Introduction to Algorithms

Meng-Tsung Tsai

09/10/2019

Course Materials

Textbook

Introduction to Algorithms (I2A) 3rd ed. by Cormen, Leiserson, Rivest, and Stein.

Reference Book

Algorithms (JfA) 1st ed. by Erickson. An e-copy can be downloaded from author's website: http://jeffe.cs.illinois.edu/teaching/algorithms/

<u>Websites</u>

http://e3new.nctu.edu.tw for slides, written assignments, and solutions.

http://oj.nctu.me for programming assignments.

Office Hours

Lecturer's

On Wednesdays 16:30 - 17:20 at EC 336.

TA. Erh-Hsuan Lu (呂爾軒) and Tsung-Ta Wu (吳宗達)

On Mondays 10:10 - 11:00 at ES 724.

More TA hours will be announced.

Grading Policy

- 1. No plagarism and cheating. You may fail this course by doing this.
- 2. Saying I don't know is better than talking nonsense. In written assignments and quizzes, the midterm exam, and the final exam, you receive 25% credits if you explictly write down "I don't know." Leaving blank or talking nonsense gives you 0 point.
- 3. Your final grade will be at least

Min(
$$(2Max(A, B) + Min(A, B) + C + D)/5, 99)$$

where A is the average of your written assignments and quizzes, B is the average of your programming assignments and quizzes, and C (resp. D) denotes your grade of the midterm exam (resp. the final exam).

4. Anyone who fails this course may ask for a make-up exam only if he/she participates in class regularly.

Important Dates

In Class

Oct 24: Quiz 1

Nov 05: Midterm Exam

Dec 31: Quiz 2

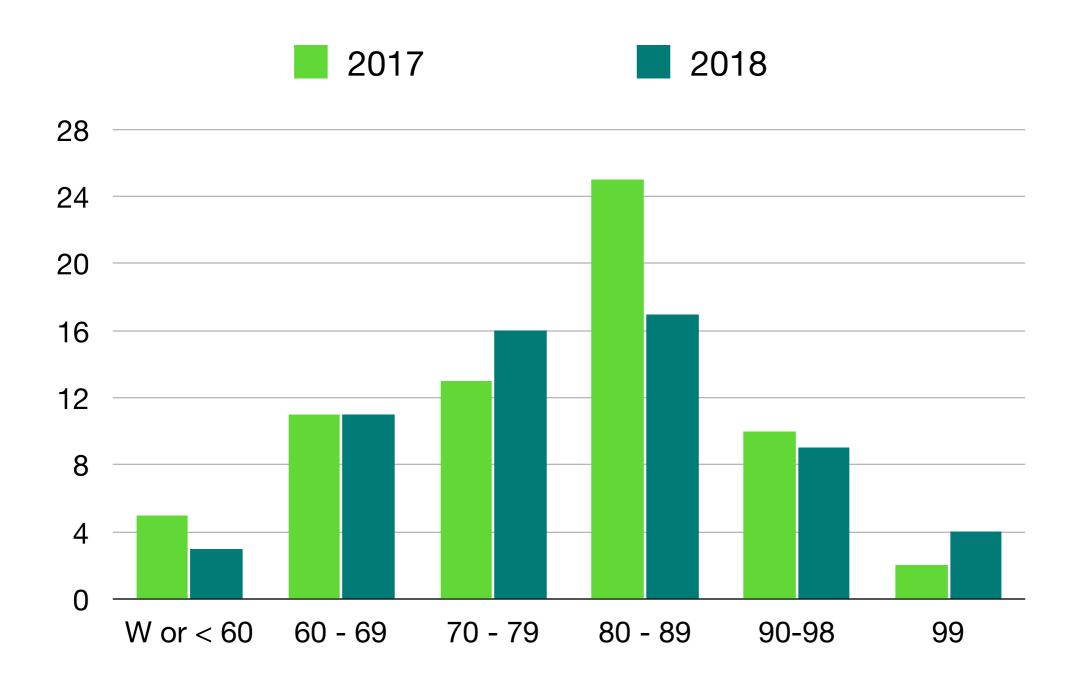
Jan 07: Final Exam

Outside of Class

Nov 16 (Sat 13:30 - 17:30): Programmign Quiz 1

Dec 28 (Sat 13:30 - 17:30): Programming Quiz 2

Distribution of Grades



What are algorithms?

