

Introduction to Algorithms

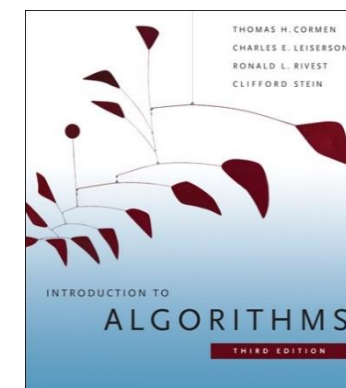
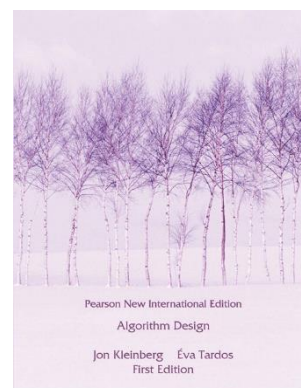
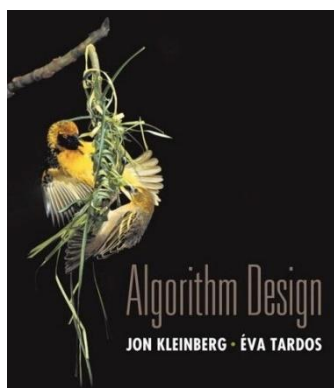
# Algorithm Analysis

# Learning Objectives

- **Overall:** To promote individual learning and **analytical thinking skills**.
  - More detailed objectives:
    1. Be able to **prove correctness** of algorithms.
      - Some people spend their whole career to prove something.
    2. Have deep understanding of differences and advantages of **iterative and recursive algorithms**.
    3. Have a knowledge base of (many) **existing algorithms**.
    4. Understand **speed vs. space trade-off**, importance of **data-structures, and data preprocessing**.
      - Search for an item: hash, tree, and traverse all.
    5. Be able to **design new algorithms**.
      - Correctness.
      - Efficiency.
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# Overview

- Acknowledgement: Dr. Zhang Wei, Dr. LIU Fang, Dr. Donny - course materials
- Appointment by email:
  - [kan.chen@singaporetech.edu.sg](mailto:kan.chen@singaporetech.edu.sg)
  - [AF.Alfred.Whang@singaporetech.edu.sg](mailto:AF.Alfred.Whang@singaporetech.edu.sg)
  - [Pohkok.Loo@singaporetech.edu.sg](mailto:Pohkok.Loo@singaporetech.edu.sg)
- References:
  - **Algorithm Design**, Jon Kleinberg and Eva Tardos.
  - **Introduction to Algorithms**, Thomas Cormen, Charles Lieserson, *et al.*
  - Introduction to the Design and Analysis of Algorithms, Anany Levitin.
  - Course materials and activities.



# Tentative Dates and Grading Policy

- **Marks:**
    - 28% - Assignment (group of 5/6 students)
    - 12% - Tutorial participation/quiz - no makeup
    - 30% - Midterm
    - 30% - Final Exam
  - Part 1: week 1 to 6 (Loo)
    - 1 assignment (group 14%), week 5, Due on 30/9 23:00
    - 5 tutorial participation/quiz (6%), during each tutorial
    - test 1 (30%), week 6, 8/10 between 13:00-17:00 (1 hour)
  - Part 2: week 8 to 13 (Chen Kan)
    - 1 assignment (group 14%), week 12, 18/11 23:00
    - 5 tutorial participation/quiz (6%), during each tutorial
    - Test 2 (30%), week 14, TBA
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# Tentative Dates and Grading Policy

- Blended learning
  - All lecture content/video/zoom recording will be posted before the tutorial.
  - The key face to face session is during the tutorial where you will work on tutorial questions, ask questions and complete quizzes.
- Form your own team using the excel file I have provided.  
[CSD3130,3131 Y2024T1 \(team\).xlsx](#)
  - Once this is done, I will need your team to enrol yourself in xSit. More information will be provided for the enrolment in xSit.

# Tentative Dates and Grading Policy

- Quizzes
  - There will be quiz in each tutorial session in class via xSit with lockdown browser. There is no makeup for the quiz.
- Assignment
  - This is done as a team of 6 students. There will be a self and peer evaluation where you will evaluate yourself and your team members to reflect the contribution of each member after submission. The final score will be scaled according to the evaluation result.



# Stable Matching: Example

- History: A self-enforcing college admission process, or job recruiting process.
  - Each student has his/her preference list of the universities.
  - Each university has her preference list of the students.
  - Normal that the preferences do not align perfectly.
  - How to make both sides satisfied?



