

# Graph Theory and Optimization

## Integer Linear Programming

Nicolas Nisse

Université Côte d'Azur, Inria, CNRS, I3S, France

October 2018

# Outline

- 1 Integer Linear Programme
- 2 Some examples
- 3 Integrality gap
- 4 Polynomial Cases
- 5 More Examples

## Linear Programme (reminder)

Linear programmes can be written under the **standard form**:

$$\begin{array}{ll}\text{Maximize} & \sum_{j=1}^n c_j x_j \\ \text{Subject to:} & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \text{for all } 1 \leq i \leq m \\ & x_j \geq 0 \quad \text{for all } 1 \leq j \leq n.\end{array}$$

- the problem is a **maximization**;
- all constraints are **inequalities** (and not equations);
- all variables  $x_1, \dots, x_n$  are **non-negative**.

Linear Programme (Real variables) can be solved in polynomial-time in the number of variables and constraints (e.g., ellipsoid method)

## Linear Programme (reminder)

Linear programmes can be written under the **standard form**:

$$\begin{array}{ll}\text{Maximize} & \sum_{j=1}^n c_j x_j \\ \text{Subject to:} & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \text{for all } 1 \leq i \leq m \\ & x_j \geq 0 \quad \text{for all } 1 \leq j \leq n.\end{array}$$

- the problem is a **maximization**;
- all constraints are **inequalities** (and not equations);
- all variables  $x_1, \dots, x_n$  are **non-negative**.

Linear Programme (Real variables) can be solved in polynomial-time in the number of variables and constraints (e.g., ellipsoid method)

# Integer Linear Programme

Integer Linear programmes:

$$\begin{aligned}
 &\text{Maximize} && \sum_{j=1}^n c_j x_j \\
 &\text{Subject to:} && \sum_{j=1}^n a_{ij} x_j \leq b_i && \text{for all } 1 \leq i \leq m \\
 &&& x_j \in \mathbb{N} && \text{for all } 1 \leq j \leq n.
 \end{aligned}$$

- the problem is a **maximization**;
- all constraints are **inequalities** (and not equations);
- all variables  $x_1, \dots, x_n$  are **Integers**.

Integer Linear Programme is NP-complete in general!

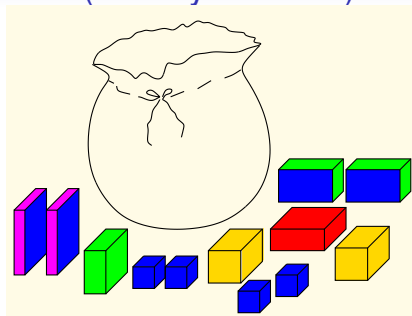
# Outline

- 1 Integer Linear Programme
- 2 Some examples
- 3 Integrality gap
- 4 Polynomial Cases
- 5 More Examples

# Knapsack Problem

(Weakly NP-hard)

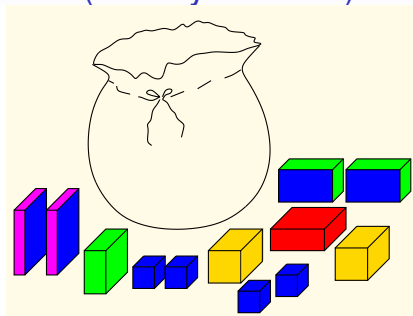
- **Data:**
  - a knapsack with maximum weight 15 Kg
  - 12 objects with
    - a weight  $w_i$
    - a value  $v_i$
- **Objective:** which objects should be chosen to maximize the value carried while not exceeding 15 Kg?



# Knapsack Problem

(Weakly NP-hard)

- **Data:**
  - a knapsack with maximum weight 15 Kg
  - 12 objects with
    - a weight  $w_i$
    - a value  $v_i$
- **Objective:** which objects should be chosen to maximize the value carried while not exceeding 15 Kg?



$$\begin{array}{ll}
 \max & \sum_{1 \leq i \leq 12} v_i x_i \\
 \text{subject to} & \sum_{1 \leq i \leq 12} w_i x_i \leq 15 \\
 & x_i \in \{0, 1\}
 \end{array}$$



# Minimum Vertex Cover

(NP-hard)

Let  $G = (V, E)$  be a graph

**Vertex Cover:** set  $K \subseteq V$  such that  $\forall e \in E, e \cap K \neq \emptyset$

*set of vertices that "touch" every edge*

# Minimum Vertex Cover

(NP-hard)

Let  $G = (V, E)$  be a graph

**Vertex Cover:** set  $K \subseteq V$  such that  $\forall e \in E, e \cap K \neq \emptyset$   
*set of vertices that "touch" every edge*

**Solution:**  $K \subseteq V$

$\Rightarrow$  variables  $x_v$ , for each  $v \in V$   
 $x_v = 1$  if  $v \in K$ ,  $x_v = 0$  otherwise.

