3 Sorting Algorithms

input: sequence (a(1), ..., a(n)) or set $\{a(1), ..., a(n)\}$

output: sequence $(a(\pi(1)), ..., a(\pi(n)))$ with $a(\pi(1)) \le \cdots \le a(\pi(n))$

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naive algorithm

1) compute minimum $min\{a(1), ..., a(n)\}$

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$$\min\{a(1),\dots,a(n)\} = \min\{\min\{a(1),a(2)\},a(3),\dots,a(n)\}$$

$$M(1) = 0$$

$$M(n) = 1+M(n-1)$$

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expand

$$M(b) = 1 + M(n-1)$$

$$= 1 + 1 + M(n-2)$$

$$= 2 + M(n-2)$$

$$\dots$$

$$= x + M(n-x)$$

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$$n-x=1 , x=n-1$$

$$M(n)=n-1$$

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$$sort(A) = min(A) \circ sort(A \setminus min(A))$$

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$$= n-1+S(n-1)$$

$$= n-1+n-2+S(n-2)$$
...
$$= \sum_{i=n-x}^{n-1} i+S(n-x)$$

$$n-x=1$$

$$S(n) = \sum_{i=1}^{n-1} i = (n-1) \cdot n/2$$

Merge Sort

input: sorted sequences

with

(n/2)

output: merged sequence

$$(c(1), \dots, c(n))$$
 sorted

$$a(1) \le \dots \le a(n/2)$$

$$b(1) \le \cdots \le b(n/2)$$

$$c(1) \le \cdots \le c(n)$$

input: sorted sequences

$$(a(1), \dots, a(n))$$
 with $(h(1))$ $h(m)$

output: merged sequence

$$(c(1), \dots, c(n+m))$$
 sorted

$$merge((a(1),...,a(n)),(b(1),...b(m))) \\ = \begin{cases} a(1) \circ merge((a(2),...,a(n)),(b(1),...b(m))) & a(1) \leq b(1) \\ b(1) \circ merge((a(1),...,a(n)),(b(2),...b(m))) & a(1) > b(1) \end{cases}$$

Merge Sort

$$a(1) \le \cdots \le a(n)$$

$$b(1) \le \cdots \le b(nm)$$

$$c(1) \le \cdots \le c(n+m)$$

$$n + m - 1$$

comparisons

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$$sort((a(1), ..., a(n)) = merge(sort(a(1) ... a(n/2)), sort(a(n/2 + 1), ..., a(n)))$$

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$$S(1) = 0$$

 $S(n) < n/2 + n/2 + 2 \cdot S(n/2)$
 $= 2 \cdot S(n/2) + n$

difference equations (and difference inequalities)

