Introduction to Algorithms and Data Structures

Lesson 4: Searching (2) Block search

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Search Problem

- Problem: S is a given set of data. For any given data x, determine efficiently if S contains x or not.
- Efficiency: Estimate the time complexity by n = |S|, the size of the set S
 - In this problem, "checking every data in S" is enough, and this gives us an upper bound O(n) in the worst case.

Roughly, "the running time is proportional to *n*."

Data structure 2 Data in the array in increasing order

- This is something like dictionary and address book...
- Q: Do you use sequential search algorithm

when you check dictionary?



Idea of m-block method

(0) Divide the array into m blocks B_0 , B_1 , ..., B_{m-1}

(1) Check the biggest item in each block,

and find the block B_i that can contain x

(2) Perform sequential search in B_i

Idea of m-block method

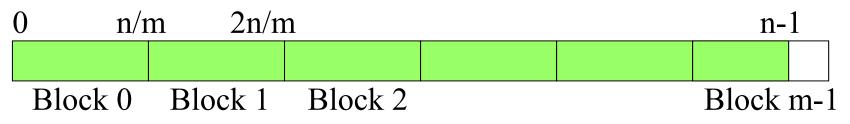
(0) Divide the array into m blocks B_0 , B_1 , ..., B_{m-1}

 (1) Check the biggest item in each block, and find the block B_i that can contain x

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Simple implementation:

divide into the blocks of same size except the last one.



- Each block has length k, where $k = \lceil n/m \rceil$
- Block B_j has items from s[jk] to s[(j+1)k-1]: $B_j = [jk, (j+1)k-1]$

Idea of m-block method

(0) Divide the array into m blocks B₀, B₁, ..., B_{m-1}
(1) Check the biggest item in each block, and find the block B_j that can contain x
(2) Perform sequential search in B_j

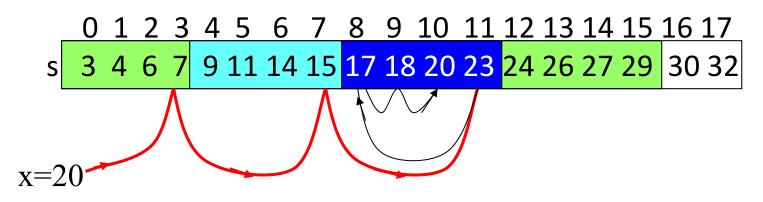
If the program exits from the loop, the variable j indicates the index of the block, and j indicates the last one otherwise.

