

Integer Programming

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Revision: Mixed Integer Programming Problems

Mixed Integer Programming (MIP) Problem:

$$\min x_0 = c^T x$$

subject to

$$Ax = b$$

$$x_j \geq 0 \quad \text{for } j \in N = \{1, \dots, n\}$$

$$x_j \in \mathbb{Z} \quad \text{for } j \in Z \subseteq N.$$

Note: $x_j \in N \setminus Z$ are continuous, $x_j \in Z$ are integral.

Additional assumption this lecture: Finite bounds $\underline{x}_j, \bar{x}_j$ for $j \in Z$: $x_j \in \{\underline{x}_j, \underline{x}_j + 1, \dots, \bar{x}_j\}$.

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Solving MIP's: Complete Enumeration

Idea: Loop through all possible values of the integer variables (without loss of generality, $\{x_1, \dots, x_z\}$ with $z \leq n$) and solve LP problems in the remaining (continuous) variables:

```
for  $x_1 \in \{\underline{x}_1, \underline{x}_1 + 1, \dots, \bar{x}_1\}$  do
  for  $x_2 \in \{\underline{x}_2, \underline{x}_2 + 1, \dots, \bar{x}_2\}$  do
    ...
    for  $x_z \in \{\underline{x}_z, \underline{x}_z + 1, \dots, \bar{x}_z\}$  do
      Solve LP in  $x_{k+1}, \dots, x_n$  with  $x_1, \dots, x_k$  fixed.
      Update tentative optimal solution if necessary.
    end for
  end for
end for
Print (final) optimal solution or report infeasibility.
```

Complexity?

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Divide and Conquer Principle

Need to *structure* search so that we touch only few solutions.

One Approach: Divide and Conquer

- ▶ *Divide* a large problem into several smaller ones.
- ▶ *Conquer* by working on the smaller problems.

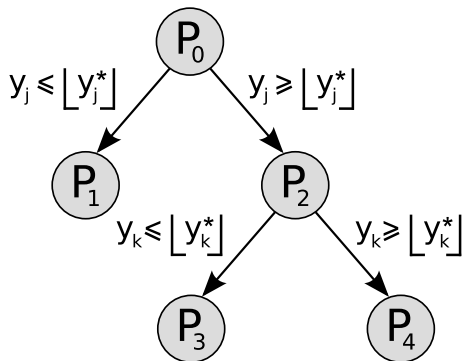
Branch & Bound:

- ▶ Solve continuous relaxation of original problem $P_0 \Rightarrow x^*(P_0)$.
- ▶ **Divide (Branch):** Choose $p \in Z$ with $x_p^* \notin \mathbb{Z}$. Create two subproblems, P_1 and P_2 , with added constraints $x_p \leq \lfloor x_p^* \rfloor$ and $x_p \geq \lceil x_p^* \rceil$, respectively.
- ▶ **Conquer (Bound/Fathom):** If optimal solution of continuous relaxation of P_i is worse than any known feasible solution for P_0 , disregard P_i .

Note: Any solution to P_0 is also feasible for *either* P_1 *or* P_2 .
Hence, by solving P_1 *and* P_2 , we solve P_0 .

Divide and Conquer Principle

Recursive application of divide and conquer principle leads to binary tree:



Terminal nodes = problems that remain to be solved.

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Branch & Bound Algorithm for MIP's

Preliminaries:

- ▶ P_0 denotes original problem:

$$\min x_0 = c^T x$$

subject to

$$Ax = b$$

$$x_j \geq 0 \quad \text{for } j \in N = \{1, \dots, n\}$$

$$x_j \in \mathbb{Z} \quad \text{for } j \in Z \subseteq N.$$

- ▶ For any problem P , $x^*(P)$ denotes optimal solution for continuous relaxation of P .
- ▶ OPT denotes objective function value of best *feasible* solution (for P_0) found so far. At beginning, $OPT = \infty$ or based on a priori knowledge (heuristic).

