Combinatorial Optimization

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Sheet 1

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General remark:

In order to obtain a bonus for the final grading, you may hand in written solutions to the exercise marked with a star at the beginning of the exercise session on October 4.

Exercise 1

Let $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$ and $c \in \mathbb{R}^n$. Show that the dual of the linear program

$$\max\{c^T x \colon x \in \mathbb{R}^n, \, Ax \le b, \, x \ge 0\}$$

can be interpreted as the linear program

$$\min\{b^T y \colon y \in \mathbb{R}^m, \, A^T y \ge c, \, y \ge 0\}.$$

In addition, mark (and argue!) the entries of the following table corresponding to possible outcomes in this primal/dual pair of linear programs:

		Dual		
		Finite optimum	Unbounded	Infeasible
Primal	Finite optimum			
	Unbounded			
	Infeasible			

Solution

First part: Understand the primal as $\max\{c^T x : x \in \mathbb{R}^n, \binom{A}{-I} x \leq \binom{b}{0}\}$, and re-interpret the dual of this LP. Second part:

- Primal and dual feasible and bounded is *possible*: Example is c = b = (0) and A = (0).
- Primal feasible and bounded, dual unbounded is *impossible*: Assume that $Ax \leq b$ is has a solution x. Then by weak duality, c^Tx a lower bound for all solutions to the dual, in contradiction to the fact that the dual is unbounded.
- Primal feasible and bounded, dual infeasible is *impossible*: If the primal has an optimal solution, the duality theorem tells us that the dual has an optimal solution as well. In particular the dual is feasible.
- Primal unbounded and dual feasible and bounded is *impossible*: Assume that $A^Ty = c$ has a solution y. Then by weak duality, b^Ty is an upper

bound for all solutions to the primal, in contradiction to the fact that the primal is unbounded.

- Primal unbounded, dual unbounded is *impossible*: As seen above, if the primal is unbounded, then the dual is infeasible.
- Primal unbounded, dual infeasible is *possible*: Example is c = (1), b = (0) and A = (0).
- Primal infeasible, dual feasible and bounded is *impossible*: With the strong duality theorem, if the dual is feasible and bounded, so is the primal.
- Primal infeasible, dual unbounded is *possible*: Example is c = (0), b = (-1) and A = (0).
- Primal and dual infeasible is *possible*: Example is c = (1), b = (-1) and A = (0).

Exercise 2

Determine a maximum weight matching of the graph below. Provide of proof of optimality by determining a feasible dual solution to the linear program

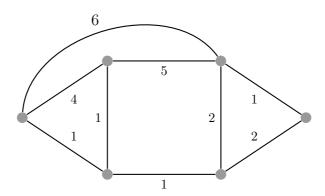
$$\max \sum_{e \in E} w(e)x(e)$$

$$v \in V : \sum_{e \in \delta(v)} x(e) \leq 1$$

$$\bigcup_{|U| \text{ odd}}^{U \subseteq V} : \sum_{e \in E(U)} x(e) \leq \lfloor |U|/2 \rfloor$$

$$e \in E : x(e) \geq 0$$

whose objective value coincides with the weight of your matching.

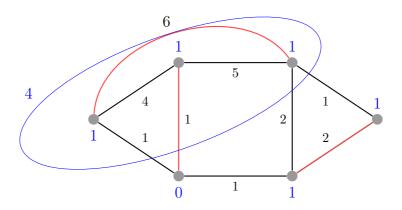


Solution

The dual is

$$\begin{array}{lll} & \min & \sum_{v \in V} y(v) + \sum_{\substack{U \subseteq V \\ |U| \text{ odd}}} \lfloor |U|/2 \rfloor z(U) \\ e \in E : & \sum_{v \in e} y(v) + \sum_{\substack{U \supseteq e \\ |U| \text{ odd}}} z(U) & \geq & w(e) \\ & & z(U) & \geq & 0 \\ & & y(v) & \geq & 0 \end{array}$$

A feasible primal solution picks the red edges and has value 9. A feasible dual solution is denoted in blue and has value 9. Thus both solutions are optimal.



Exercise $3 (\star)$

In the following, we consider undirected graphs without edge weights.

Suppose that we have an efficient algorithm to decide whether a graph contains a perfect matching (i.e. each node is incident with exactly one edge).

Use that algorithm to determine efficiently the size of a maximum matching in a given graph G = (V, E).

Solution

Let \mathcal{A} be the efficient algorithm to decide whether a given graph contains perfect matching. We will present an algorithm for determining the size of a maximum matching in G that uses O(|V|) calls to \mathcal{A} .

MAXIMUM-MATCHING-SIZE(G = (V, E))

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1 if \mathcal{A}(G) = \text{TRUE}

2 return \frac{|V|}{2}

3 \triangleright We assume that V = \{v_1, \dots, v_{|V|}\}.