# Stable Matching: Example

- **Example** to be more intuitive (in time order):
  - (May 1) Jon accepted the offer from Apple.
  - (May 2) Google offers Jon and Jon likes Google more.
  - (May 3) Jon retracts his acceptance of Apple and accepts Google's offer. (Apple has one free slot now.)
  - (May 4) Apple soon offers David and he likes Apple more.
  - (May 5) David retracts his Facebook offer and accepts Apple's offer. (Facebook has one free slot now.)
  - (May 6) Facebook moves to X, and X turns down Y and accepts Facebook.
  - (May 7) Y moves to Z ... Situation gets **out of control**.



# Stable Matching: Example

- Things can be different, possibly even worse.
  - Jon, David and the others have communication channels.
  - They hear something and accordingly do something.
  - The **network** of the communication will be **fully connected** with around  $O(n^2)$  links.
    - Suppose 50 students, 50 companies, then we have 10,000 communication links.
    - Each link can trigger a change of the offer distribution.
    - **Chaos happens**. Everyone tired and unhappy.
- Basic issue?
  - People cannot act in their **self-interest**, otherwise **system may crash**.

# Stable Matching: Example

- Basic issue?
  - People cannot act in their **self-interest**, otherwise **system may crash**.
- Our desired process – **stable situation**:
  - Use **intelligence to guide** the process.
  - **Idea: self-interest prevents retracting offers**.
    - **Company**, with its offers all being accepted, gets a call from another applicant
      - -> “No, we prefer each applicant we have accepted than you, so we **will not change**.” **Chaos avoided**.
    - **Student**, with an offer, gets a call from a university.
      - -> “No. **I am happy** where I am.” No chaos.

# Stable Matching

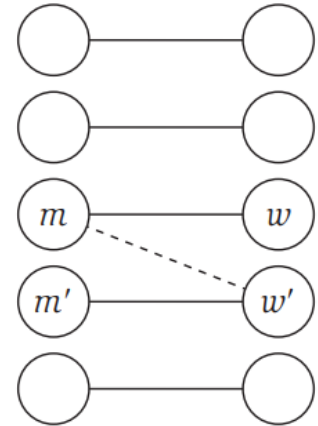
- Question: given employer  $E$  and applicant  $A \notin E$ , can we have:
  - $E$  prefers her accepted applicants than  $A$  (another applicant).
  - $A$  prefers his/her current employer over working for  $E$ .
- Obj: design algorithm -> produce **stable situations**.
- Similar application: residents and hospitals matching.
  - Even a decade earlier than the employee and employer one.
  - Still widely used today in USA.
- Intuition: personal interest & system interest.
- Problem formulation -> algorithm design -> algorithm analysis.

# Problem Formulation

- Problem: As **clean** as possible. As **simple** as possible. Improve **step by step**.
  - Apply to programming as well.
- Complex situation: a **company can hire multiple employees**.
  - 1-to-many relationship.
- Simple: **marriage problem**.
  - A man can only marry a woman.
  - A woman can only marry a man.
  - **1-to-1 relationship**.
- **Math definition**.
  - $n$  men:  $M = \{m_1, m_2, \dots, m_n\}$ .
  - $n$  women:  $W = \{w_1, w_2, \dots, w_n\}$ .
  - **Matching**:  $S$ , a set of pairs, pair  $(m, w) \in S$  has a man  $m \in M$  and a woman  $w \in W$ .
    - Requirement: each man/woman only **appears at most once in  $S$** .
  - **Perfect matching**: everyone married & married to one person. Or  $|M| = |W| = |S| = n$ .

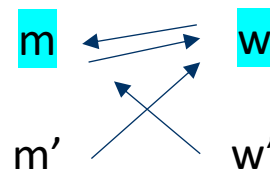
# Problem Formulation

- **Preference**: each man  $m$  ranks all women. The order is  $m$ 's preference list.
  - $m$  prefers  $w$  to  $w'$  if he ranks  $w$  higher than  $w'$ .
  - Ties not allowed.
  - Each woman  $w$  also ranks all the men, also her preference list.
- Perfect matching all we want? (All men and women in the matching)
- **Unstable pair**  $(m, w)$  and  $(m', w')$  in a perfect matching  $S$ :
  - $m$  prefers  $w'$ .
  - $w'$  prefers  $m$ .
  - $(m, w')$  is called an **instability**. The pair is not in  $S$ .
  - $m$  and  $w'$  prefer each other but they are assigned to a different person
  - Objective: find a **perfect matching, without any instability** -> **stable matching**.
- Questions:
  - Is stable matching possible?
  - If true, how to find it fast?



