# Mixed Integer Linear Programming

Which is sort of famous, for valid reasons

# Mixed Integer Linear Programming

## A Mixed Integer Linear Program is a problem in the form

$$\operatorname{argmin}_{x} \left\{ c^{T} x \mid Ax \geq b, x \geq 0, x_{I} \in \mathbb{Z} \right\}$$

- The cost function and all constraints are linear
- All variables are non-negative
- lacksquare Some variables (those with index in  $m{I}$ ) are integer

#### MILP is an extremely powerful formalism

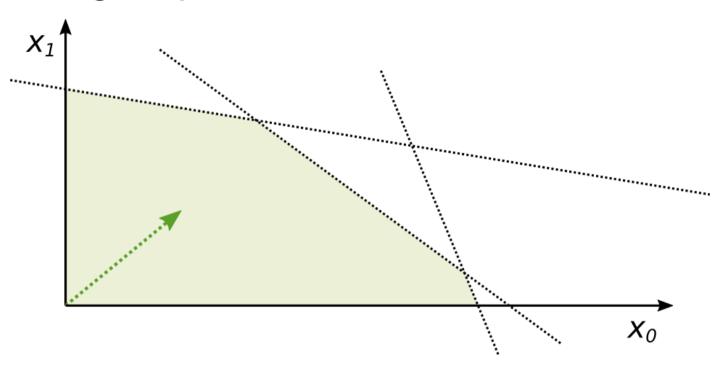
Thanks to the presence of integer variables

- ...Any combinatorial element can be modeled
- ...And non-linearity can be approximated

## MILP solvers classically rely on three main techniques

### **Linear Relaxation**

If we remove the integrality constraints from a MILP we obtain an LP



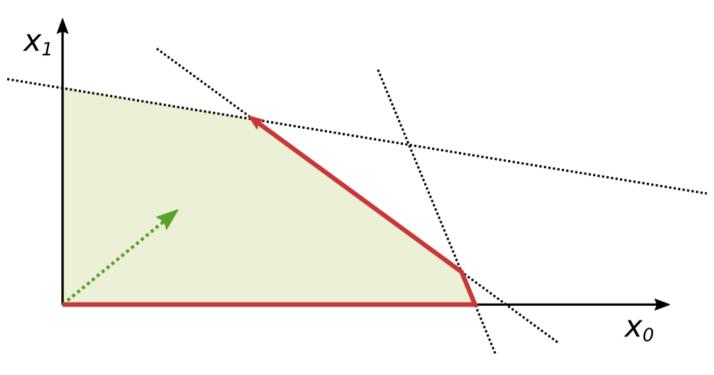
$$\operatorname{argmin}_{x} \left\{ c^{T} x \mid Ax \ge b, x \ge 0 \right\}$$

### This is called the linear (or LP) relaxation of the MILP

- The feasible space is defined via linear constraints  $\Rightarrow$  is is a polytope
- lacktriangle The cost vector  $oldsymbol{c}$  is also the gradient and determines an optimization direction

### **Linear Relaxation**

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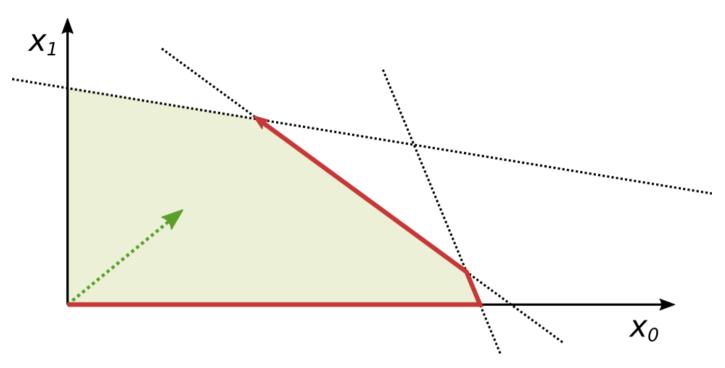
$$\operatorname{argmin}_{x} \left\{ c^{T} x \mid Ax \ge b, x \ge 0 \right\}$$

## LPs can be solved in pseudo-polynomial time via the **Simplex method**

- The method start from a polytope vertex
- ...And then moves between adjiacent vertexes until the optimum is reached

### **Linear Relaxation**

If we remove the integrality constraints from a MILP we obtain an LP



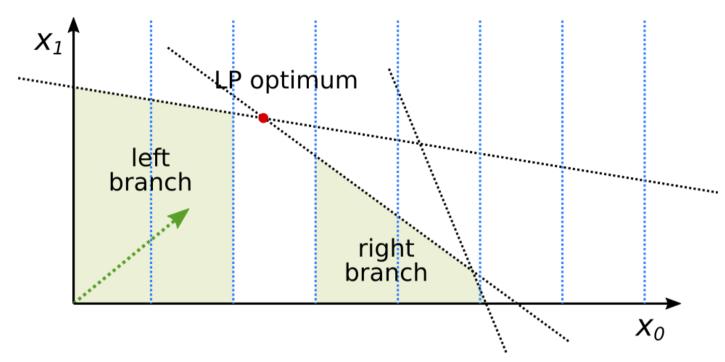
$$\operatorname{argmin}_{x} \left\{ c^{T} x \mid Ax \ge b, x \ge 0 \right\}$$

