
Lecture 9: Solving Integer Linear Programs

Recap

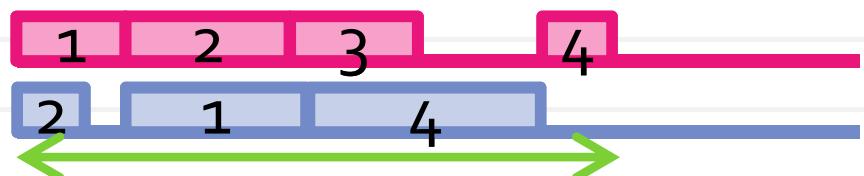
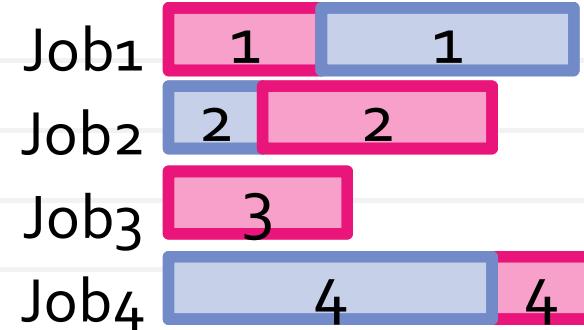


$$\min p^T x$$

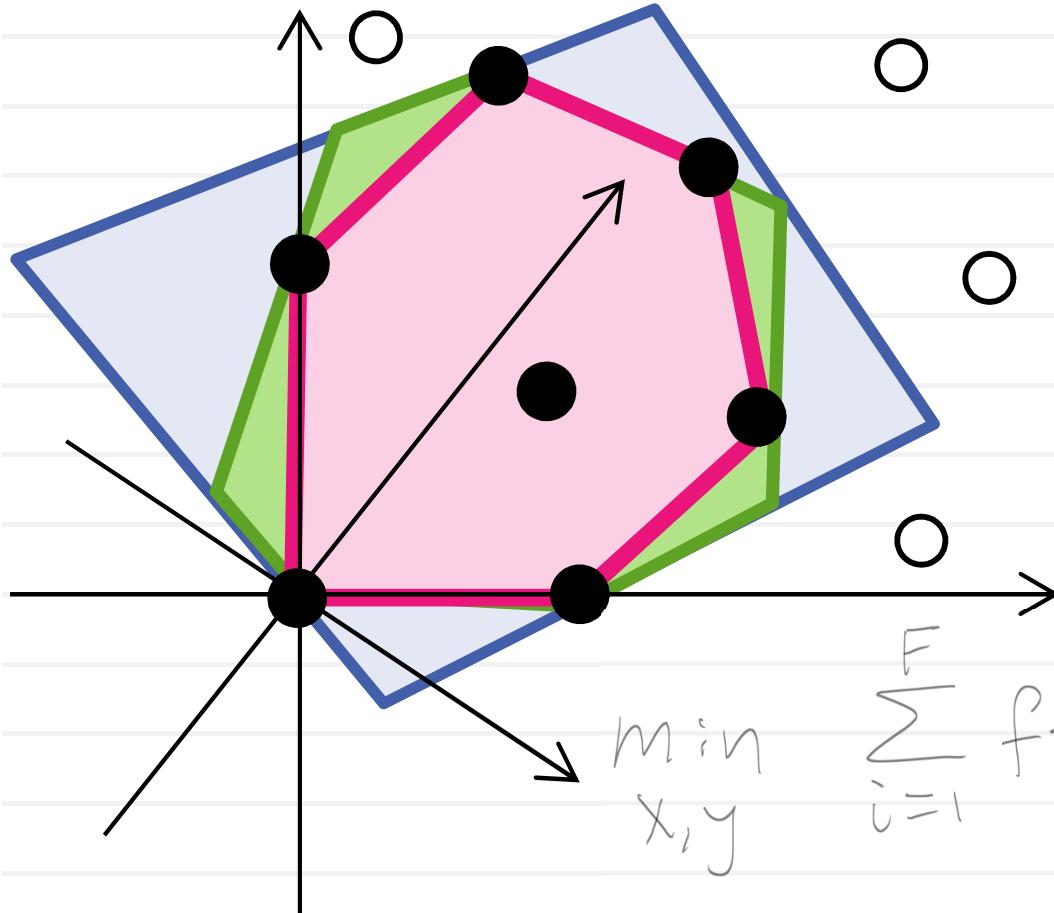
$$\text{s.t.: } Ax \leq b$$

$$Cx = d$$

$$\forall i \in I \quad x_i \in \{0; 1\}$$



Integer programs and polytopes



$$\min_{x,y} \sum_{i=1}^F f_i y_i + \sum_{i=1}^F \sum_{j=1}^C a_{ji} \cdot x_{ji}$$

$$x_i, y_i \in \{0;1\} \quad \forall j \sum_{i=1}^F x_{ji} = 1$$

$$\forall i \quad x_{ji} \leq y_i$$

$$\sum_{j=1}^C x_{ji} \leq C \cdot y_i$$

Now what?

We get some fractional solution. What to do next?

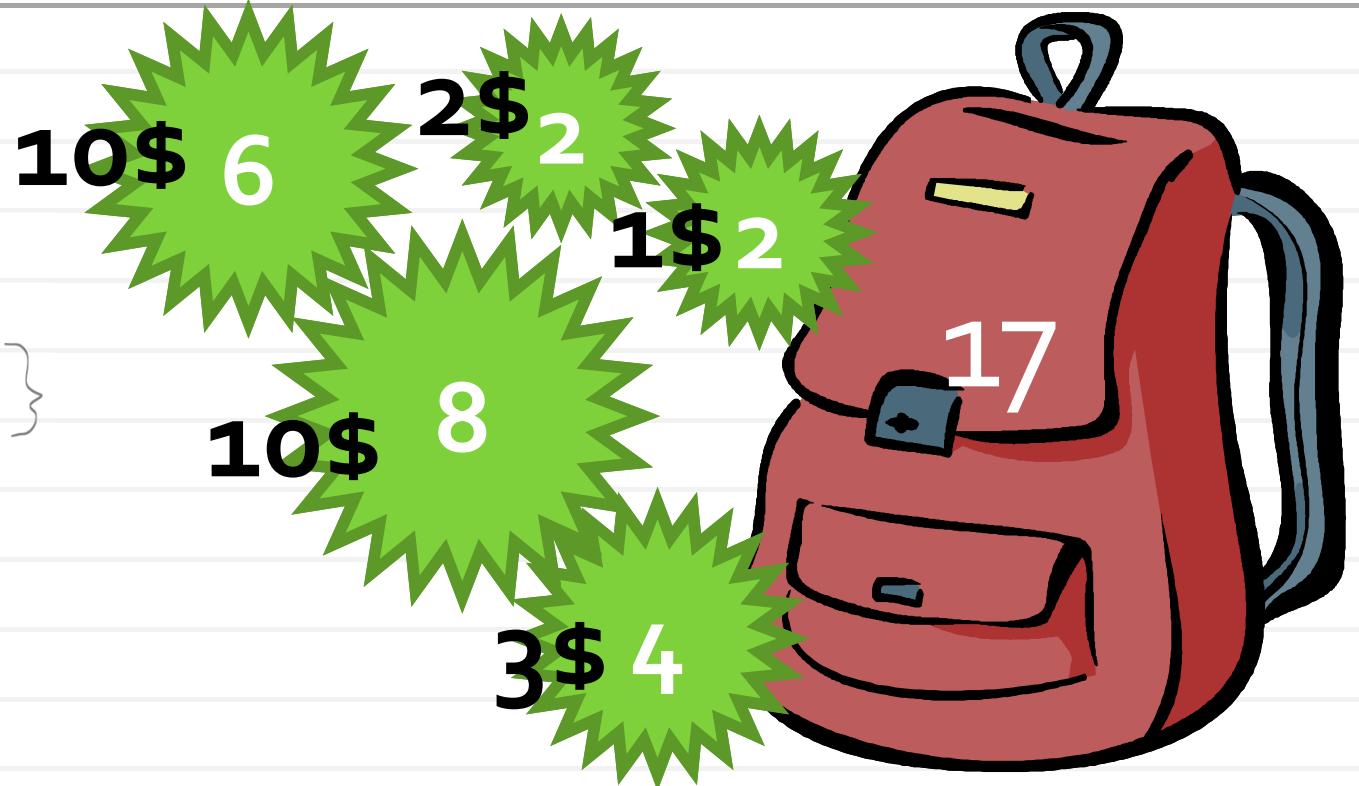
- We can round to the nearest integer solution (no feasibility guarantees!)
- We can use fractional values to guide greedy/local search
- More principled methods: branch-and-bound, cutting plane, branch-and-cut

Knapsack again

$$\min -C^T x$$

$$\text{s.t.: } S^T x \leq S$$

$$x_i \in \{0, 1\}$$



Relaxation:

$$c = [10 \ 2 \ 3 \ 1 \ 10]$$

$$s = [6 \ 2 \ 4 \ 2 \ 8]$$

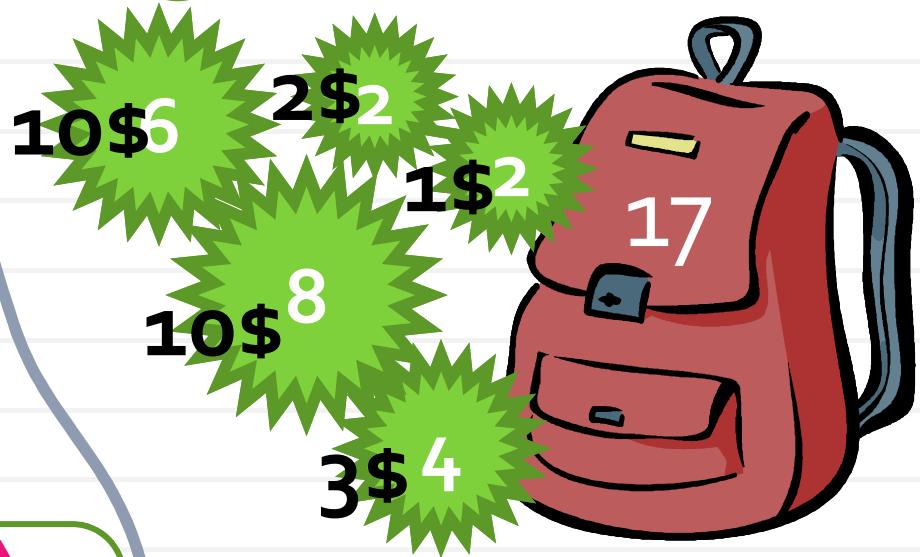
$$\begin{aligned} \min & -C^T x \\ \text{s.t.: } & S^T x \leq S \\ & 0 \leq x \leq 1 \end{aligned} \Rightarrow$$