Branch & Bound Algorithm for MIP's

Algorithm:

1. Initialization.

- ▶ Set list of problems to $\{P_0\}$. Initialize OPT .
- ▶ Solve LP relaxation of $P_0 \Rightarrow x^*(P_0)$.
- ▶ If $x^*(P_0)$ feasible for P_0 , $OPT = c^T x^*(P_0)$ and stop.

2. Problem Selection.

Choose a problem P from list whose $x^*(P)$ has $c^T x^*(P) < OPT$. If no such P exists, stop.

3. Variable Selection.

Choose $x_p \in Z$ with $x_p^*(P) \notin \mathbb{Z}$.

4. Branching.

- ▶ Create two new problems P' and P'' with $x_p \leq \lfloor x_p^*(P) \rfloor$ and $x_p \geq \lceil x_p^*(P) \rceil$, respectively.
- ▶ Solve continuous relaxations of P' and $P'' \Rightarrow x^*(P')$, $x^*(P'')$.
- ▶ **Update OPT :** If P' feasible, $x^*(P')$ feasible for P_0 and $c^T x^*(P') < OPT \Rightarrow OPT = c^T x^*(P')$. Same for P'' .
- ▶ **Further Inspection:** If P' feasible and $c^T x^*(P') < OPT \Rightarrow$ add P' to list of problems. Same for P'' .

Afterwards, go back to (2).

Branch & Bound Algorithm for MIP's

Output:

- ▶ $OPT = \infty$: P_0 is infeasible.
- ▶ $OPT < \infty$: P_0 is feasible. $OPT =$ optimal objective value.

Optimal Solution:

 Obtained via slight modification

- ▶ Store vector \hat{x} for best feasible solution (for P_0) found so far.
- ▶ Whenever OPT is updated (Steps 1+4), also update \hat{x} .

Termination: Under assumption of finite bounds $\underline{x}_j, \bar{x}_j$ for $j \in Z$, algorithm terminates in finitely many steps.

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Example

Example

Assume the following problem is given:

$$\max 2x_1 + 3x_2 + x_3 + 2x_4$$

subject to

$$5x_1 + 2x_2 + x_3 + x_4 \leq 15$$

$$2x_1 + 6x_2 + 10x_3 + 8x_4 \leq 60$$

$$x_1 + x_2 + x_3 + x_4 \leq 8$$

$$2x_1 + 2x_2 + 3x_3 + 3x_4 \leq 16.$$

The bounds are $x_1 \in [0, 3]$, $x_2 \in [0, 7]$, $x_3 \in [0, 5]$ and $x_4 \in [0, 5]$.
Furthermore, $x_j \in \mathbb{Z}$ for all $j = 1, \dots, 4$.

Example

Change to minimization objective (not necessary!):

$$\min -2x_1 - 3x_2 - x_3 - 2x_4$$

subject to

$$5x_1 + 2x_2 + x_3 + x_4 \leq 15$$

$$2x_1 + 6x_2 + 10x_3 + 8x_4 \leq 60$$

$$x_1 + x_2 + x_3 + x_4 \leq 8$$

$$2x_1 + 2x_2 + 3x_3 + 3x_4 \leq 16.$$

$x_1 \in [0, 3]$, $x_2 \in [0, 7]$, $x_3 \in [0, 5]$ and $x_4 \in [0, 5]$. $x_j \in \mathbb{Z}$ for all $j = 1, \dots, 4$.

Example

1. Initialization.

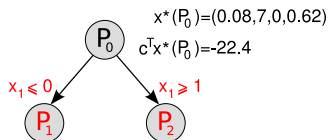
- ▶ Set list of problems to $\{P_0\}$. Initialize OPT .
- ▶ Solve LP relaxation of $P_0 \Rightarrow x^*(P_0)$.
- ▶ If $x^*(P_0)$ feasible for P_0 , $OPT = c^T x^*(P_0)$ and stop.

$$\textcircled{P_0} \quad \begin{array}{l} x^*(P_0) = (0.08, 7, 0, 0.62) \\ c^T x^*(P) = -22.4 \end{array}$$

Problem list: $\{P_0\}$, $OPT = \infty$.

