Introduction to Combinatorial Algorithms

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Introduction

Introduction to the course

What are:

- Combinatorial Structures?
- Combinatorial Algorithms?
- Combinatorial Problems?

Combinatorial Structures

Combinatorial structures are *collections* of k-subsets/k-tuple/permutations from a parent set (finite).

• Undirected Graphs:

Collections of 2-subsets (edges) of a parent set (vertices).

$$V = \{1, 2, 3, 4\}$$
 $E = \{\{1, 2\}, \{1, 3\}, \{1, 4\}, \{3, 4\}\}$

Directed Graphs:

Collections of 2-tuples (directed edges) of a parent set (vertices).

$$V = \{1, 2, 3, 4\}$$
 $E = \{(2, 1), (3, 1), (1, 4), (3, 4)\}$

Hypergraphs or Set Systems:

Similar to graphs, but hyper-edges are sets with possibly more than two elements.

$$V = \{1, 2, 3, 4\}$$
 $E = \{\{1, 3\}, \{1, 2, 4\}, \{3, 4\}\}$

Building blocks: finite sets, finite lists (tuples)

Finite Set

$$X = \{1, 2, 3, 5\}$$

undordered structure, no repeats

$$\{1, 2, 3, 5\} = \{2, 1, 5, 3\} = \{2, 1, 1, 5, 3\}$$

• cardinality (size) = number of elements, |X| = 4.

A k-subset of a finite set X is a set $S \subseteq X$, |S| = k.

For example: $\{1,3\}$ is a 2-subset of X.

Finite List (or Tuple)

$$L = [1, 5, 2, 1, 3]$$

- ordered structure, repeats allowed $[1, 5, 2, 1, 3] \neq [1, 1, 2, 3, 5] \neq [1, 2, 3, 5]$
- length = number of items, length of L is 5.

An n-tuple is a list of length n.

A permutation of an n-set X is a list of length n such that every element of X occurs exactly once.

Graphs

Definition

A graph is a pair (V, E) where:

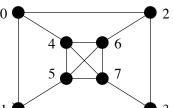
V is a finite set (of vertices).

E is a finite set of 2-subsets (called edges) of V.

$$G_1 = (V, E)$$

$$V = \{0, 1, 2, 3, 4, 5, 6, 7\}$$

$$E = \{\{0, 4\}, \{0, 1\}, \{0, 2\}, \{2, 3\}, \{2, 6\}, \{1, 3\}, \{1, 5\}, \{3, 7\}, \{4, 5\}, \{4, 6\}, \{4, 7\}, \{5, 6\}, \{5, 7\}, \{6, 7\}\}$$

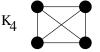


Complete graphs are graphs with all possible edges.

K₁

К2



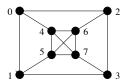


Substructures of a graph: hamiltonian cycle

Definition

A hamiltonian cycle is a closed path that passes through each vertex once.

The list [0,1,5,4,6,7,3,2] describes a hamiltonian cycle in the graph: (Note that different lists may describe the same cycle.)



Problem (Traveling Salesman Problem)

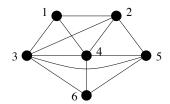
Given a weight/cost function $w: E \to R$ on the edges of G, find a smallest weight hamiltonian cycle in G.

Substructures of a graph: cliques

Definition

A clique in a graph G=(V,E) is a subset $C\subseteq V$ such that $\{x,y\}\in E$, for all $x,y\in C$ with $x\neq y$.

(Or equivalently: the subgraph induced by C is complete).



- Some cliques: $\{1, 2, 3\}, \{2, 4, 5\}, \{4, 6\}, \{1\}, \emptyset$
- Maximum cliques (largest): $\{1, 2, 3, 4\}$, $\{3, 4, 5, 6\}$, $\{2, 3, 4, 5\}$

Famous problems involving cliques

Problem (Maximum clique problem)

Find a clique of maximum cardinality in a graph.

Problem (All cliques problem)

Find all cliques in a graph without repetition.



Definition

A set system (or hypergraph) is a pair (X, \mathcal{B}) where:

X is a finite set (of points/vertices).

 \mathcal{B} is a finite set of subsets of X (blocks/hyperedges).



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- Partition of a finite set:

A partition is a set system (X, \mathcal{B}) such that

$$B_1 \cap B_2 = \emptyset$$
 for all $B_1, B_2 \in \mathcal{B}, B_1 \neq B_2$, and $\bigcup_{B \in \mathcal{B}} B = X$.

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Steiner triple system (a type of combinatorial designs):

 \mathcal{B} is a set of 3-subsets of X such that for each $x,y\in X, x\neq y$, there exists eactly one block $B\in\mathcal{B}$ with $\{x,y\}\subseteq B$.

$$X = \{0, 1, 2, 3, 4, 5, 6\}$$

 $\mathcal{B} = \{\{0, 1, 2\}, \{0, 3, 4\}, \{0, 5, 6\}, \{1, 3, 5\}, \{1, 4, 6\}, \{2, 3, 6\}, \{2, 4, 5\}\}\}$

Combinatorial algorithms

Combinatorial algorithms are algorithms for investigating combinatorial structures.