4 G' = G

5 for step \leftarrow 1 to |V|

6 V' = V' \cup x_{step}

7 E' = E' \cup \{\{v_i, x_{step}\} : v_i \in V\}

8 G' = (V', E')

9 if \mathcal{A}(G') = \text{TRUE}

10 return \frac{|V| - step}{2}
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In order to prove correctness of the algorithm we have to show the following:

- a) The algorithm terminates, and
- b) value returned by the algorithm is the size of a maximum matching of G.
- a) Let us consider the last iteration, i.e. the graph augmented by |V| artificial nodes. By the constructions in the lines 6 and 7, G' contains edges $\{v_i, x_i\}, i = 1..|V|$ and thus a perfect matching.

- b) Assume that the algorithm stops with step = t (The correctness is trivial if the algorithms stops in line 2). We show that the following holds:
- i) Matching of size $\frac{|V|-t}{2}$ exists in G, and ii) matching of the size $\frac{|V|-t}{2}$ is the maximum in G.
- i) In a perfect matching of G', each vertex $x_k, k = 1..t$, must be joined to a vertex $v_{i_k} \in V$, since by the construction, there is no edge between x_i and $x_j, 1 \leq i, j \leq t$. The remaining |V| - t vertices, all in V, are covered by a matching.
- ii) Let us assume that there exists a matching of size $S > \frac{|V|-t}{2}$. Assume wlog that the vertices v_1, \ldots, v_{2S} form that matching. Let t' = |V| 2S. Clearly t' < t. For step = t' all the vertices $v_{2S+1}, \ldots, v_{2S+t'}$ are matched to the vertices $x_1, \ldots, x_{t'}$, thus $v_{2S+1}, \ldots, v_{2S+t'}, x_1, \ldots, x_{t'}$ are perfectly covered by a matching of size t'. It implies that G' for step = t' contains a perfect matching, a contradiction.

Note that it is possible to reduce the number of calls to A in the worst case to $\frac{|V|}{2}$, since you only have to consider the cases in which the total number of nodes is even (otherwise there trivially is no perfect matching).

Exercise 4

Show the following: A face F of $P = \{x \in \mathbb{R}^n : Ax \leq b\}$ is inclusion-wise minimal if and only if it is of the form $F = \{x \in \mathbb{R}^n \mid A'x = b'\}$ for some subsystem $A'x \le b'$ of $Ax \le b$.

Solution

Let F be an inclusion-wise minimal face. Write

$$F = \{x \in P \colon A'x = b'\}$$

where $A'x \leq b'$ is the maximal possible subsystem of $Ax \leq b$ with that property, and let

$$G = \{x \in \mathbb{R}^n \colon A'x = b'\}$$

Assume $F \neq G$, then there is a point $z \in G \backslash F$. In particular, $z \notin P$. Furthermore, there exists a point $y \in F$. Consider the line segment parameterized by:

$$w(t) = (1-t)y + tz, t \in [0,1]$$

Let $a^T x \leq \beta$ be the first inequality of $Ax \leq b$ that is violated as w(t) moves from y to z, and let $t \in [0,1)$ such that $a^T w(t) = \beta$. Then

$$F' = \{ x \in P : A'x = b', a^Tx = \beta \}$$

is a face of P by the theorem seen in lecture 1, it is clearly contained in F, and it is non-empty because $w(t) \in F'$. Finally, note that $a^T x = \beta$ cannot be contained in the system A'x = b', because $a^T w(t) = \beta$ does not hold for all $t \in [0,1]$. Therefore, F' is defined by a subsystem of equations that is strictly bigger than any subsystem that defines F (remember that we chose A'x = b' to be maximal!) and so $F' \neq F$. In conclusion, F' is a proper sub-face of F, which contradicts the inclusion-wise minimality of F. So the assumption was wrong, in fact we have F = G.

Let F be a face of P such that $F = \{x \in \mathbb{R}^n : A'x = b'\}$ for a subsystem $A'x \leq b'$ of $Ax \leq b$. Assume that F is not inclusion-wise minimal, i.e. there is a proper sub-face $F' \subseteq F$. We can write

$$F' = \{ x \in P \colon A'x = b', A''x = b'' \}$$

for a subsystem $A''x \leq b''$ of $Ax \leq b$. Let $y \in F'$ and $z \in F \setminus F'$. Then the line through y and z is contained entirely in F, however there will be one inequality $a^Tx \leq \beta$ of the system $A''x \leq b''$ that is not parallel to the line through y and z. This means that the line cannot be entirely contained in P. This is a contradiction, and so F must be inclusion-wise minimal.