## LPs can be solved in polynomial time via Interior Point methods

- These used to be slower in practice than the Simplex, but not anymore
- In a MILP complex, the Simplex is still preferred (later we will see why)

# Technique #1: Branching

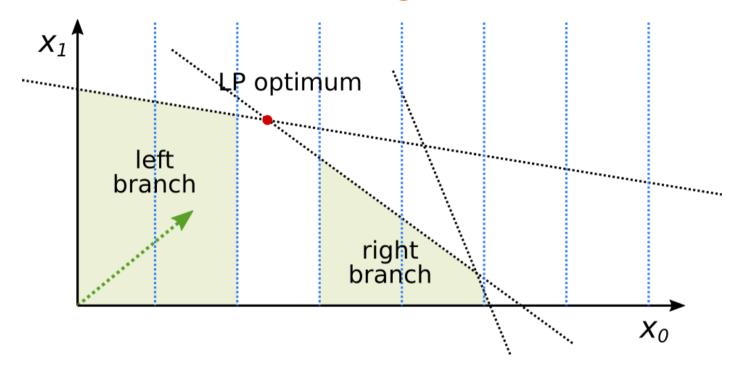
# When tackling a MILP, we start by solving its LP relaxation



- If all integrality constraints are satisfied, we have found the true optimum
- If some  $x_j$  has a fraction value  $v_j$ , we split the problem in two:
  - ullet In the first subproblem, we add the constraint  $x_j \leq \lfloor v_j 
    floor$
  - In the second subproblem, we add  $x_j \geq \lceil v_j \rceil$
- Then we can repeat the whole process

# Technique #1: Branching

## This approach is referred to as branching



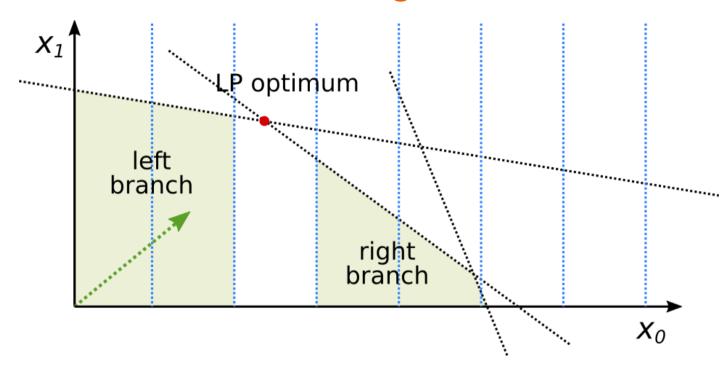
- The first subproblem is also known as the left branch
- The second as the right branch

## Branching is the main method that makes MILP solvers complete

- In the worst case, we end up with a search tree having an exponential number of nodes
- ...But that's somewhat unavoidable, since solving MILP is NP-hard

# **Technique #1: Branching**

### This approach is referred to as branching



## Branching is also the reason why the Simplex method is preferred to MILPs

- The Simplex method has a "dual" version
- ...Whose optimum can be updated efficiently when new constraints are added

...And you can guess that's a pretty common operation ;-)

# Technique #2: Bounding

## Let's look again at the LP relaxation

$$\operatorname{argmin}_{x} \left\{ c^{T} x \mid Ax \ge b, x \ge 0 \right\}$$

- The problem has the same structure
- ...But a larger feasible space (that's why it is called a relaxation)

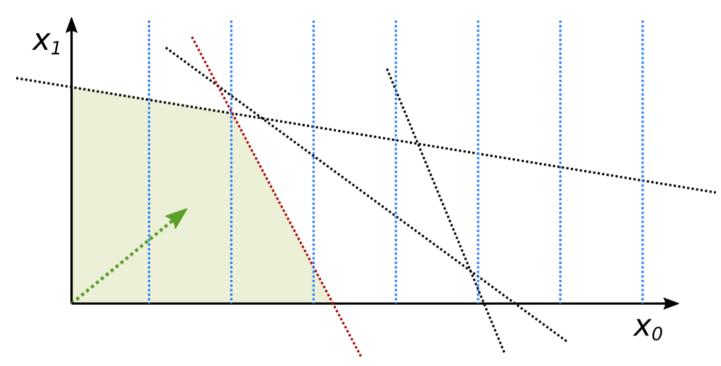
Hence, its optimal cost will be a lower bound (say lb) for the MILP