# Minimum Vertex Cover

(NP-hard)

Let  $G = (V, E)$  be a graph

**Vertex Cover:** set  $K \subseteq V$  such that  $\forall e \in E, e \cap K \neq \emptyset$

*set of vertices that "touch" every edge*

**Solution:**  $K \subseteq V$

$\Rightarrow$  variables  $x_v$ , for each  $v \in V$

$x_v = 1$  if  $v \in K$ ,  $x_v = 0$  otherwise.

**Objective function:** minimize  $|K|$

minimize  $\sum_{v \in V} x_v$

# Minimum Vertex Cover

(NP-hard)

Let  $G = (V, E)$  be a graph

**Vertex Cover:** set  $K \subseteq V$  such that  $\forall e \in E, e \cap K \neq \emptyset$

*set of vertices that "touch" every edge*

**Solution:**  $K \subseteq V$

$\Rightarrow$  variables  $x_v$ , for each  $v \in V$

$x_v = 1$  if  $v \in K$ ,  $x_v = 0$  otherwise.

**Objective function:** minimize  $|K|$

minimize  $\sum_{v \in V} x_v$

**Constraint:**  $\forall \{u, v\} \in E, u \in K \text{ or } v \in K$

$x_u + x_v \geq 1$

# Minimum Vertex Cover

(NP-hard)

Let  $G = (V, E)$  be a graph

**Vertex Cover:** set  $K \subseteq V$  such that  $\forall e \in E, e \cap K \neq \emptyset$

*set of vertices that "touch" every edge*

**Solution:**  $K \subseteq V$

$\Rightarrow$  variables  $x_v$ , for each  $v \in V$

$x_v = 1$  if  $v \in K$ ,  $x_v = 0$  otherwise.

**Objective function:** minimize  $|K|$

minimize  $\sum_{v \in V} x_v$

**Constraint:**  $\forall \{u, v\} \in E, u \in K \text{ or } v \in K$

$x_u + x_v \geq 1$

$$\begin{array}{ll} \text{Minimize} & \sum_{v \in V} x_v \\ \text{Subject to:} & x_v + x_u \geq 1 \quad \text{for all } \{u, v\} \in E \\ & x_v \in \{0, 1\} \quad \text{for all } v \in V \end{array}$$

# Vertex Coloring

(NP-hard)

Let  $G = (V, E)$  be a graph

**$k$ -Proper coloring:**  $c : V \rightarrow \{1, \dots, k\}$  s.t.  $c(u) \neq c(v)$  for all  $\{u, v\} \in E$ .  
*color the vertices  $s$  ( $\leq k$  colors) s.t. adjacent vertices receive  $\neq$  colors*

## Vertex Coloring

(NP-hard)

Let  $G = (V, E)$  be a graph

**$k$ -Proper coloring:**  $c : V \rightarrow \{1, \dots, k\}$  s.t.  $c(u) \neq c(v)$  for all  $\{u, v\} \in E$ .  
*color the vertices  $s$  ( $\leq k$  colors) s.t. adjacent vertices receive  $\neq$  colors*

**Solution:**  $c : V \rightarrow \{1, \dots, n\} \Rightarrow$  variables  $y_j$ , is color  $j \in \{1, \dots, n\}$  used?  
 variable  $c_v^j$  for color  $j$  and vertex  $v$ :  $c_v^j = 1$  if  $v$  colored  $j$ ,  $c_v^j = 0$  otherwise

## Vertex Coloring

(NP-hard)

Let  $G = (V, E)$  be a graph

**$k$ -Proper coloring:**  $c : V \rightarrow \{1, \dots, k\}$  s.t.  $c(u) \neq c(v)$  for all  $\{u, v\} \in E$ .  
*color the vertices  $s$  ( $\leq k$  colors) s.t. adjacent vertices receive  $\neq$  colors*

**Solution:**  $c : V \rightarrow \{1, \dots, n\} \Rightarrow$  variables  $y_j$ , is color  $j \in \{1, \dots, n\}$  used?  
 variable  $c_v^j$  for color  $j$  and vertex  $v$ :  $c_v^j = 1$  if  $v$  colored  $j$ ,  $c_v^j = 0$  otherwise

