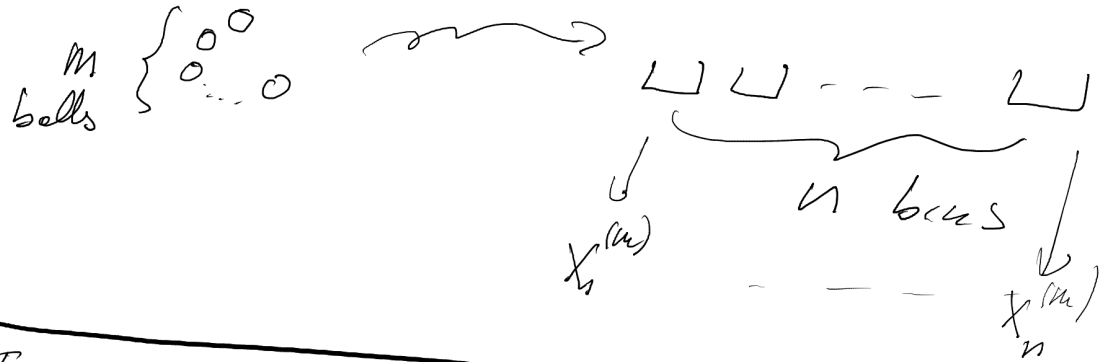


Balls & bins

Recall: we have m balls
 each is thrown unif. at random
 into one of n bins

$X_i^{(m)}$:= # of balls in the i -th bin
 ($i = 1, \dots, n$)



E.g.: $n=365$, \mathbb{Q} : $\max_{i \leq 365} X_i^{(m)} \geq 2$? ... birthday paradox
 E.g.: analyze bucket sort

Lemma $m=n \Rightarrow \Pr\left[\exists i: X_i^{(m)} > \frac{3 \ln n}{\ln \ln n}\right] < \frac{1}{n}$ (n suff. large)

Proof $\Pr[X_1^{(m)} \geq M] \leq \binom{n}{M} \frac{1}{n^M} \leq \frac{1}{M!} \leq \left(\frac{e}{M}\right)^M$ (OR Chernoff!) \rightarrow max is large

$$\Pr[\exists i \dots] \leq n \left(\frac{e}{M}\right)^M \leq \dots \leq \frac{1}{n}$$

□

Lemma $m=n \Rightarrow \Pr\left[\max X_i^{(m)} < \frac{\ln n}{\ln \ln n}\right] \leq \frac{1}{n}$ (n suff. large)

Proof The rest of this class.

Reason?

Poisson distribution

$$Pr[X_1^{(m)} = r] = \binom{m}{r} \frac{1}{n^r} \left(1 - \frac{1}{n}\right)^{m-r} \stackrel{r \text{ pevné, } n, m \gg 1}{=} \frac{m^r}{r!} \cdot \frac{1}{n^r} \left(1 - \frac{1}{n}\right)^m \stackrel{m \rightarrow \infty}{=} \frac{1}{r!} \left(\frac{m}{n}\right)^r e^{-\frac{m}{n}}$$

Def Discrete Poisson rand. var. X with par. μ ($X \sim \text{Pois}(\mu)$) i.s. def
 r.v. that sat. $Pr[X = k] = \frac{e^{-\mu} \mu^k}{k!}$ ($k = 0, 1, 2, \dots$)

1) Def. is correct

2) $E[X] = \mu$

3) $\text{Var } X = \mu$ (indep. r.v.)

