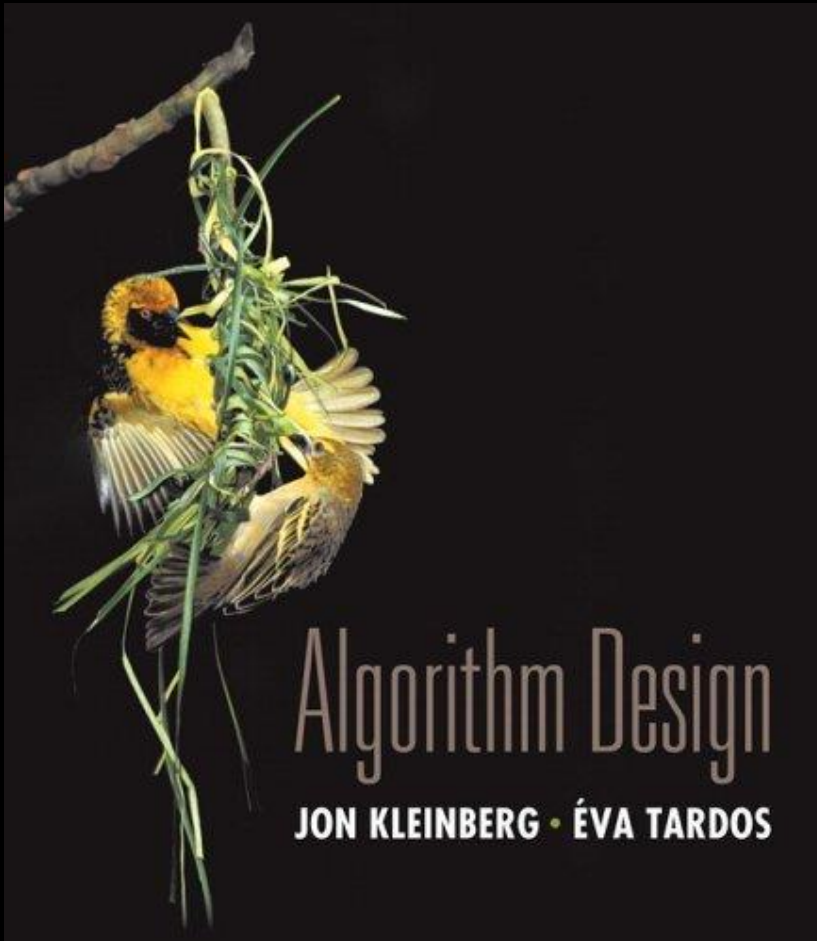


Chapter 13

Randomized Algorithms



Slides by Kevin Wayne.
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Desigualdades de Cauda

- Como gerar limites superiores para a probabilidade de uma variável aleatória se afastar da média?

Ferramenta fundamental para caracterizar o tempo de execução e/ou a probabilidade de sucesso de algoritmos aleatorizados

Desigualdade de Markov

Seja X uma V.A. que assume somente valores não negativos. Então, para todo t positivo,

$$\Pr[X \geq t] \leq \frac{E[X]}{t}$$

Prova: Considere a variável 0-1 Y que assume valor 1 se $X \geq t$ e 0, caso contrário. Note que $Y \leq X/t$. Logo,

$$\Pr[X \geq t] = E[Y] \text{ e } E[Y] \leq E[X]/t.$$

Portanto, $\Pr[X \geq t] \leq E[X]/t$

Desigualdade de Markov

Qual é a probabilidade obtermos mais de 75 caras ao jogar uma moeda justa 100 vezes ?

X : número de vezes que o resultado é cara

X_i :1 se o resultado da i -ésima tentativa é cara e 0 caso contrário.

$$E[X_i] = 1/2$$

$$E[X] = \sum_{i=1}^{100} E[X_i] = 50$$

Aplicando Markov temos

$$\Pr[X > 75] \leq 50/75 = 2/3$$

Variância de uma distribuição

A variância de uma variável aleatória X é definida como

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

- A variância mede o quanto a distribuição "foge" da média.
- O desvio padrão de X , σ_x , é a raiz quadrada da variância

Variância de uma distribuição

Temos que:

$$\begin{aligned}\text{Var}(X) &= E[(X - E[X])^2] = E[(X^2 - 2XE[X] + E[X]^2)] = \\ &E[(X^2 - XE[X] + E[X]^2)] = E[X^2] - E[X]^2\end{aligned}$$

Variância de uma distribuição

Def Duas variáveis aleatórias X e Y são independentes se para qualquer par de reais x, y

$$\Pr[X=x, Y=y] = \Pr[X=x]\Pr[Y=y]$$

Lema: Se X e Y são V.A. independentes, então

$$E[XY] = E[X]E[Y]$$

$$\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$$

Variância de uma distribuição

Definição. Uma variável binomial X com parâmetros n e p conta o número de caras em uma sequência de n lançamentos aonde a probabilidade cara ocorrer em um lançamento é p

Lema. O valor esperado de uma V.A. binomial X com parâmetros n e p é np .

