

W12 D1
13/1/15

Prob. Methods in Comb.

Reminder: Yanson's inequality:

$\{B_i\}_{i \in I}$ events of the form $A_i \subseteq R$, $P(\bigcap_{i \in I} \bar{B}_i) \approx e^{-p}$.

$X = \sum_{i \in I} \mathbb{1}_{B_i}$, then $X \sim \text{Pois}(p)$.

Bonferroni Inequalities

A_1, \dots, A_n events in prob. space. Given $I \subseteq [n]$

let $A_I = \bigcap_{i \in I} A_i$. Given $\epsilon \in \{0, 1\}^n$, $A_\epsilon = \bigcap_{i \in I} A_i^{e_i}$ where $A^0 = \bar{A}, A^1 = A$.

$P(A_{\{\epsilon_0, \dots, \epsilon_n\}}) = \sum_{r=0}^m \sum_{I \in \binom{[n]}{r}} (-1)^r P(A_I)$. Bonferroni - If $s \geq 0$,

$$\sum_{r=0}^{2s+1} (-1)^r P(A_{\{\epsilon_0, \dots, \epsilon_n\}}) \leq \sum_{r=0}^{2s+1} \sum_{I \in \binom{[n]}{r}} (-1)^r P(A_I).$$

Brun's Sieve

$n \rightarrow \infty$ in prob. space of $A_1(n), \dots, A_m(n)$ events in it.

$X(n) = \sum_{i=1}^m \mathbb{1}_{A_i(n)}$, $S^{(r)} = S^{(r)}(n) = \sum_{I \in \binom{[n]}{r}} P(A_I(n))$. Suppose $\exists p \text{ const.}$

s.t. ~~$\forall r \geq 1$~~ $\forall r \geq 1$, $\lim_{n \rightarrow \infty} S^{(r)}(n) = \frac{p^r}{r!}$. Then for every fixed $t \geq 0$,

$$\lim_{n \rightarrow \infty} P(X=t) = \frac{p^t}{t!} e^{-p} = P(\text{Pois}(p)=t).$$

pf. For $t=0$, fix $\delta > 0$. First choose S_0 s.t. $\forall s > S_0$,

$$\left| \sum_{r=0}^S \frac{p^r}{r!} (-1)^r - e^{-p} \right| < \delta/2. \text{ Choose } n_0 \text{ s.t. } \forall n > n_0, \left| S^{(r)}(n) - \frac{p^r}{r!} \right| < \frac{\delta}{2(S_0+2)}$$

a) $r \leq 2S_0 + 1$. By Bonferroni, $\sum_{r=0}^{2S_0+1} (-1)^r \frac{p^r}{r!} - \frac{\delta}{2} \leq \sum_{r=0}^{2S_0+1} S^{(r)}(n) \leq P(X=0) \leq$

$$\sum_{r=0}^{2S_0+1} (-1)^r S^{(r)}(n) \leq \sum_{r=0}^{2S_0+1} \frac{(-1)^r p^r}{r!} + \frac{\delta}{2} \leq e^{-p} + \delta.$$

For $t \neq 0$ fix $T \subseteq I$ of size t .

$P(\text{Only } A_i \text{ with } i \in T \text{ hold})$ can be estimated using the prev. methods using $X_T = \# \text{events } i \in T \text{ which hold}$, and summing over $|T|=t$ we

$$\text{get } \frac{p^t}{t!} e^{-p} \sum_{r=0}^t \frac{(-1)^{r-t}}{(r-t)!} \frac{p^{r-t}}{(r-t)!}$$

Example: Threshold func. for $\forall v \in \mathbb{N}$ copy of $K_{3,3}$ -

Let $p = p(n)$, $\nu = \nu(n)$ s.t. $e^{-p} = \frac{c}{n}$, $\nu = \binom{n-1}{2} c^3$ (so

$\nu \approx \log n$, $\nu \approx n^{-2/3} (\log n)^{1/3}$). Then $P(G_{n,p})$ has the property $\lambda = e^{-c}$.

pf. $G \sim G(n, p)$. $B_{xyz} = \{x, y, z \text{ induce a triangle}\}$. $A_x = \bigcap_{y, z \in x} \bar{B}_{xyz}$.

