

MA427 – Mathematical Optimisation

Lecture 1

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THE LONDON SCHOOL
OF ECONOMICS AND
POLITICAL SCIENCE ■

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General information

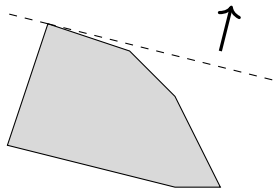
- ▶ Instructor: Giacomo Zambelli, g.zambelli@lse.ac.uk
- ▶ Office hours: Mondays 14:30-16:30
- ▶ Assessment: 90% written exam, 3 hours in ST. 10% course work (2 exercise sets)
- ▶ Formative work: 1 exercise set will be collected and feedback will be given.

Course overview

- ▶ **Linear programming (4 lectures)**
- ▶ **Linear mixed-integer programming (3 lectures)**
Totally unimodular matrices, ideal formulations
branch-and-bound, cutting planes.
- ▶ **Convex optimization (3 lectures)**
Lagrangian duality, Karush-Kuhn-Tucker conditions, gradient descent.

Linear Programming (LP)

$$\begin{array}{ll}\max & 2x_1 + 8x_2 \\ \text{s.t.} & 2x_1 + x_2 \leq 10 \\ & x_1 + 2x_2 \leq 10 \\ & x_1 + x_2 \leq 6 \\ & x_1 + 3x_2 \leq 12 \\ & 3x_1 - x_2 \geq 0 \\ & x_1 + 4x_2 \geq 4 \\ & x_1, x_2 \geq 0\end{array}$$



Linear programming: MA423 and MA427

MA423 lectures 1-3

- ▶ Basic concepts, standard forms of LP
- ▶ Duality and optimality conditions
- ▶ Simplex Method (dictionary form)

MA427 lectures 1-4

- ▶ *Fourier-Motzkin elimination, Farkas' lemma, and the duality theorem*
- ▶ Geometry of linear programs
- ▶ Simplex Method: tableau form, two phase and dual simplex

Reminder: systems of linear equations

$$Ax = b$$

$$A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m.$$

Gaussian elimination

- ▶ Efficient method to solve a system of linear equations.
- ▶ Elementary operations:
 - ▶ Multiply a row by a nonzero real number.
 - ▶ Add a multiple of one row to another one.
 - ▶ Swap two rows.

Proposition

The system $Ax = b$ is feasible, if and only if $\text{rk}(A) = \text{rk}(A|b)$.

Fourier-Motzkin elimination



Joseph Fourier (1768-1830)



Theodore Motzkin (1908-1970)

Linear feasibility problem: system of linear inequalities

$$Ax \leq b$$

$$A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m.$$

- ▶ **Goal:** try to mimic Gaussian elimination for systems of linear inequalities.
- ▶ Elementary operations that preserve feasibility:
 - ▶ Multiply a row by a **positive** real number.
 - ▶ Add a **positive** multiple of one row to another one.
 - ▶ Swap two rows.

Example

Are the following two systems feasible?

$$\begin{array}{rcl} 2x_1 + x_2 & \leq & 10 \\ -4x_1 + 2x_2 & \leq & -30 \\ x_1 - x_2 & \leq & 8 \\ -x_2 & \leq & 4 \end{array}$$

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Fourier-Motzkin elimination

$$Ax \leq b$$

$A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$, $x = (x_1, x_2, \dots, x_{n-1}, x_n)$.

- ▶ Reduce to a system of $n - 1$ variables, by eliminating x_n .
- ▶ Let $I = \{1, \dots, m\}$ and

$$I^+ = \{i \in I : a_{in} > 0\}, \quad I^- = \{i \in I : a_{in} < 0\}, \\ I^0 = \{i \in I : a_{in} = 0\}.$$

- ▶ Divide the i th row by $|a_{in}|$ for each $i \in I^+ \cup I^-$ to get $a'_{in} = \pm 1$ or 0 everywhere.

$$\begin{aligned} \sum_{j=1}^{n-1} a'_{ij} x_j + x_n &\leq b'_i, & i \in I^+ \\ \sum_{j=1}^{n-1} a'_{ij} x_j - x_n &\leq b'_i, & i \in I^- \\ \sum_{j=1}^{n-1} a'_{ij} x_j &\leq b_i, & i \in I^0 \end{aligned}$$

Fourier-Motzkin elimination

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Eliminate x_n by:

- ▶ For each pair $i \in I^+$ and $k \in I^-$, sum the two inequalities indexed by i and k .
- ▶ Remove all inequalities in $I^+ \cup I^-$.