$$x = [1 \ 1 \ 0.25 \ 0 \ 1]$$

$$\text{Obj} = -22.75$$

Branching, bounding, pruning

$x = [1 \ 1 \ 0.25 \ 0 \ 1]$
 $\text{Obj} = -22.75$



$\{x_3 = 0\}$
 $x = [1 \ 1 \ 0 \ 0.5 \ 1]$
 $\text{Obj} = -22.5$

$\{x_3 = 1\}$
 $x = [1 \ 1 \ 0.25 \ 0.5 \ 1]$
 $\text{Obj} = -21.75$

$\{x_3 = 0, x_4 = 0\}$
 $x = [1 \ 1 \ 0 \ 0 \ 1]$
 $\text{Obj} = -22$

$\{x_3 = 0, x_4 = 1\}$
 $x = [1 \ 1 \ 0 \ 1 \ 1]$
 $\text{Obj} = -22$

pruned

$$C = [10 \ 2 \ 3 \ 1 \ 10]$$

$$S = [6 \ 2 \ 4 \ 2 \ 8]$$

$$\min -C^T x$$

$$\text{s.t. } S^T x \leq S$$

$$x_i \in \{0, 1\}$$

$$0 \leq x_i \leq 1$$

Branch-and-bound parameters

- Choice one: which variable to branch on

Typical answer: the most fractional

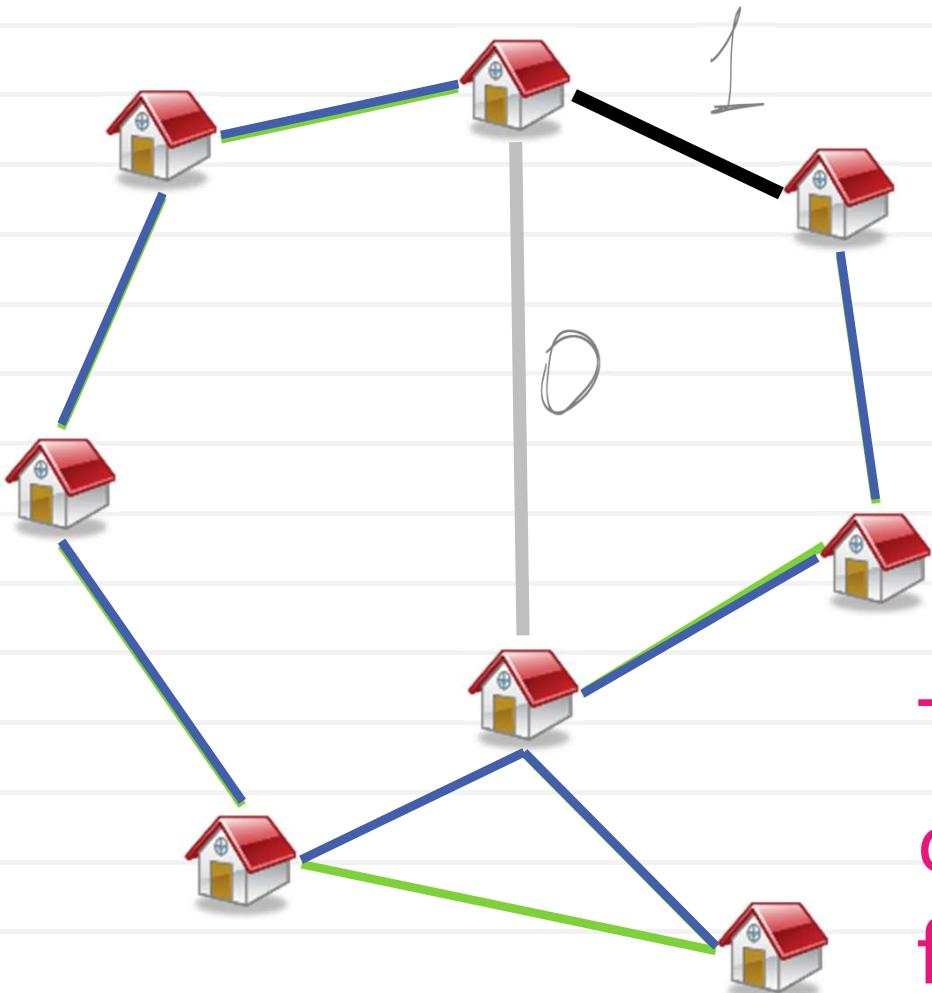
- Choice two: which node to branch

Typical answers: best-first or depth-first

Bound choices:

- Relaxation
- Lagrangian dual
- Customary bound for your task

Example: a bound for TSP

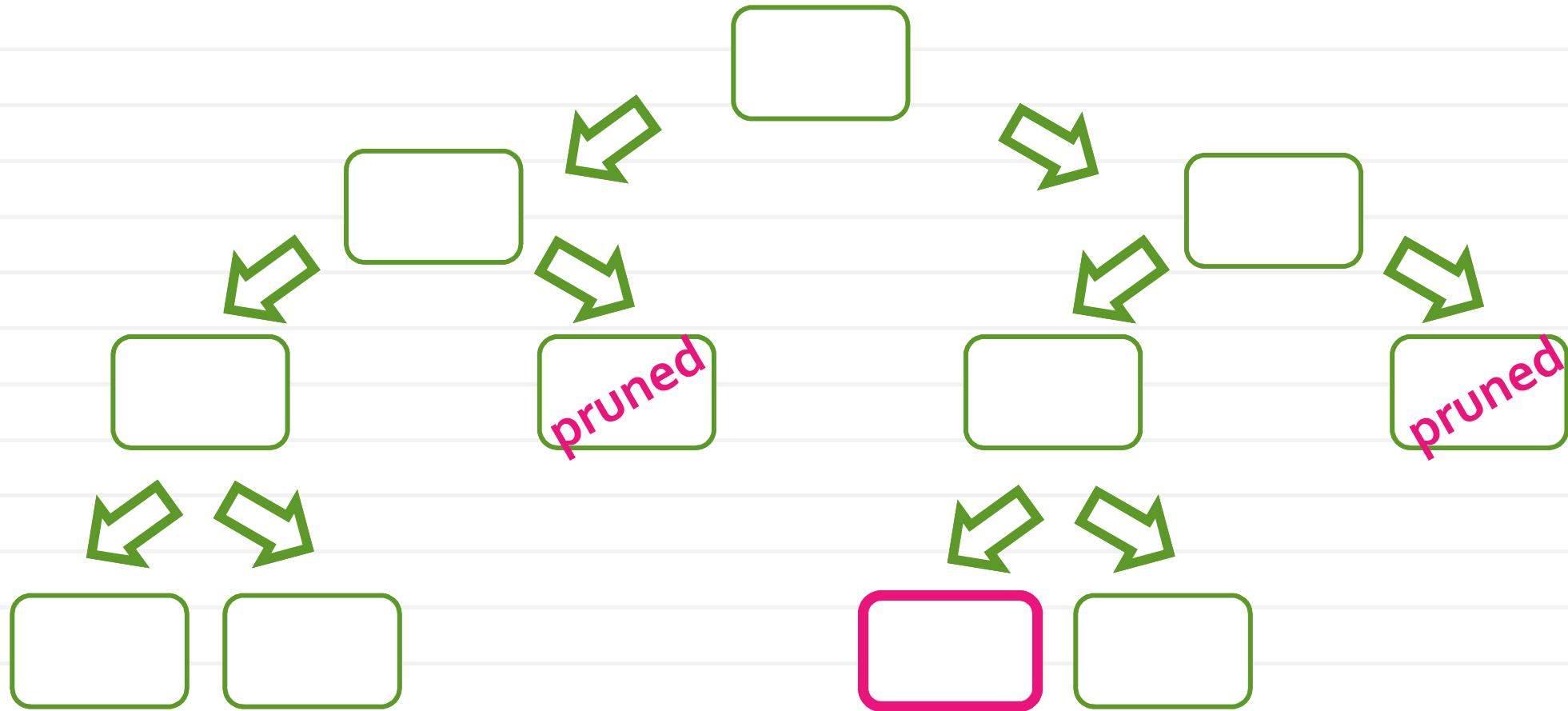


The weight of the optimal tree is a lower bound on the weight of the optimal tour

The MST algorithms give cheap and good bounds for TSP (can be made even better using Lagrange duality)

Best-first search

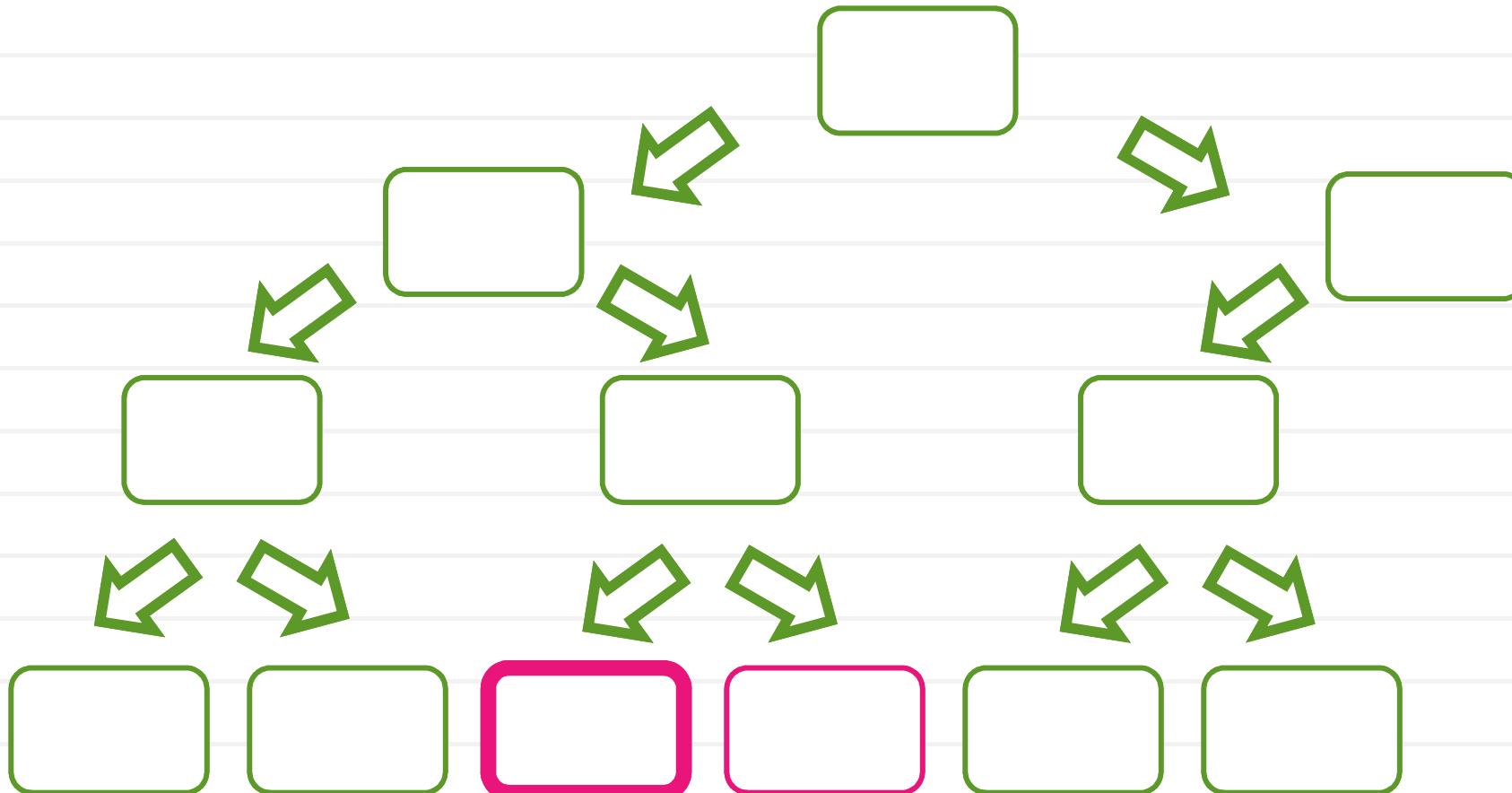
Idea: always explore the branch with the lowest bound



