# Automated Planning

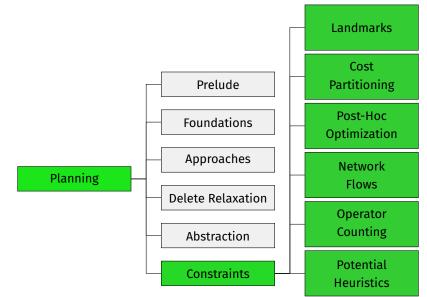
F5. Linear & Integer Programming

Jendrik Seipp

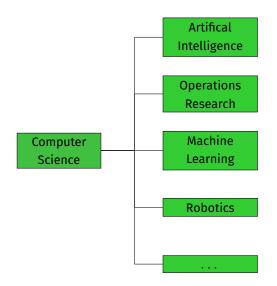
Linköping University

based on slides from the AI group at the University of Basel

# Content of this Course (IP & LP Relevance)



# Beyond this Course (Integer & Linear Programming Relevance)



Normal Forms and Duality

Summary 000

# **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
- However, even on the application side there is much more (e.g., modelling tricks or solver parameters to speed up computation)

# 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?

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

Example (Optimization Problem as Integer Program)

### $X_A \ge 0$ , $X_B \ge 0$

#### Example (Optimization Problem)

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

### 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

### $X_A \ge 0$ , $X_B \ge 0$

#### Example (Optimization Problem)

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#### Example (Optimization Problem)

"two plus twice the units of A may not be

smaller than the number of B"

### 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                  |
| $2 + 2X_A \ge X_B$                                |
| $X_A \ge 0$ , $X_B \ge 0$                         |

### Example (Optimization Problem)

"two plus twice the units of A may not be"

smaller than the number of B"

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

| Example (Optimization Problem as II | nteger Program)   |
|-------------------------------------|-------------------|
| maximize X <sub>A</sub> +           | $5X_B$ subject to |
| 2 + 2X                              | $A \geq X_B$      |
|                                     |                   |
|                                     |                   |
| $X_{A} \geq 0$ ,                    | $X_B \ge 0$       |

### Example (Optimization Problem)

"More than 12 products in total cannot be produced per day"

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

| Example (Optimization Proble | m as In     | teger                  | Program)   |  |
|------------------------------|-------------|------------------------|------------|--|
| maximize                     | $X_A + !$   | 5 <i>X<sub>B</sub></i> | subject to |  |
|                              | 2 + 2X      | > X                    | R          |  |
|                              | $X_A + X_B$ | . – .                  |            |  |
|                              |             |                        |            |  |
|                              |             |                        |            |  |
| Χ <sub>Α</sub>               | $\geq$ 0,   | X <sub>B</sub> ≥       | <u>≥</u> 0 |  |

### Example (Optimization Problem)

"More than 12 products in total cannot be produced per day"

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

| Example (Optimization Proble | m as Inte                         | ger Program) |  |
|------------------------------|-----------------------------------|--------------|--|
| maximize                     | $X_A + 5X$                        | 🕼 subject to |  |
|                              | $2 + 2X_A \ge$<br>$X_A + X_B \le$ |              |  |
|                              | VA I VB -                         | 2 12         |  |
| X <sub>A</sub>               | $\geq$ 0, )                       | $K_B \ge 0$  |  |

### Example (Optimization Problem)

"There is only material for 6 units of A"

Summary 000

### 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                  |  |  |
|   |  |  |
| $2 + 2X_A \ge X_B$                                |  |  |
| $X_A + X_B \leq 12$                               |  |  |
| $X_A \leq 6$                                      |  |  |
|   |  |  |
| $X_A \ge 0$ , $X_B \ge 0$                         |  |  |

### Example (Optimization Problem)

"There is only material for 6 units of A"

Summary 000

### 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                  |  |  |
|   |  |  |
| $2 + 2X_A \ge X_B$                                |  |  |
| $X_A + X_B \le 12$                                |  |  |
| $X_A \leq 6$                                      |  |  |
|   |  |  |
| $X_A \ge 0$ , $X_B \ge 0$                         |  |  |

#### $\rightarrow$ unique optimal solution:

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

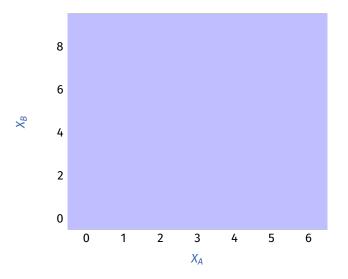
# Same Program as Input for the CPLEX Solver

### File ip. lp

```
Maximize
 obj: X_A + 5 X_B
Subject To
c1: -2 X_A + X_B <= 2
c2: X_A + X_B <= 12
Bounds
0 <= X_A <= 6
 O <= X_B
General
X_A X_B
End
```

