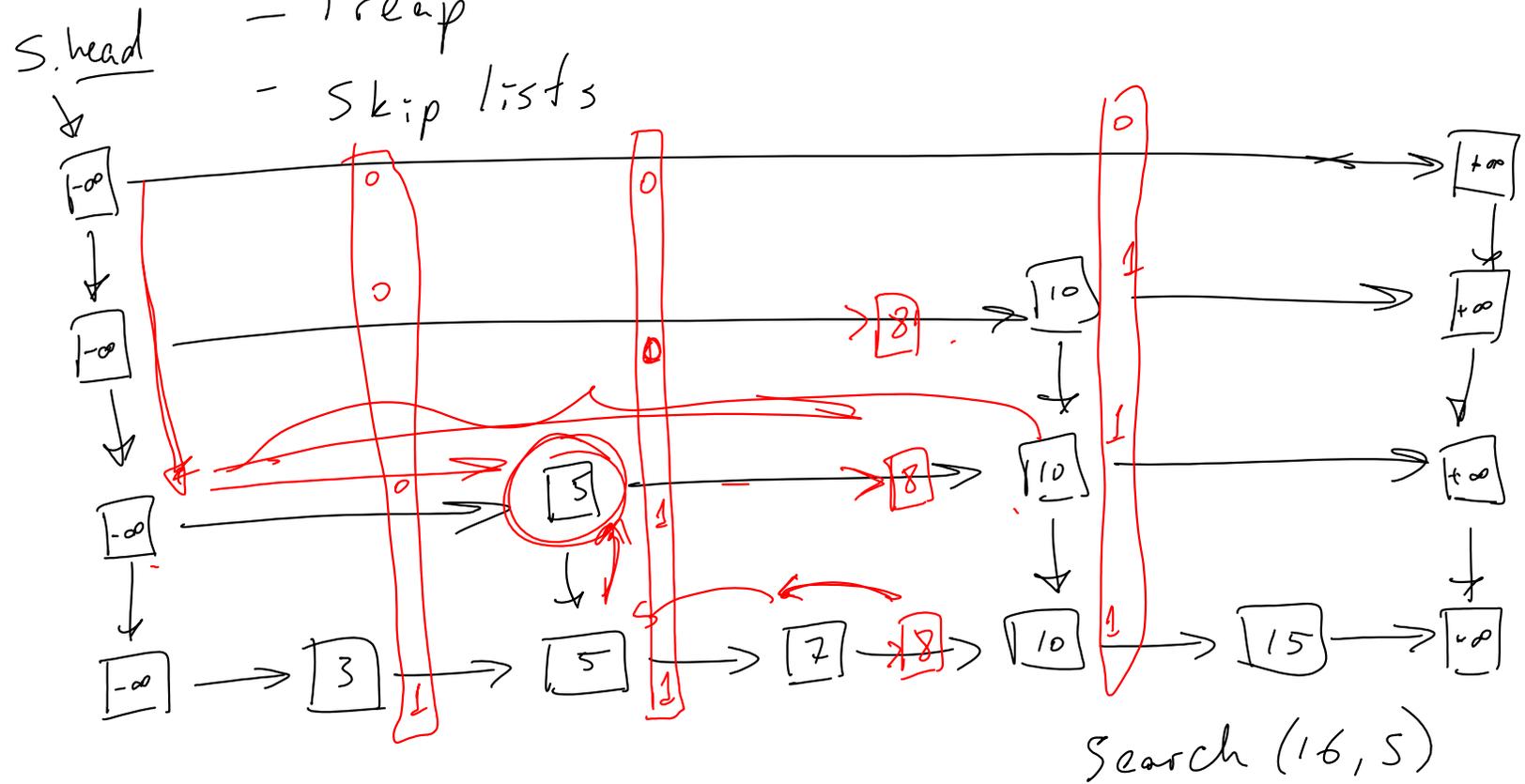


Last week:

- Treap
- Skip lists



Predecessor-Search (k, S)

$v = S.head$

while $v.down \neq nil$

while $(v.next \neq nil \ \& \ v.next.key \leq k)$

$v = v.next$

$v = v.down$

while $(v.next \neq nil \ \& \ v.next.key \leq k)$

$v = v.next$

return v

Analysis of (Randomized) Skip List Height

Lemma: The height $h(T)$ of the skip list T on n keys is at most

$(c+1) \log n$ w/ prob. $1 - \frac{1}{n^c}$

$$P_r [h(T) < (c+1) \log n] \geq 1 - \frac{1}{n^c}$$

for any constant $c > 0$

high probability

Proof: A node v will be at most at level k if k heads came up in a row.

