

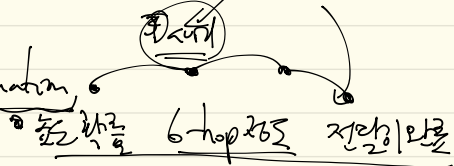
Lecture 8 (chapter 6) Small-world phenomenon

ER graph: $\frac{1}{N^2}$, $\frac{1}{N}$, giant component, diameter, connectivity
 structured graph: $\left\{ \begin{array}{l} \text{small world} \\ \text{...} \\ \text{...} \end{array} \right.$

시행: $\frac{1}{N^2}$: Milgram's experiment

• $\frac{1}{N^2}$ (x), known contact, local information
 ↓
 "시행: $\frac{1}{N^2}$ "

⇒ Destination $\left(\begin{array}{c} \frac{1}{N^2} \\ \frac{1}{N} \end{array} \right) \left(\frac{1}{N^2} \right) \dots$



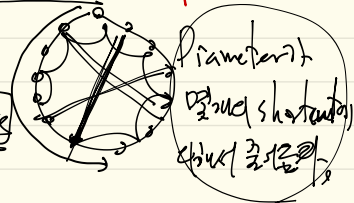
Social network graph: $\frac{1}{N^2}$ $\frac{1}{N}$ $\frac{1}{N}$

email network, citation network: Erdos number chapter 6.2

Q1: $\frac{1}{N^2}$ $\frac{1}{N}$ structure, $\frac{1}{N^2}$ $\frac{1}{N}$ (shortcut) chapter 6.3

how much information
only local information
 (navigability)

small world redox
 (?)



6.2 Strogatz and Watts (SW) SW(n, p)

Model: Given some integer $m > 0$, consider $m \times m$ grid
 Let $n = m^2$ - $\text{Larg}(n)$ regime
 $m = \sqrt{n}$

Connect nodes with two types

Type 1: local edges: just grid neighbors

Type 2: "shortcut" edges: $\frac{1}{n}$ uniformly random node \rightarrow node \rightarrow node

small world emerged(?)

induced cluster graph

$$SW(m, p) \rightarrow \mathcal{G}'$$

$$D(SW(m, p)) \leq 2\sqrt{k} (2 + D(\mathcal{G}')) \quad \underline{D(\mathcal{G}')} \text{ orderwise}$$

Homework 1. (x) 3/30/18

Homework 2. any other useful useful clusters

Lemma 6.2. $\Gamma_1(u)$: a group of nodes containing u of size $\leq \log n$, such that
 $\Gamma_2(u)$: reachable from nodes in $\Gamma_1(u)$ via all nodes in $\Gamma_1(u)$
 \vdots shortcuts generated from $\Gamma_1(u)$ are connected by only "grid edges"

Similarly, we define $\Gamma_x(u)$.

Let $d_1(u) = |\Gamma_1(u)|, d_2(u) = |\Gamma_2(u)|, \dots, d_x(u) = |\Gamma_x(u)|$. $N' \subseteq V$ set of nodes

Let $\varepsilon > 0$ be fixed, such that $\textcircled{1} k_p(1-\varepsilon) \geq 1$ and $\textcircled{2} k_p(1-\varepsilon) < k_p(1-\varepsilon)^2$

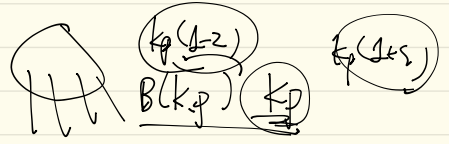
$$\frac{\log(k_p(1-\varepsilon))}{\log(k_p(1-\varepsilon)^2)} < 2 \iff \textcircled{2}$$

The constant C_0 can be chosen s.t for all $u \in N'$, with $\textcircled{1-\varepsilon}$ the following holds:

$$k_p(1-\varepsilon) \leq \frac{d_i(u)}{d_{i-1}(u)} \leq k_p(1+\varepsilon), \quad i=2, \dots, D$$

where $D = \left\lceil \frac{\log n}{2 \log(k_p(1+\varepsilon))} \right\rceil + 1$