Idea of m-block method

(0) Divide the array into m blocks B₀, B₁, ..., B_{m-1}
(1) Check the biggest item in each block, and find the block B_j that can contain x
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i=j*k; t = min{ (j+1)*k-1, n-1 };
while( i < t )
    if x s[i] then exit from the loop;
    else i=i+1; //next item in the block
if x == s[i] then return i and halt;
else return -1 and halt.</pre>
```

Example and time complexity



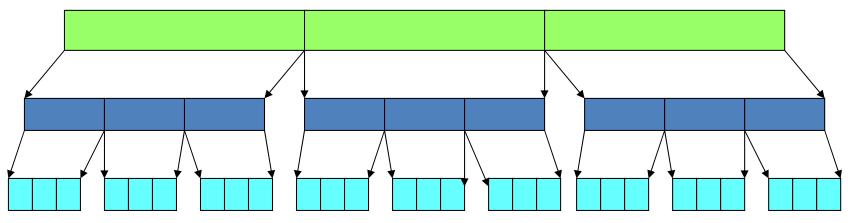
- # of comparisons # of blocks + length of block = m + n/m
- What the value of m that minimize m + n/m ?
 - Let f(m) = m + n/m, and take the differential for m
 - $f'(m) = 1 n/m^2 = 0 \rightarrow m = \sqrt{n}$
 - When m = \sqrt{n} , # of comparisons $\sqrt{n} + n/\sqrt{n} = 2\sqrt{n}$
- Time complexity: $O(\sqrt{n})$

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For example, when n=1000000,
Linear search takes n/2=500000 comparisons, but
Block search takes v1000000=1000 comparisons!!
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Algorithm 3: Double m-block method

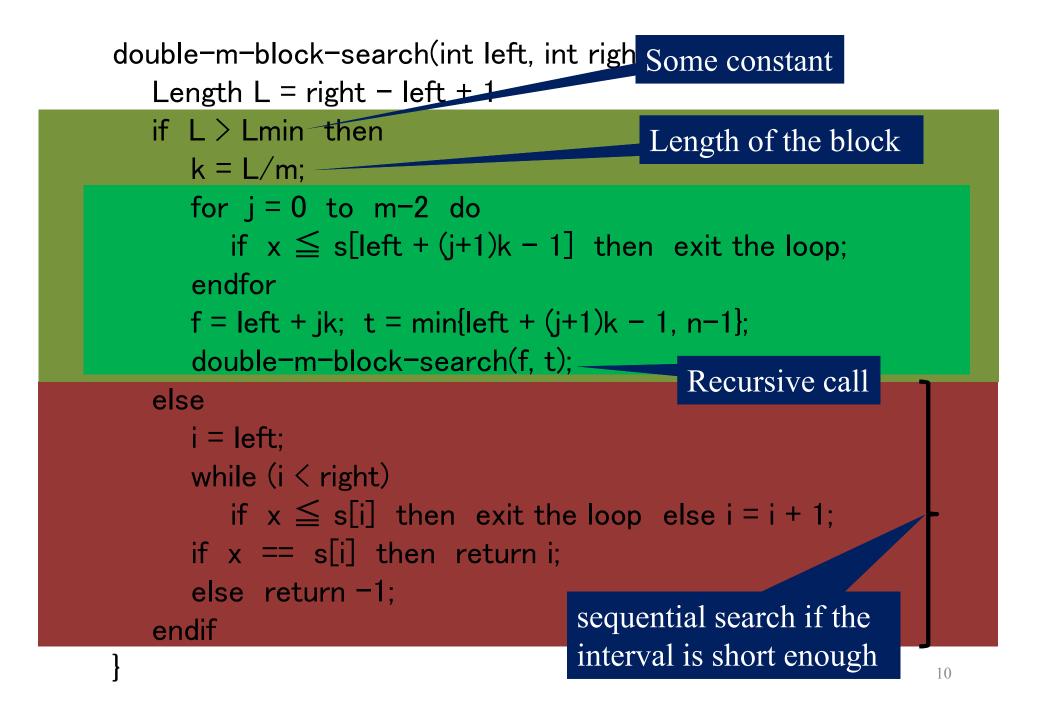
In the m-block method, we use sequential search in each block. We can use m-block method again in the block!!

Recursive call: basic and strong idea

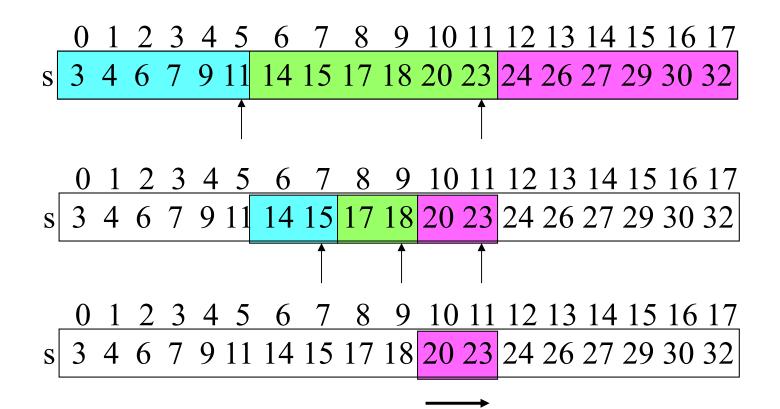


Idea of double m-block method

Divide search area into m blocks, and repeat the same process for the block that contains x, and repeat again and again up to the block has length at most some constant N



Example: find 20 (x=20) for block size 3



Analysis of time complexity

- Length of search space $n \to \left\lceil \frac{n}{m} \right\rceil \to \left\lceil \frac{\left\lceil \frac{n}{m} \right\rceil}{m} \right\rceil \to \left\lceil \frac{\left\lceil \frac{n}{m} \right\rceil}{m} \right\rceil \to \cdots$
- Let n_i be the length after the *i*-th call

. . .

$$n_1 = \left\lceil \frac{n}{m} \right\rceil \le \frac{n}{m} + 1$$
$$n_2 = \left\lceil \frac{n_1}{m} \right\rceil \le \frac{n}{m^2} + \frac{1}{m} + 1$$

$$n_i \le \frac{n}{m^i} + \sum_{j=0}^{i-1} \frac{1}{m^j} \le \frac{n}{m^i} + 2$$

Analysis of time complexity

- The length n_i after the *i*-th recursive call: $n_i \quad n/m^i + 2$
- How many recursive calls made? $n_{i} \leq \text{Lmin} \iff \text{Lmin} \geq \frac{n}{m^{i}} + 2 \iff i \geq \log_{m} \frac{n}{\text{Lmin} - 2}$
- Each recursive call make at most m-1 comparisons, so the total number of comparisons is $\leq (m-1)\log_m \frac{n}{L\min 2} + L\min$
- The time complexity is O(log *n*)

Analysis of time complexity: The best value of m

- $T(n,m) = (m-1)\log_m \frac{n}{\operatorname{Lmin} 2} + \operatorname{Lmin}$ = $\frac{m-1}{\log_2 m}\log_2 \frac{n}{\operatorname{Lmin} - 2} + \operatorname{Lmin}$
- To make T(n,m) the minimum, smaller m is better because m-1 grows faster than log₂ m (which will be checked in the big-O notation).
- Therefore, m=2 is the optimal

We will have "binary search"