Let

$$n=2^k \ k\in\mathbb{N}$$

be a power of two and assume

$$f(1) = a$$

$$f(n) = 2 \cdot f(n/2) + b \cdot n$$

$$f(n) = ?$$

Idea: expand definition until you see something

$$f(n) = 2 \cdot f(n/2) + b \cdot n$$

$$= 2 \cdot (2 \cdot f(n/4) + b \cdot n/2) + b \cdot n$$

$$= 2^2 \cdot f(n/2^2) + 2 \cdot b \cdot n$$

$$= \dots$$

$$= 2^x \cdot f(n/2^x) + x \cdot b \cdot n$$

Stop recursion at

$$n/2^x = 1 \quad , \quad x = k = \log n$$

Conjecture

$$f(n) = n \cdot f(1) + b \cdot n \cdot \log(n)$$

= $b \cdot n \cdot \log n + a \cdot n$

Merge Sort

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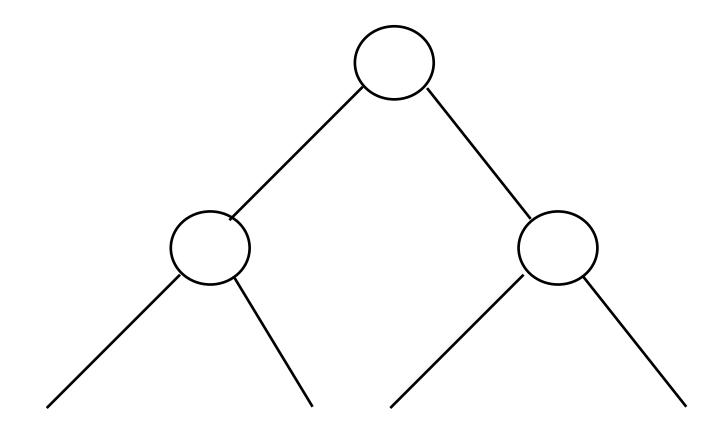
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$$S(n) < n \cdot \log n$$

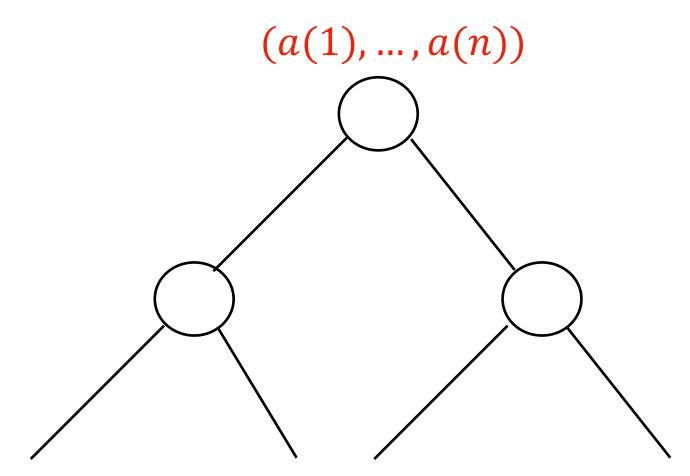
no



can represent any such algorithm as binary tree

nodes: comparisons

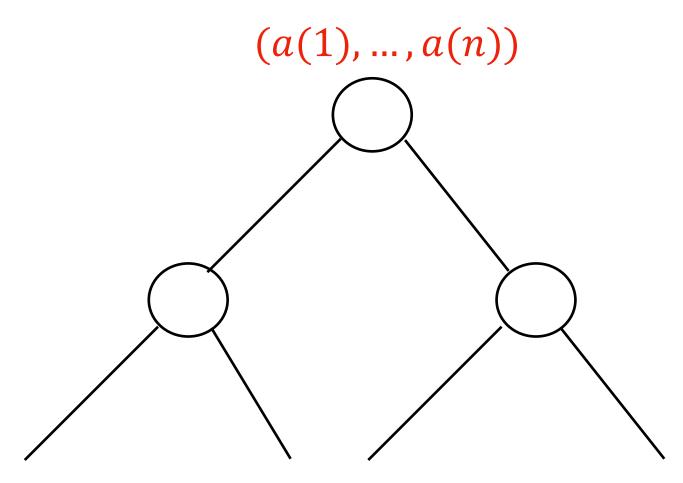
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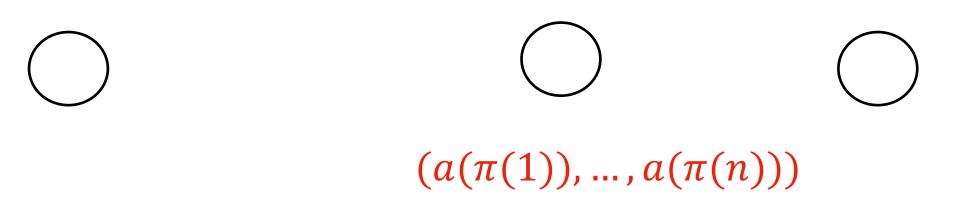