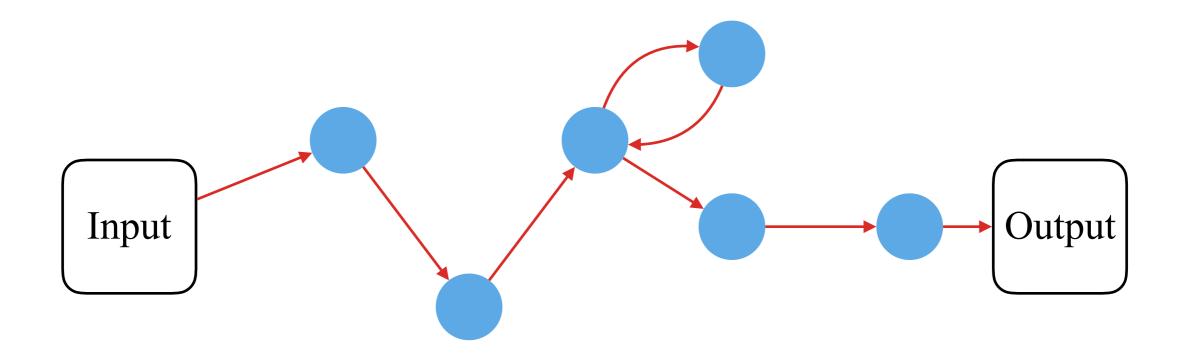
Formally, given the specification of a problem

Formally, given the specification of a problem

Input

Output

Formally, given the specification of a problem



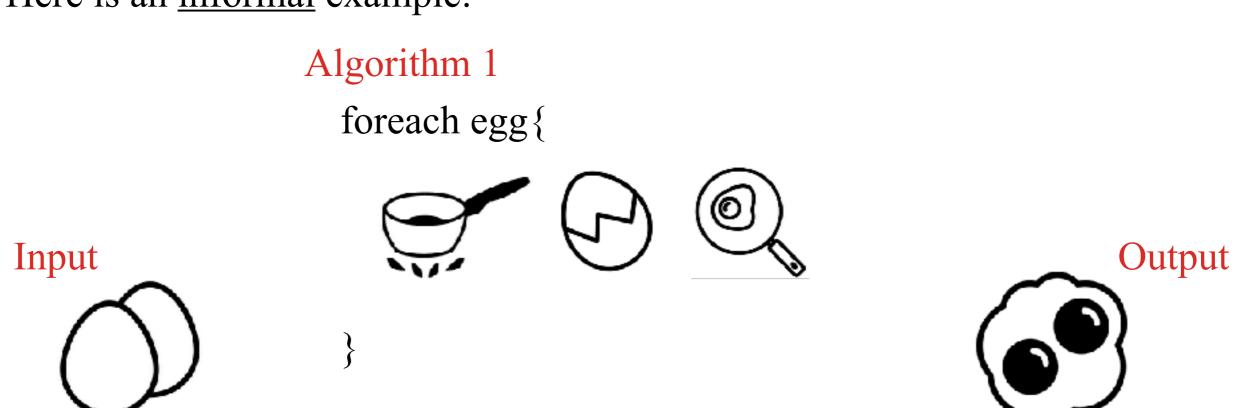
an algorithm is computational procedures that take some values as input and produce some values as output.

Here is an <u>informal</u> example:

