

# ECS 122A: Algorithm Design and Analysis

## Week 2 Discussion

Ji Wang

Fall 2020

# A bit about logistics

## Discussion Schedule:

Ji Wang → Christopher Peterson → Terry Guan

- ▶ We will rotate to lead the discussion, with the exception of Veteran Day (Wed) and the entire Thanksgiving week.
- ▶ The first discussion of the week will be live and recorded. The rest will be in the form of live Q&A or extra office hours.
- ▶ Check Canvas homepage frequently for update on discussion notes and videos.

**Office hours:** Available every weekday

Ji: M 10:30am - 12:30pm, T 5 - 6pm, F 4 - 6pm

**Contact us:** For private questions, send direct message via Canvas.





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It's the **algorithms** that work under the hood!

# Outline

- ▶ Proof technique: Mathematical Induction
- ▶ Data structure: Heap
- ▶ Design description: Pseudocode
- ▶ Analysis: Asymptotic Notation
- ▶ Example: Insertion Sort

# Mathematical Induction: Approach

**When** applicable: Prove that a property  $P(n)$  holds for every natural number  $n$ .

**How** it works:

1. Show that a property  $P(n)$  holds for the base case, usually when  $n = 0$  or  $1$ .
2. Assume that  $P(n)$  is true for  $n = k$  where  $k$  is greater than the base case.
3. Prove that  $P(n)$  is true for  $n = k + 1$ . In this step we will use the assumption above.
4. Then, by the principle of induction, we can conclude that  $P(n)$  is true for every natural number that is greater than or equal to the base case.

# Mathematical Induction: Example

**Problem Statement:** Prove the following statement  $P(n)$  is true

$$\sum_{i=1}^n i^3 = \left(\frac{n(n+1)}{2}\right)^2$$

**Build a skeleton:**

- ▶ Base Case:
- ▶ Inductive Hypothesis:
- ▶ Inductive Step:
- ▶ Thus, by the principle of induction, we can conclude that the statement above is true for every natural number  $n$ .



# Mathematical Induction: Example

$$1^3 + 2^3 + \cdots + n^3 = \left(\frac{n(n+1)}{2}\right)^2$$

► Base Case:

Verify  $P(n)$  is true when  $n = 1$ .  $1^3 = \left(\frac{1 \cdot (1+1)}{2}\right)^2$ .

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► Base Case:

Verify  $P(n)$  is true when  $n = 1$ .  $1^3 = \left(\frac{1 \cdot (1+1)}{2}\right)^2$ .

► Inductive Hypothesis:

Assume  $P(n)$  is true when  $n = k$ . Then, we have

$$1^3 + 2^3 + \cdots + k^3 = \left(\frac{k(k+1)}{2}\right)^2.$$

## Mathematical Induction: Example

$$1^3 + 2^3 + \cdots + n^3 = \left(\frac{n(n+1)}{2}\right)^2$$

- Inductive Step: We need to show  $P(n)$  is true when  $n = k + 1$ .

$$\begin{aligned} 1^3 + 2^3 + \cdots + (k+1)^3 &= 1^3 + 2^3 + \cdots + k^3 + (k+1)^3 \\ &= \left(\frac{k(k+1)}{2}\right)^2 + (k+1)^3 \\ &= \frac{k^2(k+1)^2}{4} + \frac{4(k+1)^3}{4} \\ &= \frac{(k+1)^2(k^2 + 4(k+1))}{4} \\ &= \frac{(k+1)^2(k+2)^2}{4} \\ &= \left(\frac{(k+1)((k+1)+1)}{2}\right)^2 \end{aligned}$$

## Recap: Max-heap

**Definition:** Tree-based data structure which is an almost complete tree.

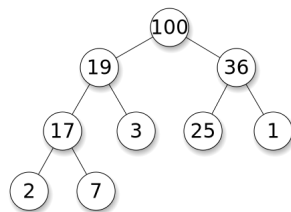
**Property:** For any given node, the key (value) of its parent node is greater than or equal to that of itself.

### Operations:

- ▶ find-max: find a maximum item of a max-heap. (peek)
- ▶ insert: add a new key to the heap. (push)
- ▶ extract-max: return the node of maximum value from a max-heap after removing it from the heap. (pop)
- ▶ increase-key or decrease-key: update a key within a max-heap.
- ▶ heapify: create a heap out of given array of elements.

**Applications:** e.g. Prim's minimal-spanning-tree algorithm and Dijkstra's shortest-path algorithm

## Max-heap: an example



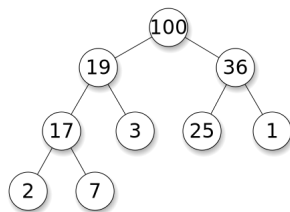
`push(10)`

`push(22)`

`pop()`

`update(7, 18)`

## Max-heap: an example



`push(10)`

`push(22)`

`pop()`

`update(7, 18)`

The time complexity of each operation is left to you.