$$P_r [h(v) \geq k] = P_r [\text{first } k \text{ flips are heads}] = \frac{1}{2^k}$$

$$P_r [h(T) \geq (c+1) \log n] = P_r \left[\max_{v \in T} h(v) \geq (c+1) \log n \right]$$

$$= P_r \left[\underline{h(v_1) \geq (c+1) \log n} \text{ or } \underline{h(v_2) \geq (c+1) \log n} \text{ or } \dots \text{ or } \underline{h(v_n) \geq (c+1) \log n} \right]$$

$$\leq P_c [h(v_1) \geq (c+1)\log n] +$$

$$P_c [h(v_2) \geq (c+1)\log n] +$$

$$\vdots$$

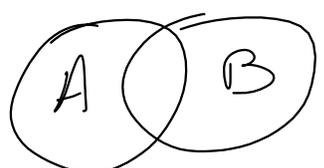
$$P_c [h(v_n) \geq (c+1)\log n]$$

$$\leq \sum_{i=1}^n \frac{1}{2^{(c+1)\log n}} = \frac{n}{(2^{\log n})^{(c+1)}}$$

$$= \frac{n}{n^{(c+1)}} = \frac{1}{n^c}$$

Union Bound

$$P_c [A \text{ or } B] \leq P_c [A] + P_c [B]$$



$$|A \cup B| \leq |A| + |B|$$

$$P_c [h(T) < \underline{(c+1)\log n}] \geq 1 - P_c [h(T) \geq (c+1)\log n]$$

$h(T) \in O(\log n)$

$$\geq 1 - \frac{1}{n^c}$$

$$1 - \frac{1}{n}$$

with high probability

$$P_c [h(T) = O(\log n)] = 1 - \frac{1}{n^{O(1)}}$$

Insert(k, S)

$v \leftarrow \text{search}(k, S)$

if $v.\text{key} \neq k$
 $v' = \text{new Node}(k)$
ListInsert(v, v')

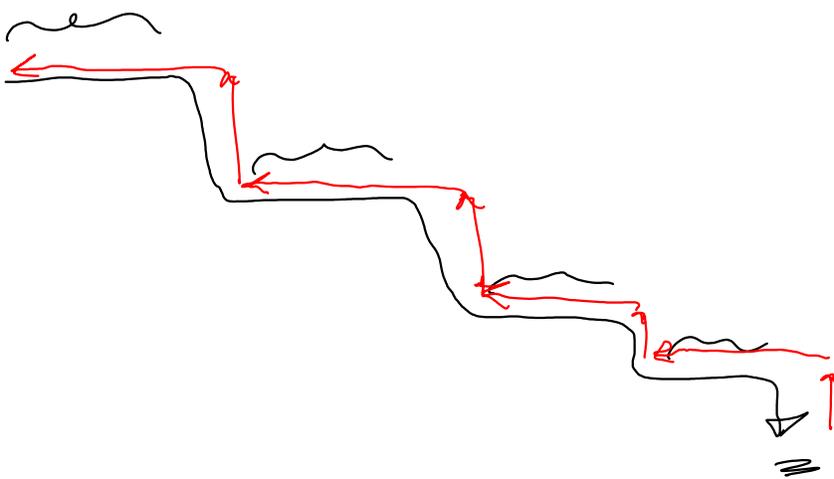
bit = coinFlip()

while(bit == 1)

promote a copy of v' to
the list above

bit = coinFlip()

(Randomized) Skip List Search
Runtime Analysis



Backward-search (v, S)

while ($v \neq S.head$)

if $v.up \neq nil$

$v = v.up$

else

$v = v.left$

if $coinFlip() = 1$

→ $v = v.up$

else

→ $v = v.left$

Expected # of tails is equal
to expected # of heads

$$\begin{aligned} E[\text{runtime}(\text{search})] &= 2 \cdot E[\text{\# of heads}] \\ &= 2 \cdot \underline{E[\text{height of skip list}]} \end{aligned}$$