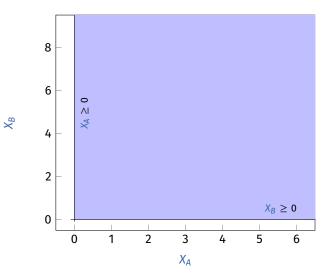
Normal Forms and Duality

Summary 000



Normal Forms and Duality

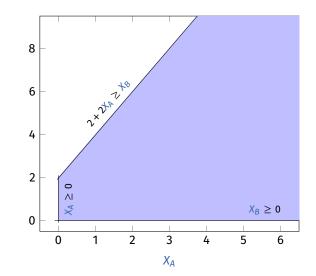
Summary 000



Normal Forms and Duality

Summary 000

# Integer Program Example: Visualization

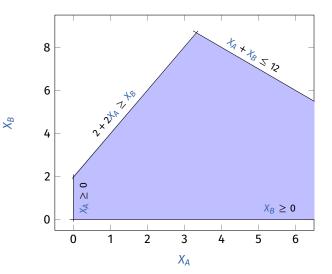


 $X_{\rm B}$ 

Normal Forms and Duality

Summary 000

# Integer Program Example: Visualization

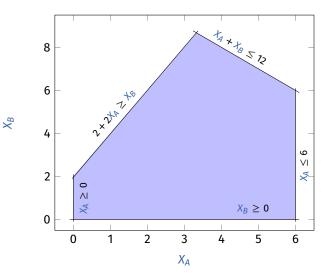


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Normal Forms and Duality

Summary 000

# Integer Program Example: Visualization



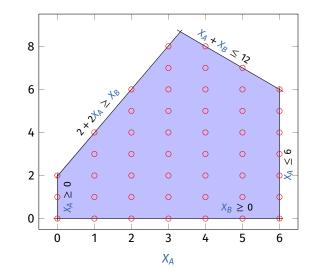
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 $X_{B}$ 

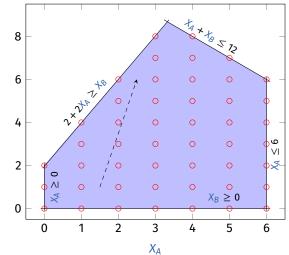
Linear Programs

Normal Forms and Duality

Summary 000



 $X_{B}$ 

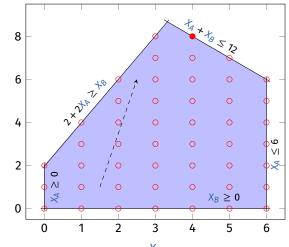


 $X_{B}$ 

Linear Programs

Normal Forms and Duality

Summary 000



### **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.

Normal Forms and Duality

### Another Example

| Example |                   |                                    |                  |
|---------|-------------------|------------------------------------|------------------|
|         | minimize          | $3X_{o_1} + 4X_{o_2} + 5X_{o_3}$   | subject to       |
|         |                   | V \ 1                              |                  |
|         |                   | $X_{o_4} \geq 1$                   |                  |
|         |                   | $X_{o_1} + X_{o_2} \ge 1$          |                  |
|         |                   | $X_{o_1} + X_{o_3} \ge 1$          |                  |
|         |                   | $X_{o_2} + X_{o_3} \ge 1$          |                  |
|         |                   |                                    |                  |
|         | $X_{o_1} \ge 0$ , | $X_{o_2} \geq 0,  X_{o_3} \geq 0,$ | $X_{o_4} \geq 0$ |

What example from a recent chapter does this IP encode?

Normal Forms and Duality

### Another Example

| Example |                   |                                    |                  |
|---------|-------------------|------------------------------------|------------------|
| n       | ninimize          | $3X_{o_1} + 4X_{o_2} + 5X_{o_3}$   | subject to       |
|         |                   | X <sub>04</sub> ≥ 1                |                  |
|         |                   | $X_{o_1} + X_{o_2} \ge 1$          |                  |
|         |                   | $X_{o_1} + X_{o_3} \ge 1$          |                  |
|         |                   | $X_{o_2} + X_{o_3} \ge 1$          |                  |
|         |                   |                                    |                  |
|         | $X_{o_1} \ge 0$ , | $X_{o_2} \geq 0,  X_{o_3} \geq 0,$ | $X_{o_4} \geq 0$ |

What example from a recent chapter does this IP encode?

 $\rightsquigarrow$  the minimum hitting set

# 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
- $\rightsquigarrow$  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
- $\rightsquigarrow$  Finding solutions for IPs is NP-complete.