+: the most economical, only explores the nodes that we must to, to assert optimality

Depth-first search

Idea: always explore the most integer branch



- + : finds some incumbents early-on. Can terminate early with an approximate solution.
- : if run to optimality, will explore more nodes than breadth-first

Branch-and-bound summary

Initialize the list L of the active nodes to the root node

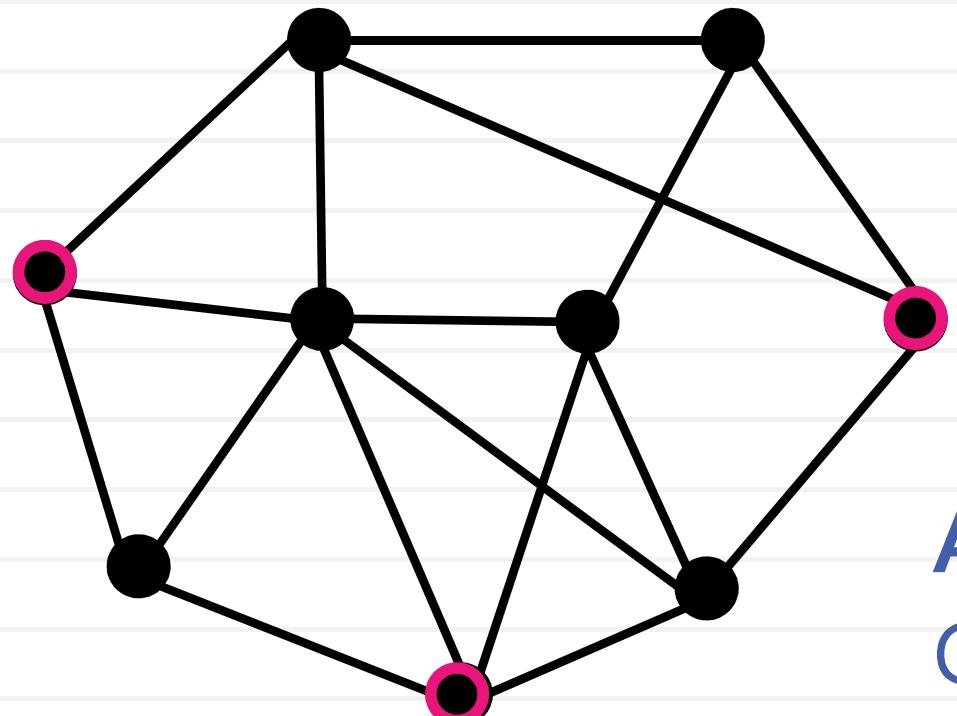
$incumbent_value = +\infty$

Until L is empty or time is out

- Pick the active node from L (*e.g. the deepest or with the smallest lower bound*)
- Pick the variable to branch
- **For each** of the two children nodes
 - $[x \ obj] = \text{Solve LP (node)}$
 - **If** $obj \geq incumbent_value$ **then continue;** //pruning
 - **If** x is integer **then**
 $incumbent = x; incumbent_value = obj;$
 - **Else** add node to L

MWIS problem

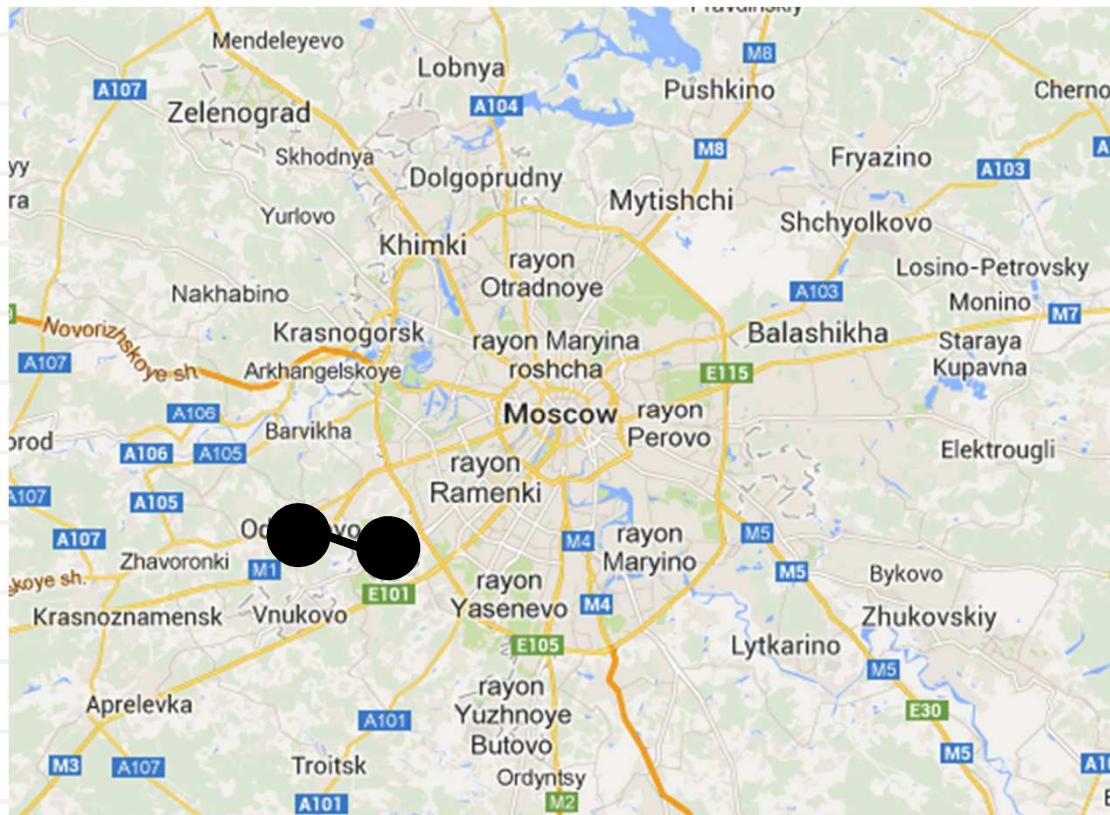
Maximum weight independent set
(node packing) problem:



$$\begin{aligned} \max \quad & \omega^\top x \\ \text{s.t.: } & \forall (i,j) \in E \quad x_i + x_j \leq 1 \\ & x_i \in \{0; 1\} \end{aligned}$$

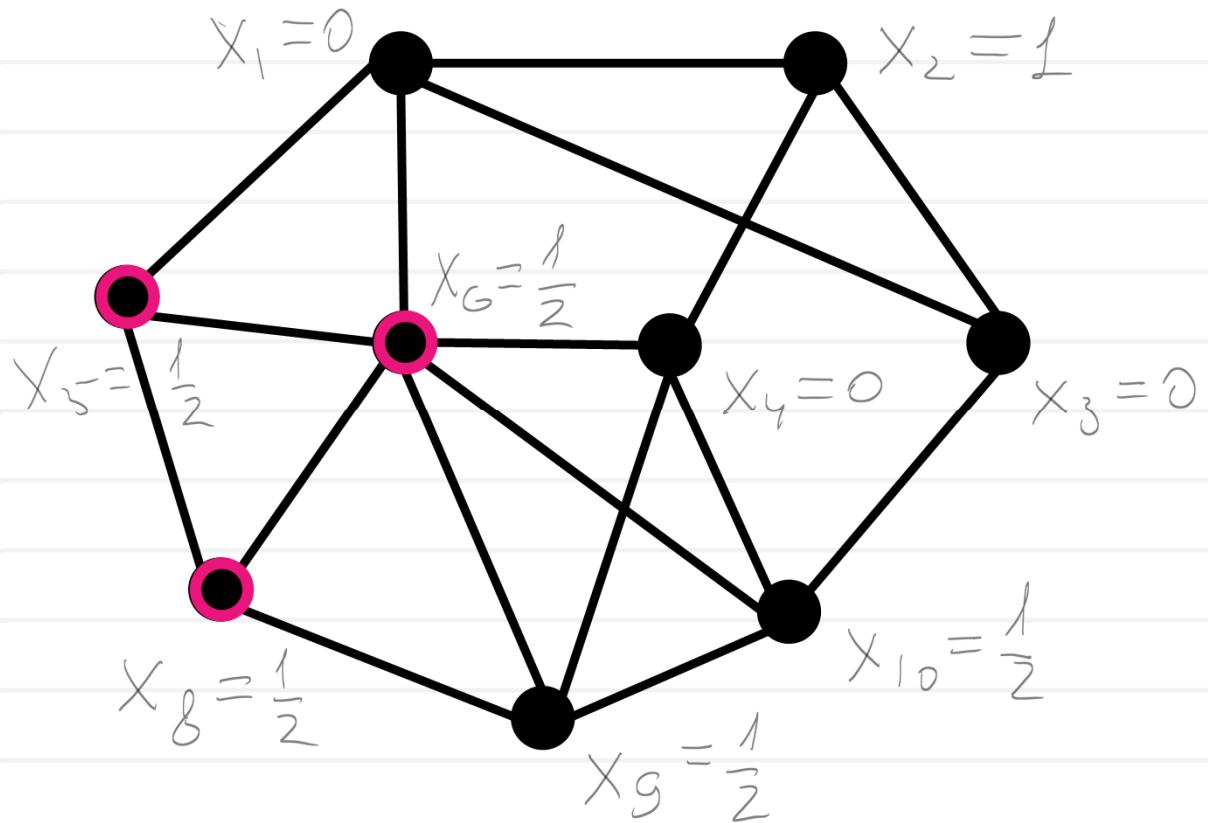
Applications:
Choosing courses
Inviting people to a party
Map coloring
Map labeling

Map labeling



$$\begin{aligned} & \max \quad \omega^T x \\ & \times \\ & \text{s.t.: } h(i,j) \in E \\ & \quad x_i + x_j \leq 1 \\ & \quad h_i \quad x_i \in \{0; 1\} \end{aligned}$$

LP relaxation for MWIS



$$\begin{aligned} & \min_{\mathbf{x}} \mathbf{\omega}^\top \mathbf{x} \\ \text{s.t.: } & \forall (i,j) \in E \quad x_i + x_j \leq 1 \end{aligned}$$