Example

- Problem Selection.** Choose a problem P from list whose $x^*(P)$ has $c^T x^*(P) < OPT$. If no such P exists, stop.
- Variable Selection.** Choose $x_p \in Z$ with $x_p^*(P) \notin \mathbb{Z}$.
- Branching.**
 - ▶ Create two new problems P' and P'' with $x_p \leq \lfloor x_p^*(P) \rfloor$ and $x_p \geq \lceil x_p^*(P) \rceil$, respectively.

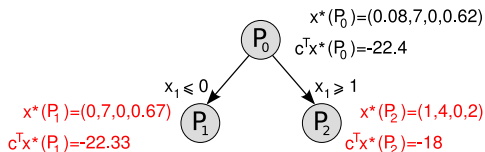


Problem list: $\{P_1, P_2\}$, $OPT = \infty$.

Example

1. Branching.

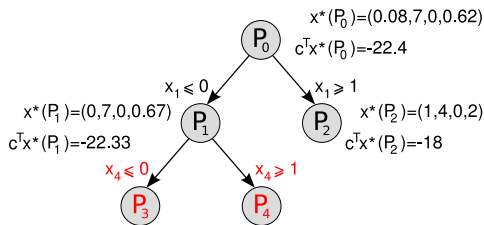
- ▶ Solve continuous relaxations of P' and $P'' \Rightarrow x^*(P')$, $x^*(P'')$.
- ▶ **Update OPT :** If P' feasible, $x^*(P')$ feasible for P_0 and $c^T x^*(P') < OPT \Rightarrow OPT = c^T x^*(P')$. Same for P'' .
- ▶ **Further Inspection:** If P' feasible and $c^T x^*(P') < OPT \Rightarrow$ add P' to list of problems. Same for P'' .



Problem list: $\{P_1\}$, $OPT = -18$.

Example

- Problem Selection.** Choose a problem P from list whose $x^*(P)$ has $c^T x^*(P) < OPT$. If no such P exists, stop.
- Variable Selection.** Choose $x_p \in Z$ with $x_p^*(P) \notin \mathbb{Z}$.
- Branching.**
 - ▶ Create two new problems P' and P'' with $x_p \leq \lfloor x_p^*(P) \rfloor$ and $x_p \geq \lceil x_p^*(P) \rceil$, respectively.

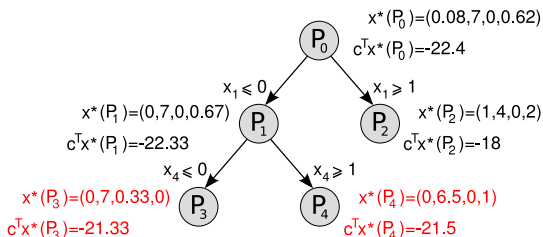


Problem list: $\{P_3, P_4\}$, $OPT = -18$.

Example

1. Branching.

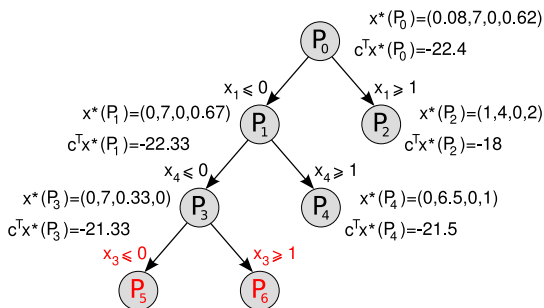
- ▶ Solve continuous relaxations of P' and $P'' \Rightarrow x^*(P')$, $x^*(P'')$.
- ▶ **Update OPT :** If P' feasible, $x^*(P')$ feasible for P_0 and $c^T x^*(P') < OPT \Rightarrow OPT = c^T x^*(P')$. Same for P'' .
- ▶ **Further Inspection:** If P' feasible and $c^T x^*(P') < OPT \Rightarrow$ add P' to list of problems. Same for P'' .



Problem list: $\{P_3, P_4\}$, $OPT = -18$.

