

Planning and Optimization

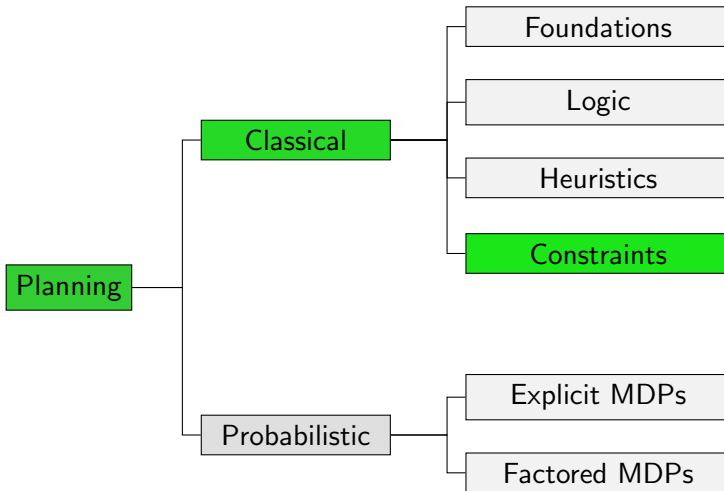
E4. Linear & Integer Programming

Malte Helmert and Thomas Keller

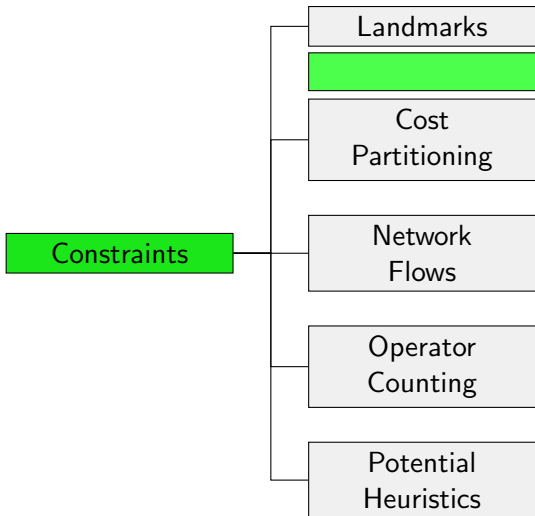
Universität Basel

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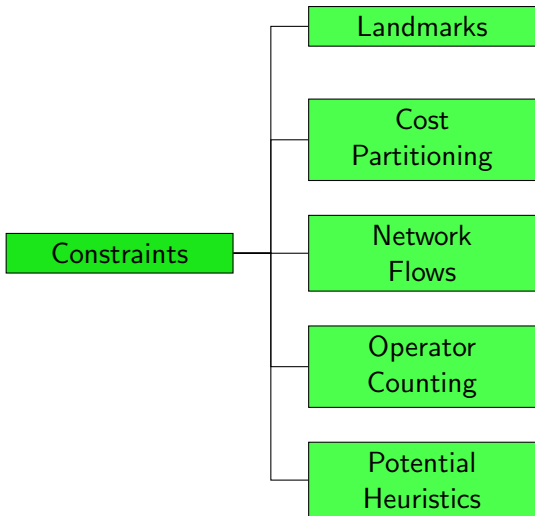
Content of this Course



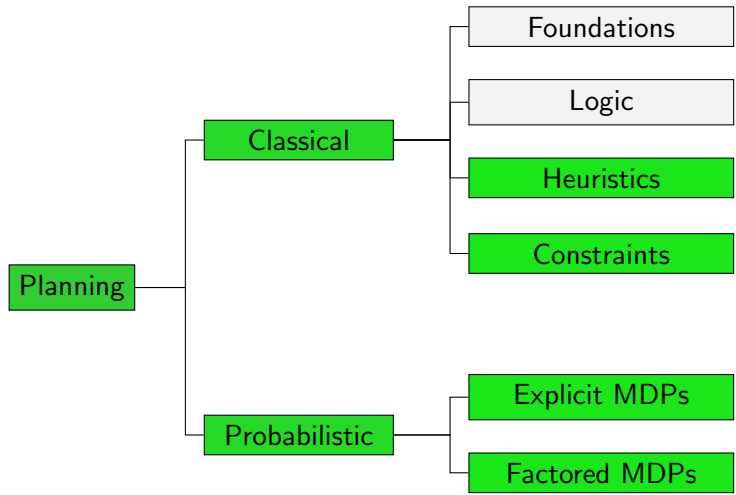
Content of this Course: Constraints (Timeline)