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

Normal Forms and Duality

Summary 000

# **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-valued, 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                 |  |  |
|  |  |  |
| $2 + 2X_A \ge X_B$                               |  |  |
| $X_A + X_B \le 12$                               |  |  |
| $X_A \leq 6$                                     |  |  |
|  |  |  |
| $X_A \ge 0,  X_B \ge 0$                          |  |  |

# 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                 |  |  |
|  |  |  |
| $2 + 2X_A \ge X_B$                               |  |  |
| $X_A + X_B \le 12$                               |  |  |
| $X_A \leq 6$                                     |  |  |
|  |  |  |
| $X_A \ge 0,  X_B \ge 0$                          |  |  |

 $\rightarrow$  unique optimal solution:

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

# Same Program as Input for the CPLEX Solver

### File lp.lp

```
Maximize

obj: X_A + 5 X_B

Subject To

c1: -2 X_A + X_B <= 2

c2: X_A + X_B <= 12

Bounds

0 <= X_A <= 6

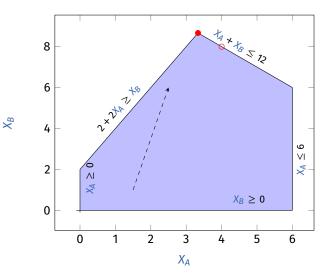
0 <= X_B

End
```

 $\rightarrow$  Demo

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Summary 000



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Summary 000

# Solving Linear Programs

#### Observation:

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

# Solving Linear Programs

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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).

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# 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.

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Summary 000

# LP Relaxation of MHS heuristic

#### Example (Minimum Hitting Set)

minimize  $3X_{o_1} + 4X_{o_2} + 5X_{o_3}$  subject to  $X_{o_4} \ge 1$   $X_{o_1} + X_{o_2} \ge 1$   $X_{o_1} + X_{o_3} \ge 1$   $X_{o_2} + X_{o_3} \ge 1$  $X_{o_2} \ge 0, \quad X_{o_2} \ge 0, \quad X_{o_4} \ge 0$ 

 $\rightarrow$  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 in polynomial time

Normal Forms and Duality

Summary 000

# Normal Forms and Duality

Normal Forms and Duality

# Standard Maximum Problem

Normal form for maximization problems:

Definition (Standard Maximum Problem)

Find values for  $x_1, \ldots, x_n$ , to maximize

 $c_1x_1 + c_2x_2 + \cdots + c_nx_n$ 

subject to the constraints

$$a_{11}x_{1} + a_{12}x_{2} + \dots + a_{1n}x_{n} \le b_{1}$$

$$a_{21}x_{1} + a_{22}x_{2} + \dots + a_{2n}x_{n} \le b_{2}$$

$$\vdots$$

$$a_{m1}x_{1} + a_{m2}x_{2} + \dots + a_{mn}x_{n} \le b_{m}$$

and  $x_1 \ge 0, x_2 \ge 0, \ldots, x_n \ge 0$ .

С

# Standard Maximum Problem: Matrix and Vectors

A standard maximum problem is often given by

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

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$$
(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  $\mathbf{A}\mathbf{x} \le \mathbf{b}$  and  $\mathbf{x} \ge \mathbf{0}$ .

Normal Forms and Duality

# 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 **A**x ≥ **b**
- All linear programs can efficiently be converted into a standard maximum/minimum problem.

# Some LP Theory: Duality

#### 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

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#### **Dual Problem**

#### Definition (Dual Problem)

The dual of the standard maximum problem

```
maximize \mathbf{c}^T \mathbf{x} subject to \mathbf{A}\mathbf{x} \leq \mathbf{b} and \mathbf{x} \geq \mathbf{0}
```

is the standard minimum problem

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

Normal Forms and Duality

# Dual Problem: Example

# Example (Dual of the Optimization Problem)maximize $X_A + 5X_B$ subject to $-2X_A + X_B \le 2$ $X_A + X_B \le 12$ $X_A + X_B \le 6$ $X_A \ge 0, \quad X_B \ge 0$

Normal Forms and Duality

# Dual Problem: Example

| Example (Dual of the Optimization Problem) |                         |  |
|--|-------------------------|--|
| maximize                                   | $X_A + 5X_B$ subject to |  |
| [Y <sub>1</sub> ]                          | $-2X_A + X_B \le 2$     |  |
| [Y <sub>2</sub> ]                          | $X_A + X_B \leq 12$     |  |
| [Y <sub>3</sub> ]                          | $X_A \leq 6$            |  |
|  |                         |  |
| $X_A \ge 0,  X_B \ge 0$                    |                         |  |

Normal Forms and Duality

# Dual Problem: Example

| Example (Dual of the Optimization Problem) |                       |                   |
|--|-----------------------|-------------------|
| minimize                                   | $2Y_1 + 12Y_2 + 6Y_3$ | subject to        |
| $[X_A]$                                    | -2Y <sub>1</sub> +    | $Y_2 + Y_3 \ge 1$ |
| $[X_B]$                                    |                       | $Y_1 + Y_2 \ge 5$ |
|  |                       |                   |
| $Y_1 \ge 0$ , $Y_2 \ge 0$ , $Y_3 \ge 0$    |                       |                   |

Normal Forms and Duality

# **Duality Theorem**

Theorem (Duality Theorem)

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

The dual provides a different perspective on a problem.

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Summary ●00

# 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.