

CMPT231

Data Structures and Algorithms

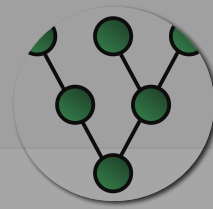
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Psalm 127:1-2 (NIV)

Unless the Lord **builds** the house,
the builders **labor** in vain.

Unless the Lord **watches** over the city,
the guards **stand watch** in vain.

In vain you **rise early** and **stay up** late,
toiling for food to eat –
for he **grants sleep** to those he loves.



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Schedule

#	Date	Topic	Text	HW Due
1	Sep 13	Analysis of Algorithms, Insertion Sort, Math Review, Asymptotic Notation	ch 1-3	-
2	Sep 20	Divide and Conquer, Solving Recurrences, Randomized Algorithms	ch 4-5	HW1: 9/22
3	Sep 27	Heaps/Queues, Quicksort	ch 6-7	HW2: 9/29
4	Oct 4	Linear-time Sort and Hash Tables	ch 8, 11	HW3: 10/6
5	Oct 11	Pointers: Linked Lists, Binary Search Trees	ch 10, 12	HW4: 10/13
	Oct			HW5:



What you need to succeed in 231

- **Explorer's** heart (self-motivated)
- Discrete **math** (e.g., MATH150)
 - Logic, **proofs**
- Comfortable **coding** environment
 - Python, C++, Java, etc.
 - but not until **later** in semester

Outline for today

- **Algorithmic** analysis
 - **Insertion** sort
- Discrete **math** review
 - **Logic** and proofs
 - Monotonicity, limits, iterated functions
 - **Fibonacci** sequence and golden ratio
- **Asymptotic** notation: V , O , Θ , o , ω
 - **Proving** asymptotic bounds

What is an algorithm?

- Precise **process** for solving a problem:
 - Input → Compute → Output
- Various languages for **expression**:
 - English, pseudocode, UML diagrams, etc.
- Programming languages for **implementation**:
 - Python, C, Java, etc.
- Focus: not **toolkits** but **problem solving**

Algorithmic complexity

- How many machine **instr** to **execute**
 - As function of input **size**
 - Ignoring **constant** factors
- Depends on machine **architecture**
 - CPUs generally **sequential**
 - GPUs are massively **parallel**
- **Running time** is more complex than this
 - Cache / **memory** very important

Basic machine model

- Our simple CPU **instruction set**:
 - **Arith**: +, -, *, /, <, >, ≠
 - **Data**: load (read), store (write), copy
 - **Control**: if/else, for/while, functions
 - **Types**: char, int, float
 - Data **Structures**: pointers, fixed-length arrays
 - but **not** Python list / STL vec!
- Assume each takes **constant** time

Problem definition: Sorting

- Input: **array** of key-value pairs
 - wlog, assume **keys** are $1 \dots n$
 - **values** (payload) can be any data
- Output: array **sorted** by key
 - **in-place**: modify original array
 - **out-of-place**: return a copy
- In standard **libraries**:
 - Python: `sort()`, `sorted()`
 - C++/Java: `sort()`
 - **How** do they do it?

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Insertion sort: hand of cards

```
insertion_sort(A, n):  
  for j = 2 to n:  
    key = A[j]  
    i = j - 1  
    while i > 0 and A[i] > key:  
      A[i+1] = A[i]  
      i = i - 1  
    A[i+1] = key
```

In: E B D F A C

j=3 B E D F A C

j=4 B D E F A C

j=5 B D E F A C

j=6 A B D E F C

Out: A B C D E F

Proof of correctness

- Loop **invariant**:
- Property that is true **before**, **during**, and **after** loop
- For insertion sort: at each iteration of loop,
 - part of array $A[1 \dots j-1]$ is in **sorted** order

```
insertion_sort(A, n):  
  for j = 2 to n:  
    key = A[j]  
    i = j - 1  
    while i > 0 and A[i] > key:  
      A[i+1] = A[i]  
      i = i - 1  
    A[i+1] = key
```

Complexity analysis

Let t_j be the number of times the loop condition is checked in the inner “while” loop:

```
insertion_sort(A, n):  
  for j = 2 to n: # c0 * n  
    key = A[j] # c1 * (n-1)  
    i = j - 1 # c2 * (n-1)  
    while i > 0 and A[i] > key: # c3 * sum (t_j)  
      A[i+1] = A[i] # c4 * sum (t_j - 1)  
      i = i - 1 # c5 * sum (t_j - 1)  
    A[i+1] = key # c6 * (n-1)
```

