CS 240 – Data Structures and Data Management

Module 6E: Dictionaries for special keys - Enriched

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Winter 2022

version 2022-02-08 13:47

Outline

- A tighter lower bound
- Improving binary search
- More on interpolation search
- More on pruned tries

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$$x_0 \quad x_1 \quad x_2 \quad x_3 \quad x_4 \quad x_5 \quad x_6 \quad \cdots \quad x_{n-1}$$

Search: $\left(\begin{array}{c} x_0 \quad x_1 \quad x_2 \quad x_3 \quad x_4 \quad x_5 \quad x_6 \quad \cdots \quad x_{n-1} \\ & x_{n-1} \quad x_{n-1}$

• **Claim:** These instances must lead to distinct leaves (assuming no equality-comparison).

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• So we require at least $\lceil \log(2n+1) \rceil$ comparisons.

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- *binary-search* uses $\approx 2 \log n$ comparisons.
- Goal: Improve it to use $\lceil \log(2n+1) \rceil \approx \log n + 1$ comparisons.
- Main ingredient: Do only one comparison per round.

```
binary-search-optimized(A, n, k)
A: Sorted array of size n, k: key
1
    \ell \leftarrow 0, r \leftarrow n-1, \chi \leftarrow 0
    while (\ell < r)
2.
            m \leftarrow \left| \frac{\ell + r}{2} \right|
3.
             if (A[m] < k) then \ell \leftarrow m + 1
4.
             else r \leftarrow m, \chi \leftarrow 1
5.
                                                              // this is different!
6.
     if (k < A[\ell]) then return "not found, between A[\ell-1] and A[\ell]"
    else if \chi = 1 or (k \leq A[\ell]) then return "found at A[\ell]"
7.
        else "not found, between A[\ell] and A[\ell+1]"
8.
```

(χ needed for optimum # of comparisons, but not normally used)

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- Claim 4: If χ is used, then # comparisons ≤ [log(2n + 1)]. (Straightforward but tedious cases. See textbook for details.)
- This uses the *optimum* number of comparisons and also in practice performs better than *binary-search*.
 - But normally omit χ (only needed in Claim 4)
 - ► Can replace two comparisons in lines 6-7 by equality-comparison.

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- Had: Average-case run-time of *interpolation-search* is $O(\log \log n)$.
- This is very complicated to prove!
 - > Study error, i.e., distance between index of k and where we probed.
 > Argue that error is in O(√n) in first round.
 > Argue that error is in O(½n) after i rounds.
 > Study the martingale formed by the errors in the rounds.
 > Argue that its expected length is O(log log n).
- Instead: Define a variant of *interpolatation-search*
 - Better worst-case run-time.
 - Easier to analyze.
- Idea: *Force* the sub-array to have size \sqrt{n}
- To do so, search for suitable sub-array with probes.
- Crucial question: how many probes are needed?



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- Observe: $\# \text{ probes} \in O(\sqrt{N})$

Interpolation-search-modified (A, n, k)A: sorted array of size n, k: key if (k < A[0] or k > A[n-1]) return "not found" 1 2. **if** (k = A[n-1]) **return** "found at index n-1" 3. $\ell \leftarrow 0, r \leftarrow n-1$ // have $A[\ell] < k < A[r]$ 4. while $(N \leftarrow (r - \ell - 1) \ge 1)$ $m \leftarrow \ell + \left\lceil \frac{k - A[\ell]}{A[r] - A[\ell]} \cdot (r - \ell - 1) \right\rceil$ 5. if (A[m] < k)// probe rightward 6 for h = 1, 2, ...7 $\ell \leftarrow m + (h-1) \lceil \sqrt{N} \rceil, r' \leftarrow \min\{r, m + h \lceil \sqrt{N} \rceil\}$ 8. if (r' = r or A[r'] > k) then $r \leftarrow r'$ and break 9. // symmetrically probe leftward 10 else ... if $(k = A[\ell])$ return "found at index ℓ " 11. else return "not found" 12

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- Use a sloppy recursion:

$$T^{ ext{worst}}(n) \leq \left\{ egin{array}{cc} c & n \leq 15 \ T^{ ext{worst}}(\sqrt{n}) + c \cdot \sqrt{n} & ext{otherwise} \end{array}
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- Easy induction proof: $T^{\text{worst}}(n) \leq 2c\sqrt{n}$.
- Therefore worst-case run-time is $O(\sqrt{n})$.

- What is the number of probes on average?
- Rephrase: If numbers are chosen uniformly at random, what is the expected number of probes?
- Claim: Expected number of probes is $c \le 2.5$.

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- Therefore the average-case # comparisons is $\leq 2.5 \lceil \log \log n \rceil$.
- Fewer than *binary-search-optimized*'s $\lceil \log n \rceil + 1$ for $n \ge 16$.

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Pruned tries and MSD-radix sort

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Pruned tries can store real numbers

If we have a generator for each bit of a real number, then we can store them in a pruned trie.

