

Today

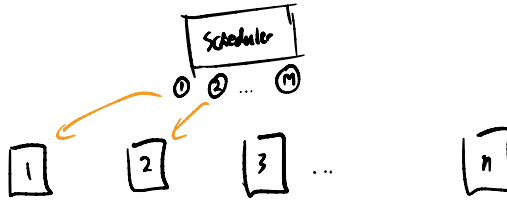
1) 3 common setups $\left\{ \begin{array}{l} \text{Random Load Balancing} \\ \text{Coupon Collector} \\ \text{Birthday Paradox} \end{array} \right\} \rightarrow \text{Balls} \rightarrow \text{Bins}$

2) Concentration Bound Paradigm

3) Solving (1) w/ (2)

3 Scenarios

1) Random load balancing



Assume schedule jobs uniformly at random.

What's MAX # jobs on a machine?

2) Coupons: 1, 2, ..., n



Assume each box contains a uniformly random coupon.

How many cereal boxes until collect all coupons
w/ good probability?

3) Birthday Paradox

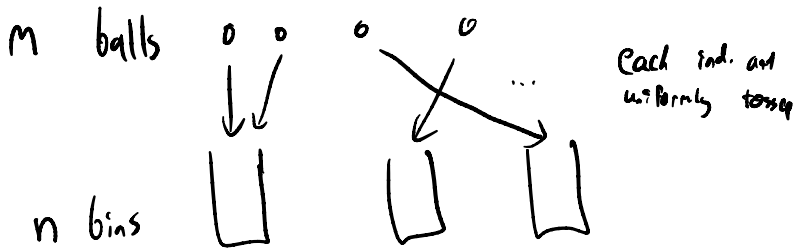


Assume bdays uniform

How many people until 2 people share
a bday w/ good probability?

Balls into Bins

Really 3 Questions about 1 Scenario



1) Max # balls in a bin (assuming $m=n$)

$$\lesssim \ln n \quad \text{w/ Pr} \geq 1 - \frac{1}{n}$$

2) How big must m be until ≥ 1 ball in all bins?

$$m \gtrsim n \cdot \ln n \quad \text{w/ Pr} \geq 1 - \frac{1}{n}$$

3) How big must m be until ≥ 2 balls in some bin?

$$m \gtrsim \sqrt{n} \quad \text{gives } \Omega(1) \text{ probability}$$

How to Solve Your Favorite Randomization Question

To show fact (*)

a) Show (*) true if all RVs near \mathbb{E}

b) Concentration: each RV at $\mathbb{E} (\pm \log n)$ w/ good probability

c) Union bound: all RVs

Let $X_{ij} :=$ ball $j \rightsquigarrow$ bin i

Let $X_i := \sum_j X_{ij}$ (# balls in bin i)

1) $\mathbb{E}[X_i] = 1$ (when $m=n$)

2) $\mathbb{E}[X_i] = \Omega(\log n)$ (when $m \geq \Omega(n \cdot \log n)$)

3) Call first $\frac{1}{2}$ of balls red, last $\frac{1}{2}$ balls blue

Let $Y_j := \begin{cases} 1 & \text{if } j\text{th ball into a bin w/ a red ball} \\ 0 & \text{o/w} \end{cases}$

Let $Y := \sum_j Y_j$

Let $R := \#$ bins w/ ≥ 2 red

$$\Pr(\geq 2 \text{ balls in a bin}) \geq \Pr(R \cup Y \geq 1) = \Pr(R) + \Pr(\bar{R}) \cdot \Pr(Y \geq 1 | \bar{R})$$

$\geq \Pr(Y \geq 1 | \bar{R})$ ↳ cuts large



But $\mathbb{E}[Y | \bar{R}] = \sum_j \mathbb{E}[Y_j | \bar{R}] = \frac{m}{2} \cdot \frac{n/2}{n} = \Omega(1)$ if $m \geq \sqrt{n}$

An OK Fact: Gaussians are "Concentrated"

Let $Z \sim N(0,1)$, then $\Pr(Z \geq t) \leq 2 \cdot \exp(-t^2/2) \quad \forall t \geq 0$

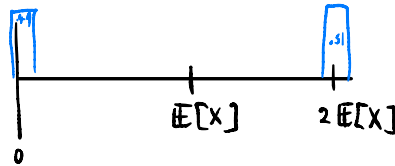
↳ will get similar bounds for general class of RVs