Summation notation: $\sum_{j=2}^n t_j = t_2 + t_3 + \dots + t_n$

Best vs. worst case

- **Best** case is if input is **pre-sorted**:
 - Still need to **scan** to verify sorted,
 - But inner **while** loop only has 1 iteration: $t_j = 1$
 - Total complexity: $T(n) = a n + b$, for some a, b
 - **Linear** in n
- **Worst** case: input is in **reverse** order!
 - Inner “while” loop always max iterations: $t_j = j$
 - Calculate **total** complexity $T(n)$:
 - Pick a line in inner loop, e.g., **line 5**: $A[i+1] = A[i]$
 - Complexity of other lines **similar**

Worst case complexity

- **Total** complexity for line 5, worst-case:

$$\begin{aligned}c_4 \sum_2^n (t_j - 1) &= c_4 \sum_2^n (j - 1) \\ &= c_4 (n - 1) \frac{n}{2} = \left(\frac{c_4}{2}\right) n^2 - \left(\frac{c_4}{2}\right) n\end{aligned}$$

- **Quadratic** in n
- **Average** case: input is random, $t_j = \frac{j}{2}$ on average
 - Still quadratic (only changes by a **constant** factor)

Theta (Θ) notation

- Insertion sort, line 5: $\left(\frac{c_4}{2}\right)n^2 - \left(\frac{c_4}{2}\right)n$
- **Constants** c_1, c_2, \dots may vary for different computers
 - As n gets **big**, constants become **irrelevant**
 - Even the n term is dominated by the n^2 term
- Complexity of insertion sort is on **order** of n^2
 - Notation: $T(n) = \Theta(n^2)$ (“theta”)
- $\Theta(1)$ means an algorithm runs in **constant time**
 - i.e., does not depend on **size** of input
- We’ll define V more **precisely** later today

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Logic notation

- $\neg A$ (or $!A$): “**not** A”
 - e.g., let $A =$ “it is Tue”: then $\neg A =$ “it is not Tue”
- $A \Rightarrow B$: “**implies**”, “if A, then B”
 - e.g., let $B =$ “meatloaf”:
 - then $A \Rightarrow B =$ “if Tue, then meatloaf”
- $A \Leftrightarrow B$: if and only if (“**iff**”):
 - **equivalence**: $(A \Rightarrow B)$ and $(B \Rightarrow A)$
- **John 14:15**: “If you love me, keep my commands”
- **v21**: “Whoever keeps my commands loves me”
- **v24**: “He who does not love me will not obey my teaching”

Logic notation: \forall and \exists

- \forall : “**for all**”,
 - e.g., “ \forall day: meal(day) = meatloaf”
 - “For all days, the meal on that day is meatloaf”
- \exists : “there **exists**” (not necessarily unique)
 - e.g., “ \exists day: meal(day) = meatloaf”
 - “There exists a day on which the meal is meatloaf”

Logic: contrapos and converse

- **Contrapositive** of “ $A \Rightarrow B$ ” is $\neg B \Rightarrow \neg A$
 - **Equivalent** to original statement
 - “If Tue, then meatloaf” \Leftrightarrow
“if not meatloaf, then not Tue”
- **Converse** of “ $A \Rightarrow B$ ” is “ $\neg A \Rightarrow \neg B$ ”
 - **Not** equivalent to original statement!
 - “if not Tue, then not meatloaf”

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Monotonicity

- $f(x)$ is **monotone increasing** iff: $x < y \Rightarrow f(x) \leq f(y)$
 - Also called “non-decreasing”
 - Can be **flat**
- $f(x)$ is **strictly increasing** iff: $x < y \Rightarrow f(x) < f(y)$
 - note inequality is **strict**
- “ $a \bmod n$ ” is the **remainder** of a when divided by n
 - e.g., $17 \bmod 5 = 2$ (in Python: `17 % 5`)