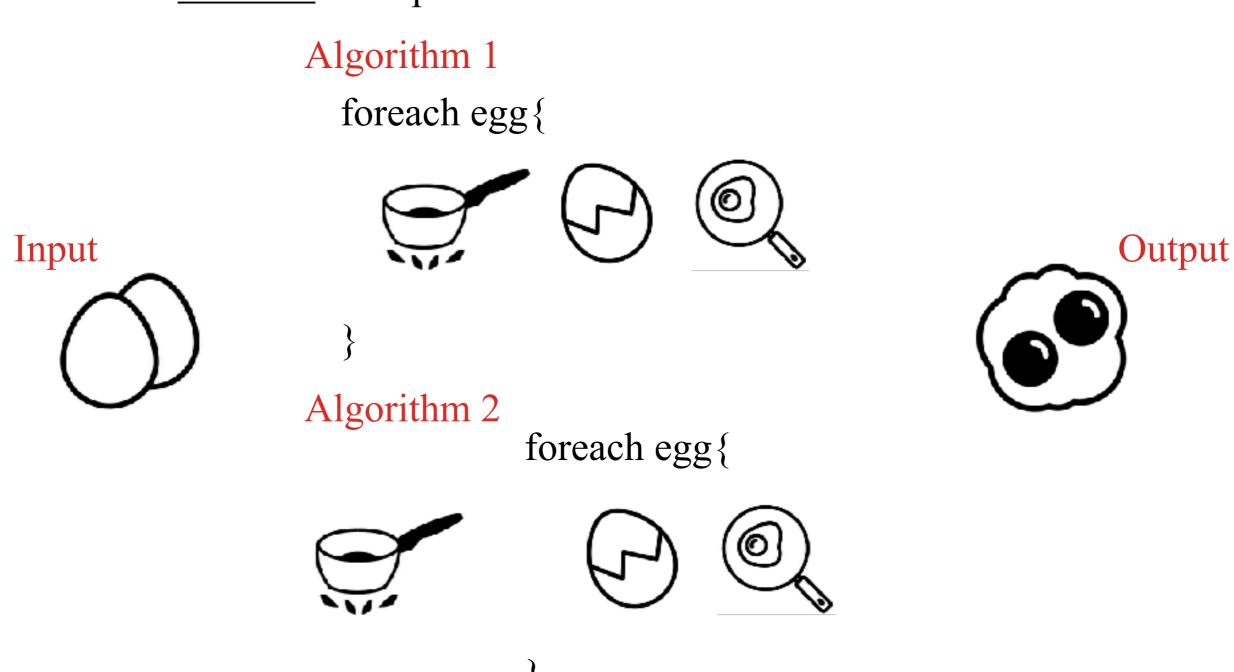
Input



Here is an <u>informal</u> example:



Here is an informal example:



Input: an array A of n integers.

Output: an index k so that A[k] is the minimum value in A.

A problem instance (an instance)



return value (ret): null

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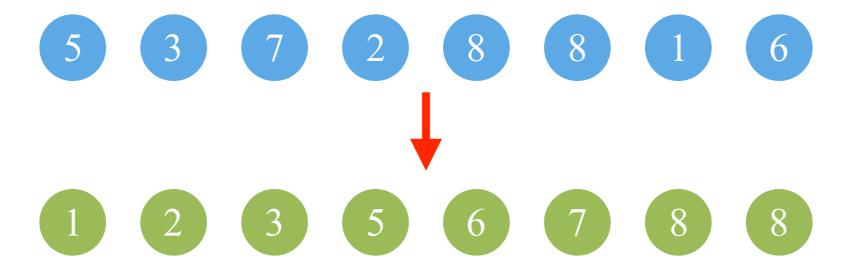
return value (ret):

```
int champion(int *s, int n){ // return -1 for empty input
  int ret = -1; // 1 assignment
  for(int i=0; i<n; ++i){ // incur 2n comparisons, \leq n-1 assignments,
     if(s[i] < s[ret]) // and n increments
       ret = i;
  return ret;
\frac{1}{2} // a constant number of operations for the overhead of function call
--- total running time ---
champion() uses at most 4n + C operations for some constant C.
```

Sorting Problem

Input: an array A of n integers.

Output: the same array with the n integers ordered nondecrementally.



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```
void selection_sort(int *s, int n){
  for(int i=0; i<n; ++i){
    int k = champion(s+i, n-i);
    int swap = s[i]; s[i] = s[k]; s[k] = swap;
}
--- about the highlight ---</pre>
```

It is called *reduction*. Reducing one problem X to another problem Y means to devise an algorithm for X using an algorithm for Y as a building block.

selection_sort() uses at most n(4n+C+3) operations for some constant C.

```
void selection_sort(int *s, int n){
    for(int i=0; i<n; ++i){
        int k = champion(s+i, n-i);
        int swap = s[i]; s[i] = s[k]; s[k] = swap;
    }
}</pre>
```

selection_sort() uses at most n(4n+C+3) operations for some constant C.

Why does the count of operations matter?

