The Probabilistic Method - Probabilistic Techniques

Lecture 7: "The Janson Inequality"

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- 1. The Lovász Local Lemma
- 2. Example Diagonal Ramsey Numbers

- 1) On the importance of stochastic independence
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1) On the importance of stochastic independence

- The local Lemma demonstrates that **rare dependencies** yield results similar to the case of stochastic independence.
- The Janson inequality actually does the same, but for the case when the total amount of dependencies is rather small.

Intuition

1 Let B_i be the undesired events.

2 non-trivial dependence \sim :

 $i \sim j \Leftrightarrow i \neq j$ and B_i, B_j dependent

B $\Delta = \sum_{i \sim j} \Pr\{B_i \land B_j\}$: measure of dependencies.

If the events were independent then the probability of the desired property is defined as follows:

$$\Pr\{\wedge \overline{B_i}\} = \prod_i \Pr\{\overline{B_i}\} = M$$

5 The Janson inequality shows that $Pr\{\wedge \overline{B_i}\}$ remains very close to M if the dependencies are small.

2) The Janson Inequality

Theorem 1

Let
$$B_i$$
 be undesired events. Define
1) $\Delta = \sum_{i \sim j} \Pr\{B_i \land B_j\}$ and
2) $M = \prod_i \Pr\{\overline{B_i}\}$
If $\Pr\{B_i\} \le \epsilon$ then $M \le \Pr\left\{\bigwedge_i \overline{B_i}\right\} \le M \cdot exp\left(\frac{1}{1-\epsilon} \cdot \frac{\Delta}{2}\right)$

<u>Remark:</u> If

a. ϵ is small (e.g. ϵ is constant or smaller i.e. undesired events are not very probable) and

b. Δ is small e.g. o(1) (i.e. there are small dependencies) Then,

$$e^{\left(\frac{1}{1-\epsilon}\cdot\frac{\Delta}{2}\right)} \to 1 \Rightarrow \Pr\left\{\bigwedge_{i}\overline{B_{i}}\right\} \simeq M = \prod_{i}\Pr\{\overline{B_{i}}\}$$

3) Example - Triangle-free sparse Random Graphs

Theorem 2

Consider the $G_{n,p}$ graph space. For every constant c,

If
$$p = \frac{c}{n} \Rightarrow \Pr\{ \not\exists K_3 \} \to e^{-c^3/6}$$

 $(G_{n,p} \text{ with } p = \frac{c}{n} \text{ is sparse because the connectivity threshold is } p = \frac{clogn}{n}$.) Proof:

- Let S be any fixed set of 3 vertices (|S| = 3).
- We define the event $B_S = \{S \text{ is } K_3 \text{ (triangle)}\}.$

•
$$\Pr\{B_S\} = p^3 = \frac{c^3}{n^3}$$

We want to prove that

$$\Pr\left\{\bigwedge_{S,|S|=3}\overline{B_S}\right\} \to e^{-c^3/6}$$

Intuition: If events B_S were independent, then:

$$M = \Pr\left\{\bigwedge_{S,|S|=3} \overline{B_S}\right\} = \prod_S \Pr\left\{\overline{B_S}\right\}$$
$$= (1 - p^3)^{\binom{n}{3}} \sim e^{-\frac{c^3}{n^3} \cdot \frac{n^3}{3!}} = e^{-c^3/6}$$

Since events are dependent, we will show that dependencies in this sparse graph are small, to get a very similar result, via applying the Janson inequality:

$$\epsilon = \Pr\{B_S\} = \frac{c^3}{n^3} \to 0$$
$$M = \prod_{S,|S|=3} \Pr\{\overline{B_S}\} = (1-p^3)^{\binom{n}{3}} \sim e^{-c^3/6}$$
$$\Delta = \sum_{S \sim T} \Pr\{B_S \wedge B_T\}$$

• Non-trivial dependence:

$$S \sim T \iff \left\{ \begin{array}{c} S \neq T \\ |S \wedge T| \geq 2 \end{array}
ight\} \iff |S \wedge T| = 2$$