Example

- Problem Selection.** Choose a problem P from list whose $x^*(P)$ has $c^T x^*(P) < OPT$. If no such P exists, stop.
- Variable Selection.** Choose $x_p \in Z$ with $x_p^*(P) \notin \mathbb{Z}$.
- Branching.**
 - ▶ Create two new problems P' and P'' with $x_p \leq \lfloor x_p^*(P) \rfloor$ and $x_p \geq \lceil x_p^*(P) \rceil$, respectively.

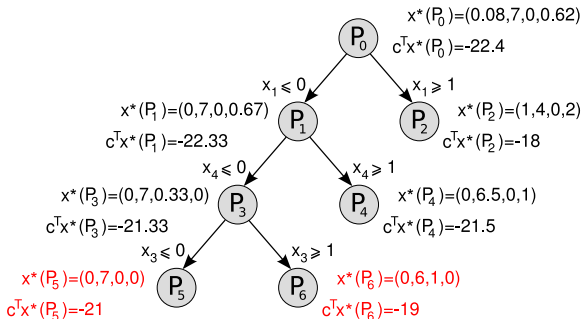


Problem list: $\{P_4, P_5, P_6\}$, $OPT = -18$.

Example

1. Branching.

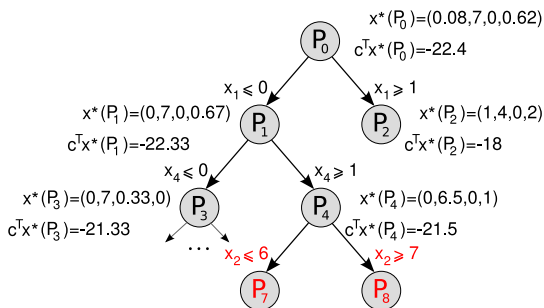
- ▶ Solve continuous relaxations of P' and $P'' \Rightarrow x^*(P')$, $x^*(P'')$.
- ▶ **Update OPT:** If P' feasible, $x^*(P')$ feasible for P_0 and $c^T x^*(P') < OPT \Rightarrow OPT = c^T x^*(P')$. Same for P'' .
- ▶ **Further Inspection:** If P' feasible and $c^T x^*(P') < OPT \Rightarrow$ add P' to list of problems. Same for P'' .



Problem list: $\{P_4\}$, $OPT = -21$.

Example

- Problem Selection.** Choose a problem P from list whose $x^*(P)$ has $c^T x^*(P) < OPT$. If no such P exists, stop.
- Variable Selection.** Choose $x_p \in Z$ with $x_p^*(P) \notin \mathbb{Z}$.
- Branching.**
 - ▶ Create two new problems P' and P'' with $x_p \leq \lfloor x_p^*(P) \rfloor$ and $x_p \geq \lceil x_p^*(P) \rceil$, respectively.

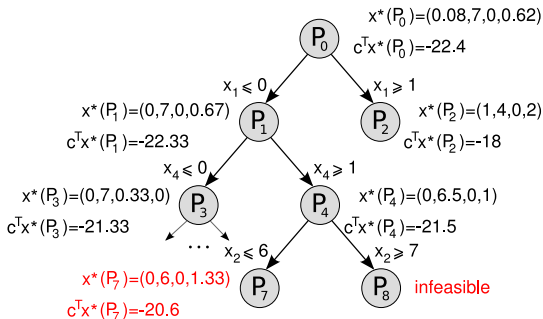


Problem list: $\{P_7, P_8\}$, $OPT = -21$.

Example

1. Branching.

- ▶ Solve continuous relaxations of P' and $P'' \Rightarrow x^*(P')$, $x^*(P'')$.
- ▶ **Update OPT:** If P' feasible, $x^*(P')$ feasible for P_0 and $c^T x^*(P') < OPT \Rightarrow OPT = c^T x^*(P')$. Same for P'' .
- ▶ **Further Inspection:** If P' feasible and $c^T x^*(P') < OPT \Rightarrow$ add P' to list of problems. Same for P'' .



Problem list: $\{\}$, $OPT = -21$. Done; $\hat{x} = (0, 7, 0, 0)$.