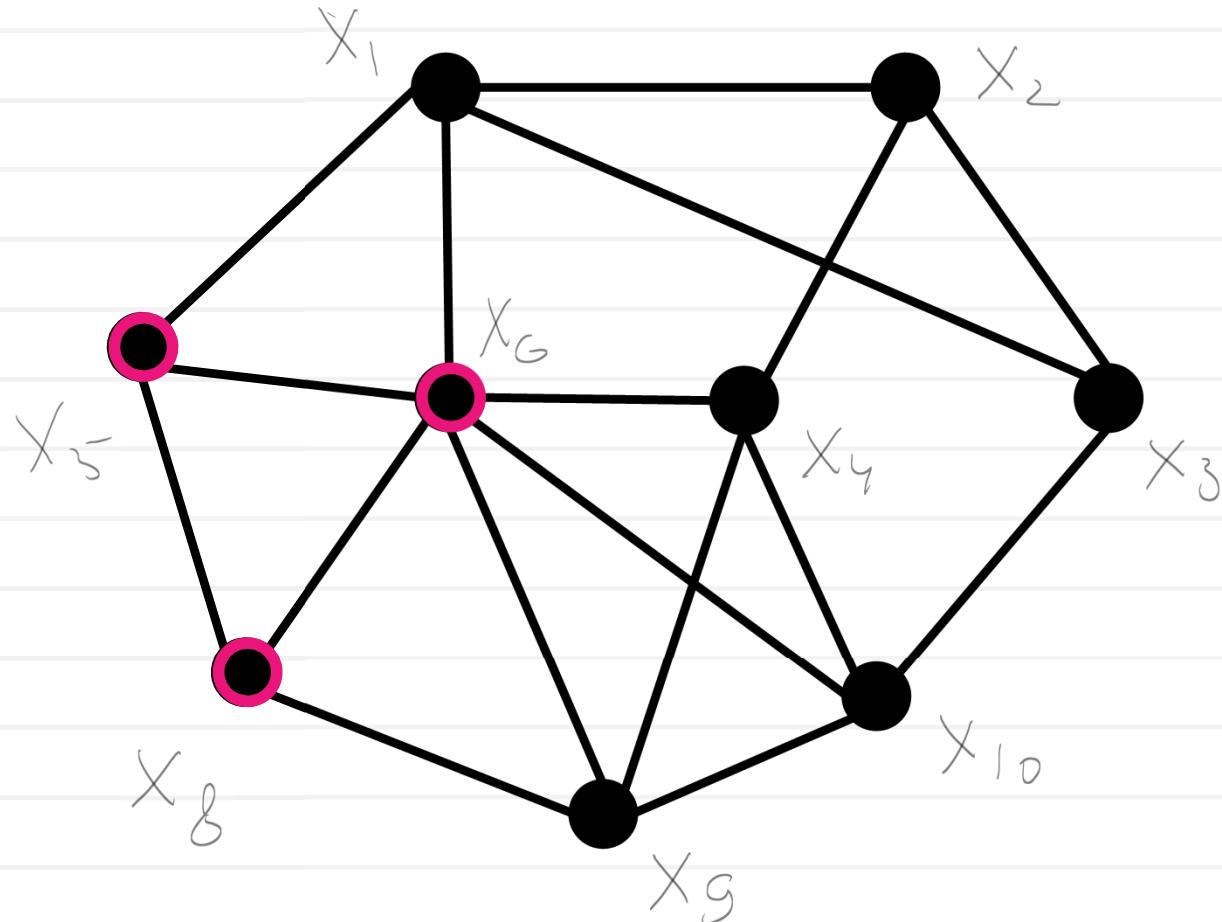
$$\forall i \quad 0 \leq x_i \leq 1$$

Is there a way to tighten the relaxation?

Yes!

$$x_5 + x_6 + x_8 \leq 1$$

MWIS problem



$$\begin{aligned} & \min \quad \omega^T x \\ & \text{s.t.: } H(i,j) \in E \\ & \quad x_i + x_j \leq 1 \\ & \forall i \quad x_i \in \{0; 1\} \end{aligned}$$

