) \leq \frac{1}{n^{\frac{1}{4}}}(\ln (n)+\alpha(n)) \leq \frac{2 \ln (n)}{n^{\frac{1}{4}}} \rightarrow 0
$$

The last approximation can be realized using the Taylor Series expansion. Henceforth, the numerator in (2) is at most a constant say c $i 0$. If $\mathrm{c} ; \mathrm{k}$, then with $\theta=\frac{c}{c+1}$ we have $B_{k} \leq \frac{c}{k} \leq \theta<1$ for any $k \geq c+1$.
So, $\sum_{k=c+1}^{\left\lfloor n^{\frac{3}{4}}\right\rfloor}\left(B_{k}\right)^{k} \leq \sum_{k=c+1}^{\left\lfloor n^{\frac{3}{4}}\right\rfloor} \theta^{k} \leq c_{1}$ for some constant $c_{1}>0$
Case 2: When $\left\lfloor n^{\frac{3}{4}}\right\rfloor \leq k \leq\left\lfloor\frac{n}{2}\right\rfloor$ then:

$$
B_{k} \leq e^{1-\frac{1}{4} \ln (n)-\left(\frac{1}{2}-\frac{1}{k}\right) \alpha(n)}
$$

Since $k \geq 2, \frac{1}{2}-\frac{1}{k} \geq 0$ so the exponent tends to infinity. So $B_{k} \leq \theta$ for some $\theta<1$
Therefore, $\sum_{k=\left\lfloor n^{\frac{3}{4}}\right\rfloor}^{\frac{n}{2}}\left(B_{k}\right)^{k}<c_{2}$ for some constant $c_{2}>0$.
Thus $E(X)=\sum_{k=1}^{\frac{n}{2}} E\left(X_{k}\right) \leq e^{-\alpha(n)}\left(c_{1}+c_{2}+c_{3}\right) \rightarrow 0$ as $n \rightarrow \infty$.

