LECTURE 9 (September 24^{th})

Midterm $1 \longrightarrow$ Mon, September 30^{th} , 7-9 pm - No lecture on Thursday – Optional Midterm Review Session Sample Exams on Jeff's website or previous years' offerings – Exam will cover material upto HW3 Pre-requisites
- Divide & Conquer, FFT Dynamic Progmmaning No Randomized Algorithms / Probability on Midterm 1 - 4 questions & 2 Hours Let the Solutions will be short should take 5-10 minutes to write LAC CHARGOMEELL FUORING TRODELOUTLY ON PILATER Clear and precise des
some justification, e. some justification, e.g., recurrence Run time - One handwritten cheat sheet – one double-sided A4 page - Conflict Exam On & Registration Form on Course Wetpage - DRES accommodation - Please send the DRES letter by - DRES accommod
- No homework t
Matching Nuts & Bolts
Introduced by Greg

We have ^a bag of ⁱ nots and in bolts. All nots are of different sizes and so are the bolts we have a bag of n nuts and n bolts. All nuts
but for every nut there is exactly one matching bolt.

Only operation allowed Compare a nut to a bolt and see if it fits or if the nut is bigger or the bolt is bigger

- No homework this week

Matching Nuts & Bolts

Introduced by Gregory Rawlins

Fask Match every not to its corresponding bolt

cannot compare two nuts with each other or two bolts with each other

How would you do it?

Brute Force Compare every nut with every bolt $- n^2$ tries One can save a factor of two since if we find a matching nut for a bolt we clon't need to compare it to others, but it is still $\theta(n^2)$ tries

This problem should remind you of sorting

If the bolts are sorted, we can pick a nut and find a matching bolt with binary search is are pous are sorred, we can put a me and you a matching pow vital boldy

This process of assuming a solution to one problem and using it to solve another is called a reduction. So, we can reduce the nut-bolt matching problem to sorting.

It turns out that this task is equivalent to sorting. If all the nuts and bolts are matched, then we can also sort in O(n logn) time because if we want to compare two nuts with each other, we can compare them to their corresponding bolt. hed, then we can also sort in O(n logn) time because if we want to
two nuts with each other, we can compare them to their corresponding bolt. So, we can run mergesort to sort in $O(n \log n)$ time.

So, to solve the task, let's try a sorting algorithm e.g. Mergesort. Can we do mergesort with nuts & bolts?

At a minimum, we can make quicksort work + pick a bolt, call it the pivot bolt .
use that bolt to partition the nuts into three parts - thase \int that are bigger than the pivot bolt, those that are smaller and the matching nut

then we can partition then we can partition the botts, using the not that matches the pivot bolt

Recall that mergesort does the following to an inpot array :

 \Box split it into two halves ω sort each half E Merge the two sorted halves

Why can't we do this here ?

Problem comes from step \Box of mergesort.

We can split the nuts into two parts but to recurse, we need to split the bolts into two parts that match the partition of the nuts. Not obvious how to do that!

So , this doesn't work , as there is no way to recurse .

What about Quicksort?

Now , we can recurse since we know in one part all of the nuts/bolts are smaller than the pivot , and in the other , larger than the pivot

basic If we try to do this cleterministically, this is $\Theta(n^2)$ in the Worst-case, just like quicksort since everytime in the recursion, we might get unlucky and the problem splits into two unbalanced parts

For quicksort, we can use median of medians to pick a clever pivot and reduce the running time to OLn. log n) but there seems to be no way of doing that here

It is possible to do it deterministically in OCn.lop n) time, but the algorithm is quite complicated and it took a lot of work to find it.

Picking a random pivot , also gives an O(n·logn) time randomized algorithm that is not so difficult to analyze.

Randomized Algorithm to Match Nuts & Bolts

& Pick ^a random pivot 11 Partition nuts/bolts into bigger a smaller parts a matched pair β Recurse on the two sub-problems

How many tests do we need to perform for ?

$$
\boxed{\phantom{\mathbf{2n-1}} \mathbf{1} \mathbf{1} \mathbf{1} \mathbf{1} \mathbf{2} \mathbf{1} \mathbf{1} \mathbf{2} \mathbf{1} \mathbf{1} \mathbf{2} \mathbf{1} \mathbf{3} \mathbf{3} \mathbf{1} \mathbf{2} \mathbf{3} \mathbf{3} \mathbf{4} \mathbf{4} \mathbf{5} \mathbf{5} \mathbf{6} \mathbf{7} \mathbf{8} \mathbf{8} \mathbf{9} \mathbf{1} \mathbf{0} \mathbf{1} \mathbf{0} \mathbf{1} \math
$$