Analysis of Expected height of Skip List

For any ^{positive integer} random variable X

$$E[X] = \sum_x x \cdot \Pr[X=x]$$

$$E[X] = \sum_{x=1}^{\infty} \Pr[X \geq x]$$

Lemma:
$$\sum_{x=1}^n x \cdot f(x) = \sum_{x=1}^n \sum_{i=x}^n f(i)$$

$$\begin{aligned} \sum_{x=1}^n x \cdot f(x) &= 1 \cdot f(1) + 2 \cdot f(2) + 3 \cdot f(3) + \dots + n \cdot f(n) \\ &= (f(1) + f(2) + \dots + f(n)) + (1 \cdot f(2) + 2 \cdot f(3) + \dots + (n-1) \cdot f(n)) \\ &= (f(1) + f(2) + \dots + f(n)) + (f(2) + f(3) + \dots + f(n)) \\ &\quad + (f(3) + f(4) + \dots + f(n)) + \dots + (f(n)) \end{aligned}$$

$$\sum_{x=1}^n \left(\sum_{i=x}^n f(i) \right)$$

Lemma: For any non-negative integer random variable X

$$E[X] = \sum_{x=1}^{\infty} \Pr[X \geq x]$$

Proof:

$$E[X] = \sum_{x=1}^{\infty} x \cdot \underbrace{\Pr[X=x]}_{f(x)} = \sum_{x=1}^{\infty} \left(\sum_{i=x}^{\infty} \underbrace{\Pr[X=i]}_{f(i)} \right)$$

$$= \sum_{x=1}^{\infty} \Pr[X \geq x]$$

$$\sum_{i=x}^{\infty} \Pr[X=i] = \Pr[X \geq x]$$

$$\Pr[X=x] + \Pr[X=x+1] + \dots + \Pr[X=\infty]$$

Expected height of Skip List

$$E[\text{height}(s)] = E\left[\max_v h(v)\right]$$

$$\text{Let } H = \max_v h(v)$$

$$E[H] = \sum_{x=1}^{\infty} P_r[H \geq x] =$$

$$= \underbrace{\sum_{x=1}^{\log n} P_r[H \geq x]}_{\text{wavy line}} + \sum_{x=(\log n)+1}^{\infty} P_r[H \geq x]$$

$$\leq \sum_{x=1}^{\log n} 1 + \sum_{x=\log n+1}^{\infty} P_r[H \geq x]$$

$$= \log n + \underbrace{\sum_{x=\log n+1}^{\infty} P_r[H \geq x]}$$

$$P_r[H \geq x] = P_r \left[\underbrace{h(v_1) \geq x}_{\text{or}} \quad \dots \quad \underbrace{h(v_n) \geq x}_{\text{or}} \right]$$

by union bound

$$\leq \sum_{i=1}^n P_r \left[\underbrace{h(v_i) \geq x} \right]$$

$$\leq n \cdot \frac{1}{2^x}$$

$$\sum_{x=\log n+1}^{\infty} P_r[H \geq x] \leq \sum_{x=\log n+1}^{\infty} n \cdot \frac{1}{2^x} = n \cdot \sum_{i=1}^{\infty} \frac{1}{2^{\log n+i}}$$

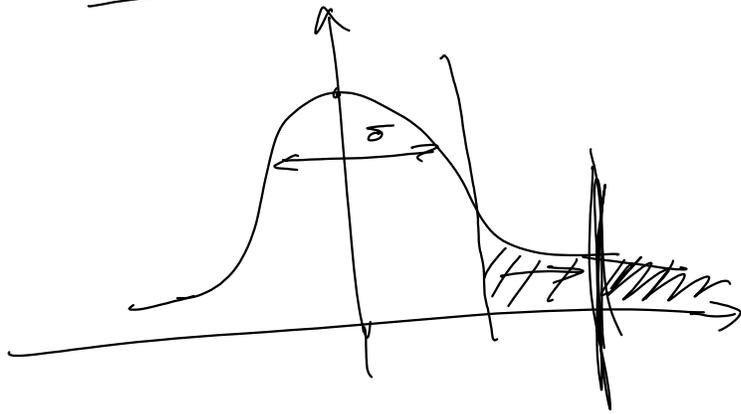
$$= n \cdot \sum_{i=1}^{\infty} \frac{1}{n \cdot 2^i} = \sum_{i=1}^{\infty} \frac{1}{2^i} = \underline{\underline{1}}$$

$$E[H] = \log n + \sum_{x=\log n+1}^{\infty} \Pr[H \geq x]$$

$$= \underline{\underline{\log n + 1}}$$

$$h < \underline{\underline{c \cdot \log n}} \quad \text{w prob} \leq \frac{1}{n^{c-1}}$$