Teorema. A variância de uma V.A. binomial X com parâmetros n e p é $np(1-p)$.

Desigualdade de Chebyshev

Seja X uma V.A. com desvio padrão σ_x . Então, para todo $t > 0$,

$$\Pr[|X - E[X]| \geq t \cdot \sigma_x] \leq \frac{1}{t^2}$$

Prova:

- $\Pr[|X - E[X]| \geq t \cdot \sigma_x] = \Pr[(X - E[X])^2 \geq t^2 \cdot \sigma_x^2]$
- Se $Y = (X - E[X])^2$, por Markov temos,
 $\Pr[(X - E[X])^2 \geq t^2 \text{Var}(X)] \leq \text{Var}(X) / (t^2 \text{Var}(X)) = 1 / t^2$

Quicksort: Expected Number of Comparisons

Theorem. Expected # of comparisons is $2n \ln n$.

Pf.

Theorem. [Knuth 1973] Stddev of number of comparisons is $\sim 0.65N$.

↑

Ex. If $n = 1$ million, the probability that randomized quicksort takes less than $4n \ln n$ comparisons is at least 99.4%.

Chebyshev's inequality. $\Pr[|X - \mu| \geq k\delta] \leq 1 / k^2$.

The result is established by setting $k = 2 \ln(n)/0.65$

Desigualdade de Chebyshev

Qual é a probabilidade obtermos mais de 75 caras ao jogar uma moeda justa 100 vezes ?

X : número de vezes que o resultado é cara

X_i :1 se o resultado da i -ésima tentativa é cara e 0, caso contrário

$$E[X] = \sum_{i=1}^{100} X_i = 50 \quad \text{e} \quad \text{Var}(X) = \sum_{i=1}^{100} \text{Var}(X_i) = 25$$

Aplicando Chebyshev temos

$$\Pr[|X-50| \geq 5 \times 5] \leq 1/25 \rightarrow \leq 4\%$$

$$\Pr[X \geq 75] \leq 2\% \quad (\text{simetria})$$

Coupon Collector

- X : number of rounds until all coupons have been collected.
- From previous class: $E[X] = nH_n$
- Markov inequality $\rightarrow \Pr[X \geq 2H_n] \leq 1/2$

Coupon Collector

- X_i : number of rounds between the appearance of the $(i-1)$ -th coupon and the i -th coupon

- X_i and X_j are independent for $i \neq j$

- Variance of X

$$\text{Var}[X] = \text{Var}[\sum_{i=1}^n X_i] = \sum_{i=1}^n \text{Var}[X_i]$$

- X_i is geometric random variable with

Coupon Collector

Lemma 3.8: If Z is a geometric random variable with parameter p then

$$\text{Var}[Z] = (1-p)/p^2 \leq 1/p^2$$

Proof

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Coupon Collector

- We have that

$$\text{Var}[X] = \sum_{i=1}^n \text{Var}[X_i] \leq \sum_{i=1}^n \left(\frac{n}{n-i+1}\right)^2 \leq \frac{\pi^2 n^2}{6}$$

By Chebyshev's

$$\Pr(|X - nH_n| \geq nH_n) \leq \frac{\pi^2 n^2}{6(n^2 H_n^2)} = O\left(\frac{1}{\ln^2 n}\right)$$

Coupon Collector

- $E(i,k)$: event that we do not obtain the i -th coupon after k rounds

$$\Pr[E[i, k]] = \left(1 - \frac{1}{n}\right)^k$$

- $E[k]$: event that we do not obtain some coupon after k rounds

$$\Pr\left[\bigcup_{k=1 \dots n} E[i, k]\right] \leq n \left(1 - \frac{1}{n}\right)^k$$

- For $k=2n \ln n$ we have probability at most $1/n$

Median Problem

Input: $S = \{a_1, a_2, \dots, a_n\}$

Output: median m of S

Median Problem

Deterministic Algorithms

Deterministic Algorithms

- Sorting Approach with a time complexity
 - $T(n) = n \log n$
- Medians of Medians:
 - Linear time but large constant