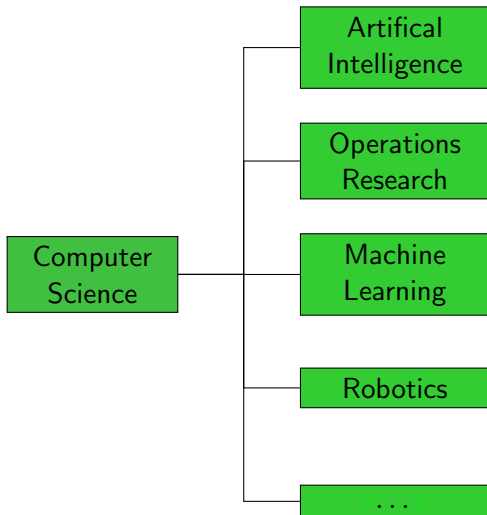
Content of this Course: Constraints (Relevance)



Content of this Course (Relevance)



Content of this Course (Relevance)



Integer Programs

Motivation

- This goes on beyond Computer Science
- Active **research** on IPs and LPs in
 - Operation Research
 - Mathematics
- Many **application** areas, for instance:
 - Manufacturing
 - Agriculture
 - Mining
 - Logistics
 - **Planning**
- As an application, we treat LPs / IPs as a **blackbox**
- We just look at **the fundamentals**

Motivation

Example (Optimization Problem)

Consider the following scenario:

- A factory produces two products A and B
- Selling one (unit of) B yields 5 times the profit of selling one A
- A client places the unusual order to “buy anything that can be produced on that day as long as two plus twice the units of A is not smaller than the number of B”
- More than 12 products in total cannot be produced per day
- There is only material for 6 units of A
(there is enough material to produce any amount of B)

How many units of A and B does the client receive if the factory owner aims to maximize her profit?

Integer Program: Example

Let X_A and X_B be the (**integer**) number of produced A and B

Example (Optimization Problem as Integer Program)

$$X_A \geq 0, \quad X_B \geq 0$$

Example (Optimization Problem)

Integer Program: Example

Let X_A and X_B be the (**integer**) number of produced A and B

Example (Optimization Problem as Integer Program)

maximize $X_A + 5X_B$ subject to

$$X_A \geq 0, \quad X_B \geq 0$$

Example (Optimization Problem)

- “one B yields 5 times the profit of one A”
- “the factory owner aims to maximize her profit”

Integer Program: Example

Let X_A and X_B be the (**integer**) number of produced A and B

Example (Optimization Problem as Integer Program)

maximize $X_A + 5X_B$ subject to

$$2 + 2X_A \geq X_B$$

$$X_A \geq 0, \quad X_B \geq 0$$

Example (Optimization Problem)

- “two plus twice the units of A may not be smaller than the number of B”

Integer Program: Example

Let X_A and X_B be the (**integer**) number of produced A and B

Example (Optimization Problem as Integer Program)

maximize $X_A + 5X_B$ subject to

$$2 + 2X_A \geq X_B$$

$$X_A + X_B \leq 12$$

$$X_A \geq 0, \quad X_B \geq 0$$

Example (Optimization Problem)

- “More than 12 products in total cannot be produced per day”

Integer Program: Example

Let X_A and X_B be the (**integer**) number of produced A and B

Example (Optimization Problem as Integer Program)

maximize $X_A + 5X_B$ subject to

$$2 + 2X_A \geq X_B$$

$$X_A + X_B \leq 12$$

$$X_A \leq 6$$

$$X_A \geq 0, \quad X_B \geq 0$$

Example (Optimization Problem)

- “There is only material for 6 units of A”