**Objective function:** minimize # of used colors      minimize  $\sum_{1 \leq j \leq n} y_j$



## Vertex Coloring

(NP-hard)

Let  $G = (V, E)$  be a graph

**$k$ -Proper coloring:**  $c : V \rightarrow \{1, \dots, k\}$  s.t.  $c(u) \neq c(v)$  for all  $\{u, v\} \in E$ .  
*color the vertices  $s$  ( $\leq k$  colors) s.t. adjacent vertices receive  $\neq$  colors*

**Solution:**  $c : V \rightarrow \{1, \dots, n\} \Rightarrow$  variables  $y_j$ , is color  $j \in \{1, \dots, n\}$  used?  
 variable  $c_v^j$  for color  $j$  and vertex  $v$ :  $c_v^j = 1$  if  $v$  colored  $j$ ,  $c_v^j = 0$  otherwise

**Objective function:** minimize # of used colors      minimize  $\sum_{1 \leq j \leq n} y_j$

**Constraints:** each vertex  $v$  has 1 color       $\sum_{1 \leq j \leq n} c_v^j = 1$

ends of each edge  $\{u, v\} \in E$  have  $\neq$  colors:  $c_v^j + c_u^j \leq 1$  for all  $j \in \{1, \dots, n\}$   
 color  $j$  used if  $\geq 1$  vertex colored with  $j$        $c_v^j \leq y_j$  for all  $v \in V$

## Vertex Coloring

(NP-hard)

Let  $G = (V, E)$  be a graph

**$k$ -Proper coloring:**  $c : V \rightarrow \{1, \dots, k\}$  s.t.  $c(u) \neq c(v)$  for all  $\{u, v\} \in E$ .  
*color the vertices  $s \leq k$  colors s.t. adjacent vertices receive  $\neq$  colors*

$$\begin{array}{ll}
 \text{Minimize} & \sum_{1 \leq j \leq n} y_j \\
 \text{Subject to:} & \sum_{1 \leq j \leq n} c_v^j = 1 \quad \text{for all } v \in V \\
 & c_v^j + c_u^j \leq 1 \quad \text{for all } j \in \{1, \dots, n\}, \{u, v\} \in E \\
 & c_v^j \leq y_j \quad \text{for all } j \in \{1, \dots, n\}, v \in V \\
 & y_j \in \{0, 1\} \quad \text{for all } j \in \{1, \dots, n\} \\
 & c_v^j \in \{0, 1\} \quad \text{for all } j \in \{1, \dots, n\}, v \in V
 \end{array}$$

# Outline

- 1 Integer Linear Programme
- 2 Some examples
- 3 Integrality gap
- 4 Polynomial Cases
- 5 More Examples

# Integer Programme vs. Linear Programme

**Integer** Linear programme (ILP):

$$\begin{array}{ll}
 \text{Max.} & \sum_{j=1}^n c_j x_j \\
 \text{s.t.:} & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \forall 1 \leq i \leq m \\
 & x_j \in \mathbb{N} \quad \forall 1 \leq j \leq n.
 \end{array}$$

**NP-hard** in general

# Integer Programme vs. Linear Programme

**Integer** Linear programme (ILP):

$$\begin{array}{ll} \text{Max.} & \sum_{j=1}^n c_j x_j \\ \text{s.t.:} & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \forall 1 \leq i \leq m \\ & x_j \in \mathbb{N} \quad \forall 1 \leq j \leq n. \end{array}$$

**NP-hard** in general

**Fractional Relaxation:** Linear Programme

$$\begin{array}{ll} \text{Max.} & \sum_{j=1}^n c_j x_j \\ \text{s.t.:} & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \forall 1 \leq i \leq m \\ & x_j \geq 0 \quad \forall 1 \leq j \leq n. \end{array}$$

**Polynomial-time** solvable

# Integer Programme vs. Linear Programme

**Integer** Linear programme (ILP):

$$\begin{array}{ll} \text{Max.} & \sum_{j=1}^n c_j x_j \\ \text{s.t.:} & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \forall 1 \leq i \leq m \\ & x_j \in \mathbb{N} \quad \forall 1 \leq j \leq n. \end{array}$$

**NP-hard** in general

**Fractional Relaxation:** Linear Programme

$$\begin{array}{ll} \text{Max.} & \sum_{j=1}^n c_j x_j \\ \text{s.t.:} & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \forall 1 \leq i \leq m \\ & x_j \geq 0 \quad \forall 1 \leq j \leq n. \end{array}$$

**Polynomial-time** solvable

What is the difference between Optimal solutions of LP and of ILP?

