

CAN WE CUT A HAMBURGER EXACTLY IN HALF SO THAT EACH HALF HAS EXACTLY 50% OF THE CHEESE, 50% OF THE MEAT, AND 50% OF THE BREAD? Yes.

RED AND BLACK POINTS

In 2D, this theorem is known as the pancake theorem. It implies that we can perfectly bisect n red points and n black points in the plane, so that each side has $n/2$ red and $n/2$ black points.

Application: We can divide m points into four sets of $n/4$ points each by two lines (Figure 2). First draw a horizontal line that divides the points in half. Color all points below the line red, and above the line black. By the pancake theorem, we can find a second line that evenly divides red and black, yielding $n/4$ points in each of the four sets of the partition.

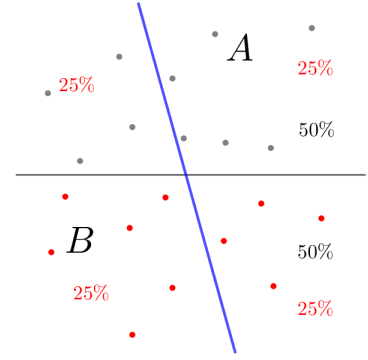


Figure 2: Illustration of the application.

Half Space Counting Problem

Let x_1, \dots, x_n be drawn from \mathbb{R}^2 . Make a data structure such that one can “efficiently” answer queries that take as input a line given by the user, and outputs the number of points on one side of the line.

DATA STRUCTURE (DIVIDE-AND-CONQUER) $\text{COUNT}(\ell, S)$ is a divide-and-conquer algorithm for computing the number of points of S on one side of ℓ , say the side that contains the origin. Assume that we partitioned our space recursively using the 25%-trick suggested by the pancake theorem.

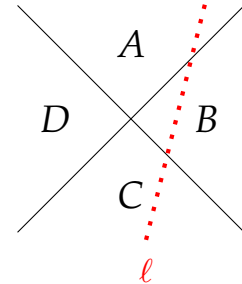
we return the count to the set S .

$\text{COUNT}(\ell, S)$

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1  if  $|S| \leq 10$ 
2      do it manually
3  else
4      determine the three sets cut by  $\ell$ , say  $A, B, C$ 
5      if the fourth set is on the good side of  $\ell$ 
6          return  $|S|/4 + \text{COUNT}(\ell, A) + \text{COUNT}(\ell, B) + \text{COUNT}(\ell, C)$ 
7      else
8          return  $\text{COUNT}(\ell, A) + \text{COUNT}(\ell, B) + \text{COUNT}(\ell, C)$ 
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We can indeed determine A, B, C in constant time. In the RAM model, $T_n = 1 + 3T_{n/4}$, which yields $T_n = \Theta(n^{\log_4 3})$ by the master theorem.

Figure 3: Illustration of the $\text{COUNT}(\ell, S)$ algorithm.*Linear time selection: finding the k^{th} smallest element*

We present a selection algorithm invented by Blum, Floyd, Pratt, Rivest and Tarjan² for finding the k^{th} smallest number in a list or array; such a number is called the k^{th} order statistic. This includes the cases of finding the minimum, maximum, and median elements.

² Blum et al. [1973]

In particular, we want to find the k^{th} smallest number in an unordered set $S = \{x_1, \dots, x_n\}$, and can use a comparison oracle. By sorting we would have time complexity $O(n \log_2 n)$.

We will give an $O(n)$ complexity solution, called $\text{SELECT}(k, S)$, where S is the collection of elements, and $1 \leq k \leq |S|$.

$\text{SELECT}(k, S)$

```

1  if  $|S| \leq 5$ 
2      SORT( $S$ )
3      return the  $k^{\text{th}}$  smallest element of  $S$ 
4  else
5      ① group all elements in groups of 5, and find the median in
        each group and call the set of medians  $M$ 
        // This costs 6 comparisons.
6      ② let  $m = \text{SELECT}(|M|/2, M)$ 
        //  $m$  is the median of the medians
7      ③ compare all elements of  $S$  with  $m$ , forming the sets  $L$  and  $R$ 
        of smaller and larger elements
8      ④ if  $|L| == k - 1$ 
9          return  $m$ 
10     elseif  $|L| \geq k$ 
11         return  $\text{SELECT}(k, L)$ 
12     else
13         return  $\text{SELECT}(k - |L| - 1, R)$ 
    
```

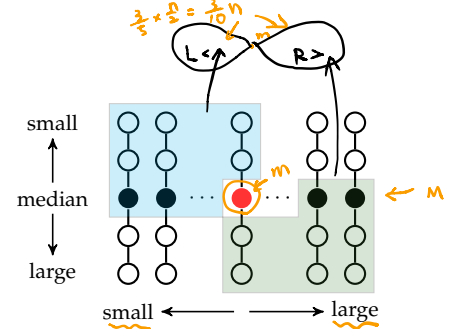


Figure 4: Medians of the median.