New system:

$$\begin{aligned}\sum_{j=1}^{n-1} (a'_{ij} + a'_{kj}) x_j &\leq b'_i + b'_k, & i \in I^+, k \in I^-, \\ \sum_{j=1}^{n-1} a_{ij} x_j &\leq b_i, & i \in I^0.\end{aligned}$$

Fourier-Motzkin elimination

Original system:

$$\begin{aligned}\sum_{j=1}^{n-1} a'_{ij} x_j + x_n &\leq b'_i, & i \in I^+ \\ \sum_{j=1}^{n-1} a'_{ij} x_j - x_n &\leq b'_i, & i \in I^- \\ \sum_{j=1}^{n-1} a_{ij} x_j &\leq b_i, & i \in I^0\end{aligned}$$

New system:

$$\begin{aligned}\sum_{j=1}^{n-1} (a'_{ij} + a'_{kj}) x_j &\leq b'_i + b'_k, & i \in I^+, k \in I^-, \\ \sum_{j=1}^{n-1} a_{ij} x_j &\leq b_i, & i \in I^0.\end{aligned}$$

Theorem

A vector $(\bar{x}_1, \dots, \bar{x}_{n-1})$ satisfies the system new system *if and only if* there exists \bar{x}_n such that $(\bar{x}_1, \dots, \bar{x}_{n-1}, \bar{x}_n)$ satisfies the original system.

Fourier-Motzkin elimination

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Proof: Let

$$\ell = \max_{k \in I^-} \left\{ \sum_{j=1}^{n-1} a'_{kj} \bar{x}_j - b'_k \right\}, \quad u = \min_{i \in I^+} \left\{ b'_i - \sum_{j=1}^{n-1} a'_{ij} \bar{x}_j \right\}$$

Selecting any \bar{x}_n such that $\ell \leq \bar{x}_n \leq u$ gives a feasible solution to the original system.

Fourier-Motzkin elimination

*Given a system of linear inequalities $Ax \leq b$, let $A^n := A$, $b^n := b$;
For $i = n, \dots, 1$, eliminate variable x_i from $A^i x \leq b^i$ to obtain
system $A^{i-1} x \leq b^{i-1}$.*

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system $A^{i-1} x \leq b^{i-1}$.*

- ▶ Final system $A^0 x \leq b^0$ has inequalities $0 \leq b_i^0$. It is feasible if and only if all b_i^0 's are nonnegative.
- ▶ Given a solution to $A^i x \leq b^i$, we can obtain a solution to $A^{i+1} x \leq b^{i+1}$ for $i = 0, 1, 2, \dots, n-1$.

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- ▶ Given a solution to $A^i x \leq b^i$, we can obtain a solution to $A^{i+1} x \leq b^{i+1}$ for $i = 0, 1, 2, \dots, n-1$.
- ▶ Number of iterations can be large: $|I^+| + |I^-|$ inequalities are replaced by $|I^+| \cdot |I^-|$ new ones: number of inequalities may grow exponentially.

Fourier-Motzkin elimination

$$\begin{array}{rcccccl} - & x_1 & & & & \leq & -1 \\ & & - & x_2 & & \leq & -1 \\ & & & & - & x_3 & \leq & -1 \\ - & x_1 & - & x_2 & & \leq & -3 \\ - & x_1 & & & - & x_3 & \leq & -3 \\ & & - & x_2 & - & x_3 & \leq & -3 \\ & x_1 & + & x_2 & + & x_3 & \leq & 6 \end{array}$$

Quiz

We have a system of linear inequalities with 20 inequalities having $+x_n$, and 2 inequalities not containing x_n . (That is, $|I^+| = 20$, $|I^-| = 0$, $|I^0| = 2$.) How many inequalities do we get after eliminating x_n ?