# Integer Programme vs. Linear Programme

**Integer** Linear programme (ILP):

$$\begin{aligned} \text{Max.} \quad & \sum_{j=1}^n c_j x_j \\ \text{s.t.:} \quad & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \forall 1 \leq i \leq m \\ & x_j \in \mathbb{N} \quad \forall 1 \leq j \leq n. \end{aligned}$$

**NP-hard** in general

**Fractional Relaxation:** Linear Programme

$$\begin{aligned} \text{Max.} \quad & \sum_{j=1}^n c_j x_j \\ \text{s.t.:} \quad & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \forall 1 \leq i \leq m \\ & x_j \geq 0 \quad \forall 1 \leq j \leq n. \end{aligned}$$

**Polynomial-time** solvable

What is the difference between Optimal solutions of LP and of ILP?

$$OPT(LP) \geq OPT(ILP) \quad (\text{for a maximization problem})$$

$$OPT(LP) \leq OPT(ILP) \quad (\text{for a minimization problem})$$

If  $OPT(LP)$  is "closed" to  $OPT(ILP)$ , then solving the **Fractional Relaxation** (in polynomial-time) gives a good bound for the ILP

# Fractional Relaxation of Vertex Coloring

**Integer** Linear programme (ILP):

$$\begin{array}{ll}
 \text{Minimize} & \sum_{1 \leq j \leq n} y_j \\
 \text{Subject to:} & \sum_{1 \leq j \leq n} c_v^j = 1 \\
 & c_v^j + c_u^j \leq 1 \\
 & c_v^j \leq y_j \\
 & y_j \in \{0, 1\} \\
 & c_v^j \in \{0, 1\}
 \end{array}$$

Fractional Relaxation (LP):

$$\begin{array}{ll}
 \text{Minimize} & \sum_{1 \leq j \leq n} y_j \\
 \text{Subject to:} & \sum_{1 \leq j \leq n} c_v^j = 1 \\
 & c_v^j + c_u^j \leq 1 \\
 & c_v^j \leq y_j \\
 & y_j \geq 0 \\
 & c_v^j \geq 0
 \end{array}$$

$$\begin{array}{l}
 y_{\text{red}} = y_{\text{blue}} = y_{\text{green}} = 1 \\
 c_a^{\text{red}} = c_b^{\text{blue}} = c_c^{\text{green}} = 1 \\
 \text{OPT(ILP)} = \sum_c y_c = 3
 \end{array}$$

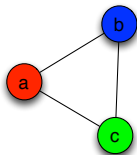
$$\begin{array}{l}
 y_{\text{red}} = y_{\text{blue}} = 1/2, y_{\text{green}} = 0 \\
 c_a^{\text{red}} = c_b^{\text{red}} = c_c^{\text{red}} = 1/2 \\
 c_a^{\text{blue}} = c_b^{\text{blue}} = c_c^{\text{blue}} = 1/2 \\
 \text{OPT(LP)} = \sum_c y_c = 1
 \end{array}$$



# Fractional Relaxation of Vertex Coloring

**Integer** Linear programme (ILP):

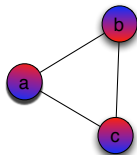
$$\begin{array}{ll}
 \text{Minimize} & \sum_{1 \leq j \leq n} y_j \\
 \text{Subject to:} & \sum_{1 \leq j \leq n} c_v^j = 1 \\
 & c_v^j + c_u^j \leq 1 \\
 & c_v^j \leq y_j \\
 & y_j \in \{0, 1\} \\
 & c_v^j \in \{0, 1\}
 \end{array}$$



$$\begin{aligned}
 y_{\text{red}} &= y_{\text{blue}} = y_{\text{green}} = 1 \\
 c_a^{\text{red}} &= c_b^{\text{blue}} = c_c^{\text{green}} = 1 \\
 \text{OPT(ILP)} &= \sum_c y_c = 3
 \end{aligned}$$

Fractional Relaxation (LP):

$$\begin{array}{ll}
 \text{Minimize} & \sum_{1 \leq j \leq n} y_j \\
 \text{Subject to:} & \sum_{1 \leq j \leq n} c_v^j = 1 \\
 & c_v^j + c_u^j \leq 1 \\
 & c_v^j \leq y_j \\
 & y_j \geq 0 \\
 & c_v^j \geq 0
 \end{array}$$



$$\begin{aligned}
 y_{\text{red}} &= y_{\text{blue}} = 1/2, y_{\text{green}} = 0 \\
 c_a^{\text{red}} &= c_b^{\text{red}} = c_c^{\text{red}} = 1/2 \\
 c_a^{\text{blue}} &= c_b^{\text{blue}} = c_c^{\text{blue}} = 1/2 \\
 \text{OPT(LP)} &= \sum_c y_c = 1
 \end{aligned}$$

