Lecture I - Welcome to Combinatorics

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Combinatorial Optimization

Definition

Mathematical programming and optimisation

Types of mathematical optimisation models

Course Structure

Graphs

Definitions

adjective

Combinatorial:

Definition

 relating to the selection of a given number of elements from a larger number without regard to their arrangement.

Optimization (or Optimisation):

- ٠ noun
- the action of making the best or most effective use of a situation or resource.



Combinatorial Optimization

Definition

Can be achieved by:

- Analysing/Visualizing properties of functions / extreme points or
- Applying numerical methods

Optimisation has important applications in fields such as

- operations research (OR);
- economics;
- statistics;
- bioinformatics;
- machine learning and artificial intelligence.





Combinatorial Optimization

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In this course, optimisation is viewed as the core element of mathematical programming.

Math. programming is a central OR modelling paradigm:

- variables → decisions/point of interest: business decisions, parameter definitions, settings, geometries, ...;
- domain \rightarrow constraints/limitations: logic, design, engineering, ...;
- function \rightarrow objective function/profit: measurement of (decision) quality.

However, math. programming has many applications in fields other than OR, which causes some confusion;

We will study math. programming in its most general form: both constraints and objectives can nonlinear or linearfunctions, but the domain is discrete.



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Types of programming

Rule of Thumb:

The simpler are the assumptions which define a type of problems, the better are the methods to solve such problems.

Some useful notation:

- $x \in \mathbb{R}^n$: vector of (decision) variables x_j , j = 1, ..., n;
- $f: \mathbb{R}^n \to \mathbb{R} \cup \{\pm \infty\}$ objective function;
- $X \subseteq \mathbb{R}^n$: ground set (physical constraints);
- $g_i, h_i : \mathbb{R}^n \to \mathbb{R}$: constraint functions;
- $g_i(x) \leq 0$ for $i = 1, \ldots, m$: inequality constraints;
- $h_i(x) = 0$ for i = 1, ..., l: equality constraints.



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Our goal will be to solve variations of the general problem P:

- $\begin{array}{ll} (P): & \min \ f(x) \\ & \texttt{s.t.} \ g_i(x) \leq 0, i=1,\ldots,m \\ & h_i(x)=0, i=1,\ldots,l \end{array}$
- Linear programming (LP): linear $f(x) = c^{\top}x$ with $c \in \mathbb{R}^n$; constraint functions $g_i(x)$ and $h_i(x)$ are affine $(a_i^{\top}x b_i)$, with $a_i \in \mathbb{R}^n$, $b \in \mathbb{R}$); $X = \{x \in \mathbb{R}^n : x_j \ge 0, j = 1, ..., n\}.$
- Nonlinear programming (NLP): some (or all) of the functions f, g_i or h_i are nonlinear;
- (Mixed-)integer programming ((M)IP): LP where (some of the) variables are binary (or integer). $X \subseteq \mathbb{R}^k \times \{0,1\}^{n-k}$
- Mixed-integer nonlinear programming (MINLP): MIP+NLP.





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Course Structure

Structure

- Classes: Lecture and Exercise Session
 - \rightarrow attendance is not mandatory;
 - \rightarrow attendance is mandatory in Guest Lecture;
- Materials: Lectures Notes and Slides
 → available on "MyCourse";



Combinatorial Optimization

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Course Structure

1 project (group or individual): 60%;
 → based on course problems;

- 3 assignments (individual): 45% (15 each). \longrightarrow unique problem.
- Guest Lecture.

Mixture of modelling, analytical mathematics and programming.