$\textcircled{10}$ $d_{i-1} \leq d_i(u) \leq d_i$



\Rightarrow Lemma 2.10 $\frac{1}{2}$ $\frac{1}{2}$ Martingale, \rightarrow Azuma-Hoeffding inequality

(proof of Theorem 6.1) $p(t) = p(A \cap B) + p(A \cap B^c)$

$$= p(A|B) \cdot p(B) + p(A|B^c) \cdot p(B^c)$$

$$\leq p(A|B) + p(B^c)$$

the event
Let $\Sigma_\varepsilon(\omega)$ be:

for any two nodes $\Sigma_\varepsilon(\omega) = \left\{ \frac{d_\varepsilon(u,v)}{d_\varepsilon(\omega)} \leq (t\varepsilon)kp \right\}$

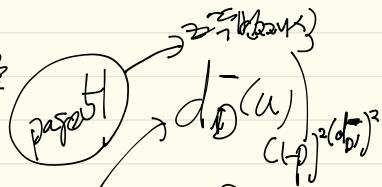
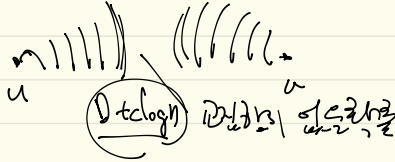
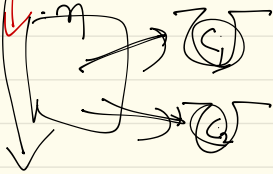
$$\Pr \left(\underbrace{d_\varepsilon(u,v)}_A > 2D + 2C \log n \right)$$

$$B: \left(\bigcap_{j=2}^D \Sigma_\varepsilon(\omega) \right) \cap \left(\bigcap_{j=2}^D \Sigma_\varepsilon(\omega) \right)$$

$$\leq \Pr \left(d_\varepsilon(u,v) > 2D + 2C \log n \mid \Sigma_2(\omega), \Sigma_3(\omega), \dots, \Sigma_D(\omega), \Sigma_2(\omega), \dots, \Sigma_D(\omega) \right)$$

$$+ \Pr \left(\bigcap_{j=2}^D \Sigma_\varepsilon(\omega) \right) + \Pr \left(\bigcap_{j=1}^D \Sigma_\varepsilon(\omega) \right)$$

$\leq o(\pi^2)$ from Lemma 6.2



$$\leq \pi$$

↳ the probability that two sets of sizes $(C \log n)(t\varepsilon)kp$

picked uniformly at random from n nodes empty intersection

tk

Using $\log_2 n \leq \log_2 (1 + \sqrt{n}) \leq \log_2 n + \frac{1}{2} \log_2 n$ $D = \lceil \frac{\log_2 n}{2 \log_2 (1 + \sqrt{n})} \rceil + 1$

$\Pr(G_n \subseteq \mathcal{C}) = \phi$ (Lemma 4.5) $(1 + o(1)) \exp\left[(n-2r) \log\left(\frac{1+r}{n-2r}\right) + n \log\left(1 - \frac{r}{n}\right) \right]$

$\Pr \leq (1 + o(1)) \exp(-c^2 \log^2 n)$

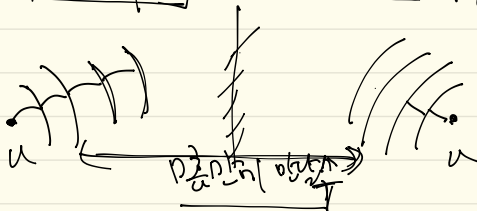
$\Pr(d_G(u,v) \geq \dots) \leq o(n^2) + \dots$

$\Pr(D(G) \geq \dots) \leq \max_{u \neq v} \dots$

$\Pr \leq o(n^2) + \dots$

Lemma 4.3 \checkmark probability of having a path of length $\geq 2D$

Theorem 4.1 \checkmark diameter $\leq f(n)$



$\Pr(d_G(u,v) \geq 2D + 2c \log n) \xrightarrow{n \rightarrow \infty} 0 \leq 2D + 2c \log n$

$\Pr(d_G(u,v) \geq f(n)) \xrightarrow{n \rightarrow \infty} 0$

$\Pr(\text{Diameter} \geq f(n)) \xrightarrow{n \rightarrow \infty} 0$

$\text{skz: } \Pr(A) = \Pr(A \cap B) + \Pr(A \cap B^c) \leq \Pr(A \cap B) + \Pr(B^c) \rightarrow 0$

$\text{Diameter} \leq f(n)$

Homework 4 $d_1 = \left\lceil \frac{d_1(u)}{2} \right\rceil \log n \approx \frac{d_1(u)}{2} \log n$ 이것이 맞을까?

