

15.083J/6.859J Integer Optimization

Lecture 5: Ideal formulations I

1 Outline

SLIDE 1

- Total unimodularity
- Dual Methods

2 Total unimodularity

SLIDE 2

- $S = \{\mathbf{x} \in \mathcal{Z}_+^n \mid \mathbf{Ax} \leq \mathbf{b}\}$, $\mathbf{A} \in \mathcal{Z}^{m \times n}$ and $\mathbf{b} \in \mathcal{Z}^m$.
- $P = \{\mathbf{x} \in \mathcal{R}_+^n \mid \mathbf{Ax} \leq \mathbf{b}\}$.
- When $P = \text{conv}(S)$ for all integral vectors \mathbf{b} ?

2.1 Cramer's rule

SLIDE 3

- $\mathbf{A} \in \mathcal{R}^{n \times n}$ nonsingular.
- $\mathbf{Ax} = \mathbf{b} \iff \mathbf{x} = \mathbf{A}^{-1}\mathbf{b} \iff \forall i: x_i = \frac{\det(\mathbf{A}^i)}{\det(\mathbf{A})}$.
- $\mathbf{A}^i: \mathbf{A}_j^i = \mathbf{A}_j$ for all $j \in \{1, \dots, n\} \setminus \{i\}$ and $\mathbf{A}_i^i = \mathbf{b}$.

2.2 Definition

SLIDE 4

- $\mathbf{A} \in \mathcal{Z}^{m \times n}$ of full row rank is **unimodular** if the determinant of each basis of \mathbf{A} is 1, or -1. A matrix $\mathbf{A} \in \mathcal{Z}^{m \times m}$ of full row rank is **unimodular** if $\det(\mathbf{A}) = \pm 1$.
- A matrix $\mathbf{A} \in \mathcal{Z}^{m \times n}$ is **totally unimodular** if the determinant of each square submatrix of \mathbf{A} is 0, 1, or -1.

2.3 Examples

SLIDE 5

- $\mathbf{A} = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1 \end{bmatrix}$ is not TU: $\det \left(\begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix} \right) = -2$.
- $\begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}$ is TU.

2.4 Proposition

SLIDE 6

- \mathbf{A} is TU if and only if $[\mathbf{A}, \mathbf{I}]$ is unimodular.

- \mathbf{A} is TU if and only if $\begin{bmatrix} \mathbf{A} \\ -\mathbf{A} \\ \mathbf{I} \\ -\mathbf{I} \end{bmatrix}$ is TU.

- \mathbf{A} is TU if and only if \mathbf{A}' is TU.

2.5 Theorem

SLIDE 7

- \mathbf{A} integer matrix of full row rank. \mathbf{A} is unimodular if and only if $P(\mathbf{b}) = \{\mathbf{x} \in \mathfrak{R}_+^n \mid \mathbf{A}\mathbf{x} = \mathbf{b}\}$ is integral for all $\mathbf{b} \in \mathcal{Z}^m$ for which $P(\mathbf{b}) \neq \emptyset$.
- \mathbf{A} integer matrix. \mathbf{A} is TU if and only if $P(\mathbf{b}) = \{\mathbf{x} \in \mathfrak{R}_+^n \mid \mathbf{A}\mathbf{x} \leq \mathbf{b}\}$ is integral for all $\mathbf{b} \in \mathcal{Z}^m$ for which $P(\mathbf{b}) \neq \emptyset$.