Programming Language: Python and Julia

Assessment



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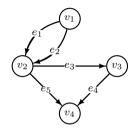
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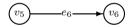
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Simple definitions I

- (undirected) graph $G = (V, E, \Psi)$
 - vertices V
 - edges E
 - function $\Psi \colon E \to \{X \subseteq V \colon |X| = 2\}$
- directed graph $G=(V,E,\Psi)$
 - vertices V
 - edges E
 - function $\Psi \colon E \to \{(v,w) \in V \times V \colon v \neq w\}$
- Edges e can have a value associated to it: $\rightarrow w \longrightarrow f_{uv}$ called weight or flow;
- in practice: $e = \{u, v\}$, e = (u, v)respectively, G = (V, E) $\mathfrak{C} E$ can contain multiple parallel edges







Combinatorial Optimization

Definitions

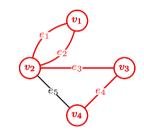
Mathematical programming and optimisation

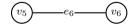
Types of mathematical optimisation models

Course Structure

Simple definitions II

- edge progression W in G from u_1 to u_{k+1} :
 - sequence $[u_1, a_1, u_2, \ldots, u_k, a_k, u_{k+1}]$ with $k \ge 0$
 - $a_i = \{u_i, u_{i+1}\} \in E(G)$
 - e.g. $[v_3, e_3, v_2, e_2, v_1, e_1, v_2, e_3, v_3, e_4, v_4]$
- walk W in G from u_1 to u_{k+1} :
 - edge progression with $a_i \neq a_j$, $1 \leq i < j \leq k$
 - e.g. $[v_2, e_2, v_1, e_1, v_2, e_3, v_3, e_4, v_4]$
- path P in G from u_1 to u_{k+1} , $u_1 u_{k+1}$ path:
 - graph $(\{u_1, \ldots, u_{k+1}\}, \{a_1, \ldots, a_k\})$ with $[u_1, a_1, u_2, \ldots, u_k, a_k, u_{k+1}]$ walk and $u_i \neq u_j$, $1 \leq i < j \leq k+1$
 - e.g. $[v_1, e_1, v_2, e_3, v_3, e_4, v_4]$







Combinatorial Optimization

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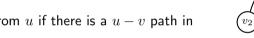
Mathematical programming and optimisation

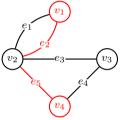
Types of mathematical optimisation model:

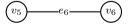
Course Structure

Simple definitions III

- v reachable from u if there is a u v path in G
- connected iff there is a u v path in G for all $u, v \in V(G)$









Combinatorial Optimization

Definitions

Testing connectivity

- $\bullet\,$ decide if G is connected
- obvious if we have a graphical representation
- not so obvious if we only have sets V = V(G), E = E(G)

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incidence matrix	adjacency matrix	adjacency list	e_1 v_1
$A \in \{0,1\}^{ V \times E },$	$A \in \mathbb{Z}^{ V \times V },$	$L = [\ell(v) \colon v \in V], (v \in V]$	$e_2 e_2 e_3 v_3$
$a_{v,e} = \begin{cases} 1, & \text{if } v \in e \\ 0, & \text{if } v \notin e \end{cases}$	$a_{v,w} = \{e = \{v,w\} \in E\} $	· · · · · · · · · · · · · · · · · · ·	
$\begin{pmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 1 & 0 \end{pmatrix}$	$\begin{pmatrix} 0 & 2 & 0 & 0 & 0 & 0 \\ 2 & 0 & 1 & 1 & 0 & 0 \end{pmatrix}$	$\ell(v_1) = [e_1, e_2] \ \ell(v_2) = [e_1, e_2, e_3, e_5]$	v_4
$\begin{bmatrix} 1 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 0 \end{bmatrix}$	$\begin{bmatrix} 2 & 0 & 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 & 0 \end{bmatrix}$	$\ell(v_3) = [e_3, e_4] \\ \ell(v_4) = [e_4, e_5] $	$5 - e_6 - v_6$
$\left(\begin{array}{cccccccccc} 0 & 0 & 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{array}\right)$	$\begin{pmatrix} 0 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{pmatrix}$	$\ell(v_5) = [e_6] \\ \ell(v_6) = [e_6]$	
O(V E)	$O(V ^2)$	$O(E \log V)$	



Combinatorial Optimization

Testing connectivity - An algorithm

Algorithm:	Depth	First	Search
(DFS)			