# Possible – Proven with Examples

- Two men  $m$ ,  $m'$  and two women  $w$ ,  $w'$ .
  - $m$  prefers  $w$  to  $w'$
  - $m'$  prefers  $w$  to  $w'$
  - $w$  prefers  $m$  to  $m'$
  - $w'$  prefers  $m$  to  $m'$
- Do we have complete agreement?
  - The men agree on the order of women.
  - The women agree on the order of men.
  - Unique stable matching:**  $(m, w)$  and  $(m', w')$ .
    - Although  $m'$  prefers  $w$ ,  $w$  didn't prefer  $m'$
- Another perfect matching  $(m, w')$  and  $(m', w)$ .
  - Is this stable? **No ( $m$  and  $w$  prefer each other)**
- A more complicated example in the next slide.



Focus is  
the pair  
where they  
want each  
other  
 $(m, w)$

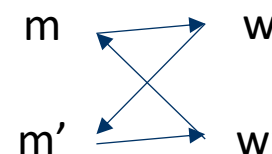
$n = 2$ ,  $n^2$  possibilities,  $2^2=4$   
 $(m, w)$ ,  $(m, w')$ ,  $(m', w)$ ,  $(m', w')$   
 6 possible combinations  
 $(m, w)$ ,  $(m, w')$  – not possible  
 $(m, w)$ ,  $(m', w)$  – not possible  
 $(m, w)$ ,  $(m', w')$  - ok  
 $(m, w')$ ,  $(m', w)$  – ok, but **unstable**  
 $(m, w')$ ,  $(m', w')$  – not possible  
 $(m', w)$ ,  $(m', w')$  – not possible

$m$  and  $w$  wants each other  
but now they are both  
refused their choice and with  
others

We have a situation where **2**  
parties want each other but  
yet they have someone else

# Possible – Proven with Examples

- A more complicated example.



- $m$  prefers  $w$  to  $w'$
- $m'$  prefers  $w'$  to  $w$
- $w$  prefers  $m'$  to  $m$
- $w'$  prefers  $m$  to  $m'$

No situation where both parties want each other but yet get someone else

- Men's preferences meshes well  $\rightarrow (m, w)$  and  $(m', w')$  perfect for men.
- Women's preferences meshes well  $\rightarrow (m, w')$  and  $(m', w)$  perfect for women.
- Any problem?
  - Both are **stable matchings**, but they **clash**.
  - No happy ending for all men and women.
- Conclusion:  $\geq 1$  **stable matching possible**.

$(m, w), (m, w'), (m', w'), (m', w)$   
 6 possible combinations  
 $(m, w), (m, w')$  – not possible  
 $(m, w), (m', w')$  – ok (fr men)  
 $(m, w), (m', w)$  – not possible  
 $(m, w'), (m', w')$  – not possible  
 $(m, w'), (m', w)$  – ok (fr woman)  
 $(m', w'), (m', w)$  – not possible



# Design the Algorithm

- Our plan:
  - Show that a **stable matching exists**.
  - Propose an **efficient algorithm** accordingly.
- Some basic ideas – 3 Stages.
  - **Free**. In the beginning.
  - **Engagement**. Intermediate state. (**Still can change**)
  - **Married**. The algorithm or matching **terminates**. Engagements declared final.
- Why engage?
  - Bob likes Alice the most and proposes to her.
  - Shall Alice accept?
    - Accepts: what if Alice loves Jon the most and Jon proposes to her later? Regret?
    - If not, what if Bob is the best she can get? Lonely in the end?
  - **Engagement** is helpful to deal with the above situation.
- Procedure: all **free in the beginning**, **eventually get engaged**, and finally married.

# Gale-Shapley Algorithm

Initially all  $m \in M$  and  $w \in W$  are free

While there is a man  $m$  who is free and hasn't proposed to every woman

    Choose such a man  $m$

    Let  $w$  be the highest-ranked woman in  $m$ 's preference list to whom  $m$  has not yet proposed

    If  $w$  is free then

$(m, w)$  become engaged

    Else  $w$  is currently engaged to  $m'$

        If  $w$  prefers  $m'$  to  $m$  then

$m$  remains free

        Else  $w$  prefers  $m$  to  $m'$

$(m, w)$  become engaged

$m'$  becomes free

    Endif

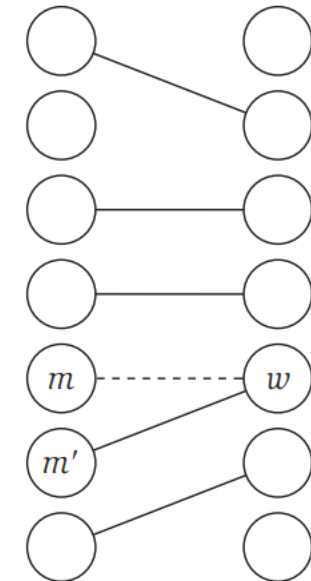
Endif

Endwhile

Return the set  $S$  of engaged pairs

- Can return a **stable matching**?
- How to **prove** our judgement?