## We can use this bound as an early stopping criterion

- Let  $x^*$  be the best (mixed-integer) solution we have found so far
- If for some node of the search tree we have  $lb>c^Tx^*$
- Then we have no hope of beating  $x^*$  and we can destroy (fathom) the node

# **Branching + Bounding = Branch & Bound**

It is also common to speed-up MILP solution by using cutting planes



# Cutting planes are linear inequalities inferred by relying on some property

- In MILP they are typically inferred based on integrality constraint
- They must be valid for any feasible solution
- They are useful if they force a fractional solution to become closer to integer

#### A common example is that of Gomory Cuts

While solving the simplex, we end up with many equalities in the form:

$$x_i + \sum_{j \in \bar{B}} \bar{a}_{ij} x_j = \bar{b}_i$$

- lacksquare Where  $x_i>0$  and  $x_j=0, \forall j\in ar{B}$
- $\blacksquare$  **B** = the set of indexes of non-zero variables in the current LP solution (base)
- $ar{B}$  = the set of indexes of zero variables in the current LP solution
- We will assume all variables are integer, for simplicity

## We can rewrite the equation as

$$x_i + \sum_{i \in \bar{B}} (\bar{a}_{ij} - \lfloor \bar{a}_{ij} \rfloor + \lfloor \bar{a}_{ij} \rfloor) x_j = \bar{b}_i - \lfloor \bar{b}_i \rfloor + \lfloor \bar{b}_i \rfloor$$

## By simple algebraic manipulation we can then get:

$$x_i + \sum_{j \in \bar{B}} \lfloor \bar{a}_{ij} \rfloor x_j - \lfloor \bar{b}_i \rfloor = -\sum_{j \in \bar{B}} (\bar{a}_{ij} - \lfloor \bar{a}_{ij} \rfloor) x_j + (\bar{b}_i - \lfloor \bar{b}_i \rfloor)$$

We will build an inequality that is valid for any feasible, integer point:

- The right-most part is necessarily < 1, since:
  - $ar{b}_i ar{b}_i$ ] is positive and fractional
  - Each  $\bar{a}_{ij}$   $\lfloor \bar{a}_{ij} \rfloor$  is positive (and fractional)
  - Each  $x_j$  must be  $\geq 0$
- The left-most part is necessarily an integer, since:
  - lackbrack lackbrac
  - Variables are integer as per our assumption

By simple algebraic manipulation we can then get:

$$x_i + \sum_{j \in \bar{B}} \lfloor \bar{a}_{ij} \rfloor x_j - \lfloor \bar{b}_i \rfloor = -\sum_{j \in \bar{B}} (\bar{a}_{ij} - \lfloor \bar{a}_{ij} \rfloor) x_j + (\bar{b}_i - \lfloor \bar{b}_i \rfloor)$$

- ullet Hence, the right-most part should be < 1 and integer
- ... Meaning that it must be  $\leq 0$

$$-\sum_{j\in\bar{B}}(\bar{a}_{ij}-\lfloor\bar{a}_{ij}\rfloor)x_j+(\bar{b}_i-\lfloor\bar{b}_i\rfloor)\leq 0$$

And from here:

$$\sum_{j \in \bar{B}} (\bar{a}_{ij} - \lfloor \bar{a}_{ij} \rfloor) x_j \ge (\bar{b}_i - \lfloor \bar{b}_i \rfloor)$$

# This inequality is the Gomory Cut

$$\sum_{j \in \bar{B}} (\bar{a}_{ij} - \lfloor \bar{a}_{ij} \rfloor) x_j \ge (\bar{b}_i - \lfloor \bar{b}_i \rfloor)$$

- Now, if we target a  $x_i$  that should be integer, i.e.  $i \in I$
- ...But it's fractional in the current solution

Then we have  $ar{b}_i - \lfloor ar{b}_i \rfloor > 0$ 

- lacktriangle Combined with the fact that  $x_j=0, \forall j\in ar{B}$  in the current solution
- We have that the cut is actually making the solution no longer feasible

# Branching + Bounding + Cutting Planes = Branch & Cut

- Using cutting planes can speed up the solution process considerably
- But it's best not to overdo it, since subsequent cuts may become weaker

#### **Some Considerations**

## We have just scratched the surface with MILP

Modern MILP solver do much more:

- Presolving
- Constraint propagations
- Symmetry breaking
- •

#### MILP methods have a long history

- There is a huge gap between the solver performance
- The best solvers (Gurobi, Cplex, Mosek) are commercial (free for academics)
- Then you have a single semi-free solver (<u>SCIP</u>)
- ...A good free solver (<u>CBC</u>)
- ...And finally there is stuff you should not touch (glpk, lpsolve)