

# CHAPTER 8 NOTES (w4-5)

Sorting in linear time; lower bounds for sorting  
Radix sort, counting...

## Lower bounds for sorting

- Q1: how fast can we sort?

- Q2: how fast can comparison sort be?

- Lower bounds

- Only based on comparing elements.  
(All we've seen so far.)

-  $\Omega(n)$ : to examine all inputs

- All sorts so far are  $\Omega(n \lg n)$

↳ We'll show: this is lower bound  
for COMPARISON SORTING.

Decision tree:

- Abstraction of comparison sort.
- Abstracts away control & data movement, shows comparisons made by a particular algorithm.
- Count only comparisons

Can't sort on non-Adversary?

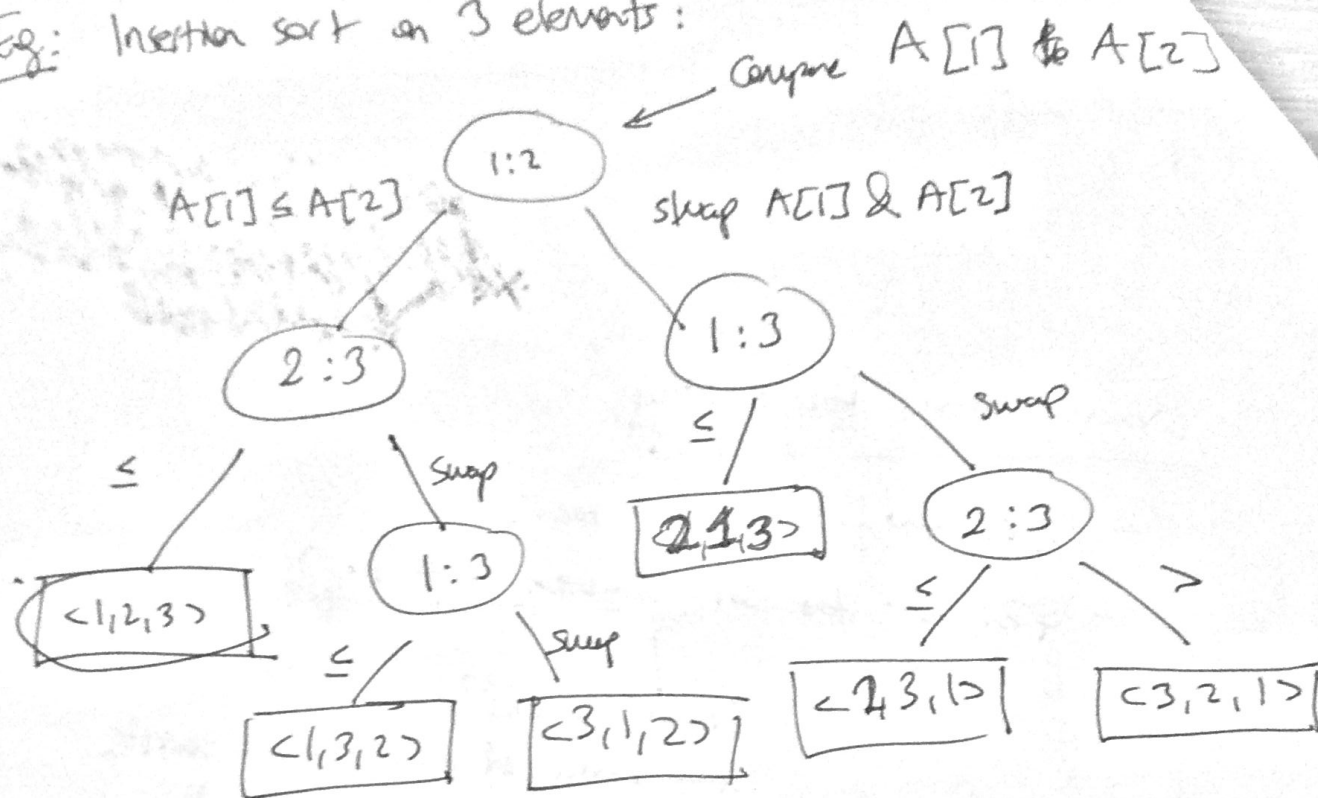
Depends on problem!