Lecture 8 (part 3) Lemma 6.2

Thm 6.1

$$\Pr(d_{\mathcal{G}}(u, v) > 20 + 2 \log n) \leq \underbrace{\Pr\left(\bigcap_{i=2}^{\log n} \mathcal{E}_x(u)\right)}_{\text{Lemma 6.2}} + \underbrace{\Pr\left(\bigcap_{i=2}^{\log n} \mathcal{E}_x(v)\right)}_{\text{Lemma 6.2}} + T$$

$\mathcal{E}_x(u)$ and $\mathcal{E}_x(v)$ are events related to clusters of size $\log n$.

Lemma 6.2.

$\Gamma_1(u)$: a group of nodes containing u of size $\log n$, such that all nodes in $\Gamma_1(u)$ are connected by only "good edges".

$\Gamma_2(u)$: reachable from nodes in $\Gamma_1(u)$ via shortcuts generated from $\Gamma_1(u)$.

\vdots

Similarly, we define $\Gamma_x(u)$.

~~$d_1(u) = \left\lceil \frac{d_1(u)}{2} \right\rceil$~~ $d_1(u) = \left\lceil \frac{d_1(u)}{2} \right\rceil$, $d_2(u) = \left\lceil \frac{d_2(u)}{2} \right\rceil$, ..., $d_i(u) = \left\lceil \frac{d_i(u)}{2} \right\rceil$.

N' \Leftarrow set of nodes

Let $\varepsilon > 0$ be fixed, such that $\textcircled{1} \log(kp(1-\varepsilon)) > 1$, and

$$\frac{\log(kp(1+\varepsilon))}{\log(kp(1-\varepsilon))} < 2 \iff \textcircled{2} \log(kp(1+\varepsilon)) < \log(kp(1-\varepsilon))^2$$

$\log(kp(1+\varepsilon)) < \log(kp(1-\varepsilon))^2$

The constant c_{20} can be chosen s.t. for all $u \in N'$, with $\textcircled{1} \log(kp(1-\varepsilon)) > 1$, the following holds:

$$\log(kp(1-\varepsilon)) \leq \frac{d_i(u)}{d_{i-1}(u)} \leq \log(kp(1+\varepsilon)), \quad i=2, \dots, D$$

where

$$\textcircled{1} = \left\lceil \frac{\log n}{2 \log(kp(1-\varepsilon))} \right\rceil + 1$$

$\log(kp(1-\varepsilon)) > 1$

$(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \dots)$ Recall $\sum_{k=1}^{\infty} (k-\epsilon) p^k \approx \frac{d_1(u)}{d_2(u)} \leq (1+\epsilon) p^k$

$\Pr(\bigcap_{i=2}^n \Sigma_i(u)) \geq \frac{1}{2}$, $\Pr(\bigcap_{i=2}^n \Sigma_i(u)) \geq \frac{1}{2}$

Approach 1.

$\geq \Pr(\Sigma_2(u)) + \Pr(\Sigma_3(u)) + \dots + \Pr(\Sigma_n(u)) \rightarrow \frac{1}{2}$

$\frac{d_2(u)}{d_1(u)}$ $\frac{d_3(u)}{d_2(u)}$ $\frac{d_n(u)}{d_{n-1}(u)}$

$d_1(u), \dots, d_n(u) \rightarrow \frac{d_1(u)}{d_2(u)} \approx \frac{1}{2} \text{ or } \frac{1}{2} \approx \frac{d_2(u)}{d_1(u)}$