Generally, any **clique** provides a *cutting plane*

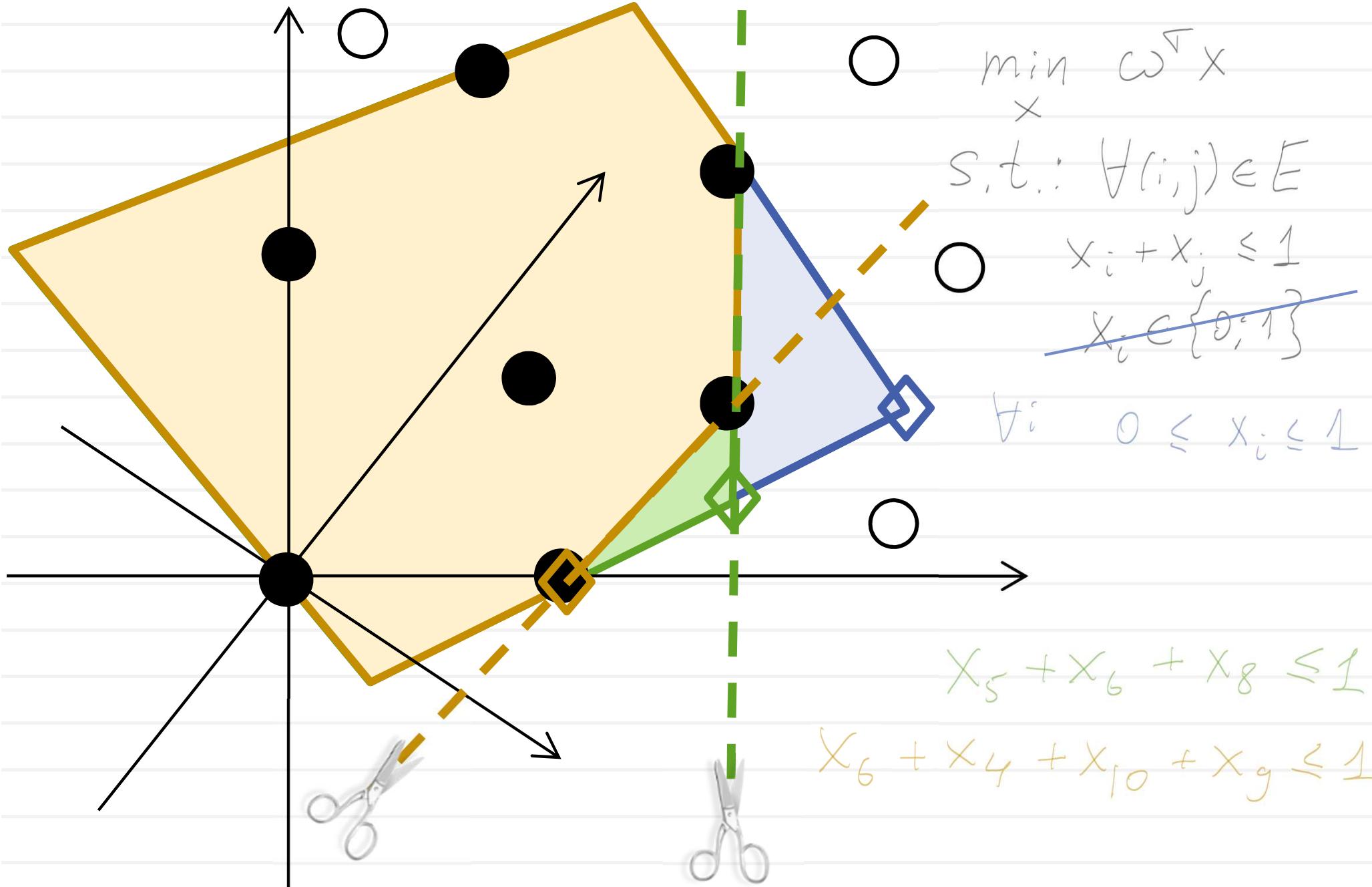
$$x_5 + x_6 + x_8 \leq 1$$

$$x_6 + x_4 + x_{10} \leq 1$$

$$x_6 + x_4 + x_{10} + x_9 \leq 1$$

$$x_4 + x_9 + x_{10} \leq 1$$

What has happened



Cutting plane algorithm

$[p, D_{ILP}] = \text{Formulate problem as an ILP}$

$[p, A, b, C, d] = \text{LPRelaxation(ILP)}$

$x = \text{SolveLP}(p, A, b, C, d)$

while x not integer

$[A', b'] = \text{FindSeparatingCut}(s)(x, D_{ILP})$

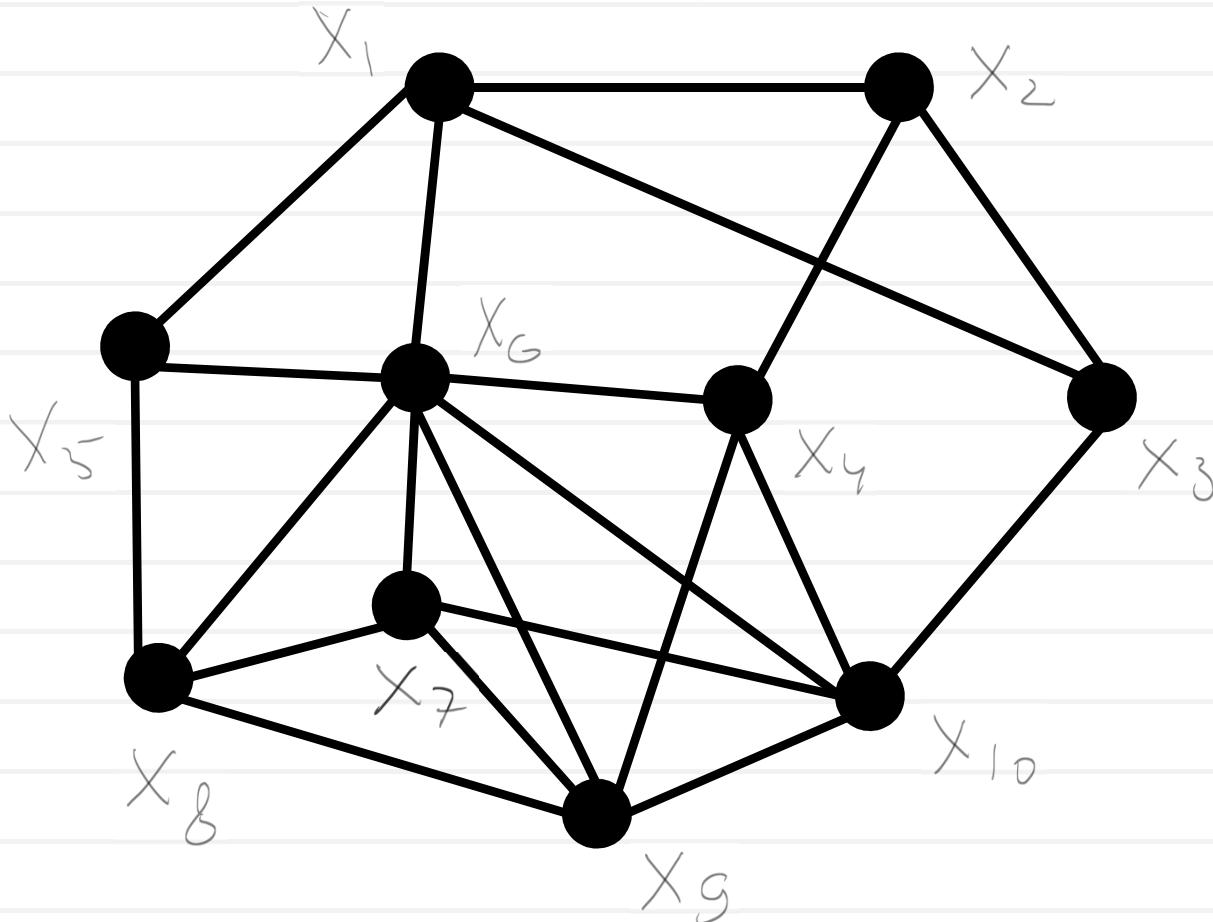
$A = [A; A'], b = [b, b']$

$x = \text{SolveLP}(p, A, b, C, d)$

end

Separation oracle for MWIS

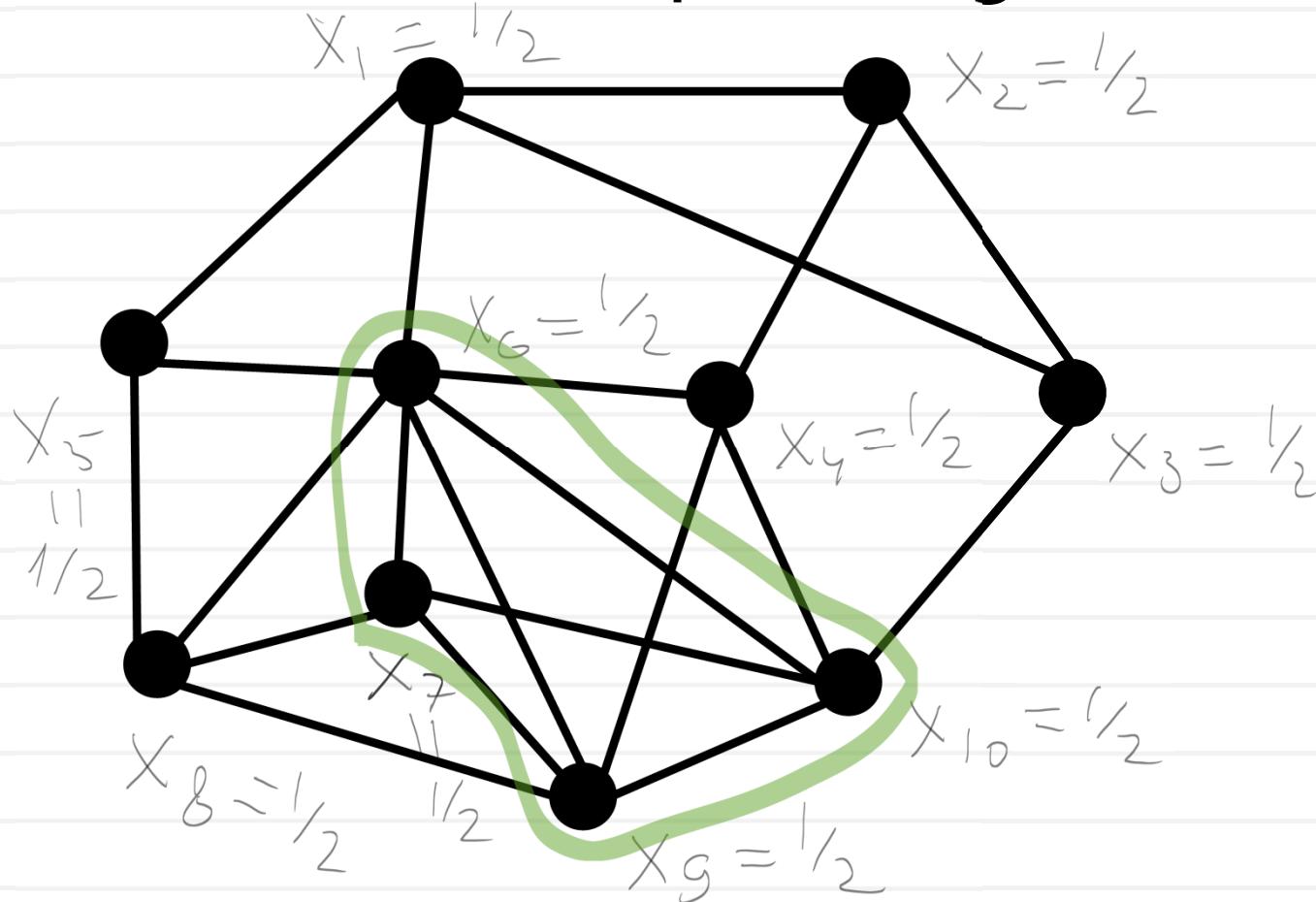
$[A' \ b'] = \text{FindSeparatingCut}(s) (x, D_{ILP})$



- Finding a clique with a sum > 1
- Can be done e.g. greedily

Separation oracle for MWIS

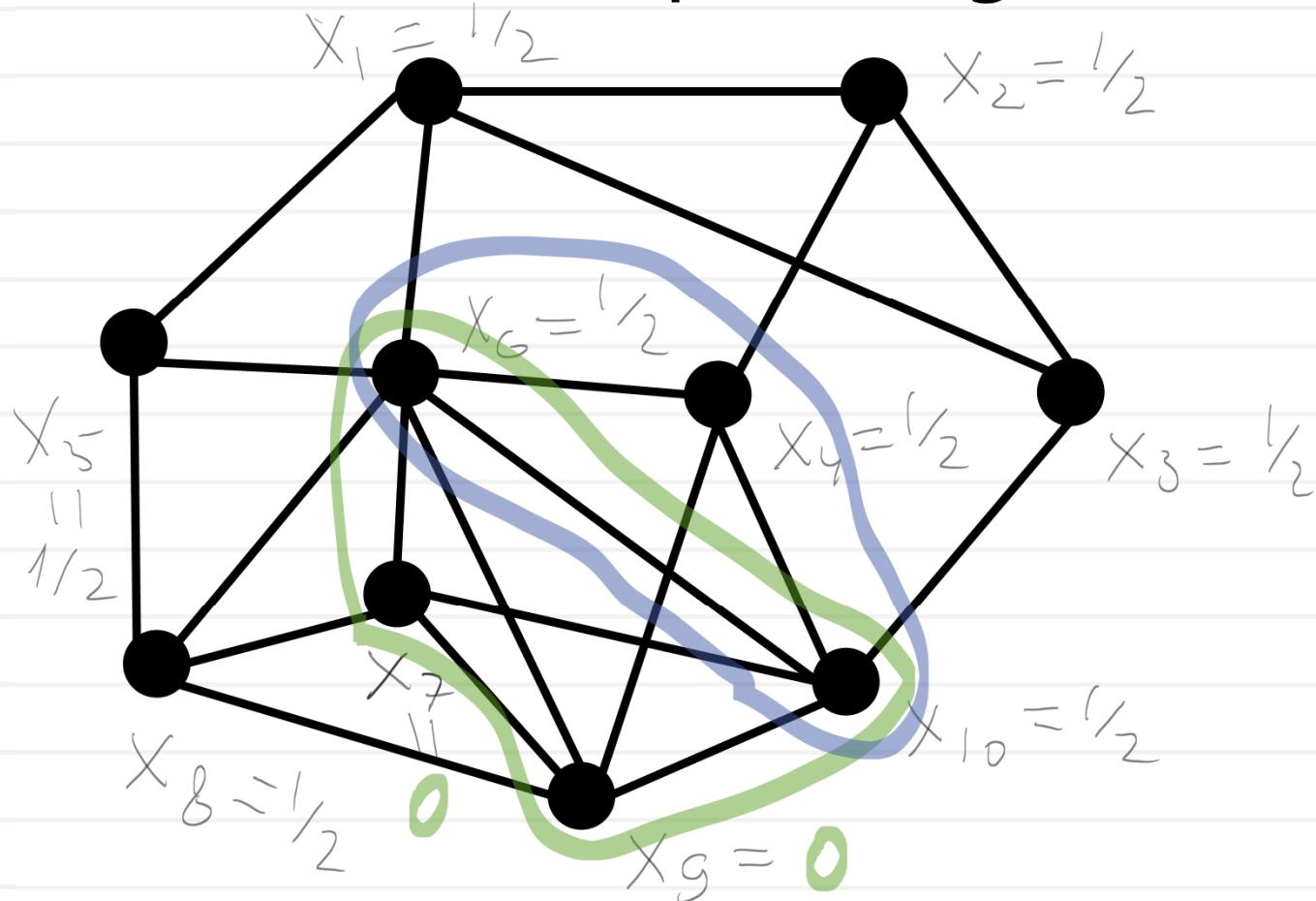
$[A' b'] = \text{FindSeparatingCut}(s)(x, D_{ILP})$



$$\begin{aligned} & \min \quad \omega^\top x \\ & \text{s.t.: } \forall (i, j) \in E \\ & \quad x_i + x_j \leq 1 \\ & \quad x_6 + x_7 + x_9 + x_{10} \leq 1 \end{aligned}$$

Separation for MWIS

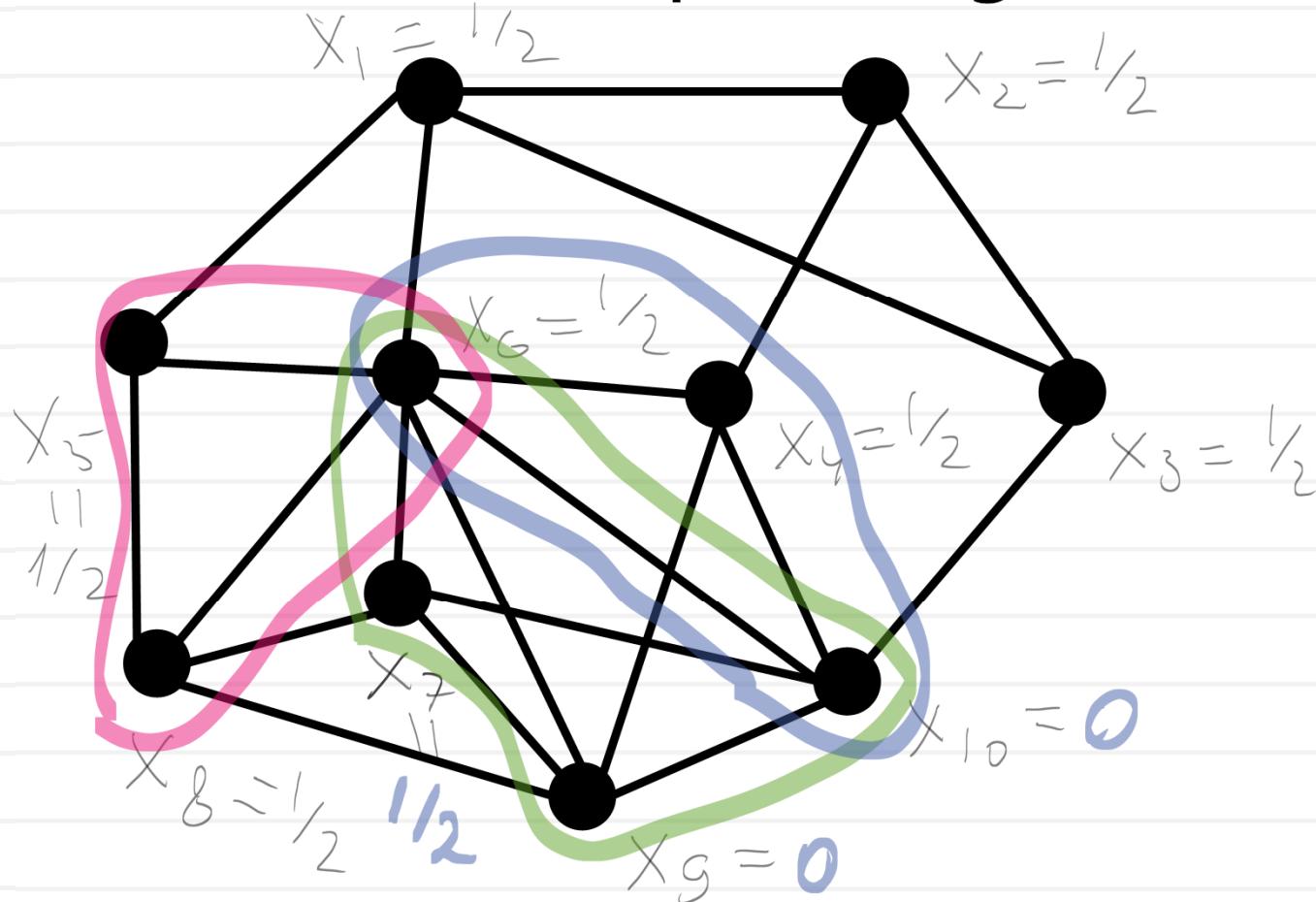
$[A' b'] = \text{FindSeparatingCut}(s)(x, D_{ILP})$



$$\begin{aligned} & \min \quad \omega^\top x \\ & \text{s.t.: } \forall (i, j) \in E \\ & \quad x_i + x_j \leq 1 \\ & \quad x_6 + x_7 + x_9 + x_{10} \leq 1 \\ & \quad x_6 + x_4 + x_{10} \leq 1 \end{aligned}$$