Woman  $w$  will become engaged to  $m$  if she prefers him to  $m'$ .



**Figure 1.2** An intermediate state of the G-S algorithm when a free man  $m$  is proposing to a woman  $w$ .

# Demo 1

Iteration	Man Proposes to	Woman	Engagement	Free Males	Reason
0				A, B, C	
1					
2					
3					
4					
5					
6					

Man	Preference List	Woman	Preference List
A	$w_1, w_2, w_3$	$w_1$	C, B, A
B	$w_1, w_2, w_3$	$w_2$	C, B, A
C	$w_1, w_2, w_3$	$w_3$	C, B, A

# Demo 1 - Result

Male proposes, but woman get to select the most preferred man

Iteration	Man Proposes to	Woman	Engagement	Free Males	Reason
0				A, B, C	
1	A	$w_1$	$(A, w_1)$	B, C	1 <sup>st</sup> proposal to $w_1$
2	B	$w_1$	$(B, w_1)$	A, C	$w_1$ likes B more than A
3	A	$w_2$	$(B, w_1), (A, w_2)$	C	1 <sup>st</sup> proposal to $w_2$
4	C	$w_1$	$(C, w_1), (A, w_2)$	B	$w_1$ likes C more than B
5	B	$w_2$	$(C, w_1), (B, w_2)$	A	$w_2$ likes B more than A
6	A	$w_3$	$(C, w_1), (B, w_2), (A, w_3)$	END	1 <sup>st</sup> proposal to $w_3$

Man	Preference List	Woman	Preference List
A	$w_1, w_2, w_3$	$w_1$	C, B, A
B	$w_1, w_2, w_3$	$w_2$	C, B, A
C	$w_1, w_2, w_3$	$w_3$	C, B, A

## Demo 2 (Let Women propose for fairness)

Iteration	Wm Proposes to	Man	Engagement	Free Wms	Reason
0				A, B, C	
1					
2					
3					
4					
5					
6					

Wms	Preference List	Man	Preference List
A	$m_1, m_2, m_3$	$m_1$	A, B, C
B	$m_1, m_2, m_3$	$m_2$	A, B, C
C	$m_1, m_2, m_3$	$m_3$	A, B, C

## Demo 2 - Result

Woman proposes, but man get to select the most preferred woman

Iteration	Wm Proposes to	Man	Engagement	Free Wms	Reason
0				A, B, C	
1	A	$m_1$	$(A, m_1)$	B, C	1 <sup>st</sup> proposal to $m_1$
2	B	$m_1$	$(A, m_1)$	B, C	$m_1$ (with A) rejects B
3	B	$m_2$	$(A, m_1), (B, m_2)$	C	1 <sup>st</sup> proposal to $m_2$ (SUC)
4	C	$m_1$	$(A, m_1), (B, m_2)$	C	$m_1$ (with A) rejects C
5	C	$m_2$	$(A, m_1), (B, m_2)$	C	$m_2$ (with B) rejects C
6	C	$m_3$	$(A, m_1), (B, m_2), (C, m_3)$	END	1 <sup>st</sup> proposal to $m_3$ (SUC)

Wms	Preference List	Man	Preference List
A	$m_1, m_2, m_3$	$m_1$	A, B, C
B	$m_1, m_2, m_3$	$m_2$	A, B, C
C	$m_1, m_2, m_3$	$m_3$	A, B, C

# Analyze the Algorithm

- (Assume men proposing)
- **Fact 1.1.** Woman  $w$  remains engaged from the point at which she receives her first proposal; and the sequence of partners to which she is engaged gets better and better.
  - From the woman's perspective.
  - Only changes her mind if a more preferred man proposes.
- **Fact 1.2.** The seq. of women to whom man  $m$  proposes gets worse and worse.
  - From the man's perspective.
  - Only proposes again if he is free or his engaged more preferred woman left him.
- **Fact 1.3.** The algorithm terminates after at most  $O(n^2)$  iterations.
  - Proof trick: focus on the terminate condition of the while loop.
    - One man proposes to one different woman (he never proposed to) in each iteration.
    - How many men? How many woman?  $n \times n$

$n = 2$ ,  $n^2$  possibilities,  $2^2=4$   
 $(m,w), (m,w'), (m',w), (m',w')$



# Analyze the Algorithm

- **Fact 1.4.** If man  $m$  is free at some point in the execution of the algorithm, then there is a woman to whom he has not yet proposed.
  - Proof idea: Contradiction analysis. Suppose the man has proposed to all, but now he is free. The only possibility is that all his proposed women are engaged with other men.
  - We have  $n$  women, engaged to  $n$  men, which does not include the man  $m$ . Now, we have  $n+1$  men. But it cannot be true.
- **Fact 1.5.** The set  $S$  returned at termination is a perfect matching.
  - Proof idea: The algorithm terminates only if there is no free man. So, all men and women are in the matching.

