

Chernof Bound - Large Deviation Bound

Theorem

Let X_1, \dots, X_n be independent, identically distributed, 0 – 1 random variables with $Pr(X_i = 1) = E[X_i] = p$. Let $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$, then for any $\delta \in [0, 1]$ we have

$$Prob(\bar{X}_n \geq (1 + \delta)p) \leq e^{-np\delta^2/3}$$

and

$$Prob(\bar{X}_n \leq (1 - \delta)p) \leq e^{-np\delta^2/2}.$$

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Let X_1, \dots, X_n be independent, 0 – 1 random variables with $Pr(X_i = 1) = E[X_i] = p_i$. Let $\mu = \sum_{i=1}^n p_i$, then for any $\delta \in [0, 1]$ we have

$$Prob\left(\sum_{i=1}^n X_i \geq (1 + \delta)\mu\right) \leq e^{-\mu\delta^2/3}$$

and

$$Prob\left(\sum_{i=1}^n X_i \leq (1 - \delta)\mu\right) \leq e^{-\mu\delta^2/2}.$$

Packet Routing on Parallel Computer

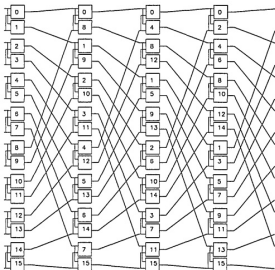
Communication network:



Packet Routing on Parallel Computer

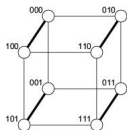
Communication network:

- nodes - processors, switching nodes;
- edges - communication links.

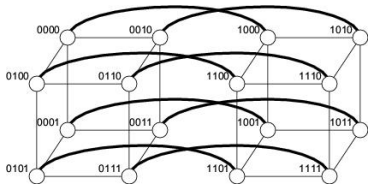


The n -cube

The 3-cube:



The 4-cube:



The n -cube

The n -cube:

$N = 2^n$ nodes: $0, 1, 2, \dots, 2^n - 1$.

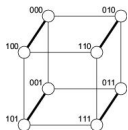
Let $\bar{x} = (x_1, \dots, x_n)$ be the number of node x in binary.

Nodes x and y are connected by an edge iff their binary representations differ in exactly one bit.

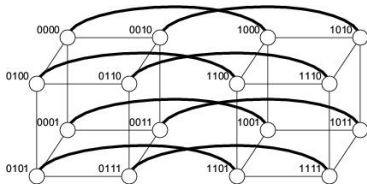
Bit-wise routing: correct bit i in the i -th transition - route has length $\leq n$.

The n -cube

The 3-cube:



The 4-cube:



Model and Computational problem

- Up to one packet can cross an edge per step, each packet can cross up to one edge per step.
- A permutation communication request: each node is the source and destination of exactly one packet.
- What is the time to route an arbitrary permutation on the n -cube?

Two phase routing algorithm:

- 1 Send packet to a randomly chosen destination.
- 2 Send packet from randomly chosen destination to real destination.

Path: Correct the bits, starting at x_1 to x_n .

Any greedy queuing method - if some packet can traverse an edge one does.

Theorem

The two phase routing algorithm routes an arbitrary permutation on the n -cube in $O(\log N) = O(n)$ parallel steps with high probability.

- We focus first on phase 1. We bound the routing time of a given packet M .
- Let e_1, \dots, e_m be the $m \leq n$ edges traversed by a given packet M in phase 1.
- Let $X(e)$ be the total number of packets that traverse edge e at that phase.
- Let $T(M)$ be the number of steps till M finished phase 1.

Lemma

$$T(M) \leq \sum_{i=1}^m X(e_i).$$

- We call any path $P = (e_1, e_2, \dots, e_m)$ of $m \leq n$ edges that follows the bit fixing algorithm a *possible packet path*.
- We denote the corresponding nodes v_0, v_1, \dots, v_m , with $e_i = (v_{i-1}, v_i)$.
- For any possible packet path P , let $T(P) = \sum_{i=1}^m X(e_i)$.

- If phase I takes more than T steps then for some possible packet path P ,

$$T(P) \geq T$$

- There are at most $2^n \cdot 2^n = 2^{2n}$ possible packet paths.
- Assume that e_k connects $(a_1, \dots, a_i, \dots, a_n)$ to $(a_1, \dots, \bar{a}_i, \dots, a_n)$.
- Only packets that started in address

$$(*, \dots, *, a_i, \dots, a_n)$$

can traverse edge e_k , and only if their destination addresses are

$$(a_1, \dots, a_{i-1}, \bar{a}_i, *, \dots, *)$$

- There are no more than 2^{i-1} possible packets, each has probability 2^{-i} to traverse e_i .