Integer Program: Example

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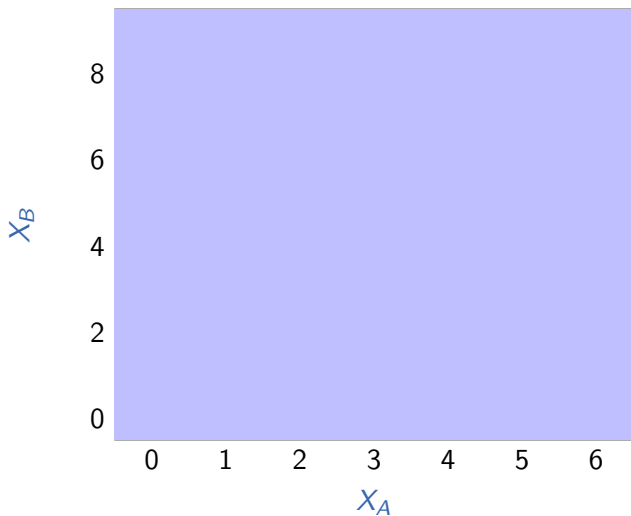
$$X_A \leq 6$$

$$X_A \geq 0, \quad X_B \geq 0$$

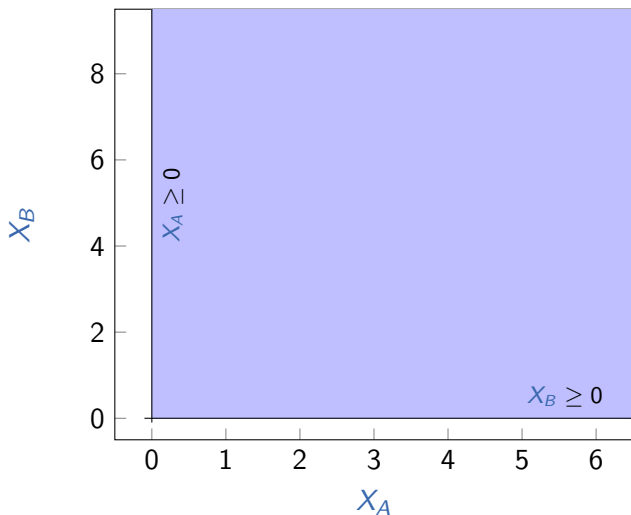
↪ unique optimal solution:

produce 4 A ($X_A = 4$) and 8 B ($X_B = 8$) for a profit of 44

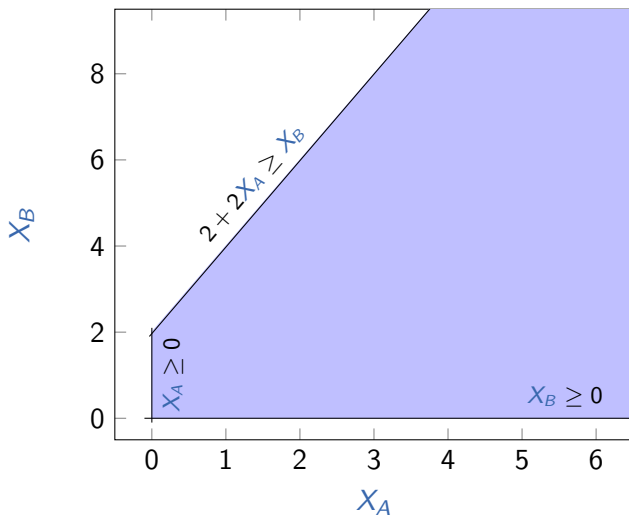
Integer Program Example: Visualization



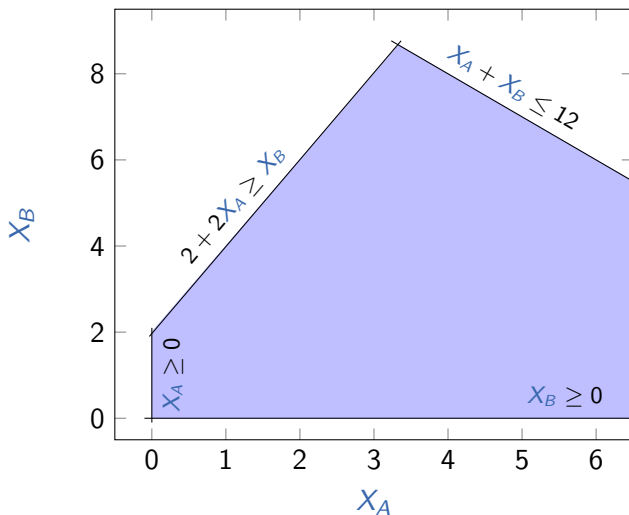
Integer Program Example: Visualization



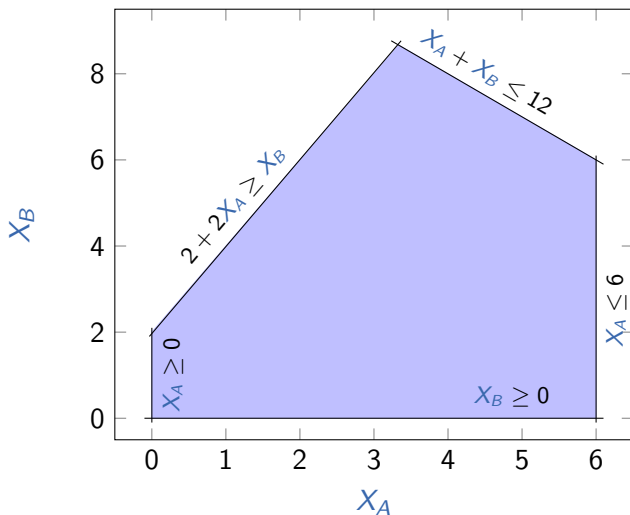
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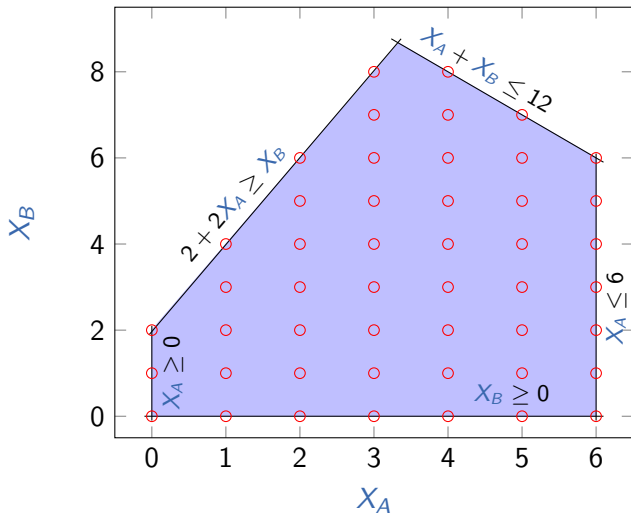
Integer Program Example: Visualization



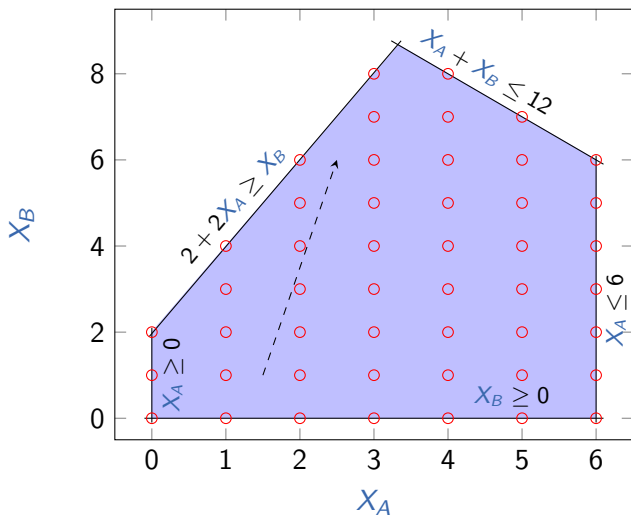
Integer Program Example: Visualization



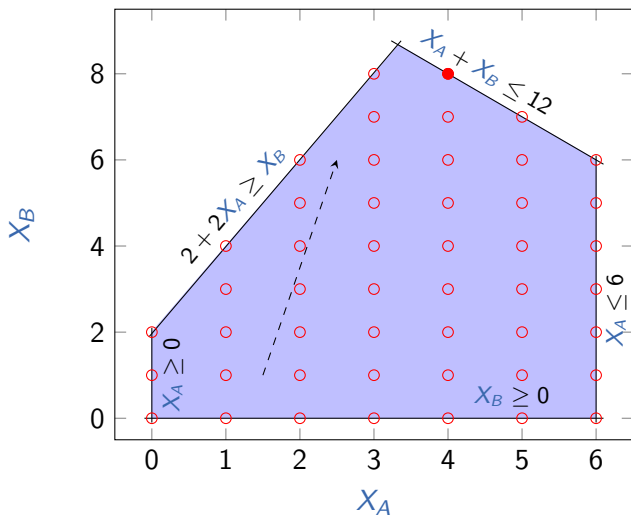
Integer Program Example: Visualization



Integer Program Example: Visualization



Integer Program Example: Visualization



Integer Programs

Integer Program

An **integer program (IP)** consists of:

- a finite set of **integer-valued variables** V
- a finite set of **linear inequalities** (constraints) over V
- an **objective function**, which is a linear combination of V
- which should be **minimized** or **maximized**.