Apply Yanson with $\bar{p} = \mathbb{E}[\#\bar{B}_{xyz}]$ that holds $\bar{p}^3 \binom{n-1}{2} = p$ and

$\lambda = n^{-\frac{1}{3} + o(1)} = o(1)$, so $P(A_x) \leq e^{-p - o(1)} = (1 - o(1)) e^{-p}$, then and by

Harri's inequality / FKG $\mathbb{P}(A_x) \geq (1-p^3)^{\binom{n-x}{2}} \geq (1-o(1))e^{-x}$. Now put $X = \sum_{x \in V} A_x$. $\mathbb{E}[X] = S^{(r)} = (1-o(1))e^{-r} \cdot n = (1-o(1))n$. We want $\mathbb{P}(X=0) \rightarrow e^{-n}$ so

it's enough $\leq^{(1)} \rightarrow \frac{c^r}{r!}$. Let $H=r$, $A_r = \bigcap_{x \in V} A_x = \bigcap_{S \subseteq V, |S|=r} \overline{B}_S$, $\mu_r = \left[r \binom{n-1}{2} - o(n) \right] p^3$, $\Delta = O(p^5) \cdot r \cdot n^3 = o(1)$, so $\mathbb{P}(A_r) = e^{-\mu_r} \cdot e^{-\frac{\Delta}{2}}$. $S^{(r)} = \binom{n}{r} e^{-\mu_r} (1-o(1)) = \frac{n^r e^{-\mu_r}}{r!} (1-o(1)) \stackrel{\text{by FKG}}{=} \frac{c^r}{r!} (1-o(1))$.

Large Deviation Ineq. for sums of "Almost Indep." r.v.s

Ω fin. $R \subseteq \Omega$ randomly $\mathbb{P}(x \in R) = p_x$ indep. $B_i = "A_i \subseteq R"$.

$i \neq j \Leftrightarrow A_i \cap A_j \neq \emptyset$. We call $J \subseteq I$ a disjoint family if J is \sim -indep. \Rightarrow Maximal disjoint family if J can't be extended.

Thm. \forall integers, $\mu = \sum_{i \in I} \mathbb{P}(B_i)$, $\mathbb{P}(\exists |J|=s \text{ disj. family}) \leq \frac{\mu^s}{s!} \leq \left(\frac{e\mu}{s} \right)^s$. In particular, if $s = (e-1)\mu$, this $\mathbb{P} \leq (1-\rho(\delta))^{\mu}$.

Pf. $A_I = \{ J \subseteq I \mid J \text{ disj.} \}$ \sim -indep. $\mathbb{P}(A_I) = \prod_{J \in A_I} \mathbb{P}(J)$. $\mathbb{P}(\exists |J|=s \text{ disj.}) \leq \sum_{J \in A_I} \mathbb{P}(J) = \sum_{J \in A_I} \prod_{j \in J} \mathbb{P}(B_j) = \frac{1}{s!} \sum_{J \in A_I} \prod_{j \in J} \mathbb{P}(B_j) \dots \mathbb{P}(B_{js}) \leq \frac{1}{s!} \sum_{J \in A_I} \prod_{j \in J} \mathbb{P}(B_j) \dots \mathbb{P}(B_{js}) = \frac{1}{s!} \left(\sum_{j \in I} \mathbb{P}(B_j) \right)^s = \frac{\mu^s}{s!}$.

Illustration - off-diag. Ramsey nums. (Krivelevich)

Fix $l \geq 3$. $R(l, k) = \max \{ n \mid \exists G \text{ on } n \text{ vxs. } G \not\cong K_l, \chi(G) < k \}$.

Prop. $R(l, k) \geq C_0 \left(\frac{k}{\log k} \right)^{\frac{l+1}{2}}$ for some $C_0 > 0$.

For n above, $\forall l \leq k$ $p = c' n^{-\frac{2}{l+1}}$, $n^l p^{\binom{l}{2}} = \Theta(n^2 p^k)$. For $S \subseteq V(G)$

$X_S = e(G[S])$, $Y_S = \# \text{ copies in } G(n, p) \text{ with } \geq 2 \text{ vxs. in } S$, $Z_S = \# \text{ max. edge disjoint such copies}$. $A_S = "X_S \geq \binom{l}{2} Z_S"$.