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-

$$\mathbf{E}[X(e_i)] \leq 2^{i-1} \cdot 2^{-i} = \frac{1}{2}.$$

-

$$\mathbf{E}[T(P)] \leq \sum_{i=1}^m \mathbf{E}[X(e_i)] \leq \frac{1}{2} \cdot m \leq n.$$

- **Problem:** The $X(e_i)$'s are not independent.

- A packet is *active* with respect to possible packet path P if it ever use an edge of P .
- For $k = 1, \dots, N$, let $H_k = 1$ if the packet starting at node k is active, and $H_k = 0$ otherwise.
- The H_k are independent, since each H_k depends only on the choice of the intermediate destination of the packet starting at node k , and these choices are independent for all packets.
- Let $H = \sum_{k=1}^N H_k$ be the total number of active packets.
-

$$\mathbf{E}[H] \leq \mathbf{E}[T(P)] \leq n$$

- Since H is the sum of independent $0 - 1$ random variables we can apply the Chernoff bound

$$\Pr(H \geq 6n) \leq \Pr(H \geq 6\mathbf{E}[H]) \leq 2^{-6n}.$$

For a given possible packet path P ,

$$\begin{aligned}\Pr(T(P) \geq 30n) &\leq \Pr(T(P) \geq 30n \mid H \geq 6n) \Pr(H \geq 6n) \\ &+ \Pr(T(P) \geq 30n \mid H < 6n) \Pr(H < 6n) \\ &\leq \Pr(H \geq 6n) + \Pr(T(P) \geq 30n \mid H < 6n) \\ &\leq 2^{-6n} + \Pr(T(P) \geq 30n \mid H < 6n).\end{aligned}$$

Lemma

If a packet leaves a path (of another packet) it cannot return to that path in the same phase.

Proof.

Leaving a path at the i -th transition implies different i -th bit, this bit cannot be changed again in that phase.

Lemma

The number of transitions that a packet takes on a given path is distributed $G\left(\frac{1}{2}\right)$.

Proof.

The packet has probability $1/2$ of leaving the path in each transition.

The probability that the active packets cross edges of P more than $30n$ times is less than the probability that a fair coin flipped $36n$ times comes up heads less than $6n$ times.

Letting Z be the number of heads in $36n$ fair coin flips, we now apply the Chernoff bound:

$$\begin{aligned} \Pr(T(P) \geq 30n \mid H \leq 6n) &\leq \Pr(Z \leq 6n) \\ &\leq e^{-18n(2/3)^2/2} = e^{-4n} \leq 2^{-3n-1}. \end{aligned}$$

$$\begin{aligned} \Pr(T(P) \geq 30n) &\leq \Pr(H \geq 6n) + \Pr(T(P) \geq 30n \mid H \leq 6n) \\ &\leq 2^{-6n} + 2^{-3n-1} \leq 2^{-3n} \end{aligned}$$

As there are at most 2^{2n} possible packet paths in the hypercube, the probability that there is *any* possible packet path for which $T(P) \geq 30n$ is bounded by

$$2^{2n}2^{-3n} = 2^{-n} = O(N^{-1}).$$

- The proof of phase 2 is by symmetry:
- The proof of phase 1 argued about the number of packets crossing a given path, no “timing” considerations.
- The path from “one packet per node” to random locations is similar to random locations to “one packet per node” in reverse order.
- Thus, the distribution of the number of packets that crosses a path of a given packet is the same.

Oblivious Routing

Definition

A routing algorithm is **oblivious** if the path taken by one packet is independent of the source and destinations of any other packets in the system.

Theorem

Given an N -node network with maximum degree d the routing time of any deterministic oblivious routing scheme is

$$\Omega \left(\sqrt{\frac{N}{d^3}} \right).$$

Set Balancing

Given an $n \times n$ matrix \mathcal{A} with entries in $\{0, 1\}$, let

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix} \begin{pmatrix} b_1 \\ b_2 \\ \dots \\ \dots \\ b_n \end{pmatrix} = \begin{pmatrix} c_1 \\ c_2 \\ \dots \\ \dots \\ c_n \end{pmatrix}.$$

Find a vector \bar{b} with entries in $\{-1, 1\}$ that minimizes

$$\|\mathcal{A}\bar{b}\|_{\infty} = \max_{i=1, \dots, n} |c_i|.$$

Theorem

For a random vector \bar{b} , with entries chosen independently and with equal probability from the set $\{-1, 1\}$,

$$Pr(\|\mathcal{A}\bar{b}\|_\infty \geq \sqrt{4n \ln n}) \leq \frac{2}{n}.$$

The $\sum_{i=1}^n a_{j,i} b_i$ (excluding the zero terms) is a sum of independent $-1, 1$ random variable. We need a bound on such sum.