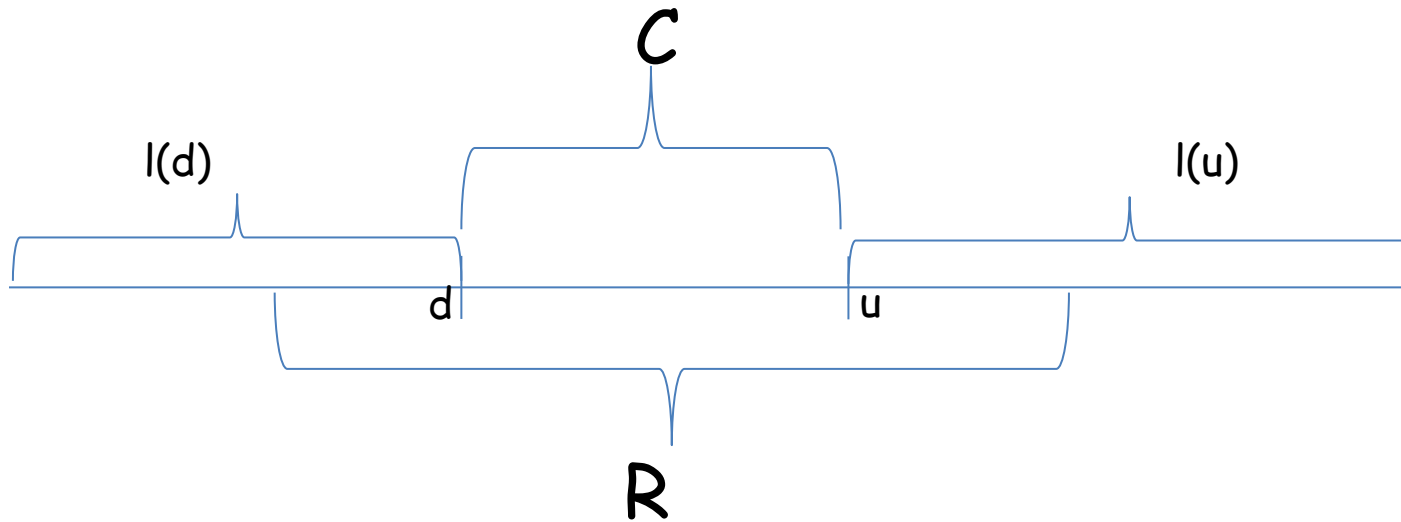
Randomized Algorithm: Approach

- Sample to find elements u and d in S such that
 - (a) $u < m < d$ and
 - (b) $T = \{x \in S \mid u \leq x \leq d\}$ has $O(n / \log n)$ elements
- Find m by sorting T and counting the number of elements in S smaller than u .

Randomized Algorithm for median

1. Sample with replacement a multiset R with $O(n^{3/4})$ elements
2. Sort R in sublinear time
3. $u \leftarrow ((1/2)n^{3/4} - n^{1/2})$ -th smallest element of R
4. $d \leftarrow ((1/2)n^{3/4} + n^{1/2})$ -th smallest element of R
5. Find $C = \{\text{elements of } S \text{ between } d \text{ and } u\}$;
 $l_d \leftarrow$ number elements in S smaller than d and
 $l_u \leftarrow$ number elements in S larger than u
6. If $l_d > n/2$ or $l_u > n/2$ **FAIL (median is not in C)**
7. If $|C| \leq 4n^{3/4}$ Sort C , otherwise **FAIL (set C is large)**
8. Output the $(n/2 - l_d)$ -th smallest element of C .

Randomized Algorithm for median



- If C is "small" and the median is between d and u we succeed.

Randomized Algorithm for median

Analysis

$$E1: Y(1) = |\{r \in R \mid r \leq m\}| < (1/2)n^{3/4} - n^{1/2}$$

$$E2: Y(2) = |\{r \in R \mid r \geq m\}| < 1/2n^{3/4} - n^{1/2}$$

$$E3: |C| > 4n^{3/4}$$

Lemma. The algorithm fails iff at least one the of the events E1, E2, E3 happens

Proof Sketch.