2.5.1 Proof

SLIDE 8

- Assume that \mathbf{A} is unimodular. $\mathbf{b} \in \mathcal{Z}^m$ and $P(\mathbf{b}) \neq \emptyset$.
- $\mathbf{x} = (\mathbf{x}_B, \mathbf{x}_N)$ extreme point of $P(\mathbf{b})$, $\mathbf{x}_B = \mathbf{A}_B^{-1}\mathbf{b}$ and $\mathbf{x}_N = \mathbf{0}$.
- Since \mathbf{A} unimodular $\det(\mathbf{A}_B) = \pm 1$. By Cramer's rule and the integrality of \mathbf{A}_B and \mathbf{b} , \mathbf{x}_B is integral.
- $P(\mathbf{b})$ is integral.
- Conversely, $P(\mathbf{b})$ integral for all $\mathbf{b} \in \mathcal{Z}^m$.
- $B \subseteq \{1, \dots, n\}$ with \mathbf{A}_B nonsingular.
- $\mathbf{b} = \mathbf{A}_B \mathbf{z} + \mathbf{e}_i$, where \mathbf{z} integral: $\mathbf{z} + \mathbf{A}_B^{-1}\mathbf{e}_i \geq \mathbf{0}$ for all i .
- $\mathbf{A}_B^{-1}\mathbf{b} = \mathbf{z} + \mathbf{A}_B^{-1}\mathbf{e}_i \in \mathcal{Z}^m$ for all i .
- i th column of \mathbf{A}_B^{-1} is integral for all i .
- \mathbf{A}_B^{-1} is an integer matrix, and thus, since \mathbf{A}_B is also an integer matrix, and $\det(\mathbf{A}_B)\det(\mathbf{A}_B^{-1}) = 1$, we obtain that $\det(\mathbf{A}_B) = 1$ or -1 .
- For second part: \mathbf{A} is TU if and only if $[\mathbf{A}, \mathbf{I}]$ is unimodular. For any $\mathbf{b} \in \mathcal{Z}^m$ the extreme points of $\{\mathbf{x} \in \mathfrak{R}_+^n \mid \mathbf{A}\mathbf{x} \leq \mathbf{b}\}$ are integral if and only if the extreme points of $\{(\mathbf{x}, \mathbf{y}) \in \mathfrak{R}_+^{n+m} \mid \mathbf{A}\mathbf{x} + \mathbf{I}\mathbf{y} = \mathbf{b}\}$ are integral.

2.6 Corollary

SLIDE 9

Let \mathbf{A} be an integral matrix.

- \mathbf{A} is TU if and only if $\{\mathbf{x} \mid \mathbf{A}\mathbf{x} = \mathbf{b}, \mathbf{0} \leq \mathbf{x} \leq \mathbf{u}\}$ is integral for all integral vectors \mathbf{b} and \mathbf{u} .
- \mathbf{A} is TU if and only if $\{\mathbf{x} \mid \mathbf{a} \leq \mathbf{A}\mathbf{x} \leq \mathbf{b}, \mathbf{l} \leq \mathbf{x} \leq \mathbf{u}\}$ is integral for all integral vectors $\mathbf{a}, \mathbf{b}, \mathbf{l}, \mathbf{u}$.

2.7 Theorems

SLIDE 10

- \mathbf{A} is TU if and only if each collection J of columns of \mathbf{A} can be partitioned into two parts so that the sum of the columns in one part minus the sum of the columns in the other part is a vector with entries 0, +1, and -1.
- \mathbf{A} is TU if and only if each collection Q of rows of \mathbf{A} can be partitioned into two parts so that the sum of the rows in one part minus the sum of the rows in the other part is a vector with entries only 0, +1, and -1.

2.8 Corollary

SLIDE 11

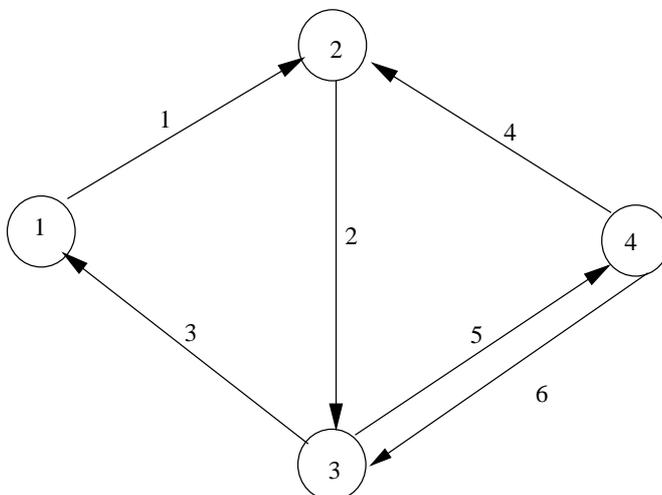
The following matrices are TU:

- The node-arc incidence matrix of a directed graph.
- The node-edge incidence matrix of an undirected bipartite graph.
- A matrix of zero-one elements, in which each column has its ones consecutively.

2.9 Example

SLIDE 12

$$\mathbf{A} = \begin{bmatrix} 1 & 0 & -1 & 0 & 0 & 0 \\ -1 & 1 & 0 & -1 & 0 & 0 \\ 0 & -1 & 1 & 0 & 1 & -1 \\ 0 & 0 & 0 & 1 & -1 & 1 \end{bmatrix}$$



2.10 Implications

SLIDE 13

Following problems can be solved as LOs:

- Network flows
- Matching in bipartite graphs
- Stable set in bipartite graphs.

3 Dual methods

SLIDE 14

-

$$\begin{aligned} Z_{LP} = \min \quad & \mathbf{c}'\mathbf{x} \\ \text{s.t.} \quad & \mathbf{x} \in P \end{aligned}$$

- Let P be a nonempty polyhedron with at least one extreme point. The polyhedron P is integral if and only if Z_{LP} is integer for all $\mathbf{c} \in \mathcal{Z}^n$.
- For converse, assume $\mathbf{x}^* \in P$, extreme point with x_j^* fractional. $\mathbf{c} \in \mathcal{Z}^n$: \mathbf{x}^* unique optimum.
- There exist $a \in \mathcal{Z}$: \mathbf{x}^* optimum for $\bar{\mathbf{c}} = \mathbf{c} + (1/a)\mathbf{e}_j$. $a\bar{\mathbf{c}}'\mathbf{x}^* - a\mathbf{c}'\mathbf{x}^* = x_j^*$, either $a\bar{\mathbf{c}}'\mathbf{x}^*$ or $a\mathbf{c}'\mathbf{x}^*$ is fractional. Contradiction.

3.1 Key idea

SLIDE 15

Construct a solution to the dual of the LP relaxation and an integer solution, feasible to IO with $Z_H = Z_D$. Since $Z_D \leq Z_{LP} \leq Z_{IP} \leq Z_H$, if $Z_H = Z_D$, $Z_{LP} = Z_{IP}$.

3.2 Submodular functions

SLIDE 16

- $f : 2^N \mapsto \mathfrak{R}_+$ is **submodular** if

$$f(S) + f(T) \geq f(S \cap T) + f(S \cup T), \quad \forall S, T \subset N.$$

- $f : 2^N \mapsto \mathfrak{R}_+$ is **supermodular** if

$$f(S) + f(T) \leq f(S \cap T) + f(S \cup T), \quad \forall S, T \subset N.$$

- It is **nondecreasing**, if

$$f(S) \leq f(T), \quad \forall S \subset T.$$

3.3 Polymatroids

SLIDE 17

$$\begin{aligned}
 & \text{maximize} && \sum_{j=1}^n c_j x_j \\
 & \text{subject to} && \sum_{j \in S} x_j \leq f(S), \quad S \subset N, \\
 & && x_j \in \mathbb{Z}_+, \quad j \in N.
 \end{aligned}$$

$$P(f) = \left\{ \mathbf{x} \in \mathbb{R}_+^n \mid \sum_{j \in S} x_j \leq f(S), \forall S \subset N \right\}.$$

3.3.1 Theorem

SLIDE 18

If the function f is submodular, nondecreasing, integer valued, and $f(\emptyset) = 0$, then $P(f) = \text{conv}(F)$, F set of feasible integer solutions.