Chernoff Bound for Sum of $\{-1, +1\}$ Random Variables

Theorem

Let X_1, \dots, X_n be independent random variables with

$$\Pr(X_i = 1) = \Pr(X_i = -1) = \frac{1}{2}.$$

Let $X = \sum_1^n X_i$. For any $a > 0$,

$$\Pr(X \geq a) \leq e^{-\frac{a^2}{2n}}.$$

de Moivre – Laplace approximation: For any k , such that $|k - np| \leq a$

$$\binom{n}{k} p^k (1-p)^{n-k} \approx \frac{1}{\sqrt{2\pi np(1-p)}} e^{-\frac{a^2}{2np(1-p)}}$$

For any $t > 0$,

$$\mathbf{E}[e^{tX_i}] = \frac{1}{2}e^t + \frac{1}{2}e^{-t}.$$

$$e^t = 1 + t + \frac{t^2}{2!} + \cdots + \frac{t^i}{i!} + \cdots$$

and

$$e^{-t} = 1 - t + \frac{t^2}{2!} + \cdots + (-1)^i \frac{t^i}{i!} + \cdots$$

Thus,

$$\begin{aligned} \mathbf{E}[e^{tX_i}] &= \frac{1}{2}e^t + \frac{1}{2}e^{-t} = \sum_{i \geq 0} \frac{t^{2i}}{(2i)!} \\ &\leq \sum_{i \geq 0} \frac{\left(\frac{t^2}{2}\right)^i}{i!} = e^{t^2/2} \end{aligned}$$

$$\mathbf{E}[e^{tX}] = \prod_{i=1}^n \mathbf{E}[e^{tX_i}] \leq e^{nt^2/2},$$

$$Pr(X \geq a) = Pr(e^{tX} > e^{ta}) \leq \frac{\mathbf{E}[e^{tX}]}{e^{ta}} \leq e^{t^2n/2 - ta}.$$

Setting $t = a/n$ yields

$$Pr(X \geq a) \leq e^{-\frac{a^2}{2n}}.$$

By symmetry we also have

Corollary

Let X_1, \dots, X_n be independent random variables with

$$\Pr(X_i = 1) = \Pr(X_i = -1) = \frac{1}{2}.$$

Let $X = \sum_{i=1}^n X_i$. Then for any $a > 0$,

$$\Pr(|X| > a) \leq 2e^{-\frac{a^2}{2n}}.$$

Application: Set Balancing

Theorem

For a random vector \bar{b} , with entries chosen independently and with equal probability from the set $\{-1, 1\}$,

$$\Pr(\|\mathcal{A}\bar{b}\|_\infty \geq \sqrt{4n \ln n}) \leq \frac{2}{n} \quad (1)$$

- Consider the i -th row $\bar{a}_i = a_{i,1}, \dots, a_{i,n}$.
- Let k be the number of 1's in that row.
- $Z_i = \sum_{j=1}^k a_{i,j} b_{ij}$.
- If $k \leq \sqrt{4n \ln n}$ then clearly $Z_i \leq \sqrt{4n \ln n}$.

If $k > \sqrt{4n \log n}$, the k non-zero terms in the sum Z_i are independent random variables, each with probability $1/2$ of being either $+1$ or -1 .

Using the Chernoff bound:

$$\Pr \left\{ |Z_i| > \sqrt{4n \log n} \right\} \leq 2e^{-4n \log n / (2k)} \leq \frac{2}{n^2},$$

where we use the fact that $n \geq k$.

The result follows by union bound (n rows).

Hoeffding's Inequality

Large deviation bound for more general random variables:

Theorem (Hoeffding's Inequality)

Let X_1, \dots, X_n be independent random variables such that for all $1 \leq i \leq n$, $E[X_i] = \mu$ and $\Pr(a \leq X_i \leq b) = 1$. Then

$$\Pr\left(\left|\frac{1}{n} \sum_{i=1}^n X_i - \mu\right| \geq \epsilon\right) \leq 2e^{-2n\epsilon^2/(b-a)^2}$$