- (A) 2
- (B) 22
- (C) 40

Characterising feasibility: Farkas' lemma



Gyula Farkas (1847-1930)

Expressing inequalities in Fourier-Motzkin

Proposition

Every inequality added during Fourier-Motzkin elimination is a *nonnegative combination* of the inequalities in the original system $Ax \leq b$. That is, for every inequality $c^\top x \leq d$ in the system $A^i x \leq b^i$, we can find a nonnegative vector $u \in \mathbb{R}^m$, $u \geq 0$, such that $c^\top = u^\top A$, and $d = u^\top b$.

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Theorem (Farkas' lemma)

Exactly one of the following two systems has a feasible solution:

- ▶ $Ax \leq b$
- ▶ $u^\top A = 0, u^\top b < 0, u \geq 0$

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Proof:

1. *Both systems cannot be simultaneously feasible:*

$$0 = (u^\top A)x = u^\top (Ax) \leq u^\top b < 0$$

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Proof:

1. *Both systems cannot be simultaneously feasible:*

$$0 = (u^T A)x = u^T (Ax) \leq u^T b < 0$$

2. *If $Ax \leq b$ is infeasible, then $u^T A = 0, u^T b < 0, u \geq 0$ is feasible:*

FM-elimination gives infeasible $A^0 x \leq b^0$, including an inequality $0 \leq b_i^0$, for $b_i^0 < 0$.

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FM-elimination gives infeasible $A^0 x \leq b^0$, including an inequality $0 \leq b_i^0$, for $b_i^0 < 0$.

$$\exists u \geq 0 \quad u^T A = 0, u^T b = b_i^0 < 0.$$

Farkas' lemma

Theorem (Farkas' lemma, standard equality form)

Exactly one of the following two systems has a feasible solution:

- ▶ $Ax = b, x \geq 0$
- ▶ $u^T A \leq 0, u^T b > 0$

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Proof: Rewrite $Ax = b, x \geq 0$ as

$$\begin{pmatrix} A \\ -A \\ -I \end{pmatrix} x \leq \begin{pmatrix} b \\ -b \\ 0 \end{pmatrix}$$

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Theorem (Farkas' lemma, standard equality form)

Exactly one of the following two systems has a feasible solution:

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- ▶ $u^\top A \leq 0, u^\top b > 0$

Proof: Rewrite $Ax = b, x \geq 0$ as

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Previous form gives $(v, v', w) \geq 0$ such that
 $v^\top A - v'^\top A - Iw = 0, v^\top b - v'^\top b - w0 < 0$. Set $u = v' - v$.

Characterising optimality: Duality theory



John von Neumann (1903-1957)



George Dantzig (1914-2005)

LP duality

Theorem (Strong Duality Theorem)

Given a matrix $A \in \mathbb{R}^{m \times n}$ and vectors $c \in \mathbb{R}^n$, $b \in \mathbb{R}^m$, let

$$P := \{x : Ax \leq b\}, \quad D := \{u : u^\top A = c, u \geq 0\}.$$

If P and D are both nonempty, then

$$\max\{c^\top x : Ax \leq b\} = \min\{u^\top b : u^\top A = c, u \geq 0\},$$

and there exist $x^ \in P$ and $y^* \in D$ such that $c^\top x^* = u^{*\top} b$.*

Proof of the Strong Duality Theorem

- Direction $\max \leq \min$

$$c^\top x = (u^\top A)x = u^\top (Ax) \leq u^\top b.$$

- Direction $\min \leq \max$: via Fourier-Motzkin elimination.