$\Pr(\bigcap_{i=2}^n \Sigma_i(u)) = \prod_{i=2}^n \Pr(\Sigma_i(u) | \Sigma_{i-1}(u), \dots, \Sigma_2(u))$

chain rule

$\Pr(A_1, A_2, A_3, A_4) = \Pr(A_1) \cdot \Pr(A_2 | A_1) \cdot \Pr(A_3 | A_1, A_2) \cdot \Pr(A_4 | A_1, A_2, A_3)$

$\Pr(\Sigma_i(u) | \Sigma_{i-1}(u), \dots, \Sigma_2(u)) \geq \frac{1}{2}$ $\frac{d_i(u)}{d_{i-1}(u)}$ $\frac{d_{i-1}(u)}{d_i(u)}$

A B

$\Pr(\Sigma_i(u) | \Sigma_{i-1}(u), \dots, \Sigma_2(u)) \geq \frac{1}{2}$

$\frac{d_i(u)}{d_{i-1}(u)} \geq \frac{1}{2} \Rightarrow d_i(u) \geq \frac{1}{2} d_{i-1}(u)$

$\frac{d_{i-1}(u)}{d_i(u)} \geq \frac{1}{2} \Rightarrow d_{i-1}(u) \geq \frac{1}{2} d_i(u)$

$\geq \frac{1}{2}$ for any $i \geq 2$

$\Pr(A|B) \cdot \Pr(B) = \Pr(A \cap B) + \Pr(A \cap B^c)$

$= \Pr(B) \Pr(A|B) + \Pr(B^c) \cdot \Pr(A|B^c)$

$\leq \Pr(A|B) + \Pr(B^c)$

이런 event C 가 있어 $P(C)$ 가 임의의 값

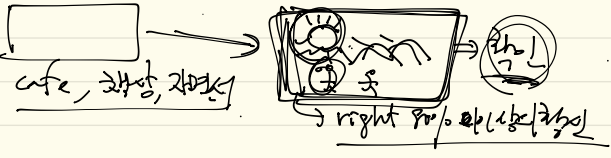
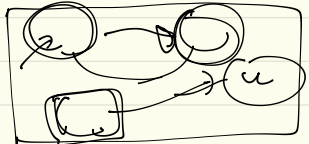
$$P(A|B) = P(A \cap C | B) + P(A \cap C^c | B)$$

$$\approx P(A|B, C) + P(C|B)$$

event C 가 어떻게 생겼는지
 → event C 가 임의의 값이냐 아니냐에 따라
 $P(C)$ 는 23.1
 왜 $\frac{1}{2}$ 가냐
 계산 $P(B)$ 가
 23.1

$$\frac{P(A \cap B, C)}{P(B)} \approx \frac{P(A|B, C) P(C|B)}{P(B)}$$

$\frac{2}{6}$ 일 = 23.1% 이



기분 좋은 사람

i) conditioned $d_{i,j}(u), d_{i,u}(u), \dots, d_{i,i}(u)$

of shortcuts from $\bar{u}(u) \sim \text{Bin}(k d_{i,u}(u), p)$

Let such a random variable be T_i .

ii) Let $d_{i,u}(u)$ etc. \Rightarrow Balls and Bins \Rightarrow $\frac{1}{2} \ln \frac{1}{1-p}$

$d_{i,u}(u)$: # of occupied bins, among $(1 - d_{i,u}(u) - \dots - d_{i,i}(u))$ after throwing T_i balls

→ ~~observed~~ ~~inequality~~

Conditioned on $T_i, T_{i,u}(u), \dots, T_{i,i}(u)$

$$\Rightarrow \Pr(d_{i,u}(u) = \frac{d_{i,u}(u)}{2} \mid T_i, T_{i,u}(u), \dots, T_{i,i}(u)) \leq \exp\left(-\frac{\chi^2}{2T}\right)$$

↓ Azuma inequality

$d_{i,u}(u)$ 의 $\frac{1}{2}$ 가냐

Let the event $C = \{T \in \mathcal{I}\}$, $\mathcal{I} = [k_p(d_{in}(u) - \frac{\epsilon}{2}), k_p(d_{in}(u) + \frac{\epsilon}{2})]$