Let $T(n)$ = running time for n nots a bolts

Then, $T(n) = 2n - 1 + T(k-1) + T(n-k)$ for some value of k

If we were looking at the worst-case behavior, then we would take a max over all k. But since we chosen the pivot randomly, we are interested in the

expected value of the running time, since k is random
\nSo,
$$
E[T(n)] = 2n - 1 + E_K [E[T(K-1)] + E[T(n-K)]]
$$

\nSince the pivot bolt is equally likely to be the smallest, second smallest or k^{th} smallest
\nfor any k,
\n $E[K=k] = \frac{1}{n}$ $\forall k$, so our recurrence becomes
\n $E[T(n)] = 2n-1 + \frac{1}{n} \sum_{k=1}^{n} E[T(k-1)] + E[T(n-k)]$
\n $= 2n-1 + \frac{1}{n} \sum_{k=0}^{n} E[T(k)]$

How to solve this recurrence? Such recurrences are called, full history recurrences and there's a trick to solving them which you can read in the Lecture notes, but here we will try something else

#ervistic calculation - Not ^a proof !

What are the bad cases for quicksort? When sub-problems are very unbalanced
Call a pivot good if
$$
\frac{n}{4} < k < \frac{3n}{4}
$$
. Then, $P\left[pivot \text{ is good}\right] = \frac{1}{2}$

So,
$$
\mathbb{E}[T(n)] = \frac{1}{2} \mathbb{E}[T(n) | good] + \frac{1}{2} \mathbb{E}[T(n) | bad]
$$

Now, we can make the unjustified assumption that the running time only increases the more unbalanced our pivot. So, even when the pivot is good, the worst-case occurs when $k = \frac{n}{4}$ or $\frac{2n}{4}$. Similarly, when pivot is bad, then the worst-case occurs When one sub-problem has size zero .

With this "heuristic" assumption,
\n
$$
\mathbb{E}[\top(n)] \leq 2n-1 + \frac{1}{2} \left[\mathbb{E}[T(\frac{n}{4})] + \mathbb{E}[T(\frac{3n}{4})] \right] + \frac{1}{2} \mathbb{E}[T(n-1)]
$$
\n
$$
\Rightarrow \mathbb{E}[T(n)] \leq 4n-2 + \mathbb{E}\left[T(\frac{n}{4})\right] + \mathbb{E}[T(\frac{3n}{4})] \leq \mathbb{E}[T(n)]
$$

We can solve this by using recursion trees

Thus, heuristically $\mathbb{E}[\top(n)] = O(n\log n) \rightarrow \pi$ is relies on an unjustified assumption which would be true if $E[T(n)]$ is a convex function of n. This turns out to be true but proving it is not easy.

So, to prove it rigorously, we introduce a different method which illustrates one of the most useful ideas in the analysis of randomized algorithms - decomposing the running time as a sum of zero-one random variables, called indicator variables.

Rigorous Analysis

Let
$$
X_{ij} = \begin{cases} 1 & \text{if } i^{th} \text{ smallest bolt is compared to } j^{th} \text{ smallest not} \\ 0 & \text{of } j^{th} \text{ smallest.} \end{cases}
$$

Then,
$$
T(n) = \sum_{ij} X_{ij}
$$

\n
$$
\mathbb{E} [T(n)] = \sum_{ij} \mathbb{E} [X_{ij}]
$$
\n
$$
= \sum_{ij} (1 \cdot \mathbb{P}[X_{ij} = 1] + 0 \cdot \mathbb{P}[X_{ij} = 0])
$$
\n
$$
= \sum_{ij} \mathbb{P}[X_{ij} = 1]
$$

Let's look at a few cases to see what $IP[X_{ij} = 1]$ is.