- For low-prec, that's fine.

Stirling's Form.  
(approx)

$\ln(n!) \approx n \lg n - n$

Eg: Insertion sort on 3 elements:



#leaves  $\geq n!$ , because every permutation appears at least once.

⑥ Tree models all possible execution traces.

Length of longest path from root to leaf:

- Depends on algorithm

- Insert. sort:  $\Theta(n^2)$

- Merge sort:  $\Theta(n \lg n)$

Lemma: Any binary tree of height  $h$  has  $\leq 2^h$  leaves.

(We all know this)

(Proof by induction.)

Theorem: Any dec. tree that  
can sort  $n$  elements must have height  $\Omega(n \lg n)$

Proof: Tree must contain  $\geq n!$  leaves,  
as there are  $n!$  possible permutations.

\* Let  $l$ : # leaves

\*  $l \geq n!$

\*  $n! \leq l \leq 2^h \Rightarrow 2^h \geq n!$

$\Rightarrow h \geq \lg(n!)$  ~~Stirling approx~~:  $n! > (\frac{n}{e})^n$

$\Rightarrow h \geq \lg(\frac{n}{e})^n$

$= n \lg(n/e)$

$= n \lg n - n \lg e$

$= \Omega(n \lg n)$

□

Corollary:

Heapsort & mergesort are asymptotically optimal  
comp. sorts.

= SORTING IN LINEAR TIME =

\* Non-comparison sorts.

## Counting sort

- Assumption: numbers to be sorted are integers in  $\{1, \dots, k\}$

Input:  $A[1 \dots n]$  where  $A[j] \in \{1, \dots, k\}$   
for  $j = 1, 2, \dots, n$ .

Array  $A$  and values  $n$  and  $k$  are given as parameters.

Output:  $B[1 \dots n]$ , sorted.  $B$  is assumed to be already allocated.

Aux storage:  $C[1 \dots k]$

### COUNTING-SORT ( $A, B, n, k$ )

Right-to-left to solve  
it STABLE SORT.

- Set  $C[1 \dots k]$  to 0.

- Store number of times each integer  $j$  appears in  $A$  in  $C$ .

- $C \leftarrow \text{cumsum}(C)$   $\rightarrow$  Now  $C$  gives the rightmost index for each integer in  $1 \dots k$
- for  $j = n \dots 1$   
 $B[C[A[j]]] \leftarrow A[j]$   
 $C[A[j]] \leftarrow C[A[j]] - 1$

$\triangleright$  Also show the exact algorithm on slides.  
AND Example

\* Stable sort.

Analysis:  $\Theta(n+k)$ , which is  $\Theta(n)$  if  $k = O(n)$ .

How big a  $k$  is practical?

- 32 bit? No.
- 16 bit? Probably no.
- 8-bit? Maybe.

$n$  grows much faster than  $k$ .

SLIDES

$\rightarrow$  This would imply increasing  $k$  by  $n$  which is nonsense!



Q: How can counting sort be faster than  $\Omega(n \log n)$ ?

Answer: not a comparison sort.

Q: Non-integer inputs?

Answer: depends on precision.

Radix sort: — **WISN** (originally w.c.b., but I was away)  
— IBM's original algorithm for sorting cards.

Key idea: sort by least significant digit first.

Algorithm:

RADIX-SORT( $A, d$ )

for  $i \leftarrow 1$  to  $d$

    use a stable sort to sort array  $A$  on  
    digit  $i$ .

▷ Show example on slides.

Correctness: Show on slides (by induction  
from least to most sign. bit.)

Analysis: Assume we're using counting sort as  
intermediate sort. Just counting  
→ sort

— Digits range in  $1..k \Rightarrow$  each pass is  $\Theta(n+k)$

— we make  $d$  passes ( $d$ : # of digits)

—  $\Theta(d(n+k))$  total.