Limits

- Formal **definitions** of limits involve \forall and \exists
- $\lim_{x \rightarrow a} f(x) = b$: “**limit** of $f(x)$ as x goes to a ”
 - $\forall z > 0, \exists y > 0: |x-a| < y \Rightarrow |f(x)-b| < z$
 - When x is “close” to a , then $f(x)$ is “close” to b
- $\lim_{n \rightarrow \infty} f(n) = b$: “**limit** of $f(n)$ as n goes to infinity”:
 - $\forall z > 0, \exists n_0: n > n_0 \Rightarrow |f(n)-b| < z$
 - When n is “big”, $f(n)$ is “close” to b

Iterated functions (recursion)

- $f^{(i)}(x)$: function **f**, **applied i** times to **x**: $f(f(f(\dots f(x) \dots)))$
 - **Not** the same as $f^i(x) = (f(x))^i$
- e.g., $\log^{(2)}(1000) = \log(\log(1000)) = \log(3) \approx 0.477$
 - But $\log^2(1000) = (\log(1000))^2 = 3^2 = 9$
- By convention, $f^{(0)}(x) = x$ (apply **f zero** times)

Iterated log: $\log^*(n)$

- $\log^*(n) = \min \left(i \geq 0 : \log^{(i)}(n) \leq 1 \right)$
- # **times log** needs to be applied to **n** until the result is **≤ 1**
- e.g., let $\lg = \log_2$:
 - then $\lg^*(16) = 3$, because
 - $\lg(\lg(\lg(16))) = \lg(\lg(4)) = \lg(2) = 1$

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Fibonacci sequence

- The n-th **Fibonacci number** is $F_n = F_{n-1} + F_{n-2}$
 - Start with $F_0 = 0, F_1 = 1$:
 - $F_n = 0, 1, 1, 2, 3, 5, 8, 13, 21, \dots$
 - (Lucas numbers start with $F_0 = 2$)
- Shows up all over **nature**
 - Num of **spirals** on sunflowers, pinecones, etc.
 - Vi Hart video: Doodling in Math



The Golden ratio ϕ

- ϕ is the **solution** to the equation $x^2 = x + 1$
 - $\phi = \frac{1 \pm \sqrt{5}}{2}$
 - Actually, two solutions: ϕ and its **conjugate**, $\bar{\phi}$
 - $\phi \approx 1.61803$, and $\bar{\phi} \approx -0.61803$
- Also shows up all over **nature**
 - Dimensions of **Nautilus** seashells, spiral **galaxies**, etc.
 - **Aspect ratio** in architecture, e.g., Parthenon

Fibonacci + golden ratio

- Can **prove** (#3.2-7) that $F_n = \frac{\phi^n - (\bar{\phi})^n}{\sqrt{5}}$
- Second term is **fractional**: $\frac{|(\bar{\phi})^n|}{\sqrt{5}} < \frac{1}{2}$
- Thus: $F_n = \left\lfloor \frac{\phi^n}{\sqrt{5}} + \frac{1}{2} \right\rfloor = \text{round} \left(\frac{\phi^n}{\sqrt{5}} \right)$
- i.e., Fibonacci grows **exponentially**!

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Asymptotic growth: Θ , O , Ω

- Behaviour “**in the limit**” (big n)
- Define V as **class** of functions: $f(n) \in V(g(n))$ iff
 - $\exists c_1, c_2, n_0: \forall n > n_0, 0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n)$
 - f is “**sandwiched**” between two multiples of g :
 - $c_1 g(n)$ and $c_2 g(n)$
- “**Big O**”: specify only **upper** bound: $f(n) \in O(g(n))$ iff
 - $\exists c_2, n_0: \forall n > n_0, 0 \leq f(n) \leq c_2 g(n)$
 - e.g., $\Theta(n^2) \subset O(n^2) \subset O(n^3)$
- “**Big Omega**”: $\Omega(g(n))$ specifies only the **lower** bound
 - Think of other examples?

