

## Second Moment Method

### Disappearance of Isolated Vertices

As  $p$  increases, there is a transition from collections of trees, to cycles, to the emergence of a giant component. As  $p$  increases further, isolated vertices are swept up by the giant component.

Let  $x$  be the number of isolated vertices.

$$p = c(\ln n)/n \quad E(x) = n^{1-c}$$

The limit of  $E(x)$ , as  $n$  goes to  $\infty$ , is 0 if  $c \geq 1$  and 1 if  $c \leq 1$ .  $\text{Prob}(x=0) = 1$

\*There is the possibility that, given a series of graphs, for the first graph  $x = \infty$  while for the remaining graphs  $x = 0$ . The Second Moment Method will be used to rule out this possibility.

\*if  $\sigma^2$  is small, we should not have the above situation ( $x = \infty$  for one graph,  $x = 0$  for the others).

\*First, a bit about Markov's Inequality and Chebyshev's Inequality:

### Theorem: Markov's Inequality

Let  $x$  be a random variable that is nonnegative. Then,  $P(x \geq a) \leq E(x)/a$

$$E(x) = \int_0^{\infty} xP(x)dx = \int_0^a xP(x)dx + \int_a^{\infty} xP(x)dx$$

-if we throw the first integral away, we get a lower bound on  $E(x)$ :  $E(x) \geq \int_a^{\infty} xP(x)dx$

- $x$  is always greater than  $a$ , so:

$$\int_a^{\infty} xP(x)dx \geq a \int_a^{\infty} P(x)dx$$

$\underbrace{\int_a^{\infty} P(x)dx}_{P(x \geq a)}$

$$P(x \geq a) \leq E(x)/a$$

-replace  $a$  by  $bE(x)$ :  $P(x \geq bE(x)) \leq E(x)/bE(x) = 1/b$

$$P(x \geq bE(x)) \leq 1/b$$

\*We can get a tighter bound by taking into account the variance....

Theorem: Chebyshev's Inequality:

Let  $x$  be a random variable. Then the probability that  $|x - E(x)| \geq t\sigma < 1/t^2$

\*With Chebyshev's Inequality, we measure distance in terms of the standard deviation.

Proof:  $\text{Prob}[|x - \mu| \geq t\sigma] = \text{Prob}[(x - \mu)^2 \geq t^2\sigma^2]$

Note: will sometimes use  $\mu$  or  $m$  in place of  $E(x)$

Apply Markov's Inequality:

$$\text{Prob}[(x - \mu)^2 \geq t^2\sigma^2] \leq E[(x - \mu)^2]/t^2\sigma^2 = \sigma^2/t^2\sigma^2 = 1/t^2 \quad (\sigma^2 = E[(x - \mu)^2])$$

- $t^2\sigma^2$  is what was 'a' back in Markov's Inequality

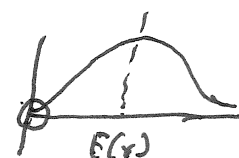
### Second Moment Method

$$\text{Prob}[|x - E(x)| \geq E(x)] \leq \sigma^2/E^2(x)$$

( $t = E(x)/\sigma$  and replace)

Now clearly for a nonnegative random variable  $x$ :

$$\text{Prob}(x = 0) \leq \text{Prob}[|x - E(x)| \geq E(x)] \leq \sigma^2/E^2(x)$$



\*We want to prove that  $P(x = 0)$ , so need to show that  $\sigma^2/E^2(x)$  goes to 0 (i.e. variance goes to 0 with respect to the expected value)

Under what conditions does  $\sigma^2/E^2(x)$  go to 0?

$$\sigma^2 = E[(x - E(x))^2] = E(x^2) - 2E(x)E(x) + E^2(x) = E(x^2) - E^2(x)$$

$$\text{Prob}[|x - E(x)| \geq E(x)] \leq \sigma^2/E^2(x) = [E(x^2) - E^2(x)]/E^2(x) = E(x^2)/E^2(x) - 1$$

If  $E(x^2)$  is asymptotically less than or equal to  $E^2(x)$ , then  $\text{Prob}(x = 0)$  goes to 0

Back to the disappearance of isolated vertices: applying the Second Moment Method:

$x$  = number of isolated vertices

$$\text{Prob}(x = 0) \longrightarrow E(x^2) \leq E^2(x)$$

-need to calculate value of  $x^2$

$$x = x_1 + x_2 + \dots + x_n \quad x_i = 1 \text{ if the vertex is isolated, } 0 \text{ otherwise}$$

indicator variables

$$E(x^2) = \sum_{i=1}^n E(x_i^2) + \sum_{i \neq j} (x_i x_j) = \sum_{i=1}^n E(x_i) + n(n-1)E(x_1 x_2) = E(x) + n(n-1)(1-p)^{2(n-1)-1}$$



Probability that  $x_1$  is isolated  $= (1-p)^{n-1}$

Probability that  $x_2$  is isolated is the same as above, except don't count the edge between  $x_1$  and  $x_2$  twice

$$\text{Now } E(x^2)/E^2(x) = 1/E(x) + [n(n-1)/n^2][(1-p)^{2(n-1)-1}/(1-p)^{2(n-1)}]$$

$$E(x^2)/E^2(x) = 1/E(x) + 1/(1-p) = 1/n(1-p)^{n-1} + 1/(1-p) \quad (E(x) = n(1-p)^{n-1})$$

For  $p = c(\ln n)/n$  and  $c < 1$ :  $E(x)$  goes to  $\infty$

$$E(x^2)/E^2(x) = 1/n^{1-c} + 1/(1-(c(\ln n)/n)) = 1$$

Thus, by the Second Moment Method, can claim that  $\text{Prob}(x = 0)$  goes to 0