Markov's Inequality

Suppose X is a non-negative RV w/ $\mathbb{E}[X] > 0$

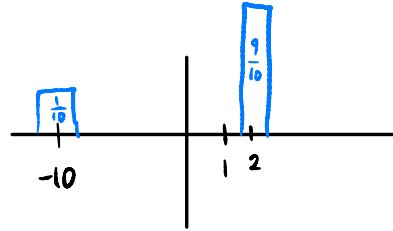
$$\Pr(X \geq a \cdot \mathbb{E}[X]) \leq \frac{1}{a} \quad \rightarrow \text{at least } \frac{1}{a} \text{ of the time} \quad \forall a \geq 1$$

Intuition



Why Non-Negative?

$$\mathbb{E}[X] = \frac{4}{5}$$



$$\begin{aligned} a \cdot \mathbb{E}[X] \cdot \Pr(X \geq a \cdot \mathbb{E}[X]) &\leq a \cdot \mathbb{E}[X] \sum_{i \geq a \cdot \mathbb{E}[X]} \Pr(X=i) \\ &\leq \sum_{i \geq a \cdot \mathbb{E}[X]} i \cdot \Pr(X=i) \\ &\leq \mathbb{E}[X] \end{aligned}$$

(non-neg)

Aside: A function $f: \mathbb{R} \rightarrow \mathbb{R}$ is non-decreasing if $a \geq b \rightarrow f(a) \geq f(b)$
For RV X have $\Pr(X \geq a) \leq \Pr(f(X) \geq f(a))$ for non-dec. f

Corollary: Chebyshev's Inequality

$$\Pr(|X - \mathbb{E}[X]| \geq a) \leq \frac{\text{Var}(X)}{a^2}$$

Let $Y = (X - \mathbb{E}[X])^2$ so $\mathbb{E}[Y] = \text{Var}(X)$

$$\begin{aligned} \Pr(|X - \mathbb{E}[X]| \geq a) &\stackrel{\text{non-dec.}}{\leq} \Pr(Y \geq a^2) \\ &= \Pr\left(Y \geq \frac{a^2 \cdot \mathbb{E}[Y]}{\mathbb{E}[Y]}\right) \\ &\leq \frac{\mathbb{E}[Y]}{a^2} = \frac{\text{Var}(X)}{a^2} \quad (\text{Markov}) \end{aligned}$$

All skipped in class

Chernoff Bound

Let X_1, X_2, \dots, X_n be independent RVs s.t. $X_i = \begin{cases} 1 & w/1r. p \\ 0 & o/w \end{cases}$

Let $X := \sum_i X_i$, $\mu := \mathbb{E}[X]$

Then $\forall \delta \geq 0$, $\Pr(X \geq (1+\delta) \cdot \mu) \leq \left(\frac{\exp(\delta)}{(1+\delta)^{1+\delta}} \right)^\mu$

$(1+\delta)^{1+\delta} = (1+\delta) \cdot (1+\delta)^\delta$

Chernoff Simplified

$$\Pr(X \geq (1+\delta) \cdot \mu) \leq \exp(-\mu \delta^2 / (2+\delta)) \quad \forall \delta \geq 0$$

$$\text{Have } \frac{2\delta}{2+\delta} \leq \log(1+\delta) \quad \forall \delta \geq 0 \iff 2\delta \leq \underbrace{(1+\delta) \cdot \log \delta + 2}_{\text{convex and LHS tangent at } \delta=1 \text{ to RHS}} \quad \forall \delta \geq 1$$

$$\begin{aligned} \text{So } \frac{\exp(\delta)}{(1+\delta)^{1+\delta}} &= \frac{\exp(\delta)}{\exp((1+\delta) \cdot \log(1+\delta))} \leq \frac{\exp(\delta)}{\exp((1+\delta) \cdot \frac{2\delta}{2+\delta})} \\ &= \exp\left(\delta - (1+\delta) \frac{2\delta}{2+\delta}\right) = \exp\left(-\frac{2\delta^2}{2+\delta}\right) \end{aligned}$$

\rightarrow If $\mu \cdot \delta \approx C \cdot \log n$, get at most $\approx \frac{1}{n^C} \rightarrow$ "with high probability"

Lower Tail

$$\Pr(X \leq (1-\delta) \cdot \mu) \leq \exp(-\delta^2 \mu / 2) \quad \forall \delta \in (0, 1)$$

Only did simplified for class

Solving Problems w/ Chernoff + Union Bound

1) Fix bin i . independent

$$X_i = \sum_j X_{ij}$$

$\begin{cases} 1 & \text{w/ Pr } \frac{1}{n} \\ 0 & \text{o/w} \end{cases}$

$$\mu = \mathbb{E}[X_i] = 1$$

$$\begin{aligned}
 \Pr(X_i \geq \mu(1 + \underbrace{2}_{\delta} \log n)) &\leq \exp(-4 \log^2 n / (2 + \log n)) \\
 &\leq \exp(-4 \log^2 n / 2 \cdot \log n) \\
 &= n^{-2}
 \end{aligned}$$