3.4 Proof

SLIDE 19

$$\begin{aligned}
 & \text{maximize} && \sum_{j=1}^n c_j x_j \\
 & \text{subject to} && \sum_{j \in S} x_j \leq f(S), \quad S \subset N. \\
 & && x_j \geq 0, \quad j \in N,
 \end{aligned}$$

dual

$$\begin{aligned}
 & \text{minimize} && \sum_{S \subset N} f(S) y_S \\
 & \text{subject to} && \sum_{\{S \mid j \in S\}} y_S \geq c_j, \quad j \in N, \\
 & && y_S \geq 0, \quad S \subset N.
 \end{aligned}$$

SLIDE 20

- $c_1 \geq c_2 \geq \dots \geq c_k > 0 \geq c_{k+1} \geq \dots \geq c_n$. $S^j = \{1, \dots, j\}$ for $j \in N$, and $S^0 = \emptyset$.

•

$$x_j = \begin{cases} f(S^j) - f(S^{j-1}), & \text{for } 1 \leq j \leq k, \\ 0, & \text{for } j > k. \end{cases}$$

$$y_S = \begin{cases} c_j - c_{j+1}, & \text{for } S = S^j, 1 \leq j < k, \\ c_k, & \text{for } S = S^k, \\ 0, & \text{otherwise.} \end{cases}$$

- \mathbf{x} is integer, $x_j \geq 0$

•

$$\begin{aligned}
\sum_{j \in T} x_j &= \sum_{\{j \mid j \in T, j \leq k\}} (f(S^j) - f(S^{j-1})) \\
&\leq \sum_{\{j \mid j \in T, j \leq k\}} (f(S^j \cap T) - f(S^{j-1} \cap T)) \\
&= f(S^k \cap T) - f(\emptyset) \\
&\leq f(T) - f(\emptyset) \\
&= f(T).
\end{aligned}$$

- \mathbf{y} is dual feasible because $y_S \geq 0$ and

$$\sum_{\{S \mid j \in S\}} y_S = y_{S^1} + \dots + y_{S^k} = c_j, \text{ if } j \leq k,$$

$$\sum_{\{S \mid j \in S\}} y_S = 0 \geq c_j, \text{ if } j > k.$$

- Primal objective value: $\sum_{j=1}^k c_j (f(S^j) - f(S^{j-1}))$
- Dual objective value:

$$\sum_{j=1}^{k-1} (c_j - c_{j+1}) f(S^j) + c_k f(S^k) = \sum_{j=1}^k c_j (f(S^j) - f(S^{j-1})).$$

3.5 Matroids

SLIDE 21

- (N, \mathcal{I}) independence system, $r(T) = \max\{|S| : S \in \mathcal{I}, S \subset T\}$.

$$\text{maximize } \sum_{j \in N} c_j x_j$$

- subject to $\sum_{j \in S} x_j \leq r(S), \quad \forall S \subset N,$
 $x_j \in \{0, 1\}.$

- Theorem: (N, \mathcal{I}) independence system. It is a matroid if and only if its rank function $r(S) = \max\{|S| : S \in \mathcal{I}, S \subset T\}$ is submodular.

3.6 Greedy algorithm

SLIDE 22

1. Given a matroid (N, \mathcal{I}) , and weights c_j for $j \in N$, sort all elements of N in decreasing order of c_j : $c_{j_1} \geq c_{j_2} \geq \dots \geq c_{j_n}$. Let $J = \emptyset$; $k = 1$.
2. For $k = 1, \dots, m$, if $J \cup \{j_k\}$ is an independent set, let $J = J \cup \{j_k\}$;
3. An optimum solution is given by the set J .

MIT OpenCourseWare
<http://ocw.mit.edu>

15.083J / 6.859J Integer Programming and Combinatorial Optimization
Fall 2009

For information about citing these materials or our Terms of Use, visit: <http://ocw.mit.edu/terms>.