 \cdot $\mathbb{P}[X_{\lambda,\tau}$ = 1]=1 since ith bolt must be compared to ith nut in the end

•
$$
\mathbb{P}[X_{1,h}=1]=\frac{2}{h} \Leftarrow
$$
 There are several cases here :

(1) pivot is the 1^{5t} bott \Rightarrow $X_{1n} \circ 1$ \Rightarrow probability $\frac{1}{h}$ (2) pivot is the n^{th} bolt \Rightarrow $X_{1n} = 1$ \Rightarrow probability $\frac{1}{n}$ (3) pivot is κ^{th} bott where \Rightarrow x_{1h} = 0 2 <k<n re are several cases
pivot is the 1^{5t}
pivot is the n^{th}
pivot is kth both v
1 k
1 k
1 k
1 k
1 k
1 k
1 k 1 botts If k is selected , the left and right problems are and right problems are

and right problems are

and right problems are

and right problems are

solved separately and no

to the top left is

to the top left is n nots solved separately and no
n bolt on the top left is compared to a nut in the bottom right

Intuitively, if a bolt and nut is closer in rank, it has a higher chance of Intuitively, if a bolt and nut is closer in rank, it has a l
being compared. For ith bolt and jth nut, if li-jl is small,

This will be divided into two subproblems in recursion - the left \bm{a} right of the pivot

So , in this case where ^K <i , we recorse $(\alpha \kappa \kappa)$ j)

For the right subproblem, the original hut bolt pair we were interested in are now

Note that $|i-j| = |(i-k) - (j-k)|$ so the difference in their ranks has not changed

Case K < i	Recursive		
Case K = i	$X_{ij} = 1$	If the first pivot is chosen from $\{i, \ldots, j\}$, then	
Case i < k < j	$X_{ij} = 0$	there are $1j - i1 + 1$ cases where we either compare	
Case k = j	$X_{ij} = 1$	Out of these there are exactly two cases where $X_{ij} = 1$	
Case K > j	Recursive	$X_{ij} = 1$	So, $\mathbb{P}[X_{ij} = 1 k \cdot i \text{ or } i < k < j \text{ or } k = j] = \frac{2}{1j - i1 + 1}$

To handle the recursive cases , we can use induction

Since
$$
xy_1 = 1
$$

\nSo, $P[X_{ij} = 1 | k-i \text{ or } i < k < j \text{ or } k = j] = \frac{2}{i j - i}$

\nTo handle the recursive cases, we can use induction

\nHypothesis for all $n \ge 2$, $P[X_{ij}] = \frac{2}{i j - i + 1}$ for $j \ne i$

\nBase Case $\frac{n = 2}{1 - 2}$ $P[X_{ij}] = 1$ for $j \ne i$

Base Case 2 IP(Xij) ⁼ ¹ for joi Inductive case p(Xij ⁼ = or) : := p +/ki) · P(ki] ⁺ [Xi ⁼ ⁺ /kj]· [k>] By induction - -- By induction : ⁼ ^v - -

$$
\text{Overall, } \qquad \mathbb{P}[X_{ij} = 1] = \frac{2}{ij - i + 1} \quad \left(\underbrace{p + q + r}_{= 1}\right) = \frac{2}{ij - i + 1}
$$

Therefore,

$$
\mathbb{E}\left[\mathcal{T}(n)\right] = \sum_{i,j} \mathbb{P}[\chi_{ij} = 1]
$$

= $n + \sum_{i \neq j} \mathbb{P}[\chi_{ij} = 1]$
= $n + 2 \sum_{i=1}^{n} \sum_{j=i+1}^{n} \frac{2}{|j-i|+1}$

 $=$ 4nH_n $=$ 7n + 4H_n Where H_n = nth Harmonic number \approx log n

Another piece of intuition to help you think about this algorithm is the following
Normally, when we think of our recursive algorithm, we think of it as follows
Pick a pivot
A
Normal Pick a pivot Normally, when we think of our recursive algorithm , we think of it as follows

But you can also consider a slightly different version of the algorithm which does the same steps in ^a different order But you can also consider
does the same steps in
The new algorithm has n

The new algorithm has n phases: in each phase, it picks a random pivot bolt from all the bolts that were not chosen as a pivot before

The first phase works the same as above , but in the second phase the algorithm choses ^a uniformly random bolt in either the green or yellow subproblem and splits that part further . For example , the following could happen:
i i st phase word
porithm chose
subproblem
happen:
i i
letter and the subset split whatever region the randomly chosen bolts happen to land in? .
<u>د با</u> Split whatever region the randomly
chosen bolts happen to land in!
Now, we can follow ith not and jth bolt down T Split whatever region the randomly

chosen bolts happen to land in!

Now, we can follow ith not and jth I

the tree and this makes it a bit more the tree and this makes it a bit more
intuitive that it only depends on (j-i)