# Analyze the Algorithm

- **Fact 1.6.** The algorithm **returns a stable matching**.
  - Proof idea: at least know it is perfect, so prove there is **no instability**.
    - **Contradiction analysis** again. Suppose we really have one:
      - $(m, w)$  and  $(m', w')$  that:
        - Man  $m$  prefers  $w'$  more && woman  $w'$  prefers  $m$  more  $\rightarrow (m, w')$  ?
        - If man  $m$  is together with  $w$ , he must have **proposed to  $w'$  before**, as he like  $w'$  more.
        - So  $w'$  **rejected  $m$  for  $m'$** ? No way, as  $w'$  like  $m$  more than  $m'$ . (Fact 1.1)

Wms	Preference List	Man	Preference List
$w$		$m$	$w', w$
$w'$	$m, m'$	$m'$	

# Computational Tractability



# Computational Tractability

- **Resources** in computer systems?
  - **Time, space, communication.**  
[<https://ieeexplore.ieee.org/document/9063714>]
  - Which one is the most important?
    - We **focus on time**, and **space** is also considered.
    - Why do we consider space?
- Obj: **quantify computational tractability**. How it changes?
  - Proposing an algorithm is easy.
  - An **efficient** (and correct) algorithm will be challenging.

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# Computational Tractability

- Informal **efficiency** def.: **runs quickly on real input**.
  - Problem 1: run the algorithm in what **environment**?
  - Problem 2: how **fast** is fast?
    - Special case: a poor-designed algorithm runs **super fast** with **tiny data input**
    - Special case: a well-designed algorithm runs **super slow** with **huge data**
  - Problem 3: what is **real input**? Crash with large input?
- We want: **platform-independent** and **instance-independent**.



<https://analyticsindiamag.com/wp-content/uploads/2019/10/computer-types-10.jpg>

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# Worst-Case Running Time

- Quantify the input size:  $N$ 
  - $n$  men and  $n$  women (eg.  $n=3$ )
  - $2n$  preference lists, each list of length  $n$  (eg.  $2 \times 3$  plist  $\times 3$ )
  - $N = 2n \cdot n = 2n^2$
- Task: analyze the running time as a function of  $N$ .
- Worst-case running time: the largest possible running time with input  $N$ .
- Benchmark: brute-force algorithm (brainless design).

Man	Preference List	Woman	Preference List
A	$w_1, w_2, w_3$	$w_1$	C, B, A
B	$w_1, w_2, w_3$	$w_2$	C, B, A
C	$w_1, w_2, w_3$	$w_3$	C, B, A

# Redefine Efficiency

- Informal efficiency def.: achieves better worst-case performance, at an analytical level, than brute-force search.
- **Polynomial time**: quantify “reasonable” running time.
  - **Search space**: often grows exponentially.
  - **Reasonable** may be: from  $n$  to  $n + 1$ , running time increases by a constant.
- Informal efficiency def.: an alg. is efficient if runs in polynomial time.
  - Extreme cases exists like the worst-case concept, i.e.,  $n^{10000000}$  ?
  - Again, in reality, the polynomial concept works well.



# Running Times

$n$  is the number of data points

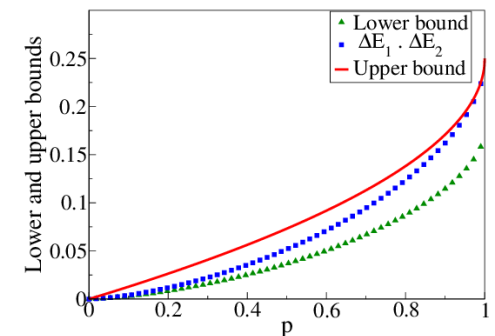
	$n$	$n \log_2 n$	$n^2$	$n^3$	$1.5^n$	$2^n$	$n!$
$n = 10$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	4 sec
$n = 30$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	18 min	$10^{25}$ years
$n = 50$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	11 min	36 years	very long
$n = 100$	< 1 sec	< 1 sec	< 1 sec	1 sec	12,892 years	$10^{17}$ years	very long
$n = 1,000$	< 1 sec	< 1 sec	1 sec	18 min	very long	very long	very long
$n = 10,000$	< 1 sec	< 1 sec	2 min	12 days	very long	very long	very long
$n = 100,000$	< 1 sec	2 sec	3 hours	32 years	very long	very long	very long
$n = 1,000,000$	1 sec	20 sec	12 days	31,710 years	very long	very long	very long

- Assumptions:
  - 1m instructions per second
  - Very long  $\rightarrow$  runs over  $10^{25}$  years.
- Intelligence does not necessarily produce polynomial algorithms.