$f(n)$

$c_1 g(n)$

Proving asymptotic growth

- (p.52 #3.1-2) $\forall a, b > 0$, prove: $(n + a)^b \in \Theta(n^b)$
- From definition: we need to **find** n_0, c_1, c_2 such that
$$\forall n > n_0 : c_1 n^b \leq (n + a)^b \leq c_2 n^b$$
- i.e., find constants so we can **sandwich** $(n + a)^b$ in between two multiples of n^b

Prove: $(n+a)^b \in \Theta(n^b)$

- **Observe** that $n+a \geq n/2$, as long as $n > 2|a|$
 - Also, $n+a \leq 2n$, as long as $n > |a|$
 - Hence, $n+a$ is **sandwiched** by $n/2$ and $2n$ (if $n > 2|a|$):
 - $n/2 \leq n+a \leq 2n$
- **Raise** to the b power (x^b is **monotone** if $x > 1$, $b > 0$)
 - Thus, $\left(\frac{n}{2}\right)^b \leq (n+a)^b \leq (2n)^b$ (for $n > 2|a|$)
- So we select $n_0 = 2|a|$, $c_1 = 2^{-b}$, $c_2 = 2^b$
 - This **proves** the Theta bound.

Asymptotic shorthand

- $V(g)$ is a **class** of functions
 - But for convenience, some **short-hand** notation:
- When V (et al) are on the **right** side of $=$:
 - It means “there **exists**“ $f \in \Theta(g)$
 - e.g., $2n^2 + 3n = \Theta(n^2)$
- When V (et al) are on the **left** side of $=$:
 - It means “**for all**“ $f \in \Theta(g)$
 - e.g., $4n^2 + \Theta(n \log(n)) = \Theta(n^2)$
 - True for **any** function in $\Theta(n \log(n))$

Asymptotic domination: o , ω

- “**Little o**“: like a strict **less than** inequality: $f \in o(g)$ iff
 - $\forall c > 0 \exists n_0: \forall n > n_0, 0 \leq f(n) < c g(n)$
 - i.e., the **limit** of $f(n)/g(n) \rightarrow 0$ as $n \rightarrow \infty$
- “**Little omega**“: like a strict **greater than**: $f \in \omega(g)$ iff
 - $\forall c > 0 \exists n_0: \forall n > n_0, 0 \leq c g(n) < f(n)$
 - i.e., the **limit** of $f(n)/g(n) \rightarrow \infty$ as $n \rightarrow \infty$

Examples of o and ω

- **Little o:** $n^{1.999} \in o(n^2)$, and $\frac{n^2}{\log(n)} \in o(n^2)$
 - but $\frac{n^2}{10000} \notin o(n^2)$
- **Little omega:** $n^{2.0001} \in \omega(n^2)$ and $n^2 \log(n) = \omega(n^2)$

Useful math identities

- All **logs** are the **same** up to a constant factor:
 - $\log_a(n) = \left(\frac{1}{\log_b(a)}\right) \log_b(n)$
 - So we often use **lg** = \log_2 for convenience
- $\Theta(1) \subset o(\log(n)) \subset o(n) \subset o(n^p) \subset o(p^n)$
 - For any constant $p > 1$
- In fact, $\forall a > 1, b: \lim_{n \rightarrow \infty} \frac{n^b}{a^n} = 0$
 - Hence, $n^b \in o(a^n)$
 - *”Exponentials dominate polynomials”*

Stirling's approximation

- **Factorial:** $n! = n(n-1)(n-2)\dots(2)(1)$
 - Number of **permutations** of n distinct objects
- **Stirling's approx:** $n! = \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \left(1 + \Theta\left(\frac{1}{n}\right)\right)$
- Hence, $\log(n!) \in \Theta(n \log(n))$

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Example asymptotic proof

- (p.62 #3-3): **Prove**: $(\log n)! \in \omega(n^3)$
- **Approach**: take **log** of both sides (log is monotone)
- **Left side**: use Stirling: $n! = \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \left(1 + \Theta\left(\frac{1}{n}\right)\right)$
 - So $\log(n!) \in V(n \log(n))$
 - Now **substitute** $\log(n)$ for n , using monotonicity of log:
 - So $\log((\log n)!) \in V((\log n) \log(\log n))$

Prove: $(\log n)! = \omega(n^3)$

- ... so: $\log((\log n)!) \in V((\log n) \log(\log n))$
- **Right side:** $\log(n^3) = 3 \log n$
 - This is **close** to the left side, with 3 instead of $\log(\log n)$
 - But we only need an **ω** bound, and $\log(\log n) \in \omega(3)$
- **Combining:** $\log((\log n)!) \in \Theta((\log n) \log(\log n))$
 - $= \omega((\log n)3) = \omega(\log(n^3))$
- So by **monotonicity**, $(\log n)! \in \omega(n^3)$

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