- Generation
 - **Construct** all combinatorial structures of a particular type.
- Enumeration
 - **Compute the number** of all different structures of a particular type.
- Search
 - **Find at least one** example of a combinatorial structures of a particular type (if one exists).
 - **Optimization problems** can be seen as a type of search problem.

Generation

Construct all combinatorial structures of a particular type.

- Generate all subsets/permutations/partitions of a set.
- ► Generate all cliques of a graph.
- Generate all maximum cliques of a graph.
- Generate all Steiner triple systems of a finite set.

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Enumeration

Compute the number of all different structures of a particular type.

- ► Compute the number of subsets/permutat./partitions of a set.
- Compute the number of cliques of a graph.
- Compute the number of maximum cliques of a graph.
- ▶ Compute the number of Steiner triple systems of a finite set.



Search

Find at least one example of a combinatorial structures of a particular type (if one exists).

Optimization problems can be seen as a type of search problem.

- ► Find a Steiner triple system on a finite set. (feasibility)
- Find a maximum clique of a graph. (optimization)
- Find a hamiltonian cycle in a graph. (feasibility)
- Find a smallest weight hamiltonian cycle in a graph. (optimization)

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 - NP = class of decision problems that can be **verified** in polynomial time. (e.g. Hamiltonian path in a graph is in NP)

Therefore, $P \subseteq NP$.

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- There are several approaches to deal with NP-hard problems.

Approaches for dealing with NP-hard problems

Exhaustive Search

- exponential-time algorithms.
- solves the problem exactly

(Backtracking and Branch-and-Bound)

Heuristic Search/Metaheuristics

- algorithms that explore a search space to find a feasible solution that is hopefully "close to" optimal, within a time limit
- approximates a solution to the problem

(Hill-climbing, Simulated annealing, Tabu-Search, Genetic Algs, etc.)

Approximation Algorithms

- polynomial time algorithm
- we have a provable guarantee that the solution found is "close to" optimal.

(Approximation algorithms not covered in this course)



Types of Search Problems

1) Decision Problem:

A yes/no problem

Problem 1: Clique (decision)

Instance: graph G = (V, E),

target size k

Question:

Does there exist a clique C of G with |C| = k?

3) **Optimal Value**:

Find the largest target size.

Problem 3: Clique (optimal value) Instance: graph G = (V, E),

Find:

the maximum value of |C|, where C is a clique

2) Search Problem

Find the guy.

Problem 2: Clique (search)

Instance: graph G = (V, E),

target size k

Find:

a clique C of G with |C| = k, if one exists.

4) Optimization:

Find an optimal guy.

Problem 4: Clique (optimization)

Instance: graph G = (V, E),

Find:

a clique C such that |C| is maximize (max. clique)

Taximize (max. clique)

Topics for the Course

"text" refers to: Kreher&Stinson, Combinatorial Algorithms: generation, enumeration& search

- Generating elementary combinatorial objects [text Ch 2,3&Knuth's TAOCP 4A] Sequential generation (successor), rank, unrank. Algorithms for subsets, k-subsets, permutations, other objects.
- 2 Exhaustive Generation and Exhaustive Search

 $[{\sf text\ Chap4+\ Kaski\&Ostergard's\ book\ Chap\ 4.1}]]$

Backtracking algorithms (exhaustive generation, exhaustive search, optimiz.) Branch-and-bound (exhaustive search, optimization)

- **Heuristic Search** [text Chap 5 + Gendreau&Potvin's Handbook of Metaheuristics Vol.2] Hill-climbing, Simulated annealing, Tabu-Search, Genetic Algs, etc.
- Computing Isomorphism and Isomorph-free Exhaustive
 Generation

 [text Chap 7 + Kaski&Ostergard's book Chap 3.4]

Graph isomorphism, isomorphism of other structures. Computing invariants. Computing certificates. Isomorph-free exhaustive generation.



Course evaluation

- 45% Assignments
 3 assignments, 15% each
 Assignments covering: theory, algorithms, implementation
- 11% Student presentation (topic can be related to project)
 10% 15-20 minute talk
 1% participation (attending talks/discussion/other method)
 Talk teaching your peers on a topic not covered in lectures.
- 44% Project: individual, chosen by student
 3% Project proposal (1-2 pages)
 11% Project presentation (15-20 minute talk) 10%, particip. 1%
 30% Project paper (10-15 page)
 - research (reading papers related to course topics),
 - original work (involving one or more of: modelling, application, algorithm design, implementation, experimentation, analysis)