# How to describe the design of an algorithm

**Objective:** Understandable (clear, precise) and concise if possible.

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<sup>1</sup>More pseudocode conventions can be found in textbook [pp.20-22]

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## 1. in English

- ▶ First (usually) points out what strategy/method used, e.g. divide-and-conquer, binary search.
- ▶ Separate paragraphs if necessary, e.g. branches.
- ▶ Bullet-point format is also a good practice.

## 2. in “real” code: whatever language you favor

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## 2. in “real” code: whatever language you favor

## 3. in Pseudocode <sup>1</sup>

- ▶ Indentation indicates block structure, similar to Python.
- ▶ Indices (often) start from 1, unlike most languages.
- ▶ Use comments as needed. (e.g. variable usage, function of a block)
- ▶ Pass parameters to a procedure *by value*.

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# How to Analyze an algorithm: Asymptotic notation

Analyzing an algorithm means predicting the resources that the algorithm needs. The primary concern in this course is to measure computational time, a.k.a. running time.

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We need a definition to evaluate the order of growth to:

- ▶ characterize the efficiency of the algorithm when input sizes are large enough.
- ▶ compare the performance with other alternative algorithms.

Thus, we:

- ▶ study a way to describe the growth of functions in the limit.
- ▶ focus on what's important (leading factor) by ignoring lower-order terms and constant factors.

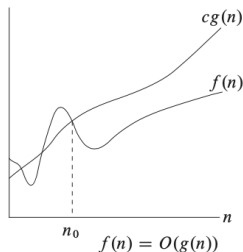
# Asymptotic notation: Big O

**Definition:**  $g(n)$  is an asymptotic upper bound for  $f(n)$ , denoted by

$$f(n) = O(g(n)),$$

if there exist constants  $c$  and  $n_0$  such that:

$$0 \leq f(n) \leq cg(n) \quad \text{for } n \geq n_0$$



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<sup>2</sup>See auxiliary notes for reviews on frequent used functions

# Asymptotic notation: Big O

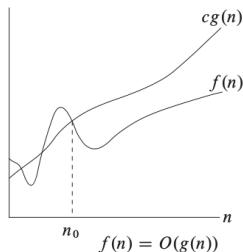
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**Example:** Show  $\lg n$  is  $O(\ln n)$



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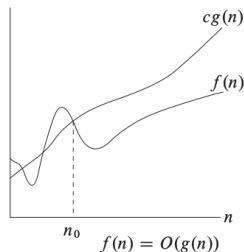
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**Example:** Show  $\lg n$  is  $O(\ln n)$

$$\begin{aligned} \lg n = \log_2 n &= \frac{\log_e n}{\log_e 2} = \frac{1}{\log_e 2} \ln n \\ &\leq 2 \ln n \quad \text{for } n \geq 1 \end{aligned}$$

it is true for  $c = 2$  and  $n_0 = 1$ .<sup>2</sup>

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## Example: Insertion Sort

**Problem Statement:** Given a list of numbers, sort them in non-decreasing order.

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**Algorithm:** We take an *incremental* approach. Split the list into sorted part and unsorted part. Insert the head of unsorted one into the appropriate position in the sorted one. Repeat until no unsorted part remaining.

**Example:** 8 3 2 7 4 1 6

8	3	2	7	4	1	6
8	3	2	7	4	1	6
3	8	2	7	4	1	6
3	8	2	7	4	1	6
2	3	8	7	4	1	6

2	3	8	7	4	1	6
2	3	7	8	4	1	6
2	3	7	8	4	1	6
2	3	4	7	8	1	6
2	3	4	7	8	1	6

1	2	3	4	7	8	6
1	2	3	4	7	8	6
1	2	3	4	6	7	8



# Insertion Sort: Pseudocode and Time Analysis

INSERTION-SORT( $A$ )

```
1  for  $j = 2$  to  $A.length$ 
2       $key = A[j]$ 
3      // Insert key into the sorted list  $A[1 \cdots j - 1]$ .
4       $i = j - 1$ 
5      while  $i > 0$  and  $A[i] > key$ 
6           $A[i + 1] = A[i]$ 
7           $i = i - 1$ 
8       $A[i + 1] = key$ 
```

► Worst case: reversed order, e.g. 5 4 3 2 1.

$$\sum_{j=2}^n \sum_{i=1}^{j-1} = O(n^2)$$

# Insertion Sort: Pseudocode and Time Analysis

INSERTION-SORT(*A*)

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7          i = i - 1
8      A[i + 1] = key
```

- ▶ Worst case: reversed order, e.g. 5 4 3 2 1.

$$\sum_{j=2}^n \sum_{i=1}^{j-1} = O(n^2)$$

- ▶ Best case: already sorted

$$\sum_{j=2}^n = O(n)$$