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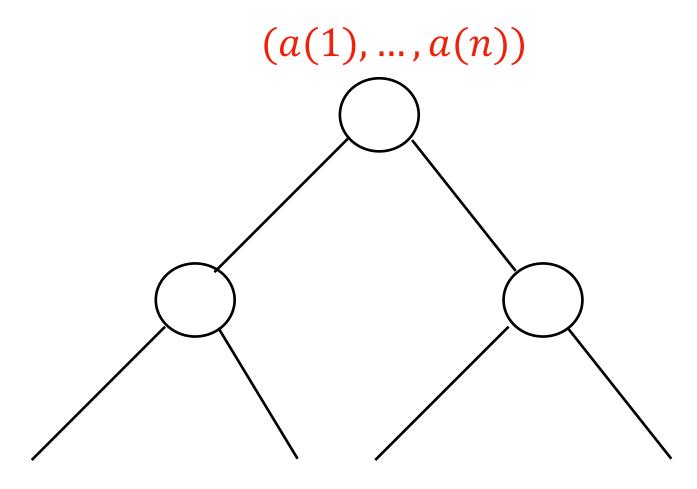
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- one leaf for each permutation

 π

depth of tree = length of longest path = worst case number of comparisons







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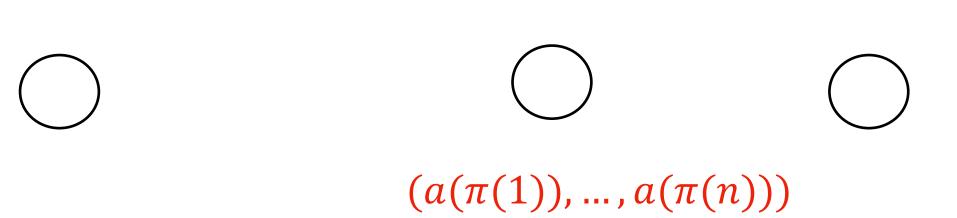
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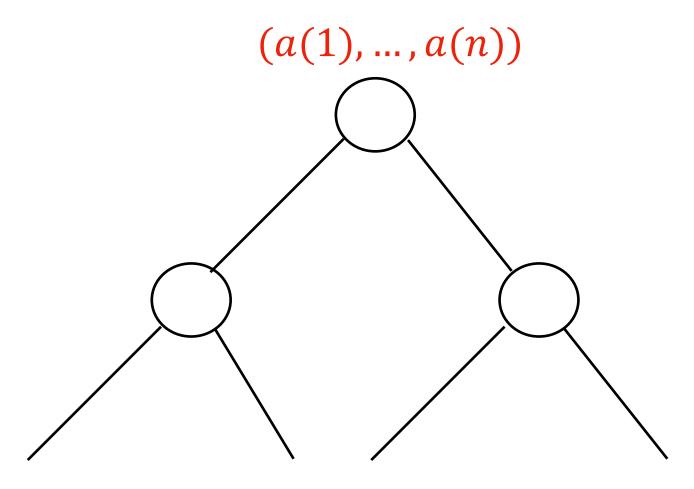
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easy induction: binary tree with depth d has at most 2^d leaves



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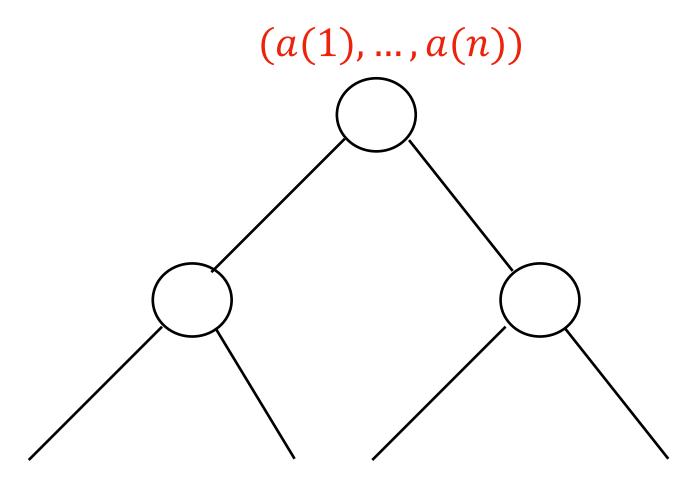
> def: number of permutations of n numbers n!