```
void selection_sort(int *s, int n){
    for(int i=0; i<n; ++i){
        int k = champion(s+i, n-i);
        int swap = s[i]; s[i] = s[k]; s[k] = swap;
    }
}</pre>
```

selection_sort() uses at most n(4n+C+3) operations for some constant C.

Why does the count of operations matter?

A: We can use it to estimate the running time of the program. 10⁸ operations takes roughly 1 second. Hence, sorting 10⁴ integers by selection sort takes roughly 4 seconds.

```
void selection_sort(int *s, int n) {
  for(int i=0; i<n; ++i) {
    int k = champion(s+i, n-i);
    int swap = s[i]; s[i] = s[k]; s[k] = swap;
  }
}
--- about the highlight ---</pre>
```

Can we replace the highlighted part with s[i] = s[k] = s[i] = s[k]?

C++ Code

```
void selection sort(int *s, int n){
  for(int i=0; i < n; ++i){
     int k = champion(s+i, n-i);
     int swap = s[i]; s[i] = s[k]; s[k] = swap;
--- about the highlight ---
Can we replace the highlighted part with s[i] = s[k] = s[i] = s[k]?
No. Why?
```

Input: a sorted array A of n integers and an integer x.

Output: a sorted array that comprises all elements in A and x.



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C++ Code

```
void insert(int *s, int n, int x){ // array s has length ≥ n+1
0: bool placed = false; // whether x has been placed in s
1: for(int i=n-1; i>=0 && !placed; --i){
2: if(s[i] > x){
3: s[i+1] = s[i];
4: }else{
5: s[i+1] = x; placed = true;
6: }
7: }
8: if(!placed) s[0] = x;
9:}
```

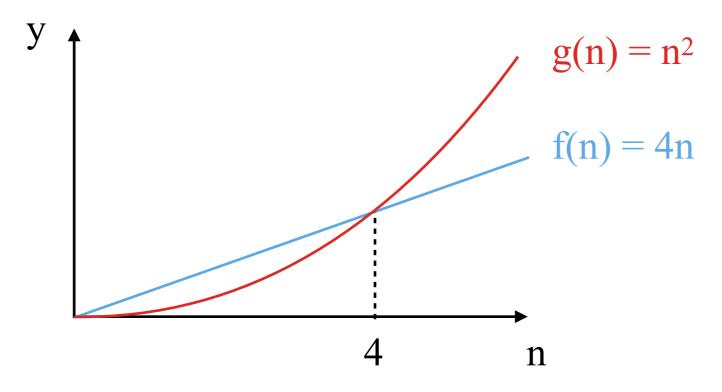
Line 1 comprises 1 assignment, n comparisons, and n decrements. Line 2 comprises n comprisons and n dereference.

Line 3 comprises n assignments, n additions, and 2n dereferences ... It is curbersome (and error-prone) to count the exact operations that an algorithm uses.

Asymptotic Notation: O-Notation

O(g(n)) is pronounced as big-Oh of g of n.

f(n) = O(g(n)) means that $f(n) \le C \cdot g(n)$ for every $n \ge n_0$ for some constants C and n_0 .



We can write 4n = O(n) by setting $(C, n_0) = (4, 1)$ or $4n = O(n^2)$ by setting $(C, n_0) = (1, 4)$.

Asymptotic Notation: O-Notation

O(g(n)) is pronounced as big-Oh of g of n.

More formally, f(n) = O(g(n)) means that f(n) is a function contained in thet set of functions $\{h(n) : \text{there exists positive constants } n_0 \text{ and } C \text{ so that } C \cdot g(n) \ge h(n) \text{ for every } n \ge n_0\}.$

Because it is curbersome to determine the constant C and we simply need to estimate the running time, we usually use asymptotic notation to denote the time complexity of an algorithm.

- --- Example ---
- 1. Selection sort runs in $O(n^2)$ time, so it can sort 10^4 integers in seconds.
- 2. Insert x into a sorted array runs in O(n), so in seconds one can complete 10^4 insertions.

Exercises

1.
$$2n^2 + 100 \text{ n} - 2000 = O(n^2)$$
?

2.
$$2n^3 - 100 n^2 = O(n^2)$$
?

3.
$$n^n = O(n!)$$
?

4.
$$n log n = O(n^{1.5})$$
?

5.
$$\log n! = O(n \log n)$$
?

Input: an array A of n integers.

Output: the same array with the n integers ordered nondecrementally.



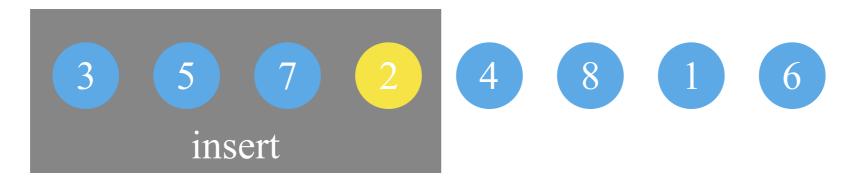
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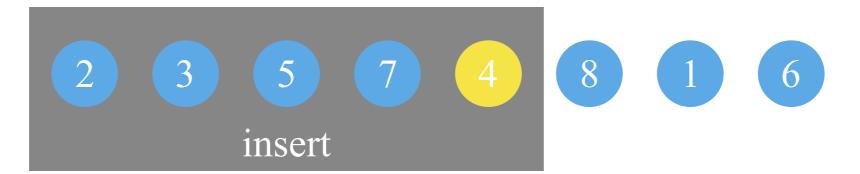
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C++ Code