Proof of the Strong Duality Theorem

- Consider the feasibility problem

$$\begin{aligned} z - c^{\top} x &\leq 0 \\ Ax &\leq b \end{aligned}$$

Proof of the Strong Duality Theorem

- ▶ Consider the feasibility problem

$$z - c^T x \leq 0$$

$$Ax \leq b$$

- ▶ Apply Fourier-Motzkin elimination to all x_i variables, but keep z .

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- ▶ Consider the feasibility problem

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- ▶ Apply Fourier-Motzkin elimination to all x_i variables, but keep z .
- ▶ The resulting system can be reduced to a single inequality

$$z \leq \beta$$

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- ▶ There exists a solution (z, \bar{x}) to the original system with $z = \beta$; we get $\max\{c^T x : Ax \leq b\} = \beta$.

Proof of the Strong Duality Theorem

- ▶ Consider the feasibility problem

$$\begin{aligned} z - c^\top x &\leq 0 \\ Ax &\leq b \end{aligned}$$

- ▶ Apply Fourier-Motzkin elimination to all x_i variables, but keep z .
- ▶ The resulting system can be reduced to a single inequality

$$z \leq \beta$$

- ▶ There exists a solution (z, \bar{x}) to the original system with $z = \beta$; we get $\max\{c^\top x : Ax \leq b\} = \beta$.
- ▶ We can express $z \leq \beta$ as a nonnegative combination (u_0, u^*) of the original system. It follows that $u^{*\top} A = c$ and $u^{*\top} b = \beta$.

Complementary slackness

Theorem

Given a matrix $A \in \mathbb{R}^{m \times n}$ and vectors $c \in \mathbb{R}^n$, $b \in \mathbb{R}^m$, let

$$P := \{x : Ax \leq b\}, \quad D := \{u : u^\top A = c, u \geq 0\}.$$

Given $x^* \in P$ and $u^* \in D$, x^* and u^* are optimal solutions for the primal and dual problem $\max\{c^\top x : x \in P\}$ and $\min\{u^\top b : u \in D\}$, respectively, if and only if the following *complementary slackness conditions* hold

$$u_i^* (a_i^\top x^* - b_i) = 0 \text{ for } i = 1, \dots, m.$$

Proof.

$$c^\top x^* = (u^{*\top} A)x^* = u^{*\top} (Ax^*) \leq u^{*\top} b.$$

Unbounded objectives

Theorem

Given a matrix $A \in \mathbb{R}^{m \times n}$ and vectors $c \in \mathbb{R}^n$, $b \in \mathbb{R}^m$, let

$$P := \{x : Ax \leq b\}, \quad D := \{u : u^\top A = c, u \geq 0\}.$$

Assume that $P \neq \emptyset$. Then the primal program $\max\{c^\top x : x \in P\}$ is *unbounded if and only if* $D = \emptyset$, which is equivalent to the existence of a vector \bar{y} with $A\bar{y} \leq 0$, $c^\top \bar{y} > 0$.

Proof.

► *Farkas' lemma:* $D = \emptyset \Leftrightarrow \exists \bar{y} : A\bar{y} \leq 0, c^\top \bar{y} > 0$.

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Proof.

- ▶ *Farkas' lemma:* $D = \emptyset \Leftrightarrow \exists \bar{y} : A\bar{y} \leq 0, c^\top \bar{y} > 0$.
- ▶ $D \neq \emptyset$: Strong duality gives an upper bound on $\max\{c^\top x : x \in P\}$.

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Proof.

- ▶ *Farkas' lemma:* $D = \emptyset \Leftrightarrow \exists \bar{y} : A\bar{y} \leq 0, c^\top \bar{y} > 0$.
- ▶ $D \neq \emptyset$: Strong duality gives an upper bound on $\max\{c^\top x : x \in P\}$.
- ▶ $D = \emptyset$: For any $\bar{x} \in P$, $\lambda > 0$,

$$\bar{x} + \lambda \bar{y} \in P, \quad \lim_{\lambda \rightarrow \infty} c^\top (\bar{x} + \lambda \bar{y}) = \infty.$$