- Line 7 FAIL iff E3 happens
- Line 6 FAILS iff E1 or E2 happens

Randomized Algorithm for median

Lemma 3.11 $\Pr(E1)=\Pr(E2) \leq 1/ (4 n^{1/4})$

Proof.

$$X_i = \begin{cases} 1, & \text{if sample } i \text{ is } \leq \text{median } m \\ 0, & \text{otherwise} \end{cases}$$

We have $P[X_i=1] = 1/2 + 1/2n$

Event E1 is equivalent to

$$Y = \sum_{i=1}^{n^{3/4}} X_i \geq \frac{n^{3/4}}{2} - n^{1/4}$$

Randomized Algorithm for median

Lemma 3.11 $\Pr(E1)=\Pr(E2) \leq 1/ (4 n^{1/4})$

We use Chebyshev to give an upper bound on

$$\Pr \left[Y = \sum_{i=1}^{n^{3/4}} X_i \geq \frac{n^{3/4}}{2} - n^{1/4} \right]$$

Sampling without replacement implies that variable X 's are independent

Y is a binomial random variable with parameters $n^{3/4}$ and $(1/2 + 1/2n)$

Randomized Algorithm for median

Lemma 3.11 $\Pr(E1)=\Pr(E2) \leq 1/ (4 n^{1/4})$

Y is a binomial random variable with parameters $n^{3/4}$ and $(1/2 + 1/2n)$

$$E[Y] = n^{3/4}/2 + n^{3/4}/2n$$

$$\text{VAR}[Y] = n^{3/4}(1/2 + 1/2n)(1/2 - 1/2n) < 1/4 n^{3/4}$$

Applying Chebyshev we get that

$$\Pr[E1]=\Pr[E2] \leq 1/ (4 n^{1/4})$$

Randomized Algorithm for median

Lemma 3.12 $\Pr(E3) \leq 1/(2 n^{1/4})$

Proof. If event E3 occurs than one of the following events also occurs

Event A : at least $2n^{3/4}$ elements of C are larger than m

Event B : at least $2n^{3/4}$ elements of C are smaller than m

Thus,

$$\Pr[E3] \leq \Pr[A] + \Pr[B]$$

Randomized Algorithm for median

Lemma 3.12 $\Pr(E3) \leq 1/(2n^{1/4})$

Proof.

If Event A occurs \Rightarrow at least $2n^{3/4}$ elements of C are larger than m

\Rightarrow Rank of u in the list S is at least $1/2n + 2n^{3/4}$

\Rightarrow Set R has at least $(1/2)n^{3/4} - (1/2)n^{1/2}$ samples among the $1/2n - 2n^{3/4}$ largest elements of S

We can bound the probability of the occurrence of the last event.

Randomized Algorithm for median

Lemma 3.12 $\Pr(E3) \leq 1/(2 n^{1/4})$

$$X_i = \begin{cases} 1, & \text{if the } i\text{th sample } i \text{ is among the } \frac{n}{2} - 2n^{3/4} \text{ largest of } S \\ 0, & \text{otherwise} \end{cases}$$

Let

$$X = \sum_{i=1}^{n^{3/4}} X_i$$

We have to give an upper bound on

$$\Pr \left[X > \frac{n}{2} - 2n^{3/4} \right]$$

Randomized Algorithm for median

X is a binomial random variable with parameters $n^{3/4}$ and

$\frac{1}{2} - 2n^{-1/4}$. Thus,

$$E[X] =$$

$$\text{VAR}[X] =$$

Applying Chebyshev

$$\Pr[B] = \Pr[A] \leq \Pr\left[X > \frac{n}{2} - 2n^{3/4}\right] \leq 1/(4n^{1/4})$$

$$\Pr[E_3] \leq \Pr[B] + \Pr[A] \leq 1/(2n^{1/4})$$

Randomized Algorithm for median

The probability that randomized median fails is at most

$$\Pr[E1] + \Pr[E2] + \Pr[E3] \leq 1/n^{1/4}$$

Chernoff Bounds (above mean)

Theorem. Suppose X_1, \dots, X_n are independent 0-1 random variables. Let $X = X_1 + \dots + X_n$. Then for any $\mu \geq E[X]$ and for any $\delta > 0$, we have

$$\Pr[X > (1 + \delta)\mu] < \left[\frac{e^\delta}{(1 + \delta)^{1 + \delta}} \right]^\mu$$

↑
sum of independent 0-1 random variables
is tightly centered on the mean

Pf. We apply a number of simple transformations.