# Fractional Relaxation of Knapsac

**Integer** Linear programme (ILP):

$$\begin{array}{ll}\max & \sum_{1 \leq i \leq n} v_i x_i \\ \text{subject to} & \sum_{1 \leq i \leq n} w_i x_i \leq W \\ & x_i \in \{0, 1\}\end{array}$$

Fractional Relaxation (LP):

$$\begin{array}{ll}\max & \sum_{1 \leq i \leq n} v_i x_i \\ \text{subject to} & \sum_{1 \leq i \leq n} w_i x_i \leq W \\ & x_i \geq 0\end{array}$$

**Example:**

- Sac:  $W = n$
- Objects:
  - one object ( $O_1$ ) of weight  $n + 0,1$  and value  $n$
  - $n - 1$  objects ( $O_2, \dots, O_n$ ) of weight 1 and value  $1/n$

$$x_1 = 0, x_2 = \dots = x_n = 1$$

$$OPT(ILP) = \sum_c v_i x_i = (n-1)/n$$

$$x_1 = \frac{n}{n+0,1}, x_2 = \dots = x_n = 0$$

$$OPT(LP) = \sum_c v_i x_i = \frac{n^2}{n+0,1}$$

⇒ the ratio between the LP optimal solution and the Integral opt. solution may be arbitrary large

# Fractional Relaxation of Knapsac

**Integer** Linear programme (ILP):

$$\begin{array}{ll}\max & \sum_{1 \leq i \leq n} v_i x_i \\ \text{subject to} & \sum_{1 \leq i \leq n} w_i x_i \leq W \\ & x_i \in \{0, 1\}\end{array}$$

Fractional Relaxation (LP):

$$\begin{array}{ll}\max & \sum_{1 \leq i \leq n} v_i x_i \\ \text{subject to} & \sum_{1 \leq i \leq n} w_i x_i \leq W \\ & x_i \geq 0\end{array}$$

**Example:**

- Sac:  $W = n$
- Objects:
  - one object ( $O_1$ ) of weight  $n+0,1$  and value  $n$
  - $n-1$  objects ( $O_2, \dots, O_n$ ) of weight 1 and value  $1/n$

$$x_1 = 0, x_2 = \dots = x_n = 1$$

$$OPT(ILP) = \sum_c v_i x_i = (n-1)/n$$

$$x_1 = \frac{n}{n+0,1}, x_2 = \dots = x_n = 0$$

$$OPT(LP) = \sum_c v_i x_i = \frac{n^2}{n+0,1}$$

⇒ the ratio between the LP optimal solution and the Integral opt. solution may be arbitrary large

# Fractional Relaxation of Knapsac

**Integer** Linear programme (ILP):

$$\begin{array}{ll}\max & \sum_{1 \leq i \leq n} v_i x_i \\ \text{subject to} & \sum_{1 \leq i \leq n} w_i x_i \leq W \\ & x_i \in \{0, 1\}\end{array}$$

Fractional Relaxation (LP):

$$\begin{array}{ll}\max & \sum_{1 \leq i \leq n} v_i x_i \\ \text{subject to} & \sum_{1 \leq i \leq n} w_i x_i \leq W \\ & x_i \geq 0\end{array}$$

**Example:**

- Sac:  $W = n$
- Objects:
  - one object ( $O_1$ ) of weight  $n+0,1$  and value  $n$
  - $n-1$  objects ( $O_2, \dots, O_n$ ) of weight 1 and value  $1/n$

$$x_1 = 0, x_2 = \dots = x_n = 1$$

$$OPT(ILP) = \sum_c v_i x_i = (n-1)/n$$

$$x_1 = \frac{n}{n+0,1}, x_2 = \dots = x_n = 0$$

$$OPT(LP) = \sum_c v_i x_i = \frac{n^2}{n+0,1}$$

$\Rightarrow$  the ratio between the LP optimal solution and the Integral opt. solution may be arbitrary large

# Outline

- 1 Integer Linear Programme
- 2 Some examples
- 3 Integrality gap
- 4 Polynomial Cases
- 5 More Examples

# No integrality gap

**Integer** Linear programme (ILP):

$$\begin{aligned} \text{Max.} \quad & \sum_{j=1}^n c_j x_j \\ \text{s.t.:} \quad & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \forall 1 \leq i \leq m \\ & x_j \in \mathbb{N} \quad \forall 1 \leq j \leq n. \end{aligned}$$

**NP-hard** in general

**Fractional Relaxation:** Linear Programme

$$\begin{aligned} \text{Max.} \quad & \sum_{j=1}^n c_j x_j \\ \text{s.t.:} \quad & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \forall 1 \leq i \leq m \\ & x_j \geq 0 \quad \forall 1 \leq j \leq n. \end{aligned}$$

**Polynomial-time** solvable

In some cases:  $OPT(ILP) = OPT(LP)$ .