Separation for MWIS

$[A' b'] = \text{FindSeparatingCut}(s)(x, D_{ILP})$



$$\begin{aligned} & \min \quad \omega^T x \\ & \text{s.t.: } \forall (i, j) \in E \\ & \quad x_i + x_j \leq 1 \end{aligned}$$

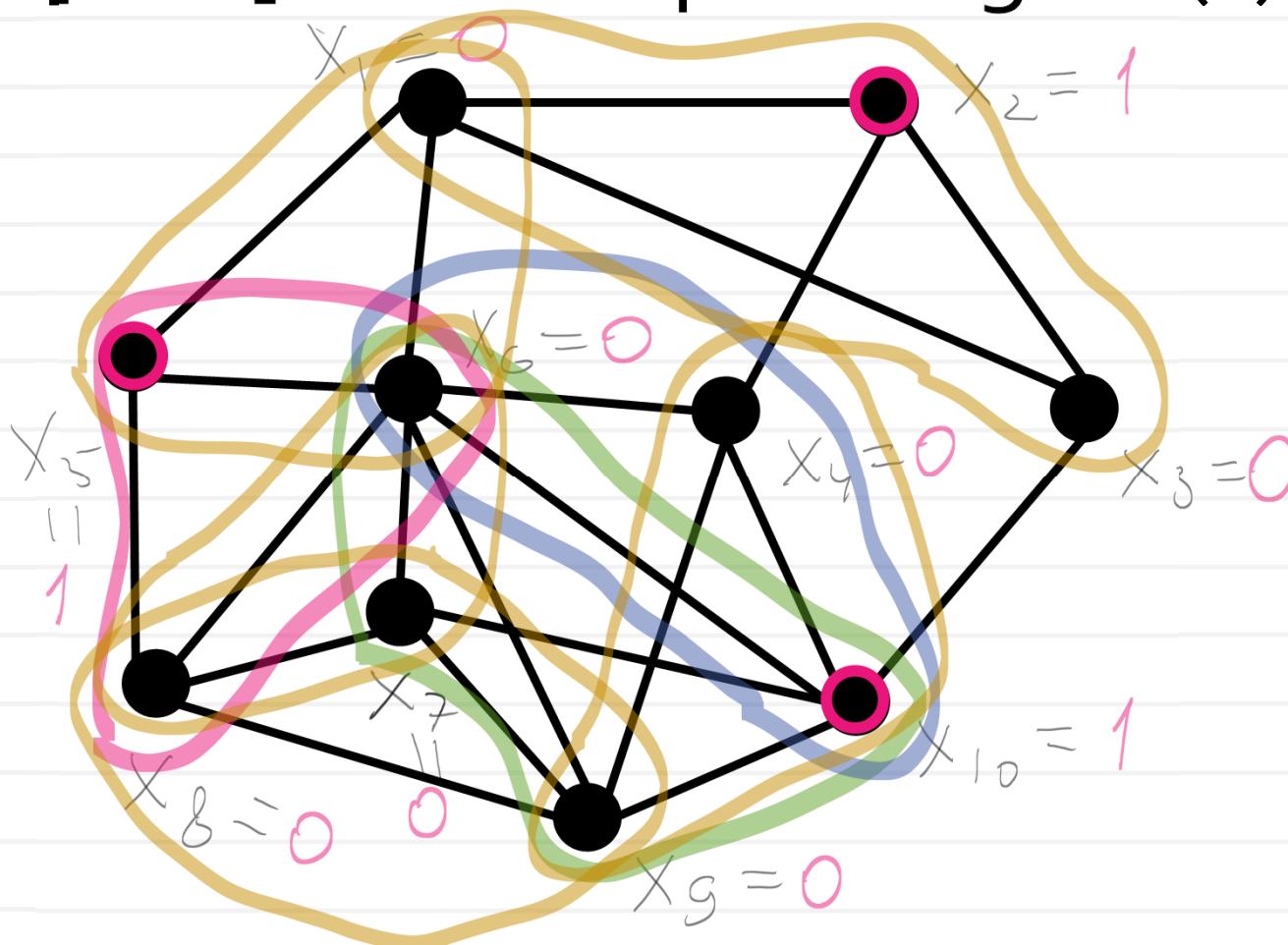
$$x_6 + x_7 + x_8 + x_{10} \leq 1$$

$$x_6 + x_4 + x_{10} \leq 1$$

$$x_5 + x_6 + x_8 \leq 1$$

Separation for MWIS

$[A' b'] = \text{FindSeparatingCut}(s)(x, D_{ILP})$



$$\begin{aligned} & \min \quad \omega^\top x \\ & \text{s.t.: } \forall (i, j) \in E \\ & \quad x_i + x_j \leq 1 \end{aligned}$$

$$x_6 + x_7 + x_9 + x_{10} \leq 1$$

$$x_6 + x_4 + x_{10} \leq 1$$

$$x_5 + x_6 + x_8 \leq 1$$

Orange cliques represent non-trivial constraints that were not needed

Travelling salesman problem

$$\min \sum_{i,j} w_{ij} x_{ij}$$

$$\text{s.t.: } x_{ij} \in \{0;1\}$$

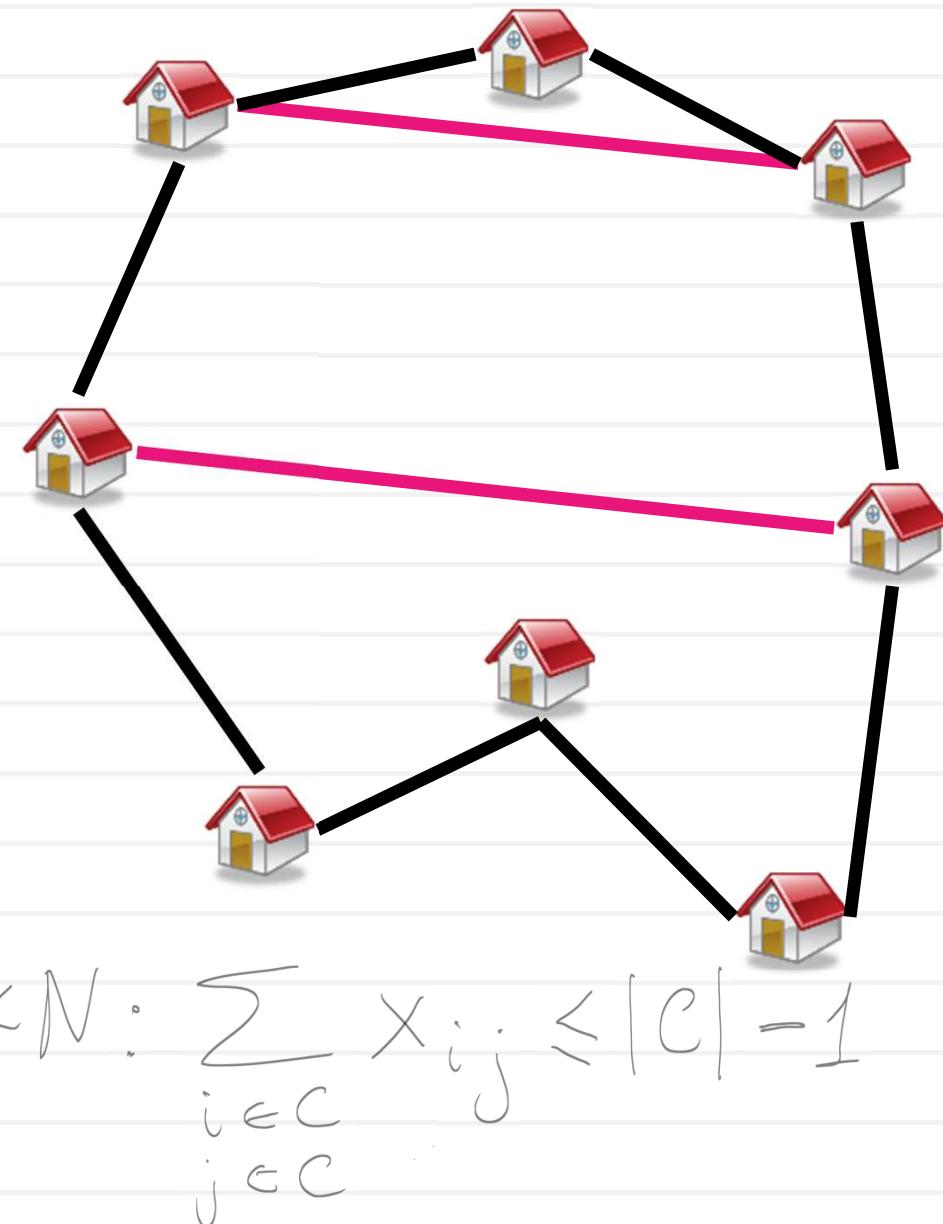
$$\forall i: \sum_j x_{ij} = 2$$

$$\sum_{i,j} x_{ij} = N$$

$$\forall c \subset \{1..N\} \ 0 < |c| < N: \sum_{\substack{i \in c \\ j \in c}} x_{ij} \leq |c| - 1$$

Exponentially many
constraints!