1,000,000 instructions per second (1mins = 60secs)  
 $n^3 = 1,000^3 = 1,000,000,000$   
 $1,000,000,000 / 1,000,000 = 1,000\text{secs} \times 1/60 = 17\text{mins}$

# Asymptotic Order of Growth – Upper Bound

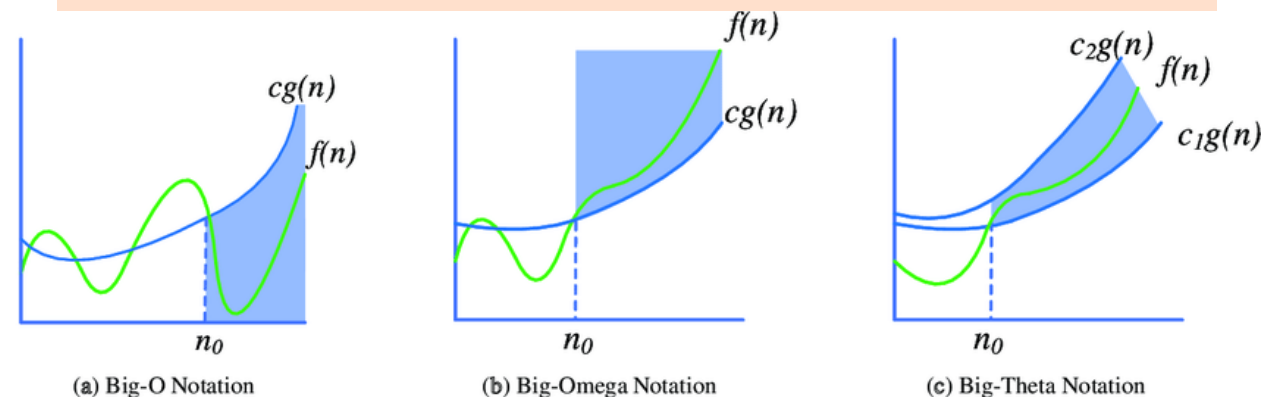
- Target: **worst-case** time, given input  $n$ , grows at a rate **at most proportional to  $f(n)$** .
  - We also call  $f(n)$  a **bound**.
- **Too specific expressions unnecessary**, i.e., runs in  $1.62n^2 + n + 4$  steps.
  - Constants vary for different architectures
  - Determined by the highest order (e.g.  $n^2$ )
- How do you plan to quantify the above terms concisely?
  - We **ignore the constants** and We **ignore the lower order items**.
- **Asymptotic Upper Bounds  $O(\cdot)$** :
  - The **worst-case running time  $T(n)$**  of a certain algorithm on an input of size  $n$ .
    - Expresses only an **upper bound**.
  - $f(n) = pn^2 + qn + r$  for constants  $p$ ,  $q$ , and  $r \rightarrow O(n^2)$ .



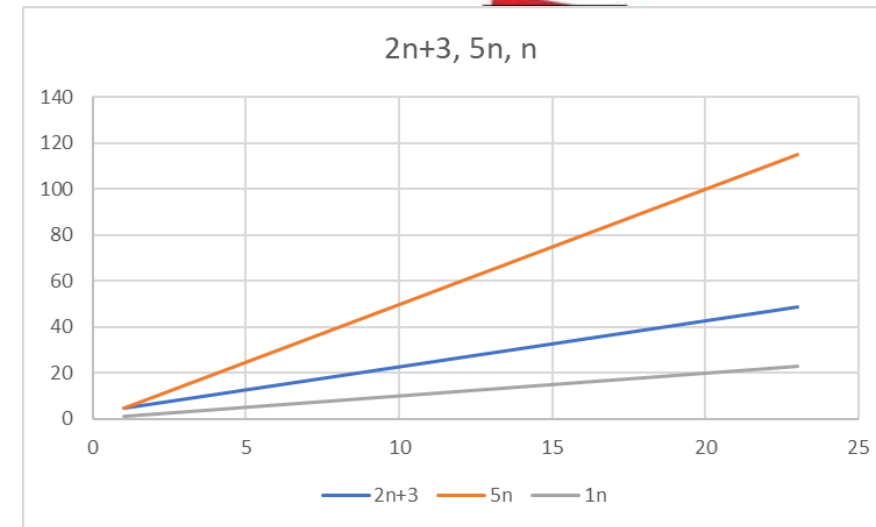
# Asymptotic Order of Growth – More bounds

