

CSE 241 Class 14

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Today: skip list analysis!

1 Things to Show

- Let H be the height of the tallest pillar in a skip list of size n .
- Want to show that H is “almost certainly” $O(\log n)$.
- Implies that likely cost of **delete** is $O(\log n)$.
- Moreover, will show that expected number of steps when traversing the list during a find is proportional to H .
- Will conclude that expected cost of **find** is “almost certainly” $O(\log n)$
- Similar argument holds for **insert**.

2 Bound on List Height

- Let h_j be height of j th node’s pillar
- From before,

$$\Pr(h_j = t) = \left(\frac{1}{2}\right)^t$$

- What is $\Pr(h_j > t)$?

$$\begin{aligned}\Pr(h_j > t) &= \sum_{i=t+1}^{\infty} \left(\frac{1}{2}\right)^i \\ &= \left(\frac{1}{2}\right)^{t+1} \sum_{i=0}^{\infty} \left(\frac{1}{2}\right)^i \\ &= \left(\frac{1}{2}\right)^{t+1} \cdot \left(\frac{1}{1-1/2}\right) \\ &= \left(\frac{1}{2}\right)^t.\end{aligned}$$

- What can we now say about $\Pr(H > t)$?

$$\begin{aligned}\Pr(H > t) &= \Pr(h_1 > t \text{ or } h_2 > t \text{ or } \dots \text{ or } h_n > t) \\ &\leq \Pr(h_1 > t) + \Pr(h_2 > t) + \dots + \Pr(h_n > t) \\ &= n \cdot \left(\frac{1}{2}\right)^t.\end{aligned}$$

Why did we use \leq instead of $=$ above? This is example of the union bound, which is the simplest of the Bonferroni inequalities:

$$\begin{aligned}\Pr(h_1 > t \text{ or } h_2 > t) &= \Pr(h_1 > t) + \Pr(h_2 > t) - \Pr(h_1 > t \text{ and } h_2 > t) \\ &\leq \Pr(h_1 > t) + \Pr(h_2 > t)\end{aligned}$$

Graphically, can illustrate as

- OK, so in what sense is it “very likely” that H is $O(\log n)$?
- Note that

$$\begin{aligned}\Pr(h_j > c \log n) &= \left(\frac{1}{2}\right)^{c \log n} \\ &= \frac{1}{n^c}.\end{aligned}$$

- Moreover, same substitution gives

$$\Pr(H > c \log n) \leq \frac{1}{n^{c-1}}.$$

- In other words, the probability of H exceeding $c \log n$ goes down polynomially with n for any fixed c .
- Hence, large deviations of H above $c \log n$ for small c are very unlikely and only get less likely as n grows.

- In this sense, we say that H is “very likely” $O(\log n)$.
- **Defn:** let E_α be a probabilistic event parameterized by a number α . We say that E_α occurs *with high probability* if for $\alpha > 1$,

$$\Pr(E_\alpha) \geq 1 - \frac{k_\alpha}{n^\alpha}$$

where k_α is a constant that depends only on α , not on n .

- (In above example, E_α is “ $H \leq (\alpha + 1) \log n$ ”.)
- High-probability bounds are a stronger result than expected times, and they are preferred when studying randomized algorithms.
- (Skip list handout also proves directly that $E[H] = O(\log n)$; math is a bit icky, so we don’t reproduce it here.)

3 Cost of Search

We will compute *expected* cost of a search, as full WHP result is again somewhat icky.

- Consider the trajectory taken by a search in the skip list.
- Can divide trajectory into horizontal and vertical “steps”

- Total search cost is proportional to number of steps.
- Sum of all vertical steps is $H - 1$ (start at height H , end at height 1).
- Can we get a bound on the number of horizontal steps?

To make things easier to see, we will run the search “backwards”!

- WLOG, we will assume that search always goes to the bottom of the skip list. (If we find the key early, drop down to the bottom level.)
- Backward traversal begins with a run of 0 or more horizontal moves from nodes of height 1.
- At some point, we encounter a node of height > 1 , at which point we *must* take a vertical step up.

- (Why? If we pass a node of height > 1 at level 0 and only later jump up to a higher level, that would imply that the traversal algorithm went down when it could have moved forward to a node with a key less than the target.)

- Traversal then takes 0 or more horizontal moves from nodes of height 2, until it sees a node of height > 2 .
- Repeat this pattern until we reach the head while traversing at height H .

Now, here's the key question.

- How many horizontal steps do we expect to take at a given level before taking a vertical step?
- Equivalently, how many nodes of height exactly t do we expect to encounter before the first node of height $> t$?
- Recall that heights are determined independently for every node by a geometric distribution with parameter $1/2$.
- Given that a node has height at least t , the probability that it has height $> t$ is given by

$$\begin{aligned}
 \Pr(h > t \mid h \geq t) &= \frac{\Pr(h > t \cap h \geq t)}{\Pr(h \geq t)} \\
 &= \frac{\Pr(h > t)}{\Pr(h > t - 1)} \\
 &= \frac{(1/2)^t}{(1/2)^{t-1}} \\
 &= 1/2.
 \end{aligned}$$

- Conceptually, we can model the sequence of nodes of height $\geq t$ as a sequence of coin flips. Each tail is a node of height $= t$, while the first head is a node of height $> t$ that ends the sequence.
- Each tail causes one horizontal step.
- What is the number n_τ of tails before the first head?
- Chance that $n_\tau = j$ is $(1/2)^{j+1}$ for $j \geq 0$.

- Hence,

$$\begin{aligned} E[n_\tau] &= \sum_{j=0}^{\infty} j \left(\frac{1}{2}\right)^{j+1} \\ &= \frac{1}{2} \sum_{j=0}^{\infty} j \left(\frac{1}{2}\right)^j \end{aligned}$$

- Time to bust out a summation formula:

$$\sum_{j=0}^{\infty} jx^j = \frac{x}{(1-x)^2}$$

for $0 < x < 1$.

- Plugging in $x = 1/2$, we get

$$\begin{aligned} E[n_\tau] &= \frac{1}{2} \frac{1/2}{(1 - 1/2)^2} \\ &= \frac{1}{2} \frac{1/2}{1/4} \\ &= 1. \end{aligned}$$

- Finally, the total number of steps N over the whole algorithm is given by $N = H - 1 + \sum_{t=1}^H n_\tau$ (since the analysis of horizontal steps holds at every level).
- Conclude that

$$\begin{aligned} E[N] &= H - 1 + \sum_{t=1}^H E[n_\tau] \\ &= H - 1 + \sum_{t=1}^H 1 \\ &= 2H - 1. \end{aligned}$$

- Since $H = O(\log n)$ WHP, we have that WHP, the expected number of steps taken by search is also $O(\log n)$.

One final thing: where did that summation formula come from?

- We know that

$$\sum_{j=0}^{\infty} x^j = \frac{1}{1-x}$$

for $0 < x < 1$.

- Differentiate both sides w/r to x :

$$\sum_{j=0}^{\infty} jx^{j-1} = \frac{1}{(1-x)^2}$$

- Multiply both sides by x :

$$\sum_{j=0}^{\infty} jx^j = \frac{x}{(1-x)^2}$$