(assuming that x_{ij} and x_{ji} are the same)



Travelling salesman problem

$$\min \sum_{i,j} w_{ij} x_{ij}$$

$$\text{s.t.: } x_{ij} \in \{0;1\}$$

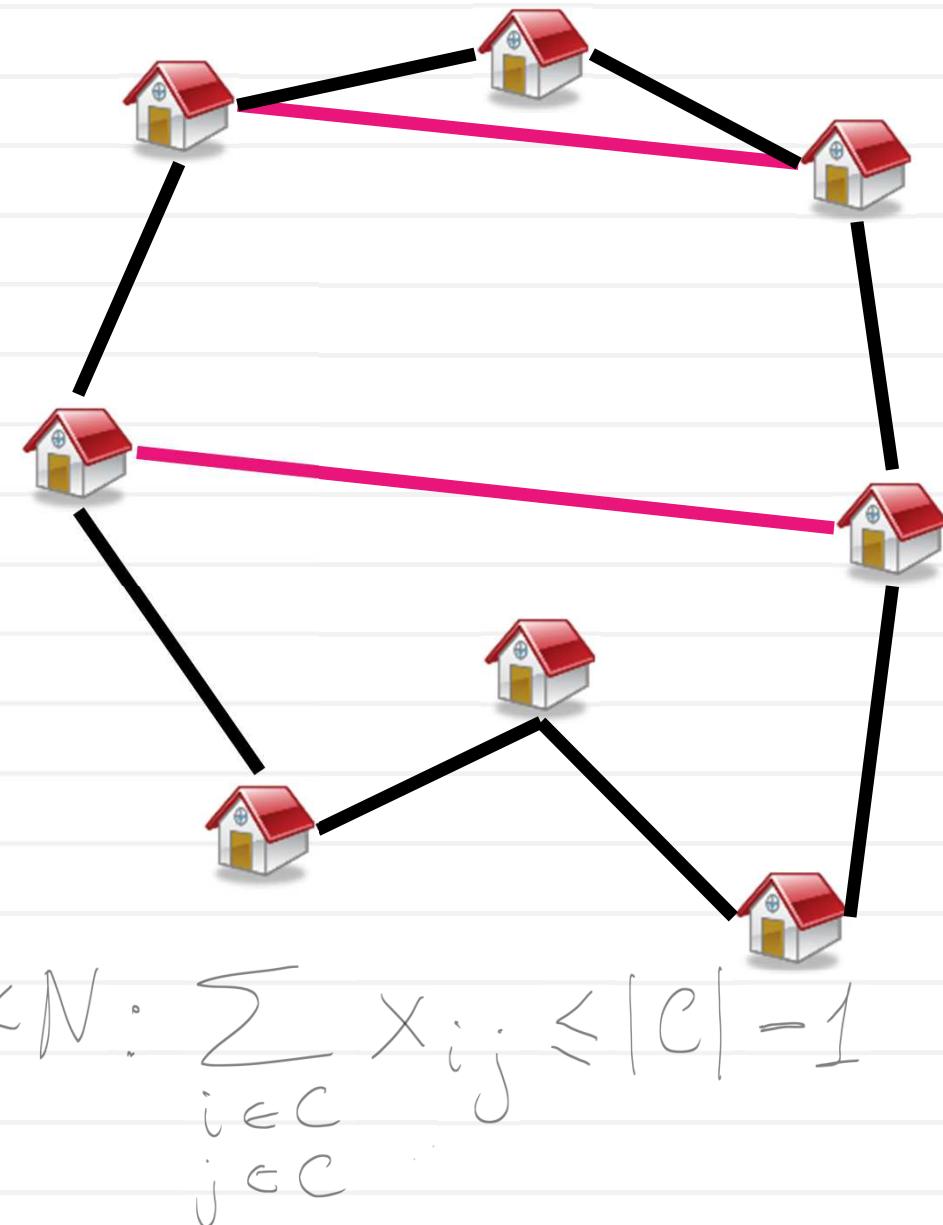
$$\forall i: \sum_j x_{ij} = 2$$

$$\sum_{i,j} x_{ij} = N$$

$$\forall c \subset \{1..N\} \quad 0 < |c| < N: \sum_{\substack{i \in c \\ j \in c}} x_{ij} \leq |c| - 1$$

Exponentially many
constraints!

(assuming that x_{ij} and x_{ji} are the same)



Travelling salesman problem

$$\min \sum_{i,j} w_{ij} x_{ij}$$

$$\text{s.t.: } x_{ij} \in \{0;1\}$$

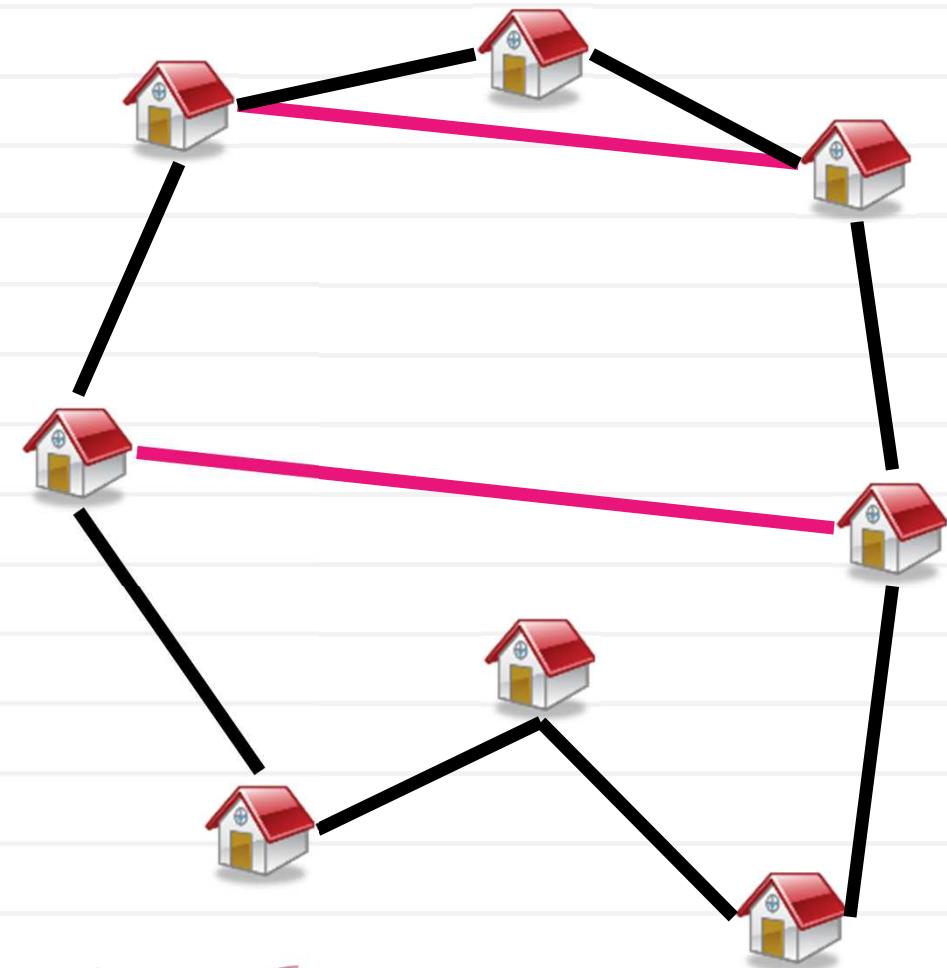
$$\forall i: \sum_j x_{ij} = 2$$

$$\sum_i x_{ij} = N$$

$$\forall C \subset \{1 \dots N\} \quad 0 < |C| < N: \sum_{\substack{i \in C \\ j \notin C}} x_{ij} \geq 2$$

Exponentially many
constraints!

(assuming that x_{ij} and x_{ji} are the same)

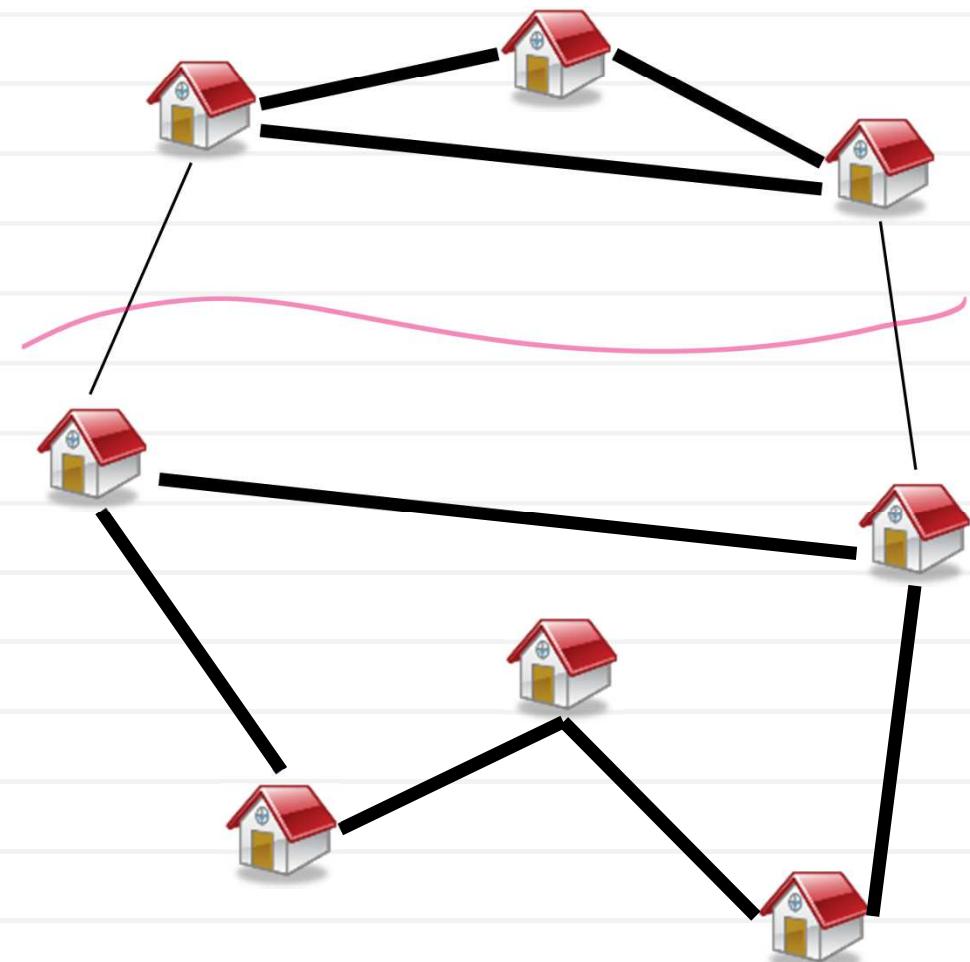


Finding Maximally Violated Constraint

$$\forall c \subset \{1 \dots N\} \quad 0 < |c| < N: \sum_{\substack{i \in c \\ j \notin c}} x_{ij} \geq 2$$

$$\min_c \sum_{\substack{i \in c \\ j \notin c}} x_{ij}$$

- (NP-)hard for arbitrary x_{ij}
- For non-negative x_{ij} polynomial algorithms exist



TSP vs MWIS

- In MWIS the integer program has a “small” number of constraints, we add extra constraint to tighten the LP relaxation
- In TSP, the integer program and its LP relaxation already has a huge number of constraints
- Still, we can use the cutting plane algorithm to solve the LP relaxation (*delayed constraint generation*)

Delayed constraint generation

This pseudocode solves the LP relaxation of the TSP
(not the TSP itself):

```
[p,A,b,C,d] = LPRelaxation(TSP(G),  
                           some subtour constraints)
```

```
x = SolveLP(p,A,b,C,d)
```

```
while x not integer && x not a tour
```

```
    Set = MinCut(x, G)
```

```
    if cut(Set,x) ≥ 2
```

```
        break
```

```
    [A,b] = AddCutSetConstraint(A,b);
```

```
    x = SolveLP(p,A,b,C,d)
```

```
end
```

To solve the TSP, we then need to add further cutting planes that will tighten the relaxation (e.g. *comb* cuts)