Lemma

(Hoeffding's Lemma) Let X be a random variable such that $\Pr(X \in [a, b]) = 1$ and $E[X] = 0$. Then for every $\lambda > 0$,

$$\mathbf{E}[e^{\lambda X}] \leq e^{\lambda^2(a-b)^2/8}.$$

Proof of the Lemma

Since $f(x) = e^{\lambda x}$ is a convex function, for any $\alpha \in (0, 1)$ and $x = \alpha a + (1 - \alpha)b$,

$$f(X) \leq \alpha f(a) + (1 - \alpha)f(b).$$

For $\alpha = \frac{b-x}{b-a} \in (0, 1)$, $x = \alpha a + (1 - \alpha)b$, thus,

$$e^{\lambda x} \leq \frac{b-x}{b-a} e^{\lambda a} + \frac{x-a}{b-a} e^{\lambda b}.$$

Taking expectation, and using $\mathbf{E}[X] = 0$, we have

$$\mathbf{E}[e^{\lambda X}] \leq \frac{b}{b-a} e^{\lambda a} - \frac{a}{b-a} e^{\lambda b} \leq e^{\lambda^2(b-a)^2/8}.$$

[Proof of the second inequality in the next slide.]

$$E[e^{\lambda X}] \leq \frac{b}{b-a} e^{\lambda a} - \frac{a}{b-a} e^{\lambda b} = e^{\phi(t)},$$

for

- $\phi(t) = -\theta t + \log(1 - \theta + \theta e^t)$,
- $\theta = \frac{-a}{b-a}$
- $t = \lambda(b-a)$

$\phi(0) = \phi'(0) = 0$, and $\phi''(t) \leq 1/4$ for all t .

By Taylor's theorem, for any $t > 0$ there is a $t' \in [0, t]$ such that

$$\phi(t) = \phi(0) + t\phi'(0) + \frac{1}{2}t^2\phi''(t') \leq \frac{1}{8}t^2.$$

Thus,

$$E[e^{\lambda X}] \leq e^{\phi(t)} \leq e^{t^2/8} = e^{\lambda^2(b-a)^2/8}.$$

Proof of the Bound

Let $Z_i = X_i - \mathbf{E}[X_i] = X_i - \mu$ and $Z = \frac{1}{n} \sum_{i=1}^n Z_i$.

$$\Pr(Z \geq \epsilon) \leq e^{-\lambda\epsilon} \mathbf{E}[e^{\lambda Z}] \leq e^{-\lambda\epsilon} \prod_{i=1}^n \mathbf{E}[e^{\lambda Z_i/n}] \leq e^{-\lambda\epsilon + \frac{\lambda^2(b-a)^2}{8n}}$$

Set $\lambda = \frac{4n\epsilon}{(b-a)^2}$ gives

$$\Pr\left(\left|\frac{1}{n} \sum_{i=1}^n X_i - \mu\right| \geq \epsilon\right) = \Pr(Z \geq \epsilon) \leq 2e^{-2n\epsilon^2/(b-a)^2}$$

A More General Version

Theorem

Let X_1, \dots, X_n be independent random variables with $\mathbf{E}[X_i] = \mu_i$ and $\Pr(B_i \leq X_i \leq B_i + c_i) = 1$, then

$$\Pr\left(\left|\sum_{i=1}^n X_i - \sum_{i=1}^n \mu_i\right| \geq \epsilon\right) \leq 2e^{-\frac{2\epsilon^2}{\sum_{i=1}^n c_i^2}}$$

Proof.

Homework!



Application: Job Completion

We have n jobs, job i has expected run-time μ_i . We terminate job i if it runs $\beta\mu_i$ time. When will the machine will be free of jobs?

X_i = execution time of job i . $0 \leq X_i \leq \beta\mu_i$.

$$\Pr\left(\left|\sum_{i=1}^n X_i - \sum_{i=1}^n \mu_i\right| \geq \epsilon \sum_{i=1}^n \mu_i\right) \leq 2e^{-\frac{2\epsilon^2(\sum_{i=1}^n \mu_i)^2}{\sum_{i=1}^n \beta^2 \mu_i^2}}$$

Assume all $\mu_i = \mu$

$$\Pr\left(\left|\sum_{i=1}^n X_i - n\mu\right| \geq \epsilon n\mu\right) \leq 2e^{-\frac{2\epsilon^2 n^2 \mu^2}{n\beta^2 \mu^2}} = 2e^{-2\epsilon^2 n/\beta^2}$$

Let $\epsilon = \beta\sqrt{\frac{\log n}{n}}$, then

$$\Pr\left(\left|\sum_{i=1}^n X_i - n\mu\right| \geq \beta\mu\sqrt{n \log n}\right) \leq 2e^{-\frac{2\beta^2 \mu^2 n \log n}{n\beta^2 \mu^2}} = \frac{2}{n^2}$$