Tail Bounds



Markov's Inequality

Lemma: Non-negative random variable X

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

Proof:

$$\begin{aligned} \underline{E[X]} &= \sum_{x=0}^{a-1} x \cdot \Pr[X=x] + \sum_{x=a}^{\infty} x \cdot \Pr[X=x] \\ &\geq \sum_{x=a}^{\infty} x \cdot \Pr[X=x] \\ &\geq a \cdot \sum_{x=a}^{\infty} \Pr[X=x] \end{aligned}$$

$$= a \cdot \Pr[X \geq a]$$

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

The Height of Treaps

For what values of h

$$\Pr[\text{Height}(\text{Treap}) \geq h] \leq \frac{1}{h}$$

$$\Pr[h(v_1) \geq h \text{ or } h(v_2) \geq h \text{ or } \dots \text{ or } h(v_n) \geq h] \leq$$

$$\leq \Pr[h(v_1) \geq h] + \Pr[h(v_2) \geq h] + \dots + \Pr[h(v_n) \geq h]$$

$$= n \cdot \Pr[h(v_k) \geq h]$$

Consider the k -th smallest key node v_k

$$n \cdot \Pr[h(v_k) \geq h] \leq \frac{1}{h}$$

$$\Rightarrow \Pr[h(v_k) \geq h] \leq \frac{1}{n^2}$$

By Markov's inequality:

$$P_c [h(v_k) \geq h] \leq \frac{\mathbb{E}[h(v_k)]}{h} = \frac{\log n}{h} \leq \frac{1}{n^2}$$

$$\Rightarrow h \geq n^2 \log n$$

$$P_c [\text{height of treep} \geq \underline{\underline{n^2 \log n}}] \leq \frac{1}{n}$$

Chebyshev's inequality

Def'n; Two random variables X & Y
are independent iff

$$P_c [X=x \ \& \ Y=y] = P_c [X=x] \cdot P_c [Y=y]$$

$$P_c [X=x \mid Y=y] = P_c [X=x]$$

Def'n: A set of random variables

X_1, X_2, \dots, X_n are k -wise independent if every subset of k variables are independent.

Example: X & Y are ^{independent} binary random variables

$$Z = (X + Y) \bmod 2$$

X, Y, Z are pair-wise independent but not 3-wise independent.

Corollary: if X_1, \dots, X_n are k -wise independent, then they are also k' -wise for any $2 \leq k' \leq k$

Theorem (Chebyshev's inequality):

Let X_1, X_2, \dots, X_n be pairwise independent non-negative indicator random variables, and

$$X = \sum_{i=1}^n X_i \quad \text{with} \quad E[X] = \mu$$

then

$$Pr[(X - \mu)^2 \geq a] < \frac{\mu}{a}$$

for any $a > 0$.

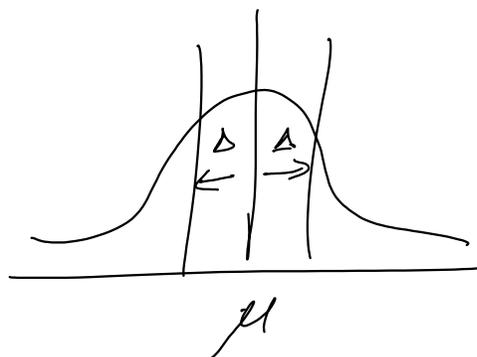
Alternatively: $k^2 = a$

$$Pr[(X - \mu)^2 \geq k^2] = Pr[|X - \mu| \geq k]$$

$$Pr[|X - \mu| \geq k] < \frac{\mu}{k^2}$$

Let $\Delta = k$:

Additive bounds:



$$P_r [X \geq \mu + \Delta] \leq \frac{\mu}{\Delta^2}$$

$$P_r [X \leq \mu - \Delta] \leq \frac{\mu}{\Delta^2}$$

Let $d\mu = k$

multiplicative bounds:

$$P_r [X \geq \underline{(1+d)\mu}] \leq \frac{1}{d^2 \mu}$$

$$P_r [X \leq (1-d)\mu] \leq \frac{1}{d^2 \mu}$$

Chernoff Bound

Theorem:

If indicator random variables $X_1 \dots X_n$ are fully independent, and $X = \sum_{i=1}^n X_i$ then

$$P_C [X \geq a] \leq e^{a-\mu} \left(\frac{\mu}{a} \right)^a$$

for all $a \geq \mu = E[X]$

Additive bounds

$$P_C [X \geq \mu + \Delta] \leq e^{-\Delta} \left(\frac{\mu}{\mu + \Delta} \right)^{\mu + \Delta}$$

$$P_C [X \leq \mu - \Delta] \leq e^{-\Delta} \left(\frac{\mu}{\mu - \Delta} \right)^{\mu - \Delta}$$

Multiplicative Bounds

$$P_C [X \geq (1 + \delta)\mu] \leq \left(\frac{e^\delta}{(1 + \delta)^{1 + \delta}} \right)^\mu$$

$$P_C [X \leq (1 - \delta)\mu] \leq \left(\frac{e^{-\delta}}{(1 - \delta)^{1 - \delta}} \right)^\mu$$

If $0 < \delta < 1$:

$$P_r [X \geq (1+\delta)\mu] \leq e^{-\frac{\delta^2 \mu}{3}}$$

$$P_r [X \leq (1-\delta)\mu] \leq e^{-\frac{\delta^2 \mu}{2}}$$

Examples

Flip a fair coin N times.