$$\frac{3n}{10} \leq |L|, |R| \leq \frac{7n}{10}$$

$\frac{n}{5} \times 6$ 5 数选 M . 6 个 (pair)

$T_{n/5}$

$\frac{4n}{5} - 1$

$\frac{3n}{10} \times 2$ 已经了.

$$T_n \leq \begin{cases} T_{n/5} + T_{\frac{7n}{10}} + \frac{11}{5}n, & n > 5 \\ 7, & n \leq 5 \end{cases}$$

The bound on the cost for every critical step of the $\text{SELECT}(k, S)$ algorithm is listed in the table below:

Line number	Bound on the cost
1 - 3	≤ 7
5	$\leq 6n/5$
6	$= T_{n/5}$
7	$= n$
11	$= T_{ L } \leq T_{7n/10}$
13	$= T_{ R } \leq T_{7n/10}$

A POSSIBLE IMPROVEMENT. As shown in Figure 5, L and R each have at least $3|S|/10$ and at most $7|S|/10$ elements. Hence, to identify L and R , if we program correctly, we only require at most $4n/10$ new comparisons. Therefore, the recurrence that calculates the complexity can be reduced to $T_n \leq 8n/5 + T_{n/5} + T_{7n/10}$. We will prove by induction that $T_n \leq Cn$ for all $n \in \mathbb{N}$.

Proof. Base case: If $n \leq 5$, then $C \geq 7$ is a safe choice.

Induction hypothesis: Assume $T_k \leq Ck$ for all $k < n$.

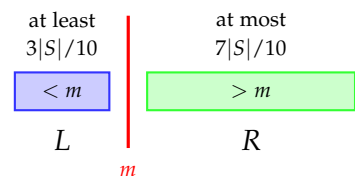


Figure 5: Illustration of L and R .

0 0
0 0
2

Inductive step: $T_n \leq 8n/5 + T_{n/5} + T_{7n/10} \leq 8n/5 + C \cdot 9n/10$.
 This should be $\leq Cn$, so we have the requirement $8n/5 \leq C \cdot n/10$, or $C \geq 16$. We have thus shown that $T_n \leq 16n$ for all $n \in \mathbb{N}$. \square

MEDIAN-OF-3 RECURRENCE. When we replace the median-of-5 step by a median-of-3 step, then one can see that $T_n = \Theta(n) + T_{n/3} + T_{2n/3}$. This equation is analyzed via a recursion tree. Note that in each level the work adds up to n . We have $(\frac{2}{3})^k n = 1$, where k is the height of the recursion tree, so $k = \log_{3/2} n / \log_2(3/2)$. Therefore, $T_n = \Theta(n \log_2 n)$.

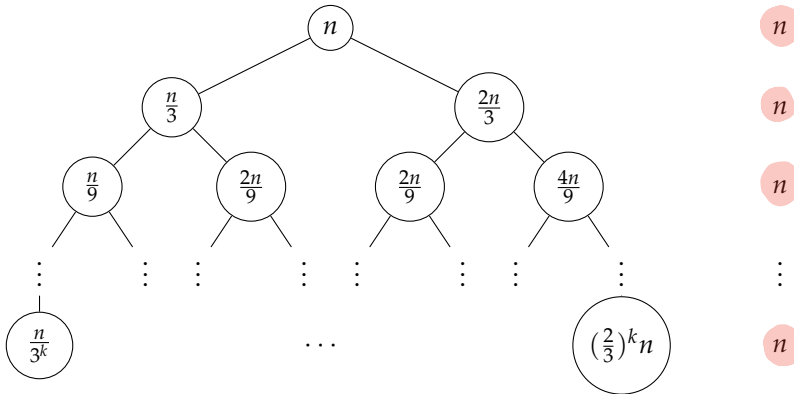


Figure 6: Recursion tree.

Observe that the recursion tree is not balanced. Nevertheless, the work at each level is precisely n .

References

M. Blum, R. W. Floyd, V. R. Pratt, R. L. Rivest, and R. E. Tarjan. Time bounds for selection. *Journal of Computer and System Sciences*, pages 448–461, Aug 1973. DOI: 10.1016/S0022-0000(73)80033-9. URL <http://people.csail.mit.edu/rivest/pubs/BFPRT73.pdf>.

Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein. *Introduction to Algorithms*. The MIT Press, 3rd edition, 2009. ISBN 978-0-262-53305-8.

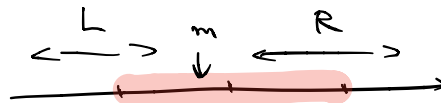
Brian Libgober. The Borsuk-Ulam and ham-sandwich theorems. May 2008. URL <http://www.math.uchicago.edu/~may/VIGRE/VIGRE2008/REUPapers/Libgober.pdf>.