— If  $k = O(n) \Rightarrow \text{total} = \Theta(dn)$

-  $n$  words (= ~~max~~ array length)

-  $b$  bits/word (determines max integer for RADIX SORT)

- Use  $r$ -bit digits  $\Rightarrow d = \lceil b/r \rceil$  (max for encoding)

- Use counting sort,  $k = 2^r - 1$ .   
 MAX INT. FOR EACH COUNTING SORT CALL (ie. AUX STORAGE LENGTH)

eg, 32-bit integers, 8-bit digits

$b=32, r=8, d = \lceil 32/8 \rceil = 4, k = 2^8 - 1 = 255$

$\hookrightarrow$  # of digits, ie. # of COUNT SORT CALLS

Time =  $\Theta\left(\frac{b}{r}(n + 2^r)\right)$

- Fewer passes
- As  $r \gg \lg n$ ,

$r < \lg n$  and  $r > \lg n$  both are ~~seem to be~~ worse choices, hence  $r = \lg n$  is our best choice in general.

Choose  $r$  to balance  $b/r$  and  $n + 2^r$ . time grows exponentially

Choosing  $r \approx \lg n$  gives us  $\Theta\left(\frac{b}{\lg n}(n + n)\right) = \Theta\left(\frac{bn}{\lg n}\right)$

Why  $\lg n$ ? Answer:

- If we choose  $r < \lg n \Rightarrow \frac{b}{r} > \frac{b}{\lg n}$  and  $n + 2^r$  terms doesn't ~~improve~~   
  $\hookrightarrow$  worse
- If we choose  $r > \lg n \Rightarrow n + 2^r$  gets big  $\Rightarrow$  (So, overall a worse choice)   
 eg.  $r = 2 \lg n \Rightarrow 2^r = 2^{2 \lg n} = (2^{\lg n})^2 = n^2$

To sort  $2^{20}$  - many 32-bit numbers,

$r = \lg 2^{20} = 20$  bits,  $\lceil b/r \rceil = 2$  passes   
  $\sim 65$  over all numbers per digit

$\rightarrow$  merge sort  $\lg n = 20$  passes! (at least)   
 (over all numbers)

NOTE: If we use  $r=b$ ? (we make a single call to counting sort!)

$\Theta\left(\frac{b}{r}(n + 2^r)\right) = \Theta(n + 2^b) \Rightarrow \Theta(n)$  if  $2^b < n (\Rightarrow b < \lg n)$

$\Rightarrow$  If we know that we're dealing with small numbers, (special case!)   
 we can directly call counting sort, as size of aux. storage will not dominate.

To sum up,

→ (Compared to  $n$ )

Chp 8, ~~re~~  
Cont'd

... if dealing w/ "small numbers",

we r2b, i.e., directly counting sort.

- In general, use  $r2lg(n)$ ,

that is, choose increasingly larger bases to

balance between

- making too many counting sort calls

AND

- having too slow counting sort calls

How does Radix sort work faster than  
comparison sort?

A: - We use keys as array indices  
directly

( $\Rightarrow$  Integer assumption is the key point!)

Q: What if  
 $lg n$  is larger  
than  $b$ ?

A: That is  
exactly the case  
where  $r2b$   
and  $b < lg n$ .



## BUCKET SORT (Briefly)

• Assume: input is generated by sampling from  $[0,1)$  uniform dist.

- Idea:
- divide  $[0,1)$  into  $n$  equal-sized buckets,  $n \approx \# \text{ inputs}$ .
  - distribute  $n$  input into buckets
  - sort each bucket by insertion sort.
  - go through buckets, listing elements in each one.

• Analysis: ~~Assume number of buckets~~

- on average, each bucket has  $n/2$  elements, by assumption.
- each insert. sort takes constant  $\Theta(1)$  time.
- in total,  $\Theta(n) + n \Theta(1) = \Theta(n)$  time.

★ SEE BOOK FOR A ~~DETAILED~~ FULL ANALYSIS.