How many heads do I expect?

$$X_i = \begin{cases} 0 & \text{if } i\text{-th flip is tails} \\ 1 & \text{if } i\text{-th flip is heads} \end{cases}$$

$$\mu = E[X_i] = P_r [X_i = 1] = \frac{1}{2}$$

X = total # of heads

$$X = \sum_{i=1}^N X_i \Rightarrow E[X] = E\left[\sum_{i=1}^N X_i\right] = \sum_{i=1}^N E[X_i] = N \cdot \frac{1}{2} = \frac{N}{2}$$

Q: What's the probability that # of heads is $\geq \frac{3}{4}N$?

$$\geq \alpha N \quad \frac{1}{2} < \alpha < 1$$

Markov's inequality:

$$\Pr[X \geq a] \leq \frac{\mu}{a}$$

$$\Pr[X \geq \alpha N] \leq \frac{\frac{N}{2}}{\alpha N} = \frac{1}{2\alpha}$$

$$\text{E.g. } \alpha = \frac{3}{4} : \Pr[X \geq \frac{3}{4}N] \leq \frac{1}{2 \cdot \frac{3}{4}} = \frac{2}{3}$$

Chebyshev's inequality

Additive: $\Pr[X \geq \mu + \Delta] \leq \frac{\mu}{\Delta^2}$

$$\Pr[X \geq \alpha N] = \Pr[X \geq \frac{1}{2}N + \underbrace{(\alpha - \frac{1}{2})N}_{\Delta}] \leq \frac{\frac{N}{2}}{(\alpha - \frac{1}{2})^2 N^2} = \frac{1}{2(\alpha - \frac{1}{2})^2 N}$$

$$\alpha = \frac{3}{4} : P_r [X \geq \frac{3}{4} N] \leq \frac{1}{2 \left(\frac{3}{4} - \frac{1}{2}\right)^2 N} = \frac{8}{N}$$

Multiplicative: $P_r [X \geq (1+\delta)\mu] \leq \frac{1}{\delta^2 \mu}$

$$P_r [X \geq \underline{\alpha} N] \leq \frac{1}{\delta^2 \mu}$$

$$= \frac{1}{(2\alpha - 1)^2 \cdot \frac{N}{2}} = \frac{2}{(2\alpha - 1)^2 N}$$

$$\left(\frac{1+\delta}{2} \right) N = \alpha N$$

$$\delta = 2\alpha - 1$$

$$\frac{1}{2} < \alpha < 1$$

$$0 < \delta = 2\alpha - 1 < 1$$

$$\alpha = \frac{3}{4} : P_r [X \geq \frac{3}{4} N] = \frac{2}{\left(2 \cdot \frac{3}{4} - 1\right)^2 N}$$

$$= \frac{2}{\left(\frac{1}{2}\right)^2 N} = \frac{8}{N}$$

Chernoff Bound

$$0 < \delta = 2\alpha - 1 < 1$$

$$P_r [X \geq (1+\delta)\mu] \leq e^{-\frac{\delta^2 \mu}{3}}$$

$$P_r [X \geq \alpha N] \leq e^{-\frac{(2\alpha - 1)^2 \cdot \frac{N}{2}}{3}} = e^{-\frac{(2\alpha - 1)^2 N}{6}}$$

$$\alpha = \frac{3}{4} : P_r [X \geq \frac{3}{4} N] \leq e^{-\frac{\left(2 \cdot \frac{3}{4} - 1\right)^2 \cdot N}{6}} = e^{-\frac{1}{4} N / 6}$$

$$= e^{-\frac{N}{24}} = \frac{1}{e^{N/24}}$$

Markov: $P_r \left[X \geq \frac{3}{4} N \right] \leq \frac{2}{3} \leftarrow$

Chebyshev: $P_r \left[X \geq \frac{3}{4} N \right] \leq \frac{8}{N} \leftarrow$

Chernoff: $P_r \left[X \geq \frac{3}{4} N \right] \leq \frac{1}{e^{N/24}} \leftarrow$

\uparrow
of heads