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 $(a(\pi(1)), ..., a(\pi(n)))$

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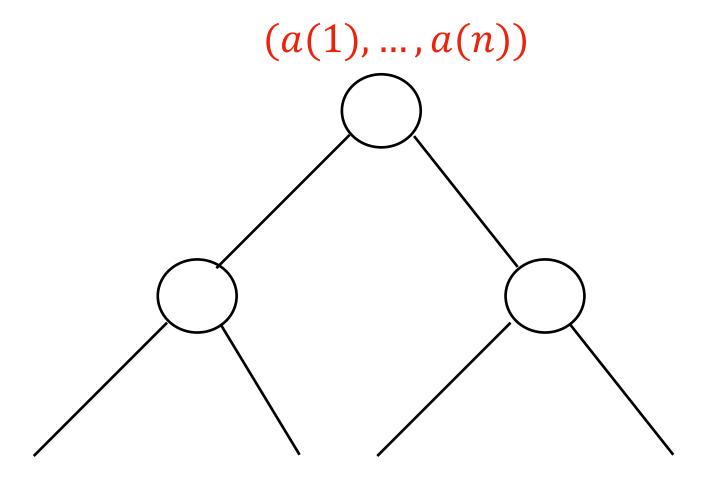
 $1! = 1 \qquad \qquad n! = n \cdot (n-1)!$

choices of first element possible permutations of remaining elements

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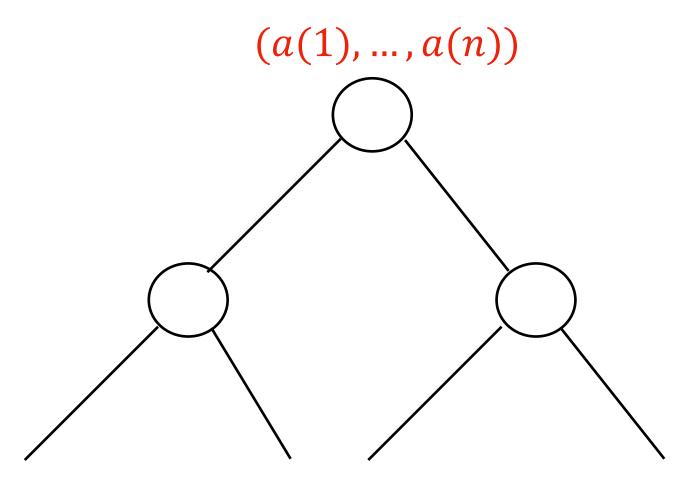
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$$2^d \ge n! > n \cdot (n-1) \cdot \dots \cdot n/2 > (n/2)^{n/2}$$

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$$d > (n/2) \cdot \log(n/2)$$

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more in ,theoretical computer science' or - time permitting - here

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two flavors

- random input
 - modeling: our lack of understanding of what future inputs we will see
 - bad: what if an adversary constructs the input when it matters?
 - the algorithm generates and branches on random data
 - good: this is under the programmer's control
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the miracle

- you program with a pseudo random number generator
- you analyze as if the generated data was truly random
- the measured run time (usually) matches the analysis

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why does probabilistic modeling of a completely deterministic -provably not random - algorithm give the right results??

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here: the algorithm does the random experiments

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Quicksort. 1960, Tony Hoare

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T(n) expected run time for sorting n elements

$$T(n) \le n + (1/n) \cdot \sum_{i=1}^{n} (T(i-1) + T(n-i))$$

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this should have a proof!