```
void insertion_sort(int *s, int n){
    for(int i=1; i<n; ++i){
        insert(s, i, s[i]);
    }
}</pre>
```

--- about the highlight ---

Again, we use a reduction here.

The running time is $O(n) \cdot O(n) = O(n^2)$. Why does this equality hold?

Exercises

Let $f(n) = O(n^a)$ and $g(n) = O(n^b)$ for some constants a, b > 0. Prove that

- 1. $f(n) + g(n) = O(n^c)$ for any $c \ge max(a, b)$.
- 2. $f(n) \cdot g(n) = O(n^c)$ for any $c \ge a+b$.

Prove or disprove that

$$\sum_{i=1}^{n} f_i(n) = O(n) \text{ where } f_i = O(n) \text{ for every } i \in \{1, 2, \dots, n\}.$$

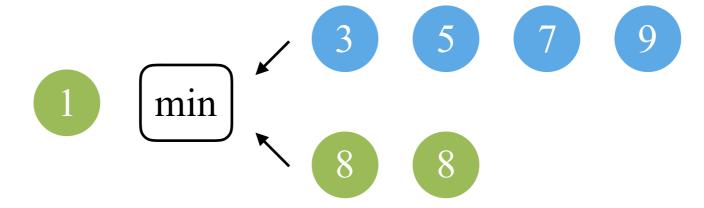
Input: two sorted arrays A and B of integers.

Output: a sorted array that comprises all elements in A and B.



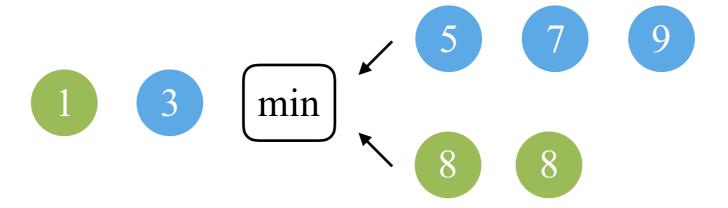
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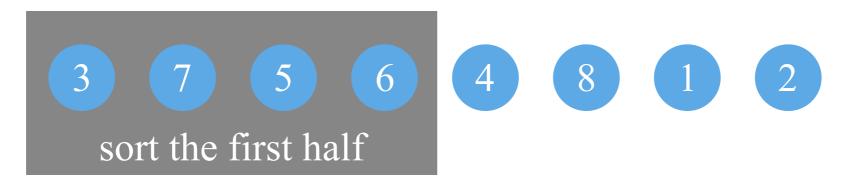
C++ Code