Terminology

- An integer assignment to all variables in V is **feasible** if it satisfies the constraints.
- An integer program is **feasible** if there is such a feasible assignment. Otherwise it is **infeasible**.
- A feasible maximum (resp. minimum) problem is **unbounded** if the objective function can assume arbitrarily large positive (resp. negative) values at feasible assignments. Otherwise it is **bounded**.
- The **objective value** of a bounded feasible maximum (resp. minimum) problem is the maximum (resp. minimum) value of the objective function with a feasible assignment.

Another Example

Example

minimize $3X_{o_1} + 4X_{o_2} + 5X_{o_3}$ subject to

$$X_{o_4} \geq 1$$

$$X_{o_1} + X_{o_2} \geq 1$$

$$X_{o_1} + X_{o_3} \geq 1$$

$$X_{o_2} + X_{o_3} \geq 1$$

$$X_{o_1} \geq 0, \quad X_{o_2} \geq 0, \quad X_{o_3} \geq 0, \quad X_{o_4} \geq 0$$

What example from a previous chapter does this IP encode?

Another Example

Example

minimize $3X_{o_1} + 4X_{o_2} + 5X_{o_3}$ subject to

$$X_{o_4} \geq 1$$

$$X_{o_1} + X_{o_2} \geq 1$$

$$X_{o_1} + X_{o_3} \geq 1$$

$$X_{o_2} + X_{o_3} \geq 1$$

$$X_{o_1} \geq 0, \quad X_{o_2} \geq 0, \quad X_{o_3} \geq 0, \quad X_{o_4} \geq 0$$

What example from a previous chapter does this IP encode?

↪ the minimum hitting set from Chapter E2

Complexity of solving Integer Programs

- As an IP can compute an MHS, solving an IP must be **at least as complex** as computing an MHS
 - Reminder: **MHS** is a “classical” NP-complete problem
 - Good news: Solving an IP is **not harder**
- ↪ Finding solutions for IPs is **NP-complete**.

Complexity of solving Integer Programs

- As an IP can compute an MHS, solving an IP must be **at least as complex** as computing an MHS
- Reminder: **MHS is a “classical” NP-complete problem**
- Good news: Solving an IP is **not harder**

↪ Finding solutions for IPs is **NP-complete**.

Removing the requirement that solutions must be **integer-valued** leads to a simpler problem

Linear Programs

Linear Programs

Linear Program

A **linear program** (LP) consists of:

- a finite set of **real-valued variables** V
- a finite set of **linear inequalities** (constraints) over V
- an **objective function**, which is a linear combination of V
- which should be **minimized** or **maximized**.

We use the introduced IP terminology also for LPs.

Mixed IPs (MIPs) are something between IPs and LPs: some variables are integer-value, some are real-valued.

Linear Program: Example

Let X_A and X_B be the (**real-valued**) number of produced A and B

Example (Optimization Problem as **Linear Program**)

maximize $X_A + 5X_B$ subject to

$$2 + 2X_A \geq X_B$$

$$X_A + X_B \leq 12$$

$$X_A \leq 6$$

$$X_A \geq 0, \quad X_B \geq 0$$

Linear Program: Example

Let X_A and X_B be the (**real-valued**) number of produced A and B

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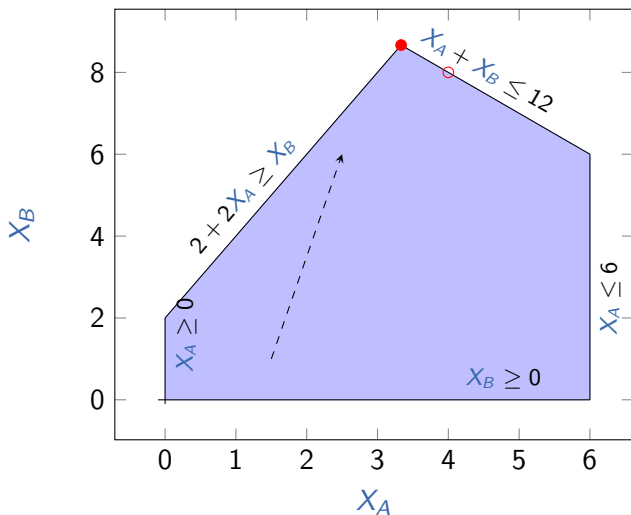
$$X_A \leq 6$$

$$X_A \geq 0, \quad X_B \geq 0$$

↪ unique optimal solution:

$$X_A = 3\frac{1}{3} \text{ and } X_B = 8\frac{2}{3} \text{ with objective value } 46\frac{2}{3}$$