$\Leftrightarrow$  there exists an integral solution with value  $OPT(LP)$ .

In such case, Polynomial-time solvable: solve the Fractional Relaxation

## Integer Programme Example: Shortest path

$D = (V, A)$  be a digraph with length  $\ell : A \rightarrow \mathbb{R}^+$ , and  $s, t \in V$ .

**Problem:** Compute a shortest directed path from  $s$  to  $t$ .

## Integer Programme Example: Shortest path

$D = (V, A)$  be a digraph with length  $\ell : A \rightarrow \mathbb{R}^+$ , and  $s, t \in V$ .

**Problem:** Compute a shortest directed path from  $s$  to  $t$ .

**Solution:** A path  $P$  from  $s$  to  $t$   $\Rightarrow$  variables  $x_a$  for each  $a \in A$   
 $x_a = 1$  if  $a \in A(P)$ ,  $x_a = 0$  otherwise.



# Integer Programme Example: Shortest path

$D = (V, A)$  be a digraph with length  $\ell : A \rightarrow \mathbb{R}^+$ , and  $s, t \in V$ .

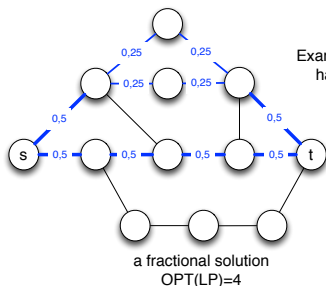
**Problem:** Compute a shortest directed path from  $s$  to  $t$ .

**Solution:** A path  $P$  from  $s$  to  $t$   $\Rightarrow$  variables  $x_a$  for each  $a \in A$   
 $x_a = 1$  if  $a \in A(P)$ ,  $x_a = 0$  otherwise.

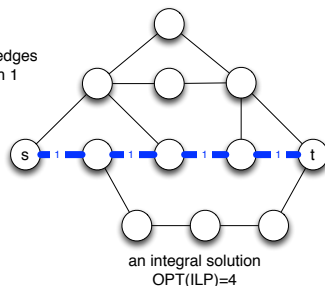
$$\begin{array}{ll}
 \text{Minimize} & \sum_{a \in A} \ell(a) x_a \\
 \text{Subject to:} & \sum_{u \in N^+(s)} x(su) = 1 \\
 & \sum_{u \in N^-(t)} x(tu) = 1 \\
 & \sum_{u \in N^+(v)} x(uv) = \sum_{u \in N^-(v)} x(vu) \quad \text{for all } v \in V \setminus \{s, t\} \\
 & x(a) \in \{0, 1\} \quad \text{for all } a \in A
 \end{array}$$

# Integer Programme Example: Shortest path

$$\begin{array}{ll}
 \text{Minimize} & \sum_{a \in A} \ell(a) x_a \\
 \text{Subject to:} & \sum_{u \in N^+(s)} x(su) = 1 \\
 & \sum_{u \in N^-(t)} x(tu) = 1 \\
 & \sum_{u \in N^+(v)} x(uv) = \sum_{u \in N^-(v)} x(vu) \quad \text{for all } v \in V \setminus \{s, t\} \\
 & x(a) \geq 0 \quad \text{for all } a \in A
 \end{array}$$



Example: all edges  
have length 1



**Exercise:** Prove that this LP always admits an integral optimal solution

# Integer Programme Example: Maximum Matching

$G = (V, E)$  be a graph

**Problem:** Compute a maximum matching

**Solution:** a set  $M \subseteq E$  of pairwise disjoint edges

$\Rightarrow$  variables  $x_e$  for each  $e \in E$

$x_e = 1$  if  $e \in M$ ,  $x_e = 0$  otherwise.