```
int* merge(int *s, int n, int *r, int m){
  int *ret = new int [n+m];
  int i = 0, j = 0, k = 0;
  while(1){
     if(i < n \&\& j < m){ // when both arrays are not empty
        ret[k++] = ((s[i] < r[j]) ? s[i++] : r[j++]);
     }else{
       ret[k++] = ((i < n) ? s[i++] : r[j++]);
```

Merging two sorted arrays takes O(n+m) time.

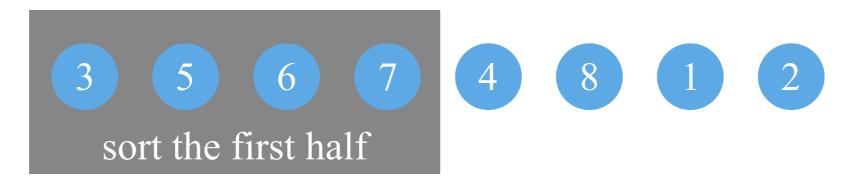
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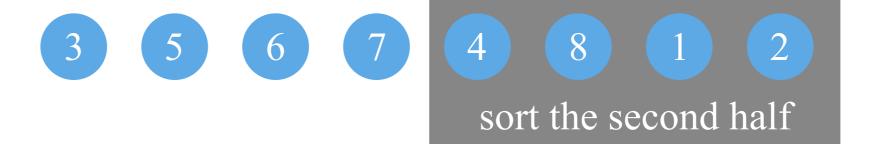
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C++ Code

```
void merge sort(int *s, int n){
  if(n == 1) return;
  int k = n/2;
  merge sort(s, k);
  merge sort(s+k, n-k);
  int *r = merge(s, k, s+k, n-k);
  memcpy(s, r, sizeof(int)*n);
--- about the highlight ---
```

A reduction from a problem to itself is called *recursion*. A recursion usually requires that the instance size decreases monotonically. Why?

C++ Code

```
void merge sort(int *s, int n){
  if(n == 1) return;
  int k = n/2;
  merge sort(s, k);
  merge sort(s+k, n-k);
  int *r = merge(s, k, s+k, n-k);
  memcpy(s, r, sizeof(int)*n);
```

Merge sort needs at most $T(n) = T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + c_1 n$ operations where $T(1) = c_2$.

Substitution Method

Merge sort needs at most $T(n) = T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + c_1 n$ operations where $T(1) = c_2$. How to represent T(n) in terms of O(g(n))?

Guess $T(n) \le d n \log n + d n$ for some constant d > 0.

// We will see how to guess.

For n = 1, $T(1) = c_2 \le d \log 1 + d$ if we pick $d \ge c_2$.

Suppose $T(n) \le d n \log n + d n$ for every $n \le k$.

For n = k,
$$T(k) = d(\lfloor k/2 \rfloor \log \lfloor k/2 \rfloor + \lceil k/2 \rceil \log \lceil k/2 \rceil) + c_1 k$$

$$\leq d(\lfloor k/2 \rfloor + \lceil k/2 \rceil) \log k + c_1 k$$

$$= dk \log k + c_1 k$$

$$\leq dk \log k + dk \qquad \text{(if we pick d} \geq c_1)$$

By induction on k, our guess is correct. Thus, $T(n) = O(n \log n)$.

Merge sort needs at most $T(n) = T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + c_1 n$ operations where $T(1) = c_2$. We have seen how to verify the guess $T(n) = O(n \log n)$. How to come up with a guess?

We simply need a guess, so we may drop the floor and the ceiling functions, and ignore the constants. We get:

$$T(n) = \begin{cases} 2T(n/2) + n & \text{if } n > 1\\ 1 & \text{otherwise} \end{cases}$$

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T(n)

T(n/2)

T(n/2)

T(n/4)

T(n/4)

T(n/4)

T(n/4)

$$T(n) = \begin{cases} 2T(n/2) + n & \text{if } n > 1\\ 1 & \text{otherwise} \end{cases}$$

 $\begin{array}{c|c} & & & \\ & & \\ \hline n/2 & & \\ \hline n/4 & & \\ \hline n/4 & & \\ \hline n/4 & & \\ \hline \end{array}$

Analyze the cost for each subproblem T(k) without considering its recursive calls.

$$T(n) = \begin{cases} 2T(n/2) + n & \text{if } n > 1\\ 1 & \text{otherwise} \end{cases}$$

 $\begin{array}{c|c} & & & \\ & & \\ \hline n/2 & & \\ \hline n/4 & & \\ \hline \end{array}$

There are log_2 n layers, and for each layer the sum of cost is n. Consequently, the total cost is O(n log n).

Exercises

- 1. To merge k sorted length-n arrays into a single sorted one, can one do this in O(k n log k) time?
- 2. Tower of Honai is a classical example for recursion. An interesting variation can be found in Ex 4, Chap 1, JfA.