→ So $X_i \geq 1 + 2 \cdot \log n$ w/ $\Pr \leq n^{-2}$

$$\Pr(X_1 \geq 1 + 2 \log n \cup X_2 \geq 1 + 2 \log n \cup \dots) \leq \sum_i \Pr(X_i \geq 3 \cdot \log n)$$

\uparrow
 UB
 $\leq \frac{1}{n}$

missed

$\mathbb{E}[X_i] = O(1)$ so $X_i \leq \log n$ w/ $\Pr \leq n^{-2}$ so all $X_i \leq \log n$

2) Let $M = 16n \cdot \ln n$, Fix i .

$$\mu = \mathbb{E}[X_i] = 16 \ln n$$

$$\begin{aligned}
 \Pr(X_i \leq (1 - \underbrace{\frac{1}{2}}_{\delta}) \cdot 16 \ln n) &\leq \exp(-\frac{1}{4} 16 \ln n / 2) \\
 &= n^{-2}
 \end{aligned}$$

$$\Pr(X_1 \leq 8 \ln n \cup X_2 \leq 8 \ln n \cup \dots) \leq \frac{1}{n}$$

$\mathbb{E}[X_i] = \Theta(\log n)$
 so $X_i = \Theta(\log n)$ w/ $\Pr \leq n^{-2}$
 so all $X_i = \Theta(\log n)$ w/ $\Pr \leq n^{-2}$

3) For blue ball j

skipped

$$\text{Let } M = 8\sqrt{n}$$

Condition on \bar{R}

$$Y_j = \begin{cases} 1 & \text{w/ pr } \frac{M/2}{n} = \frac{4}{\sqrt{n}} \\ 0 & \text{o/w} \end{cases}$$

All independent

$$\text{Have } \mu = \mathbb{E}[Y | \bar{R}] = \frac{M}{2} \cdot \frac{4}{\sqrt{n}} = 16$$

$$\Pr(Y=0 | \bar{R}) \leq \Pr(Y \leq (1 - \frac{1}{2}) \cdot 16 | \bar{R})$$

$$\leq \exp\left(-\frac{1}{4} \cdot 16/2\right)$$

$$\leq \frac{1}{e^2}$$

$$\Pr(\geq 2 \text{ balls in bin}) \geq \Pr(Y \geq 1 | \bar{R}) = 1 - \frac{1}{e^2}$$

$\mathbb{E}[Y | \bar{R}] = O(1)$ so $Y \geq 1$ w/ $\Omega(1)$ probability

Hoeffding's Inequality

Let X_1, X_2, \dots, X_n be independent RVs s.t. $X_i \in [a_i, b_i]$

Let $X := \sum_i X_i$, $\mu := \mathbb{E}[X]$

$$\Pr(|X - \mu| \geq t) \leq 2 \cdot \exp\left(-\frac{2t^2}{\sum_i (b_i - a_i)^2}\right)$$

→ Additive

→ Doesn't assume $X_i \in \{0, 1\}$

Skipped

Chernoff Proof

← sketched in class

Let $s = \ln(1+\delta)$ and $a = (1+\delta)u$

$$\Pr(X \geq a) \leq \Pr(e^{sX} \geq e^{sa}) \leq \frac{\mathbb{E}[e^{sX}]}{e^{sa}} = \frac{\mathbb{E}[\prod_i e^{sX_i}]}{e^{sa}}$$

↑
monotone
↑
Markov
↑
defn.

$$= \frac{\prod_i \mathbb{E}[e^{sX_i}]}{e^{sa}} = \frac{\prod_i p_i e^s + (1-p_i)}{e^{sa}}$$

↑
independence

$$e^{sX_i} = \begin{cases} e^s & \text{w/ Pr } p_i \\ 1 & \text{w/ Pr } 1-p_i \end{cases}$$

$$= \prod_i \frac{1 + p_i(e^s - 1)}{e^{sa}} \leq \prod_i \frac{\exp(p_i(e^s - 1))}{e^{sa}}$$

↑
 $1+x \leq \exp(x)$

$$= \frac{\exp(\sum_i p_i(e^s - 1))}{e^{sa}} = \frac{\exp(u(e^s - 1))}{e^{sa}}$$

$$= \left(\frac{\exp(\delta)}{(1+\delta)^{1+\delta}} \right)^u$$

$s = \ln(1+\delta)$

$a = (1+\delta)u$