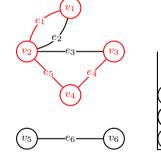
- **Input:** undirected graph G, vertex $s \in V(G)$
- **Output:** tree $(R,T) \subseteq G$, R reachable from s

$$\mathfrak{l} \hspace{0.1 in} \mathsf{set} \hspace{0.1 in} R := \{s\}, \hspace{0.1 in} Q := \{s\} \hspace{0.1 in} \mathsf{and} \hspace{0.1 in} T = \emptyset;$$

- 2 if $Q = \emptyset$ then return R, T;
- 3 else v := last vertex added to <math>Q;
- 4 choose $w \in V(G) \setminus R$ with $\{v,w\} \in E(G);$
- 5 if there is no such w then
- 6 $\ \ \$ set $Q:=Q\setminus\{v\}$ and **go to** 2

7 set
$$R := R \cup \{w\}, Q := Q \cup \{w\},$$

 $T := T \cup \{\{v, w\}\},$ go to 2;





Combinatorial Optimization



 v_3

 v_4

 v_1

 e_4

 e_5

 v_3

 v_4

Correctness



Combinatorial Optimization

- Algorithm: DEPTH FIRST SEARCH (DFS) Input: undirected graph G, vertex $s \in V(G)$
- **Output:** tree $(R,T) \subseteq G$, R reachable from s
- 1 set $R := \{s\}$, $Q := \{s\}$ and $T = \emptyset$;
- 2 if $Q = \emptyset$ then return R, T;
- 3 else v := last vertex added to Q;
- 4 choose $w \in V(G) \setminus R$ with $\{v, w\} \in E(G)$;
- ${\bf 5}~{\bf if}~{\it there}~{\it is}~{\it no}~{\it such}~w$ then
- 6 $\ \$ set $Q := Q \setminus \{v\}$ and **go to** 2

7 set
$$R := R \cup \{w\}, Q := Q \cup \{w\},$$

 $T := T \cup \{\{v, w\}\},$ go to 2;

Idea

- suppose $w \in V(G) \setminus R$ is reachable from s
- $\Rightarrow \ P \text{ is } s w \text{ path with} \\ \{x, y\} \in E(P), \ x \in R, \\ y \in V(G) \setminus R$
- $\Rightarrow x$ is added to Q in line 7
- $\Rightarrow Algorithm does not stop before$ x is removed from Q (line 6)
- $\Rightarrow \text{ there is no } w \in V(G) \setminus R \text{ with } \\ \{v, w\} \in E(G) \text{ {\it f}}$

Running time



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Algorithm: DEPTH FIRST SEARCH (DFS) **Input:** undirected graph G, vertex $s \in V(G)$ **Output:** tree $(R, T) \subseteq G$, R reachable from s **1** set $R := \{s\}, Q := \{s\}$ and $T = \emptyset$; 2 if $Q = \emptyset$ then return R, T: 3 else v := last vertex added to Q; 4 choose $w \in V(G) \setminus R$ with $\{v, w\} \in E(G)$: 5 if there is no such w then 6 set $Q := Q \setminus \{v\}$ and go to 2 7 set $R := R \cup \{w\}, Q := Q \cup \{w\},$ $T := T \cup \{\{v, w\}\}, \text{ go to } 2;$

- for each node the incident edges are considered
- runtime depends on the storage of graph
- if adjacency lists are used, the runtime is O(m) = O(|E(G)|)

5

7

8



Combinatorial Optimization

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Algorithm: BREADTH FIRST SEARCH (BFS)
```

```
Input: undirected graph G, vertex s \in V(G)
  Output: tree T \subset G
1 set Q := \{s\} and T = \{s\}:
2 while Q \neq \emptyset do
      v := first vertex in Q
3
     set Q := Q \setminus \{v\}
4
      while v has a neighbour not in T do
          w := first neightbour of v not in T
6
          set Q := Q \cup \{w\}
          set T := T \cup \{\{v, w\}\}
```



Combinatorial Optimization

