

# CMPT307: Tutorial 2

Week 14-1

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- ▷ dynamic sets:
  - \* linked lists
  - \* hash tables: hashing with chaining and open addressing
  - \* binary search trees: rotations
  - \* red-black trees
- ▷ binary heap/priority queues: heap-sort, MST, shortest paths
- ▷ disjoint sets: connected components, testing cycles, MST
- ▷ amortized analysis
  - \* aggregate analysis
  - \* accounting method
  - \* potential function method

# Sorting/Order Statistics

- ▷ insertion sort: running time is not asymptotically optimal
- ▷ merge sort: divide-and-conquer, not in place
- ▷ heap sort: binary heap data structure
- ▷ quick sort: divide-and-conquer, asymptotically optimal on average
- ▷ lower bound of comparison sort
- ▷ radix sort: beat  $O(n \log n)$  for  $k$  constant (using stable sort)
- ▷ order statistics: divide-and-conquer

# Graph Problems

- ▷ elementary graph algorithms
  - \* BFS: unweighted shortest path, matching
  - \* DFS: topological sorting, strongly connected components
- ▷ MST: greedy (graph cut)
- ▷ shortest paths: RELAX, dynamic programming
- ▷ maximum bipartite matching: augmenting paths (optimal condition)

- ▷ divide-and-conquer: sorting, order statistics, maximum subarray, (square) matrix multiplication
- ▷ dynamic programming: rod cutting, matrix-chain multiplication, longest common subsequence, optimal binary search tree
- ▷ greedy algorithm: activity selection, Huffman code, MST, matroids

- ▷ polynomial time algorithms
- ▷ complexity classes:  $\mathcal{P}$ ,  $\mathcal{NP}$ ,  $\text{co-}\mathcal{NP}$  and NP-completeness
- ▷ polynomial time reductions
- ▷ basic NP-complete problems
- ▷  $\mathcal{P} = \mathcal{NP}$ ? open, but most believe not

# Approximation Algorithms

- ▷  $\mathcal{A}$  returns a feasible solution in polynomial time
- ▷ no need to prove optimality, feasibility is often easy
- ▷ often easy to show that the running time is polynomial
- ▷ want approximation ratio as close to **lower bound** as possible  
important and often hard

# About the Final Exam

| COURSE        | DAY    | DATE   | TIME        | ROOM     | TYPE        |
|---------------|--------|--------|-------------|----------|-------------|
| cmpt 307-d300 | Friday | Dec 16 | 08:30-11:30 | SWH10081 | closed exam |

- ▷ final exam procedures
- ▷ office hours on Dec 12, starting from 13:00
  - \* if no students come, it ends at 14:00
  - \* other time by appointment
- ▷ coverage: all lecture contents (chapters/sections not covered by lectures are not required)

## Q. Types

- running time analysis
- data structures
- algorithm design
- complexity

# Running Time Analysis

- ▷ amortized analysis: analyze the amortized cost
- ▷ worst-case: may need to use the amortized cost
- ▷ average-case
  1. make reasonable assumption on input (often uniform distribution)
  2. define suitable **indicator random variable**  $X_i$ ,  $P(X_i = 1) = ?$
  3. represent the running time  $T$  as a function of  $X_i$ , say  
 $T = f(X_1, X_2, \dots)$
  4. compute  $\mathbb{E}[T]$

- ▷ worst-case running time
- ▷ average-case running time (probabilistic analysis)
- ▷ amortized costs
- ▷ (balanced) binary search trees: rotations
- ▷ heaps/priority queues
- ▷ disjoint sets

▷ divide-and-conquer:

1. describe an algorithm (or write pseudocode)
2. the correctness is often straightforward, but you must analyze the running time
3. for **recurrences**, you should be able to use the three approaches (master theorem will be presented as appendix to you)

▷ dynamic programming

1. derive a recursive formula (including boundary conditions)
2. optimal substructure must be formally proved (need verify feasibility when using cut-and-paste argument)
3. analyze the running time based on the recursive formula
4. no need to write pseudocode or to construct optimal solution if not indicated

- ▷ greedy algorithm (matroids)
  1. describe a greedy strategy and analyze the running time (often straightforward)
  2. prove the correctness (important)
  3. you may also be asked to formulate the problem as a standard matroid maximization problem, so you also need to prove that the system is a matroid
- ▷ approximation algorithm
  1. describe an algorithm and analyze the running time
  2. analyze the approximation ratio (important)

- ▷  $L \in \mathcal{NP}$  or  $\text{co-}\mathcal{NP}$ ? short and checkable certificate
- ▷ show that decision problem  $B$  is NP-complete
  1. show that  $B \in \mathcal{NP}$
  2. you may use any known NP-complete problem  $A$  and show that  $A \leq_P B$ : **must check** that the reduction is correct and can be done in polynomial time