Maximize  $\sum_{e \in E} x_e$

Subject to:  $\sum_{e \in E, v \in e} x_e \leq 1$  for all  $v \in V$

$x_e \in \{0, 1\}$  for all  $e \in E$

**Exercise:** Prove that the fractional relaxation of this ILP always admits an integral optimal solution

# Totally unimodular matrices

**unimodular matrix:** square matrix with determinant  $+1$  or  $-1$

**totally unimodular matrix:** every square non-singular submatrix is unimodular

**Integer** Linear programme (ILP):

$$\begin{aligned} \text{Max.} \quad & \sum_{j=1}^n c_j x_j \\ \text{s.t.:} \quad & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \forall 1 \leq i \leq m \\ & x_j \in \mathbb{N} \quad \forall 1 \leq j \leq n. \end{aligned}$$

**NP-hard** in general

**Fractional Relaxation:** Linear Programme

$$\begin{aligned} \text{Max.} \quad & \sum_{j=1}^n c_j x_j \\ \text{s.t.:} \quad & \sum_{j=1}^n a_{ij} x_j \leq b_i \quad \forall 1 \leq i \leq m \\ & x_j \geq 0 \quad \forall 1 \leq j \leq n. \end{aligned}$$

**Polynomial-time** solvable

## Theorem

[Hoffman, Kruskal, 1956]

If the matrix  $A = [a_{ij}]$  is totally unimodular then every *basic* feasible solution (the "corner" of the polytope) is integral

$\Rightarrow$  exist integral optimal solution of the LP

$\Rightarrow \text{OPT(ILP)}$  can be computed by solving the Fractional relaxation

# Outline

- 1 Integer Linear Programme
- 2 Some examples
- 3 Integrality gap
- 4 Polynomial Cases
- 5 More Examples

# Integer Programme Example: Minimum Spanning Tree

$G = (V, E)$  be a graph with weight  $w : E \rightarrow \mathbb{R}^+$ , and  $s, t \in V$ .

**Problem:** Compute a minimum spanning tree

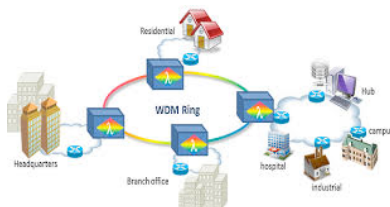
**Solution:** A spanning tree  $T \Rightarrow$  variables  $x_e$  for each  $e \in E$   
 $x_e = 1$  if  $e \in E(T)$ ,  $x_e = 0$  otherwise.

$$\begin{array}{ll}
 \text{Minimize} & \sum_{e \in E} w(e)x_e \\
 \text{Subject to:} & \sum_{e=\{u,v\} \in E, u \in S, v \notin S} x_e \geq 1 \quad \text{for all } S \subseteq V \\
 & x_e \in \{0, 1\} \quad \text{for all } e \in E
 \end{array}$$

**Remark:** The number of constraints is exponential

# Optical Networks (WDM)

**Optical network:** optical fiber connecting e.g. routers

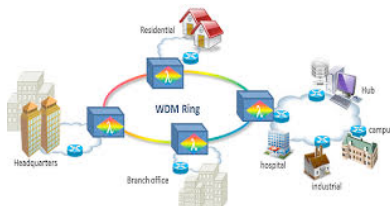


**Wavelength-division Multiplexing (WDM):** technology which multiplexes a number of optical carrier signals onto a single optical fiber by using different wavelengths (i.e., colors) of laser light [Wikipedia]

⇒: **different signals** on the same link must have **different wavelengths** (colors)

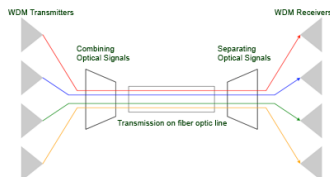
# Optical Networks (WDM)

**Optical network:** optical fiber connecting e.g. routers



**Wavelength-division Multiplexing (WDM):** technology which multiplexes a number of optical carrier signals onto a single optical fiber by using different wavelengths (i.e., colors) of laser light [Wikipedia]

⇒: **different signals** on the same link must have **different wavelengths** (colors)





# Optical Networks (WDM)

# RWA problem

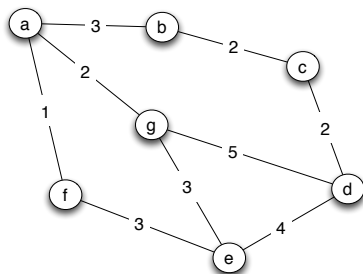
## RWA: Routing and Wavelength Assignment Problem