Linear Program Example: Visualization



Solving Linear Programs

- Observation:

Here, LP solution is an **upper bound** for the corresponding IP.

Solving Linear Programs

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Here, LP solution is an **upper bound** for the corresponding IP.
- **Complexity:**
LP solving is a **polynomial-time** problem.

Solving Linear Programs

- **Observation:**
Here, LP solution is an **upper bound** for the corresponding IP.
- **Complexity:**
LP solving is a **polynomial-time** problem.
- **Common idea:**
Approximate IP solution with corresponding LP
(**LP relaxation**).

LP Relaxation

Theorem (LP Relaxation)

The *LP relaxation* of an integer program is the problem that arises by removing the requirement that variables are integer-valued.

For a *maximization* (resp. minimization) problem, the objective value of the LP relaxation is an *upper* (resp. lower) *bound* on the value of the IP.

Proof idea.

Every feasible assignment for the IP is also feasible for the LP.

LP Relaxation of MHS heuristic

Example (Minimum Hitting Set)

minimize $3X_{o_1} + 4X_{o_2} + 5X_{o_3}$ subject to

$$X_{o_4} \geq 1$$

$$X_{o_1} + X_{o_2} \geq 1$$

$$X_{o_1} + X_{o_3} \geq 1$$

$$X_{o_2} + X_{o_3} \geq 1$$

$$X_{o_1} \geq 0, \quad X_{o_2} \geq 0, \quad X_{o_3} \geq 0, \quad X_{o_4} \geq 0$$

↪ optimal solution of LP relaxation:

$X_{o_4} = 1$ and $X_{o_1} = X_{o_2} = X_{o_3} = 0.5$ with objective value 6

↪ LP relaxation of MHS heuristic is **admissible**
and can be computed **polynomial time**

Normal Form

Standard Maximum Problem

Normal form for maximization problems:

Definition (Standard Maximum Problem)

Find values for x_1, \dots, x_n , to maximize

$$c_1x_1 + c_2x_2 + \cdots + c_nx_n$$

subject to the constraints

$$a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n \leq b_1$$

$$a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n \leq b_2$$

$$\vdots$$

$$a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \leq b_m$$

and $x_1 \geq 0, x_2 \geq 0, \dots, x_n \geq 0$.

Standard Maximum Problem: Matrix and Vectors

A standard maximum problem is often given by

- an m -vector $\mathbf{b} = \langle b_1, \dots, b_m \rangle^T$ (**bounds**),
- an n -vector $\mathbf{c} = \langle c_1, \dots, c_n \rangle^T$ (**objective coefficients**),
- and an $m \times n$ matrix

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} \text{ (coefficients)}$$

- Then the problem is to find a vector $\mathbf{x} = \langle x_1, \dots, x_n \rangle^T$ to maximize $\mathbf{c}^T \mathbf{x}$ subject to $\mathbf{A}\mathbf{x} \leq \mathbf{b}$ and $\mathbf{x} \geq \mathbf{0}$.

Standard Problems are a Normal Form

All linear programs can be converted into a standard maximum problem:

- To transform a **minimum problem** into a maximum problem, multiply the objective function by -1 .
- Replace **equality constraints** $a_{i1}x_1 + \cdots + a_{in}x_n = b_i$ with $a_{i1}x_1 + \cdots + a_{in}x_n \geq b_i$ and $a_{i1}x_1 + \cdots + a_{in}x_n \leq b_i$.
- Multiply constraints $a_{i1}x_1 + \cdots + a_{in}x_n \geq b_i$ with -1 (careful, this leads to $(-a_{i1})x_1 + \cdots + (-a_{in})x_n \leq -b_i$)
- If a variable x can be **negative**, introduce variables $x' \geq 0$ and $x'' \geq 0$ and replace x everywhere with $x' - x''$.

