

# Computational Integer Programming

## Universidad de los Andes

### Lecture 3

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## References for This Lecture

- Wolsey Chapter 2
- Nemhauser and Wolsey Sections II.3.1, II.3.6, II.4.1, II.4.2, II.5.4

## Relaxation

For simplicity, we now consider a pure integer program  $IP$  defined by

$$\begin{aligned}z_{IP} &= \max\{cx \mid x \in S\}, \\ S &= \{x \in \mathbb{Z}_+^n \mid Ax \leq b\}.\end{aligned}$$

**Definition 1.** A **relaxation** of  $IP$  is a maximization problem defined as

$$z_R = \max\{z_R(x) \mid x \in S_R\}$$

*with the following two properties:*

$$S \subseteq S_R \tag{1}$$

$$cx \leq z_R(x), \forall x \in S. \tag{2}$$

## Importance of Relaxations

- The main purpose of a relaxation is to obtain an **upper bound** on  $z_{IP}$ .
- Relaxation is used as a method of bounding in branch and bound.
- The idea is to choose a relaxation that is much easier to solve than the original problem.
- Note that the relaxation **must be solved to optimality** to yield a valid bound.
- We first consider three basic types of relaxations.
  - LP relaxation
  - Combinatorial relaxation
  - Lagrangian relaxation
- Relaxations are also used in some other bounding schemes we will look at.

## Aside: How Do You Spell “Lagrangian?”

- Some spell it “Lagrangean.”
- Some spell it “Lagrangian.”
- We ask [Google](#).
- In 2002:
  - “Lagrangean” returned 5,620 hits.
  - “Lagrangian” returned 14,3000 hits.
- In 2007:
  - “Lagrangean” returns 208,000 hits.
  - “Lagrangian” returns 5,820,000 hits.
- In 2010:
  - “Lagrangean” returns 110,000 hits (and asks “Did you mean: **Lagrangian?**”)
  - “Lagrangian” returns 2,610,000 hits.
- “Lagrangian” still wins!

## Obtaining and Using Relaxations

- An important additional property of relaxations is that if a relaxation of  $IP$  is infeasible, then so is  $IP$ .
- If the objective function of the relaxation is the same as the original, then if the optimal solution to the relaxation is feasible for  $IP$ , it is optimal for  $IP$ .
- The easiest way to obtain relaxations of  $IP$  is to **drop some of the constraints** defining the feasible set  $S$ .
- We have two choices
  - LP relaxation: Drop the integrality constraints to obtain an LP.
  - Combinatorial relaxation: Drop a set of inequality constraints that make the resulting IP “easy.”
- It is “obvious” how to obtain an LP relaxation, but combinatorial relaxations are not as obvious.

## Example: Traveling Salesman Problem

The TSP is a combinatorial problem  $(E, \mathcal{F})$  whose ground set is the edge set of a graph  $G = (V, E)$ .

- $V$  is the set of customers.
- $E$  is the set of travel links between the customers.

A feasible solution is a subset of  $E$  consisting of edges of the form  $\{i, \sigma(i)\}$  for  $i \in V$ , where  $\sigma$  is a simple permutation  $V$  specifying the order in which the customers are visited.

### IP Formulation:

$$\begin{aligned} \sum_{j=1}^n x_{ij} &= 2 \quad \forall i \in N^- \\ \sum_{\substack{i \in S \\ j \notin S}} x_{ij} &\geq 2 \quad \forall S \subset V, |S| > 1. \end{aligned}$$

where  $x_{ij}$  is a binary variable indicating whether  $\sigma(i) = j$ .

## Combinatorial Relaxations of the TSP

- The Traveling Salesman Problem has several well-known combinatorial relaxations.
- Assignment Problem
  - The problem of assigning  $n$  people to  $n$  different tasks.
  - Can be solved in polynomial time.
  - Obtained by dropping the subtour elimination constraints.
- Minimum 1-tree Problem
  - A *1-tree* in a graph is a spanning tree of nodes  $\{2, \dots, n\}$  plus exactly two edges incident to node one.
  - A minimum 1-tree can be found in polynomial time.
  - This relaxation is obtained by dropping all subtour elimination constraints involving node 1 and also all degree constraints not involving node 1.



## Lagrangian Relaxation

- The idea is again based on relaxing a set of constraints from the original formulation.
- We try to push the solution towards feasibility by **penalizing violation** of the dropped constraints.
- Suppose our *IP* is defined by

$$\begin{aligned} \max \quad & cx \\ \text{s.t.} \quad & A^1x \leq b^1 \\ & A^2x \leq b^2 \\ & x \in \mathbb{Z}_+^n \end{aligned}$$

where optimizing over  $Q = \{x \in \mathbb{Z}_+^n \mid A^2x \leq b^2\}$  is “easy.”

- Lagrangian Relaxation:

$$LR(\lambda) : z_{LR}(\lambda) = \max_{x \in Q} \{(c - \lambda A^1)x + \lambda b^1\}.$$

## Properties of the Lagrangian Relaxation

- For any  $\lambda \geq 0$ ,  $LR(\lambda)$  is a relaxation of  $IP$  (why?).
- Solving  $LR(\lambda)$  yields an upper bound on the value of the optimal solution.
- Because of our assumptions,  $LR(\lambda)$  can be solved easily.
- Recalling LP duality, one can think of  $\lambda$  as a vector of “dual variables.”
- If the solution to the relaxation is integral, it is optimal if the primal and dual solutions are complementary, as in LP.
- What is the obvious next step?