$\Pr(C) \rightarrow \frac{\text{weight}(C)}{\text{weight}}$ $E(C) = k_p d_{in}(u)$

$\Pr(\bar{C} | \Sigma_1(u), \dots, \Sigma_{in}(u)) \rightarrow 0$ em. C is event $T \in \mathcal{I}$ is not

$\leq \Pr(T \notin \mathcal{I} | \Sigma_1(u), \dots, \Sigma_{in}(u))$

$+ \Pr(\bar{C} | \Sigma_1(u), \dots, \Sigma_{in}(u), T \in \mathcal{I})$

$\Rightarrow \Pr(T \notin \mathcal{I} | \Sigma_1(u), \dots, \Sigma_{in}(u))$ ①

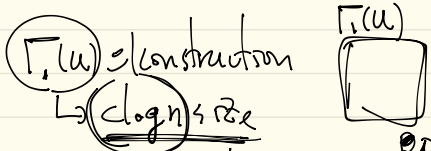
$\Pr(\frac{d_{in}(u)}{d_{in}(u)} \geq k_p(1+\epsilon) | \Sigma_1(u), \dots, \Sigma_{in}(u), T \in \mathcal{I})$ ②

$\Pr(\frac{d_{in}(u)}{d_{in}(u)} \leq k_p(1-\epsilon) | \Sigma_1(u), \dots, \Sigma_{in}(u), T \in \mathcal{I})$ ③

(Q) em. $\frac{d_{in}(u)}{d_{in}(u)} \geq k_p(1+\epsilon)$ $\Rightarrow T \in \mathcal{I}$ event C is not satisfied
 em. $\frac{d_{in}(u)}{d_{in}(u)} \leq k_p(1-\epsilon)$ $\Rightarrow T \in \mathcal{I}$ event C is not satisfied

① & ② are events that are not satisfied

① $T \sim \text{Bin}(k_p d_{in}(u), p)$



$|T(u)| = \log n$ $\Pr(k_p(1-\epsilon) > 1)$ conditioned on $\Sigma_1(u)$

$d_{in}(u) = |T(u)| \cdot (C \log n)$ $\frac{d_{in}(u)}{d_{in}(u)} \leq k_p(1-\epsilon) \Rightarrow k_p(1-\epsilon) > 1$

① $\leq 2 \exp(-(\log n)^k)$ $\rightarrow C$ is chosen such that n^k for any desired $k > 0 \Rightarrow$ check! Homework 2.5

Lemma 2.4

(2) We choose $T \in \mathbb{I}$, such that $\exp\left(-\frac{\eta^2}{2T}\right)$ is maximized
 i.e., $\sup_{T \in \mathbb{I}} \exp\left(-\frac{\eta^2}{2T}\right)$

$\Pr(d_i(u) \geq k_p(1+\epsilon)) \cdot d_{i+1}(u)$
 $\Pr(d_i(u) - \bar{d}_i(u) \geq \eta)$

$\bar{d}_i(u) = (1 - d_i(u) - d_{i+1}(u) - \dots - d_{i+k}(u)) \cdot \left(1 - \left(1 - \frac{1}{q_i}\right)^T\right)$

$\eta = d_{i+1}(u)k_p(1+\epsilon) - d_{i+1}(u)$

$\Pr(d_i(u) - \bar{d}_i(u) \geq d_{i+1}(u)k_p(1+\epsilon) - d_{i+1}(u) \mid T \in \mathbb{I}, \epsilon_1(u), \dots, \epsilon_0(u))$

Using Chernoff condition (6.2) $k_p(1+\epsilon) < (k_p(1+\epsilon))^2$
 $\Rightarrow d_{i+1}(u) = o(m)$

with (9) $\Rightarrow d_{i+1}(u)k_p(1+\epsilon) - (1+o(m))T$

Chernoff inequality \Rightarrow Chernoff (2) $\leq \exp(-\epsilon' / (C \log n))$ for some constant $\epsilon' > 0$

Homework 6 \Rightarrow Chernoff density \Rightarrow Chernoff

$\leq n^{-k}$, for any desired $k > 0$.