"Find"

T_n random. $(L) \quad m \quad (R)$
 random

$$T_n = n - 1 + \begin{cases} T_{|L|} \\ T_{|R|} \\ 0 \end{cases}$$

$T_{|L|} \leq T_{\frac{3n}{4}}$ half the time
 \uparrow
 $\leq T_n$ half the time.



$E\{T\}$ expected time. avg

$$E T_n \leq \frac{1}{2} (E T_n + E T_{\frac{3n}{4}}) + n.$$

$$2 E T_n \leq E T_n + E T_{\frac{3n}{4}} + 2n.$$

worst case n^2

$$E T_n \leq E T_{\frac{3n}{4}} + 2n \leq Cn$$

$$\frac{3n}{4} C + 2n \leq Cn$$

$$C \geq 8.$$

$T_m = \text{max time taken on any problem of size } \underline{\text{max less.}}$
 $T_m \uparrow \text{ as } m$

$$T_m \leq \begin{cases} T_{m/5} + T_{m/10} + \frac{11}{5}m, & m > 5 \\ 7, & m \leq 5 \end{cases}$$

Will show: " $T_m \leq Cn$ " for some C for all m

INDUCTION

$m \leq 5$ $T_m \leq 7 \leq Cn$ $\forall 1 \leq m \leq 5$ should be

$m > 5$, assume truth up to $m-1$. ($C \geq 7$ needed)

$$T_m \leq T_{m/5} + T_{m/10} + \frac{11}{5}m \leq Cn \left(\frac{1}{5} + \frac{1}{10} \right) + \frac{11}{5}m$$

$$= \frac{9C}{10} + \frac{11}{5}m \leq Cn$$

$\Rightarrow C \geq 22$

BIN SEARCH ($x; i, j$) $i=1$
 $j=n$

Cases:

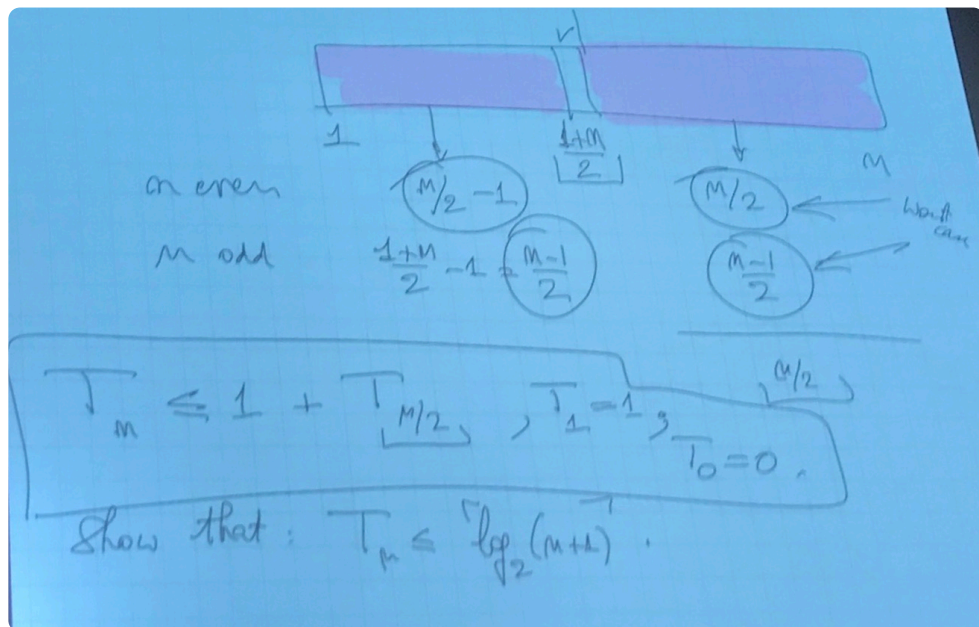
- $i > j$: return "x not present"
- $i = j$: Ternary Oracle (x, a_i)
return either " $x = a_i$ " or "x not present".
- $i < j$: $k = \lfloor \frac{i+j}{2} \rfloor$
Ternary Oracle (x, a_k)

Cases:

- $x < a_k$: return BIN SEARCH ($x; i, k-1$)
- $x = a_k$: return " $x = a_k$ ".
- $x > a_k$: return BIN SEARCH ($x; k+1, j$)

Ternary Oracle

列最后
index 以下
之后再分
析 base
case



Induction proof.
 $n=0, n=1: \checkmark$
 assume true up to $n-1$. ($n \geq 2$)

$T_n \leq 1 + T_{\lceil n/2 \rceil}$

n even $= 1 + T_{n/2} \leq 1 + \lceil \log_2(\frac{n}{2} + 1) \rceil$

$\lceil \log_2 23 \rceil$
 $= \lceil \log_2 24 \rceil$

$= 1 + \lceil \log_2(n+2) \rceil - \lceil \log_2 2 \rceil$

$= \lceil \log_2(n+2) \rceil$

should be $\lceil \log_2(n+1) \rceil$