- How to justify our proven bound  $O(\cdot)$  is good?
  - i.e., we are almost always true if we say it is  $O(n^{100000n})$ .
- **Asymptotic Lower Bounds :  $\Omega(n)$ .**
  - The func.  $T(n)$  is **at least** a constant “c” multiple of  $f(n) = c \cdot g(n)$ .
  - $T(n) = pn^2 + qn + r$  for constants  $p, q$ , and  $r \rightarrow \Omega(n^2)$ .
- **Asymptotically Tight Bounds  $\Theta(n)$ :**  $f(n) = \Omega(g(n)) = O(g(n))$  (optimal bound)
  - i.e.,  $f(n) = pn^2 + qn + r = \Omega(n^2) = O(n^2) = \Theta(n^2)$ .
- Note: consider  $n \rightarrow \infty$ .
  - Only matters when  $n$  is large.
  - $f(n) = \Theta(g(n))$  if  $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = c$ .
  - $c$  is a positive constant.

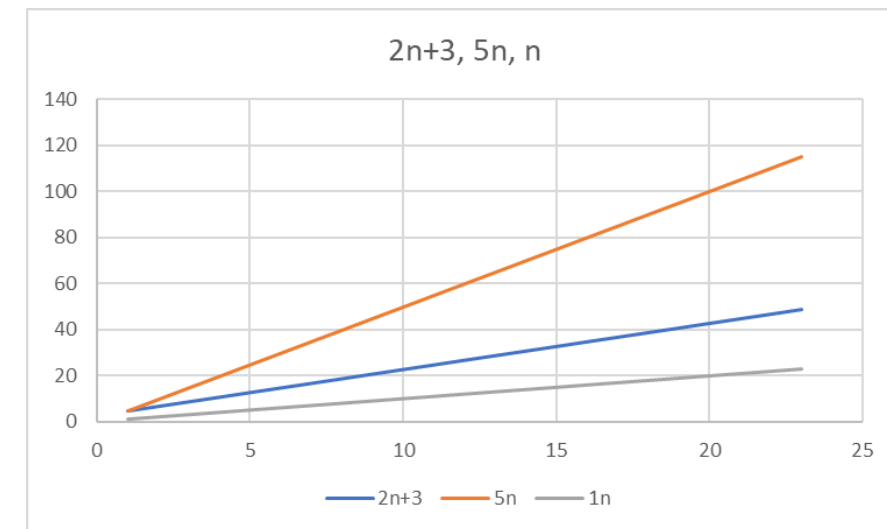
$c \cdot g(n)$  is the estimated bound



- $f(n) = 2n + 3$
- $f(n) = O(g(n))$  where  $g(n) = n$
- $f(n) \leq c_1 \cdot g(n) = c_1 \cdot n$  where  $n > n_0$ ,  $c_1 > 0$ ,  $n_0 \geq 1$
- $2n+3 \leq c_1 \cdot n$  where the RHS, big-O is larger than actual function
- If  $n_0=1$  and  $c_1 = 5$ , then  $5 \leq 5$ : ok
- If  $n_0=2$  and  $c_1 = 5$ , then  $7 \leq 10$ : ok
- Conclude that the big-oh of the function  $2n+3$  is  $O(n)$  where  $c_1=5$ ,  $n_0 \geq 1$ . This is the best worst case of this algorithm
- $f(n) = \Omega(g(n))$  where  $g(n) = n$
- $f(n) \geq c_2 \cdot g(n) = c_2 \cdot n$  where  $n > n_0$ ,  $c_2 > 0$ ,  $n_0 \geq 1$
- $2n+3 \geq c_2 \cdot n$  where RHS, big-omega is smaller than the actual function
- If  $n_0=1$  and  $c_2 = 1$ , then  $5 \geq 1$ : ok
- If  $n_0=2$  and  $c_2 = 1$ , then  $7 \geq 2$ : ok
- Conclude that the big-omega of the function  $2n+3$  is  $\Omega(n)$  where  $c_2=1$ ,  $n_0 \geq 1$ . This is the best case of this algorithm (the algo will never be better than this)