Standard Maximum Problem: Example

Example (Optimization Problem in Normal Form)

maximize $X_A + 5X_B$ subject to

$$2 + 2X_A \geq X_B$$

$$X_A + X_B \leq 12$$

$$X_A \leq 6$$

$$X_A \geq 0, \quad X_B \geq 0$$

Standard Maximum Problem: Example

Example (Optimization Problem in Normal Form)

maximize $X_A + 5X_B$ subject to

$$-2 - 2X_A \leq -X_B$$

$$X_A + X_B \leq 12$$

$$X_A \leq 6$$

$$X_A \geq 0, \quad X_B \geq 0$$

Standard Maximum Problem: Example

Example (Optimization Problem in Normal Form)

maximize $X_A + 5X_B$ subject to

$$-2X_A + X_B \leq 2$$

$$X_A + X_B \leq 12$$

$$X_A \leq 6$$

$$X_A \geq 0, \quad X_B \geq 0$$

Standard Maximum Problem: Example

Example (Optimization Problem in Normal Form)

maximize $1X_A + 5X_B$ subject to

$$-2X_A + 1X_B \leq 2$$

$$1X_A + 1X_B \leq 12$$

$$1X_A + 0X_B \leq 6$$

$$X_A \geq 0, \quad X_B \geq 0$$

Standard Minimum Problem

- there is also a **standard minimum problem**
- it's form is identical to the standard maximum problem, except that
 - the aim is to minimize the objective function
 - subject to $\mathbf{Ax} \geq \mathbf{b}$

Duality

Some LP Theory: Duality

Some LP theory: Every LP has an alternative view (its **dual** LP).

Primal	Dual
maximization (or minimization)	minimization (or maximization)
objective coefficients	bounds
bounds	objective coefficients
bounded variable	\geq -constraint
\leq -constraint	bounded variable
free variable	$=$ -constraint
$=$ -constraint	free variable

dual of dual: original LP

Dual Problem

Definition (Dual Problem)

The **dual** of the standard maximum problem

$$\text{maximize } \mathbf{c}^T \mathbf{x} \text{ subject to } \mathbf{Ax} \leq \mathbf{b} \text{ and } \mathbf{x} \geq \mathbf{0}$$

is the standard minimum problem

$$\text{minimize } \mathbf{b}^T \mathbf{y} \text{ subject to } \mathbf{A}^T \mathbf{y} \geq \mathbf{c} \text{ and } \mathbf{y} \geq \mathbf{0}$$

Dual Problem: Example

Example (Dual of the Optimization Problem)

maximize $X_A + 5X_B$ subject to

$$-2X_A + X_B \leq 2$$

$$X_A + X_B \leq 12$$

$$X_A \leq 6$$

$$X_A \geq 0, \quad X_B \geq 0$$

Dual Problem: Example

Example (Dual of the Optimization Problem)

maximize $X_A + 5X_B$ subject to

$$[Y_1] \quad -2X_A + X_B \leq 2$$

$$[Y_2] \quad X_A + X_B \leq 12$$

$$[Y_3] \quad X_A \leq 6$$

$$X_A \geq 0, \quad X_B \geq 0$$

Dual Problem: Example

Example (Dual of the Optimization Problem)

minimize $2Y_1 + 12Y_2 + 6Y_3$ subject to

$$[X_A] \quad -2Y_1 + Y_2 + Y_3 \geq 1$$

$$[X_B] \quad Y_1 + Y_2 \geq 5$$

$$Y_1 \geq 0, \quad Y_2 \geq 0, \quad Y_3 \geq 0$$

Duality Theorem

Theorem (Duality Theorem)

*If a standard linear program is **bounded feasible**, then so is its dual, and their **objective values are equal**.*

(Proof omitted.)

The dual provides a different perspective on a problem.

Summary

Summary

- Linear (and integer) programs consist of an objective function that should be maximized or minimized subject to a set of given linear constraints.
- Finding solutions for integer programs is NP-complete.
- LP solving is a polynomial time problem.
- The dual of a maximization LP is a minimization LP and vice versa.
- The dual of a bounded feasible LP has the same objective value.

Further Reading

The slides in this chapter are based on the following excellent tutorial on LP solving:



Thomas S. Ferguson.

Linear Programming – A Concise Introduction.

UCLA, unpublished document available online.