If $\bigwedge_{|S|=k} A_S$ holds, $G(n, p)$ witnesses $R(k, l) \geq n$. Let H be a family in the def. of Z_S . Delete from G edges in those cliques-

$e_G(S) \geq X_S - \binom{l}{2} Z_S > 0$ for any $S \in \binom{[k]}{2}$ in the resulting graph, and it's also K_l -free. Now, for fixed $S \in \binom{[k]}{2}$, $\mathbb{P}(A_S) \leq \mathbb{P}(X_S \geq \mathbb{E} X_S / 2) +$

$\mathbb{P}\left(\binom{l}{2} Z_S \geq \mathbb{E} X_S / 2\right)$, $\mathbb{E}[X_S] = \binom{k}{2} p$, so $\leftarrow e^{-\frac{(k-1)p}{8}}$. For 2nd part, large dev.

$\mathbb{E}(Z_S) \leq \mathbb{E}(Y_S) = \mu_S \leq \binom{k}{2} (l-2) p^{\binom{l}{2}}$, $\mu_S \geq \binom{k}{2} (l-2) p^{\binom{l}{2}}$ but $k = o(n)$ so they're both $\Theta\left(\binom{k}{2} n^{l-2} p^{\binom{l}{2}}\right)$. By lemma $\mathbb{P}(Z_S \geq 5\mu_S) \leq \left(\frac{e}{5}\right)^{5\mu_S}$. So,

$$\frac{C_\ell}{\binom{n}{\ell} p^{\binom{n}{\ell}}} \leq \frac{\mathbb{E}[X_s]}{\mu_s} \leq \frac{\binom{k}{\ell} p}{\binom{n}{\ell} p^{\binom{n}{\ell}}} = \frac{1}{C_\ell (c)^{\binom{n}{\ell}}} \text{ for small enough } C_\ell > 0. \text{ For some}$$

small enough c' , $\mathbb{E}[X_s] \leq \frac{\mathbb{E}[X_s]}{\mu_s} \leq D = \Theta(\ell)$. With this choice of c' , $\mathbb{P}\left(\binom{n}{\ell} Z_s \geq \mathbb{E}X_s/\varepsilon\right) \leq \mathbb{P}(Z_s \geq 5\mu_s) \leq (\varepsilon/5)^{\frac{\mathbb{E}[X_s]}{D}} = e^{-\varepsilon \binom{k}{\ell} p}$.

Now that we know $\mathbb{P}(\bar{A}_s) \leq 2e^{-\varepsilon \binom{k}{\ell} p}$. So, $\mathbb{P}\left(\bigcap_{s \in k} \bar{A}_s\right) \geq 1 - \sum_{s \in k} \mathbb{P}(\bar{A}_s) \geq 1 - \binom{n}{k} e^{-\varepsilon \binom{k}{\ell} p} \geq 1 - \left(\frac{en}{k} e^{-\varepsilon \frac{k-\ell}{2} p}\right)^k$. (2) follows from $p k \geq \frac{2\ell}{\varepsilon} \log k \Leftrightarrow p \geq \frac{2\ell}{\varepsilon k} \log \frac{k}{\ell} = \frac{2\ell}{\varepsilon k} \left(\frac{n}{\ell}\right)^{\frac{2}{\ell-1}} = \frac{2\ell}{\varepsilon c^{\frac{2}{\ell-1}}} \frac{p}{c}$ and we can just choose c sufficiently small.

Pois. concentration of measure - $\mathbb{P}(P \leq (1-\varepsilon)p) \leq e^{\varepsilon^2 p/2}$, $\mathbb{P}(P \geq (1+\varepsilon)p) \leq [e^{\varepsilon(1+\varepsilon)^{-\frac{1}{1-\varepsilon}}}]^N$. Define $s \geq 0$, $\mu_s = \min \left\{ \sum_{i \in J} \mathbb{P}(B_i) \mid |J|=s, J \subseteq I \right\}$.

Letting $v = \max_{i \in I} \sum_{j \in J, i \in J} \mathbb{P}(B_j)$, then $\mu_s \geq N - sv$.

Lemma $\mathbb{P}(\max_{i \in I} \text{disj. family of sizes}) \leq \frac{\mu_s^s}{s!} e^{-\mu_s + A/2}$.