- $f(n) = 2n + 3$
- $f(n) = \theta(g(n))$  where  $g(n) = n$
- $c_2 \cdot n = c_2 \cdot g(n) \leq f(n) \leq c_1 \cdot g(n) = c_1 \cdot n$  where  $n > n_0$ ,  $c_1, c_2 > 0$ ,  $n_0 \geq 1$
- $2n+3 \leq c_1 \cdot n$  where the RHS is the worst case
- Conclude that the worst case of the function  $2n+3$  is  $\theta(n)$  where  $c_1=5$ ,  $n_0 \geq 1$ . This is the best worst case of this algorithm
- $c_2 \cdot n \leq 2n+3$  where LHS, is the best case
- Conclude that the best case of the function  $2n+3$  is  $\theta(n)$  where  $c_2=1$ ,  $n_0 \geq 1$ . This is the best case of this algorithm (the algo will never be better than this)
- **theta is the average case of the algorithm**



## Common Times - $O(n)$

- Long-time goal: recognize **common styles** of analyzing the running time.
- **Linear time**:  $O(n)$ , time scales by a constant factor times the input size  $n$ .
- **Find the maximum**:  $a_1, a_2, \dots, a_n$ .
  - Each needs a comparison, in constant time.
  - $O(n) * O(1) = O(n)$ .
- **Merge sorted lists**. The same  $O(n)$ , but more complex.
  - List  $a_1, a_2, \dots, a_n$  and list  $b_1, b_2, \dots, b_n$ , both in ascending order.
  - Objective:  $c_1, c_2, \dots, c_{2n}$ .
    - i.e., 2, 3, 11, 19 and 4, 9, 16, 25  $\rightarrow$  2, 3, 4, 9, 11, 16, 19, 25.
  - A non-intelligent solution: concatenate the lists  $\rightarrow$  sort all the items.
  - Record  $pa$ ,  $pb$ , and  $pc$ , where  $pa = pb = pc = 1$  in the beginning.
  - While  $pa \leq n$  and  $pb \leq n$ :
    - If  $a_{pa} > b_{pb} \rightarrow$  smaller  $b_{pb}$  to  $c_{pc} \rightarrow pb++$  and  $pc++$ .
    - If  $a_{pa} \leq b_{pb} \rightarrow$  smaller  $a_{pa}$  to  $c_{pc} \rightarrow pa++$  and  $pc++$ .
  - Whatever left, append to  $c$ .

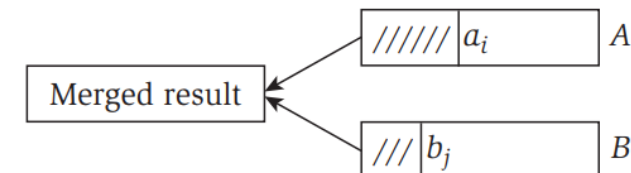
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max = a1
For i = 2 to n
    If ai > max then
        set max = ai
    Endif
Endfor
  
```

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Append the smaller of  $a_i$  and  $b_j$  to the output.



# Common Times - $O(n \log n)$ , $O(n^2)$ , $O(n^3)$ , $O(n^k)$



- **$O(n \log n)$** : the time of any alg. **splits input into equal-sized pieces**, **solves each recursively**, and **combines** the two **solutions** in linear time.
  - Popular example: **sorting**.
  - Analyze the basic idea: sorting is expensive  $\rightarrow$  sort less  $\rightarrow$  combine.
    - i.e.,  $100^2 = 10,000 \rightarrow 2 \times 50^2 = 5,000 \rightarrow 5,000 + 100 = 5,100$ , around half time.
- **$O(n^2)$** : i.e.,  **$n$  points in a plane**, what the **nearest points**?
  - Brute-force sol.: for each pt, compute the **dist. between it and another pt**.
  - $O(n)$  points, each  $O(n)$  distances, each distance  $O(1) \rightarrow O(n^2)$ .
- **$O(n^3)$** : nested loops, i.e., **3-variable solver**:  $2x + 5y + 9z = 100$ .
  - For  $x \rightarrow$  for  $y \rightarrow$  for  $z \rightarrow$  check the sum.
- **$O(n^k)$** : Any idea on getting  $O(n^k)$ ?
  - Independent sets in a graph with  $k$  nodes.
  - Brute-force, **choose  $k$  from  $n$** :  $\binom{n}{k} = \frac{n(n-1)\cdots(n-k+1)}{1 \cdot 2 \cdots k} \leq \frac{n^k}{k!} \leq n^k$ .



# Beyond Polynomial Time

- $O(2^n)$  vs.  $O(n!)$ . Which one is faster?
  - $O(2^n)$ : the number of the subsets of  $n$  nodes.
  - $O(n!)$ : the number of perfect matchings; traveling salesman problem.
- **Sublinear time**: better than linear, i.e.,  $O(\log n)$ .
  - Binary search.
  - **Query** the inputs instead of traversing them.
  - Half  $\rightarrow$  half  $\rightarrow$  half  $\rightarrow$  half ...