Given a graph  $G = (V, E)$  with capacity on links, and a traffic-demand matrix  $T$ , where  $T[u, v]$  is the amount of traffic that must transit from  $u$  to  $v$ , for any  $u, v \in V$ . Find a set of paths and one wavelength assignment for each path such that:

- all demands are routed
- capacity of each link cannot be exceeded
- total number of wavelength is as small as possible

demand matrix:

	a	b	c	d	e	f	g
a	0	0	1	1	0	1	0
b	0	0	0	0	0	2	0
c	0	0	0	0	0	0	0
d	0	0	0	0	0	0	5
e	0	0	0	0	0	0	0
f	0	0	0	0	0	0	0
g	0	1	0	0	0	0	0



## RWA problem

# Optical Networks (WDM)

# RWA problem

Let us simplify the problem

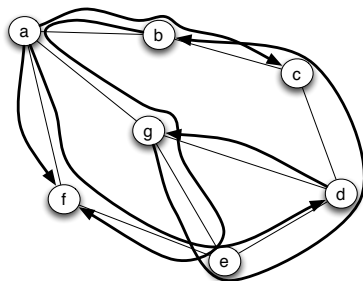
⇒ consider only Wavelength assignment

## WA: Wavelength Assignment Problem

Given a graph  $G = (V, E)$  with capacity on links, and a set of paths

Give One color to each path s.t. no two paths with the same color cross a same link

Minimize the number of colors



# Optical Networks (WDM)

# RWA problem

Let us simplify the problem

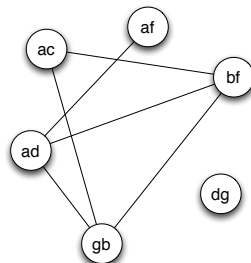
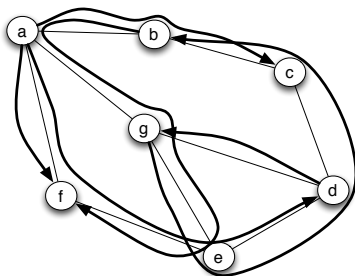
⇒ consider only Wavelength assignment

## WA: Wavelength Assignment Problem

Given a graph  $G = (V, E)$  with capacity on links, and a set of paths

Give One color to each path s.t. no two paths with the same color cross a same link

Minimize the number of colors



It is the problem of PROPER COLORING in graphs !!

the “simplified” problem is already NP-complete :(

# Optical Networks (WDM)

# RWA problem

Let us simplify the problem

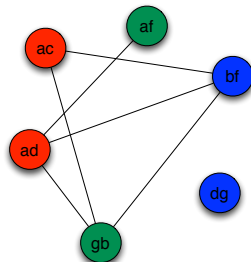
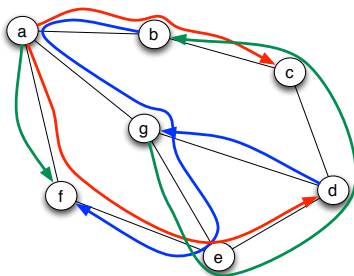
⇒ consider only Wavelength assignment

## WA: Wavelength Assignment Problem

Given a graph  $G = (V, E)$  with capacity on links, and a set of paths

Give One color to each path s.t. no two paths with the same color cross a same link

Minimize the number of colors



It is the problem of PROPER COLORING in graphs !!

the "simplified" problem is already NP-complete :(

# Optical Networks (WDM)

# RWA problem

Let us simplify the problem

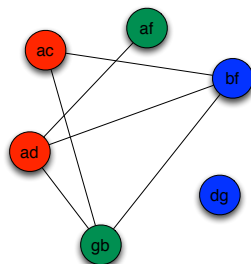
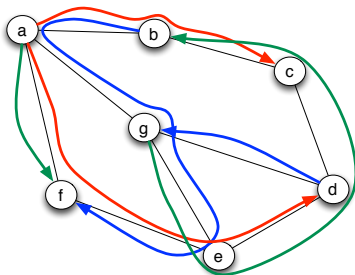
⇒ consider only Wavelength assignment

## WA: Wavelength Assignment Problem

Given a graph  $G = (V, E)$  with capacity on links, and a set of paths

Give One color to each path s.t. no two paths with the same color cross a same link

Minimize the number of colors



**Exercise:** Write a (Integer) Linear Programme that solves the RWA problem

## Summary: To be remembered

- ILP allow to model many problems
- there may be a huge **integrality gap**  
(between  $OPT(LP)$  and  $OPT(ILP)$ ).
- if no integrality gap (e.g., **totally unimodular matrices**)  
⇒ Fractional Relaxation gives Optimal Integral Solution