LP relaxation

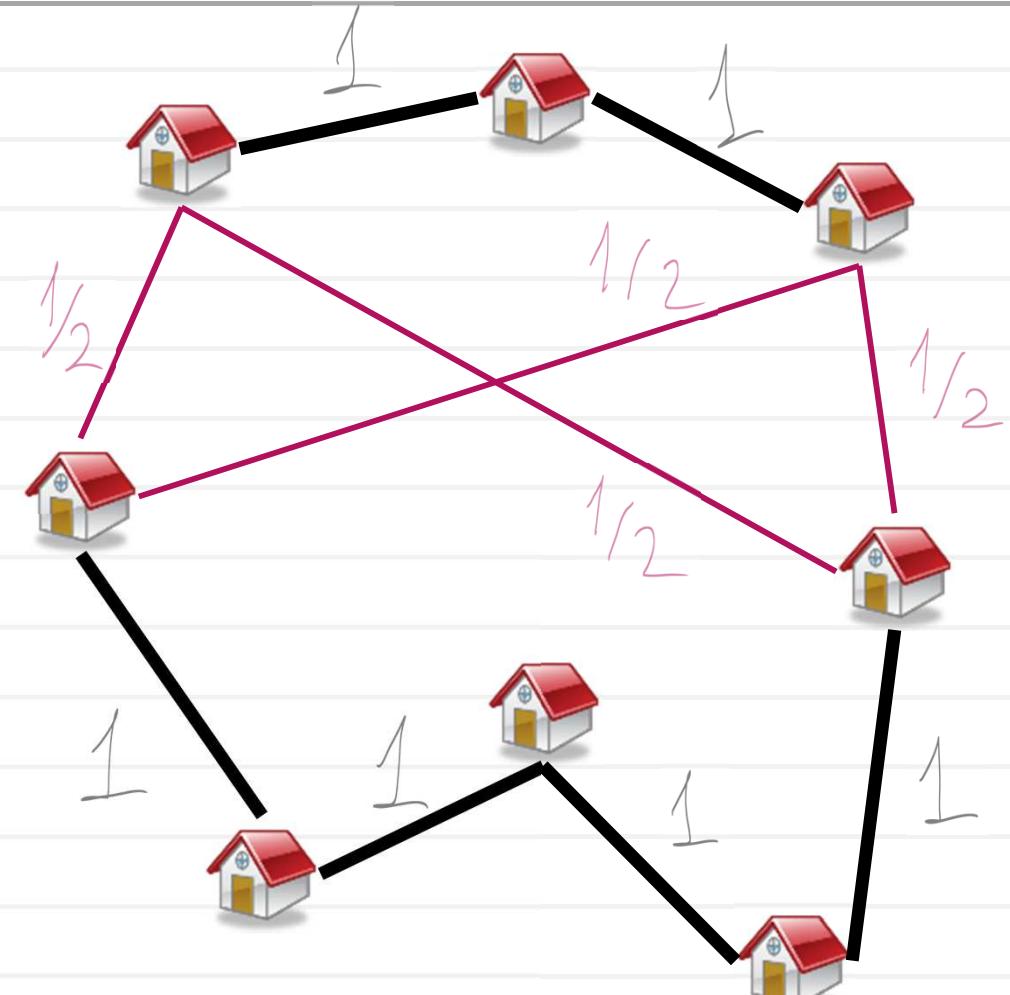
Even with all constraints in, the LP relaxation might give a fractional solution:

$$0 \leq x_{ij} \leq 1$$

$$\forall i: \sum_j x_{ij} = 2$$

$$\sum_i x_{ij} = N$$

$$\forall c \subset \{1..N\} \quad 0 < |c| < N: \sum_{\substack{i \in c \\ j \notin c}} x_{ij} \geq 2$$



Types of the tightening cuts

Problem specific (work best!):

- Clique cuts for MWIS
- (Subtour elimination constraints for TSP)
- *Comb* cuts for TSP
-

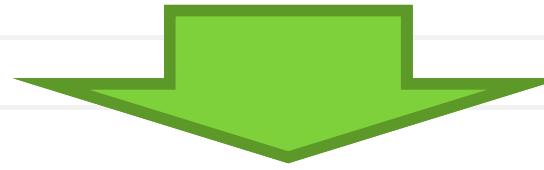
Generic cuts:

- Gomory cuts
- Cover cuts
-

Cover cut

Generic family of cutting planes, that can be derived directly from the problem

$$1x_1 + 5x_2 + 4x_3 + 3x_4 \leq 7 \quad (*)$$
$$x_i \in \{0;1\}$$


$$5 + 4 > 7$$
$$5 + 3 > 7$$

$$x_2 + x_3 \leq 1$$

$$x_2 + x_4 \leq 1$$

Furthermore: $x_1 + x_3 + x_4 \leq 2$

$\{2;3\}, \{2;4\}, \{1;3;4\}$ are *covers* w.r.t. (*)

Cover cut

Given original inequality:

$$1x_1 + 5x_2 + 4x_3 + 3x_4 \leq 7$$

Any $\alpha_1, \alpha_2, \alpha_3, \alpha_4$: $\alpha_1 + 5\alpha_2 + 4\alpha_3 + 3\alpha_4 \geq 8$
 $\alpha_i \in \{0; 1\}$

α is a cover

Gives a cutting plane:

not all variables are on

$$\alpha_1 x_1 + \alpha_2 x_2 + \alpha_3 x_3 + \alpha_4 x_4 \leq \alpha_1 + \alpha_2 + \alpha_3 + \alpha_4 - 1$$

$$\alpha_1 (x_1 - 1) + \alpha_2 (x_2 - 1) + \alpha_3 (x_3 - 1) + \alpha_4 (x_4 - 1) \leq -1$$

Cover cut

Given original inequality:

$$1x_1 + 5x_2 + 4x_3 + 3x_4 \leq 7$$

Any $\alpha_1, \alpha_2, \alpha_3, \alpha_4$: $\alpha_1 + 5\alpha_2 + 4\alpha_3 + 3\alpha_4 \geq 8$

$$\alpha_i \in \{0; 1\} \quad \beta_i = 1 - \alpha_i \in \{0; 1\}$$

$$1 - \beta_1 + 5(1 - \beta_2) + 4(1 - \beta_3) + 3(1 - \beta_4) \geq 8$$

$$\beta_1 + 5\beta_2 + 4\beta_3 + 3\beta_4 \leq 13 - 8 = 5$$

Gives a cutting plane:

$$\alpha_1(x_1 - 1) + \alpha_2(x_2 - 1) + \alpha_3(x_3 - 1) + \alpha_4(x_4 - 1) \leq -1$$

$$(1 - x_1)\beta_1 + (1 - x_2)\beta_2 + (1 - x_3)\beta_3 + (1 - x_4)\beta_4 \leq 3 - \sum x_i$$

Separation oracle for cover cuts

If $\beta_1 + 5\beta_2 + 4\beta_3 + 3\beta_4 \leq 5$ Then:

$$(1-x_1)\beta_1 + (1-x_2)\beta_2 + (1-x_3)\beta_3 + (1-x_4)\beta_4 \leq 3 - \sum x_i$$

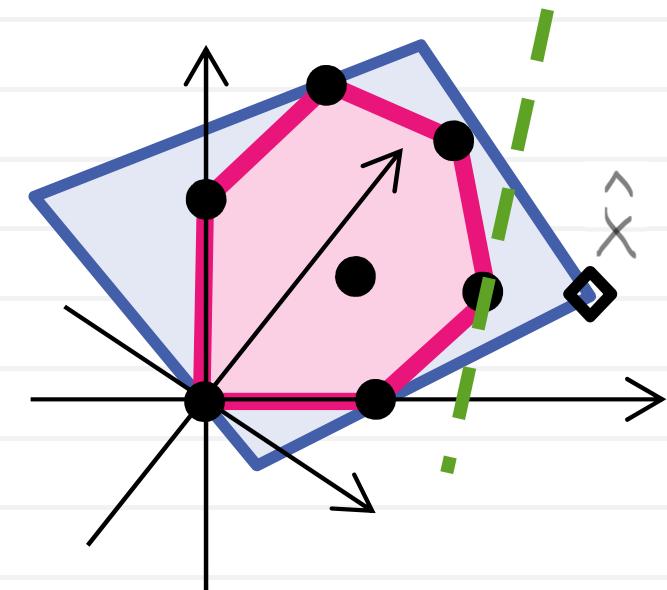
Let \hat{x} be the current solution. Let us find the “deepest cut” that separates it from the bounding box:

$$\max_{\beta} (1-\hat{x}_1)\beta_1 + (1-\hat{x}_2)\beta_2 + (1-\hat{x}_3)\beta_3 + (1-\hat{x}_4)\beta_4$$

$$\text{s.t. } \beta_1 + 5\beta_2 + 4\beta_3 + 3\beta_4 \leq 5$$

$$\beta_i \in \{0; 1\}$$

If objective $\geq 3 - \sum \hat{x}_i$
then we have cut the
current solution away



Gomory cut example

$$1.5x_1 + 4.3x_2 + 2.2x_3 + 7.5x_4 = 111.3$$

$$x_i \in \mathbb{Z}_+$$

$$1x_1 + 4x_2 + 2x_3 + 7x_4 +$$

$$(0.5x_1 + 0.3x_2 + 0.2x_3 + 0.5x_4) = 111 + 0.3$$

The red part cannot be less than 0.3, or otherwise the inequality cannot hold.

Gomory cut (in this case):

$$0.5x_1 + 0.3x_2 + 0.2x_3 + 0.5x_4 \geq 0.3$$

Gomory cuts: general formula

$$\alpha_1 x_1 + \alpha_2 x_2 + \alpha_3 x_3 + \dots + \alpha_n x_n = \beta$$
$$x_i \in \mathbb{Z}_+$$
$$\lfloor \alpha_1 \rfloor x_1 + \lfloor \alpha_2 \rfloor x_2 + \dots + \lfloor \alpha_n \rfloor x_n +$$
$$+ (\alpha_1 - \lfloor \alpha_1 \rfloor) x_1 + \dots + (\alpha_n - \lfloor \alpha_n \rfloor) x_n =$$
$$= \lfloor \beta \rfloor + (\beta - \lfloor \beta \rfloor)$$

General Gomory cut:

$$(\alpha_1 - \lfloor \alpha_1 \rfloor) x_1 + \dots + (\alpha_n - \lfloor \alpha_n \rfloor) x_n \geq \beta - \lfloor \beta \rfloor$$

Important: the equality can be a linear combination of initial equalities

Cutting plane algorithm

```
function x = SolveILPCuttingPlane(p, DILP)
[p,A,b,C,d] = LPRelaxation(ILP)
x = SolveLP(p,A,b,C,d)
while x not integer
    [A' b'] = FindSeparatingCut(s) (x, DILP)
    if [A' b'] empty
        break
    A = [A;A'], b = [b,b']
    x = SolveLP(p,A,b,C,d)
end
```

Branch and Bound

Initialize the list L to the root node

$incumbent_value = +\infty$

Until L is empty or time is out

 Pick the active node from L

 Pick the variable to branch

For each of the two children nodes

$[x \ obj] = \text{Solve LP (node)}$

If $obj \geq incumbent_obj$ **then continue;**

If x is integer **then**

$incumbent = x; incumbent_value = obj;$

Else add node to L

end

end

Branch-and-Cut

Initialize the list L to the root node

$incumbent_value = +\infty$

Until L is empty or time is out

 Pick the active node from L

 Pick the variable to branch

For each of the two children nodes

$[x \ obj] = \text{SolveILPCuttingPlane}(\text{node})$

If $obj \geq incumbent_obj$ **then continue;**

If x is integer **then**

$incumbent = x; incumbent_value = obj;$

Else add node to L

end

end