- For any $t > 0$,

$$\Pr[X > (1 + \delta)\mu] = \Pr[e^{tX} > e^{t(1 + \delta)\mu}] \leq e^{-t(1 + \delta)\mu} \cdot E[e^{tX}]$$

↑
 $f(x) = e^{tx}$ is monotone in x

↑
Markov's inequality: $\Pr[X > a] \leq E[X] / a$

- Now $E[e^{tX}] = E[e^{t \sum_i X_i}] = \prod_i E[e^{tX_i}]$

↑
definition of X

↑
independence

Chernoff Bounds (above mean)

Pf. (cont)

- Let $p_i = \Pr[X_i = 1]$. Then,

$$E[e^{tX_i}] = p_i e^t + (1 - p_i) e^0 = 1 + p_i(e^t - 1) \leq e^{p_i(e^t - 1)}$$

↑
for any $\alpha \geq 0$, $1 + \alpha \leq e^\alpha$

- Combining everything:

$$\Pr[X > (1 + \delta)\mu] \leq e^{-t(1+\delta)\mu} \prod_i E[e^{tX_i}] \leq e^{-t(1+\delta)\mu} \prod_i e^{p_i(e^t - 1)} \leq e^{-t(1+\delta)\mu} e^{\mu(e^t - 1)}$$

↑
previous slide

↑
inequality above

↑
 $\sum_i p_i = E[X] \leq \mu$

- Finally, choose $t = \ln(1 + \delta)$. ▀

Chernoff Bounds (below mean)

Theorem. Suppose X_1, \dots, X_n are independent 0-1 random variables. Let $X = X_1 + \dots + X_n$. Then for any $\mu \leq E[X]$ and for any $0 < \delta < 1$, we have

$$\Pr[X < (1 - \delta)\mu] < e^{-\delta^2 \mu / 2}$$

Pf idea. Similar.

Remark. Not quite symmetric since only makes sense to consider $\delta < 1$.

Chernoff Bounds

Qual é a probabilidade obtermos mais de 75 caras ao jogar uma moeda justa 100 vezes ?

X : número de vezes que o resultado é cara

X_i :1 se o resultado da i -ésima tentativa é cara e 0 caso contrário

$$E[X] = \sum_{i=1}^{100} X_i = 50 \quad \text{e} \quad \text{Var}(X) = \sum_{i=1}^{100} \text{Var}(X_i) = 25$$

Aplicando Chernoff temos

$$\Pr[|X-50| > (1+0.5) \times 50] \leq 0.007 \rightarrow \leq 0.7\%$$

13.10 Load Balancing

Load Balancing

Load balancing. System in which m jobs arrive in a stream and need to be processed immediately on n identical processors. Find an assignment that balances the workload across processors.

Centralized controller. Assign jobs in round-robin manner. Each processor receives at most $\lceil m/n \rceil$ jobs.

Decentralized controller. Assign jobs to processors uniformly at random. How likely is it that some processor is assigned "too many" jobs?

Load Balancing

Analysis.

- Let X_i = number of jobs assigned to processor i .
- Let $Y_{ij} = 1$ if job j assigned to processor i , and 0 otherwise.
- We have $E[Y_{ij}] = 1/n$
- Thus, $X_i = \sum_j Y_{ij}$, and $\mu = E[X_i] = 1$.
- Applying Chernoff bounds with $\delta = c - 1$ yields $\Pr[X_i > c] < \frac{e^{c-1}}{c^c}$
- Let $\gamma(n)$ be number x such that $x^x = n$, and choose $c = e \gamma(n)$.

$$\Pr[X_i > c] < \frac{e^{c-1}}{c^c} < \left(\frac{e}{c}\right)^c = \left(\frac{1}{\gamma(n)}\right)^{e\gamma(n)} < \left(\frac{1}{\gamma(n)}\right)^{2\gamma(n)} = \frac{1}{n^2}$$

- Union bound \Rightarrow with probability $\geq 1 - 1/n$ no processor receives more than $e \gamma(n) = \Theta(\log n / \log \log n)$ jobs.

Fact: this bound is asymptotically tight: with high probability, some processor receives $\Theta(\log n / \log \log n)$

Load Balancing: Many Jobs

Theorem. Suppose the number of jobs $m = 16n \ln n$. Then on average, each of the n processors handles $\mu = 16 \ln n$ jobs. With high probability every processor will have between half and twice the average load.

Pf.

- Let X_i, Y_{ij} be as before.
- Applying Chernoff bounds with $\delta = 1$ yields

$$\Pr[X_i > 2\mu] < \left(\frac{e}{4}\right)^{16n \ln n} < \left(\frac{1}{e}\right)^{\ln n} = \frac{1}{n^2} \qquad \Pr[X_i < \frac{1}{2}\mu] < e^{-\frac{1}{2}\left(\frac{1}{2}\right)^2(16n \ln n)} = \frac{1}{n^2}$$

- Probability of a processor has load larger than 2μ or smaller than $1/2 \mu$ is at most $2/n^2$.
- Union bound \Rightarrow Probability of some processor has load larger than 2μ or smaller than $1/2 \mu$ is at most $2/n$
- every processor has load between half and twice the average with probability $\geq 1 - 2/n$. ▪

13.11 Packet Routing