4) **Lemma** $X \sim \text{Pois}(\lambda), Y \sim \text{Pois}(\mu) \Rightarrow X + Y \sim \text{Pois}(\lambda + \mu)$ (also for finitely many terms)

PF $Pr[X + Y = n] = \sum_{k=0}^n Pr[X = k] \cdot Pr[Y = n - k] = \sum_k e^{-\lambda} \frac{\lambda^k}{k!} \cdot e^{-\mu} \frac{\mu^{n-k}}{(n-k)!} = \frac{e^{-(\lambda+\mu)}}{n!} \sum_k \binom{n}{k} \lambda^k \mu^{n-k} \square$

5) **Theorem** (Limit of Bin. distr.) $X_n \sim \text{Bin}(n, \lambda/n)$; $\lim_{n \rightarrow \infty} \lambda = \lambda \in \mathbb{R}, k$ fixed
 $\Leftrightarrow \lim_{n \rightarrow \infty} Pr[X_n = k] = e^{-\lambda} \frac{\lambda^k}{k!} \Rightarrow$ Applications

Poisson approximation

$X_1^{(m)}, \dots, X_n^{(m)}$ — each $\overset{\text{approx.}}{\sim} \text{Pois}(\frac{m}{n})$
 — not independent! (as $\sum X_i^{(m)} = m$)

$Y_1^{(m)}, \dots, Y_n^{(m)}$ — indep. r.v.'s, each $\sim \text{Pois}(\frac{m}{n})$

WANT: approximate $(X_1^{(m)}, \dots, X_n^{(m)})$ by $(Y_1^{(m)}, \dots, Y_n^{(m)}) \longrightarrow Y$

Thm $\#k, m, n$: distribution of $(Y_1^{(m)}, \dots, Y_n^{(m)})$ conditioned on $\sum_{i=1}^n Y_i^{(m)} = k$
 is the same as $(X_1^{(k)}, \dots, X_n^{(k)}) \sim \text{Pois}(m)$

Proof Just count prob (k_1, \dots, k_n) s.t. $\sum_{i=1}^n k_i = k$

$$P_r[X=k] = \binom{k}{k_1, \dots, k_n} \cdot \frac{1}{n^k}$$

$$P_r[Y=k] = \prod_{i=1}^n \left(e^{-\frac{m}{n}} \cdot \frac{(\frac{m}{n})^{k_i}}{k_i!} \right) = e^{-m} \cdot \frac{(\frac{m}{n})^k}{k!} \cdot \binom{k}{k_1, \dots, k_n}$$

$$P_r[\sum Y_i = k] = P_r[\text{Pois}(m) = k] = \frac{e^{-m} m^k}{k!}$$

$$P_r[Y=k \mid \sum Y_i = k] = \frac{P_r[Y=k]}{P_r[\sum_{i=1}^n Y_i = k]} = P_r[X=k]$$

Thm $f(x_1, \dots, x_n) \geq 0$ — any f from

$$\mathbb{E}[f(X_1^{(\mu)}, \dots, X_n^{(\mu)})] \leq e\sqrt{\mu} \mathbb{E}[f(Y_1^{(\mu)}, \dots, Y_n^{(\mu)})]$$

Proof $\mathbb{E}[f(Y)] = \sum_{k=0}^{\infty} \mathbb{E}[f(Y) | \sum Y_i^{(\mu)} = k] \cdot \mathbb{P}_0[\sum Y_i^{(\mu)} = k]$

$$\geq \mathbb{E}[f(Y) | \sum Y_i^{(\mu)} = m] \cdot \mathbb{P}_0[\sum_{i=1}^n Y_i^{(\mu)} = m]$$

$$= \mathbb{E}[f(X)] \cdot \mathbb{P}_0[\text{Pois}(\mu) = m]$$

$$= e^{-\mu} \frac{\mu^m}{m!} > \frac{1}{e\sqrt{\mu}} \quad \text{--- Strategy}$$

IF IN ADDITION $\mathbb{E}[f(X_1^{(\mu)}, \dots)]$ is monotone in μ

--- THEN $\mathbb{E}[f(X_1^{(\mu)}, \dots)] \leq 2 \mathbb{E}[f(Y_1^{(\mu)}, \dots)]$