•
$$\Pr\{B_S \wedge B_T\} = p^5$$

$$\begin{split} \Delta &= \binom{n}{3} \binom{3}{2} \binom{n-3}{1} p^5 \\ &= O(n^4 p^5) = O\left(n^4 \cdot \frac{c^5}{n^5}\right) = \Theta\left(\frac{1}{n}\right) = o(1) \end{split}$$

Since $\epsilon = o(1)$ and $\Delta = o(1)$ we apply the basic case of Janson inequality that gives us:

$$\lim_{n \to \infty} \Pr\left\{\bigwedge_{S} \overline{B_s}\right\} = \lim_{n \to \infty} M = e^{-c^3/6}$$

4) Example - Paths of length 3 in $G_{n,p}$

Theorem 3

Define the event $B = \{ \exists a \text{ path of length } 3 \text{ between } \underline{any \text{ pair}} \text{ of vertices} \\ in G_{n,p} \}.$ For every constant $c \geq 2$:

If
$$p = \left(\frac{c \cdot \ln n}{n^2}\right)^{1/3} \Rightarrow \Pr\{B\} \to 1$$

Proof:

We define the event:

• $B_{u,v} = \{ \not\exists a \text{ path of length } 3 \text{ between vertices } u \text{ and } v \}.$

$$\Pr\{\overline{B}\} = \Pr\left\{\bigcup_{u,v} B_{u,v}\right\}$$
$$\leq \sum_{u,v} \Pr\{B_{u,v}\} = O(n^2) \Pr\{B_{u,v}\}$$

In order to prove that $\Pr{\overline{B}} \to 0$ or, equivalently $\Pr{B} \to 1$, we must prove that

$$\Pr\{B_{u,v}\} = o(n^{-2})$$

We apply Janson Inequality to prove that $\Pr\{B_{u,v}\} = o(n^{-2})$.

We define the undesired event:

• $A_{w_1,w_2} = \{ \text{the edges } (u, w_1), (w_1, w_2), (w_2, v) \text{ exist} \}$

and express the event $B_{u,v}$ as follows:

$$\Rightarrow B_{u,v} = \left\{ \bigwedge_{w_1,w_2} \overline{A_{w_1,w_2}} \right\}$$

$$\epsilon = \Pr\left\{A_{w_1, w_2}\right\} = p^3 = \left(\left(\frac{c \cdot \ln n}{n^2}\right)^{1/3}\right)^3 \to 0$$

$$M = \prod_{w_1, w_2} \Pr\left\{\overline{A_{w_1, w_2}}\right\} = (1 - p^3)^{(n-2)(n-3)}$$
$$\leq e^{-p^3 n^2} = e^{-\frac{c \cdot \ln n}{n^2} \cdot n^2} = n^{-c} = o(n^{-2})$$

So, the result would hold if there were no dependencies.

$$\Delta = \sum_{(w_1, w_2) \sim (w'_1, w'_2)} \Pr\left\{A_{w_1, w_2} \land A_{w'_1, w'_2}\right\}$$



 $w_1 = w'_2$ $w_2 = w'_1$ case 3: Contribution to Δ :

 \blacksquare case 1:

$$\binom{n-2}{3} \cdot p^5$$

 \blacksquare case 2:



case 3:

 $\binom{n-2}{2} \cdot p^5$

$$\Rightarrow \Delta = O(n^3)p^5 + O(n^3)p^5 + O(n^2)p^5 = O(n^3)p^5 = O(n^3) \left(\frac{c \cdot \ln n}{n^2}\right)^{5/3} = O\left(\frac{(c \cdot \ln n)^{5/3}}{n^{1/3}}\right) = o(1)$$

Since $\epsilon = o(1)$ and $\Delta = o(1)$ we apply the basic case of Janson inequality that gives us:

$$\Rightarrow \lim_{n \to \infty} \Pr\left\{\bigwedge_{w_1, w_2} \overline{A_{w_1, w_2}}\right\} = \lim_{n \to \infty} M = o(n^{-2})$$