For $\mathbb{E}(f(X))$ increasing in μ

... "same proof"

$$\mathbb{E}[f(Y)] \geq \sum_{k \geq m} \underbrace{\mathbb{E}[f(Y) | \sum Y_i = k]}_{\mathbb{E}[f(X^{(k)})]} \cdot \underbrace{\mathbb{P}_0[\sum Y_i = k]}_{\geq \frac{1}{2}} \geq \mathbb{E}[f(X^{(m)})] \cdot \underbrace{\mathbb{P}_0[\sum Y_i \geq m]}_{\geq \frac{1}{2}}$$

$$\mathbb{E}[f(X^{(k)})]$$

$$\mathbb{P}[\text{Pois}(\mu) \geq m] \geq \frac{1}{2} \quad \text{--- } \textcircled{2}$$

X — exact case

Y — Poisson case

Corollary Any event that takes place with probability p in the Poisson case takes place with probability $\leq pe^{tm}$ in the exact case.

Proof

I — indic. func. of the event $\begin{cases} 1 & \text{— event occurs} \\ 0 & \end{cases}$ OR $\leq p$ for monotone I

$$\text{prob. in the Poisson case} = E(I(Y)) = p$$

$$\text{prob. in the exact case} = E(I(X)) \leq e^{tm} E(I(Y)) = pe^{tm}$$



Thm $\Pr \left[\max_{i \leq n} X_i^{(n)} < \frac{\ln n}{\ln \ln n} \right] < \frac{1}{4}$ (n suff. large)

($m = n$)

? $\mathbb{E} F(X_1, \dots)$
monot. in n ?

Proof $M := \frac{\ln n}{\ln \ln n} \approx \frac{m}{n}$

Poisson case: $\Pr \{ Y_1^{(n)} \geq M \} \geq \Pr \{ \text{Pois}(M) = M \} = \frac{1}{e \cdot M!}$

$$\Pr \left[\max_i Y_i^{(n)} < M \right] \leq \left(1 - \frac{1}{eM!} \right)^n \leq e^{-\frac{n}{eM!}}$$

$P(X_1, \dots, X_n) = 1$
 \Downarrow
 $\max \{X_1, \dots, X_n\} < M$

pick M s.t. $M! \leq \frac{n}{2e \ln n} \Rightarrow \text{Prob. in Poisson} \leq \frac{1}{n^2}$

\Rightarrow Prob. in exact case $\leq \frac{e\sqrt{n}}{n^2} \leq \frac{1}{4}$

$$M! \leq e\sqrt{M} \cdot \left(\frac{M}{e}\right)^M \leq M \left(\frac{M}{e}\right)^M$$

$$\ln M! \leq M \ln M - M + \ln M$$

$$= \frac{\ln n}{\ln \ln n} (\ln \ln n - \ln \ln \ln n) - \frac{\ln n}{\ln \ln n} + (\ln \ln n - \ln \ln \ln n)$$

$$\leq \ln n - \frac{\ln n}{\ln \ln n}$$

$$\leq \ln n - \ln \ln n - \ln(2e)$$

~~$-\ln n, \ln \ln \ln n + \ln \ln n$~~
 ≤ 0
 $\forall \ln n > \ln \ln n$ \square

Power of 2 choices

So, max load is $\Theta\left(\frac{\ln n}{\ln \ln n}\right)$ w.h.p

----- but average is just 1 ball per bin!

Different model: if 2 bins chosen rand. at random

We put ball in the one that is less full

Thus n balls, n bins, rule \uparrow

$$\Pr\left[\max_i X_i^{(n)} \geq \frac{\ln n}{\ln \ln n} + \alpha(n)\right] = o\left(\frac{1}{n}\right)$$

Application: hashing

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$f: U \rightarrow \{0, 1, \dots, n-1\}$ "random" hashing f from
representation of a subset of U :

linked lists $L_0 \rightarrow \dots \rightarrow L_{n-1}$

add x : add x to $L_{f(x)}$ ----- time $O(1)$

find x : linear search in $L_{f(x)}$ ----- time $O(|L_{f(x)}|) + O(1)$

set of size m ----- m balls
 n bins ----- $|L_i| = X_i^{(m)}$

$$E[\text{time to find}] = \frac{m}{n} + O(1)$$

$(m \ll n)$ max. time to find ----- $O\left(\frac{\ln n}{\ln \ln n}\right)$ whp