

Data-Driven Decision-Making In Enterprise Applications

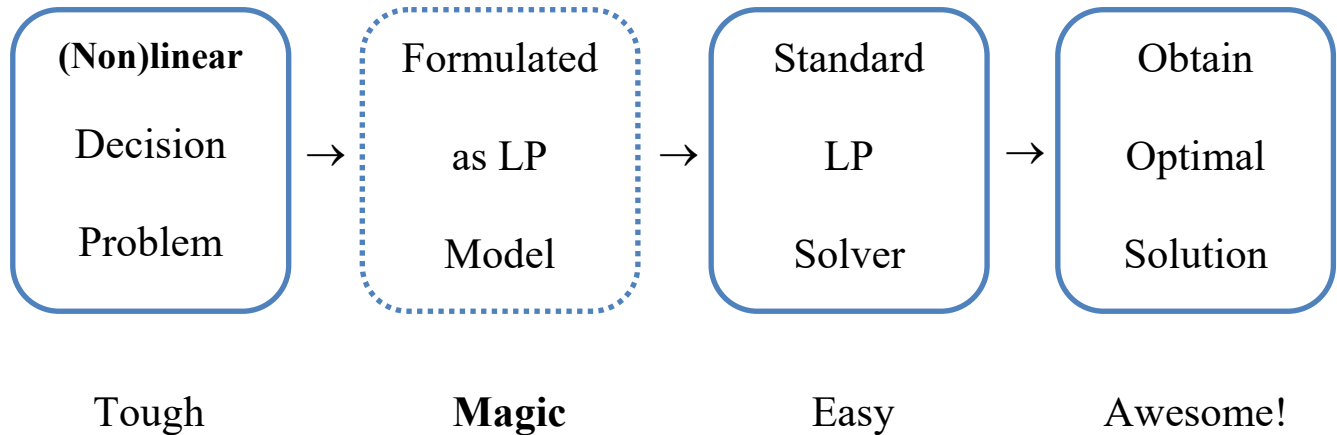
Linear Programming II

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Decision-Making Using Linear Programming



Linear Programming

- Questions regarding last week?
- Implemented/Solved Examples (using AMPL)?
- Today: Tricks to Circumvent Non-Linearities
- Penalty Approaches & Continuous Relaxations
- Description of the HPI Master Project Assignment Problem

Nonlinear Programming Models

- Often non-linear expressions are needed within a model
- (–) Linear solvers cannot be used anymore
- (–) NL solvers often cannot guarantee optimality
- (+) So-called “mild” nonlinearities can be expressed linearly
- (+) This is very valuable as we can exploit LP solvers and their optimality
- The price of such transformations is acceptable:
 - More variables and constraints

I Linearization of “and” in the Constraints

Objective: $\min_{x_1, x_2 \in \{0,1\}} 2 \cdot x_1 + x_2$

Constraints NL: ...

$$x_1 = 1 \text{ and } x_2 = 1 \quad (\text{e.g. needed as joint condition})$$

Objective: $\min_{x_1, x_2 \in \{0,1\}} 2 \cdot x_1 + x_2$

Constraints LIN: ...

$$x_1 + x_2 = 2$$

II Linearization of “or” in the Constraints

Objective:
$$\min_{x_1, x_2 \in \{0,1\}, x_3 \in [0, M]} 2 \cdot x_1 + x_2 + x_3$$

Constraints NLa: $x_1 = 1 \text{ or } x_2 = 1$ (e.g. needed as joint condition)

Constraints NLb: $x_1 = 1 \text{ or } x_2 = 0$

Constraints NLc: $x_3 = 0 \text{ or } x_3 \geq 3$

Objective:
$$\min_{x_1, x_2 \in \{0,1\}, x_3 \in [0, M], z \in \{0,1\}} 2 \cdot x_1 + x_2 + x_3$$

Constraints LINA: $x_1 + x_2 \geq 1$

Constraints LINb: $x_1 + (1 - x_2) \geq 1$

Constraints LINc: $x_3 \leq M \cdot z, \quad x_3 \geq 3 \cdot z$

III Linearization of “max” in the Objective

Objective NL: $\min_{x_1, \dots, x_N \in \mathbb{R}} \left\{ \max_{i=1, \dots, N} x_i \right\}$

Constraints: . . .

Objective LIN: $\min_{x_1, \dots, x_N \in \mathbb{R}, z \in \mathbb{R}} z$

Constraints: . . .

new $z \geq x_i$ for all $i=1, \dots, N$

IV Linearization of “min” in the Objective

Objective NL: $\max_{x_1, \dots, x_N \in \mathbb{R}} \left\{ \min_{i=1, \dots, N} x_i \right\}$

Constraints: . . .

Objective LIN: $\max_{x_1, \dots, x_N \in \mathbb{R}, z \in \mathbb{R}} z$

Constraints: . . .

new $z \leq x_i$ for all $i=1, \dots, N$

V Linearization of “min” in the Constraints

Objective: $\min_{x_1, x_2 \in [0, M]} 2 \cdot x_1 + x_2$

Constraints NL: $4 \leq \min(x_1, x_2) \leq 7$

Objective: $\min_{x_1, x_2 \in [0, M], z_1, z_2 \in \{0, 1\}} 2 \cdot x_1 + x_2$

Constraint LIN: $4 \leq x_i$ for all $i=1, 2$

new $M \cdot z_i \geq x_i - 7$ for all $i=1, 2$

new $z_1 + z_2 \leq 1$

VI Linearization of “abs” in the Objective

Objective NL: $\min_{x_1, x_2 \in \mathbb{R}} 2 \cdot x_1 + \text{abs}(3 - x_2)$

Constraints: ...

Objective LIN: $\min_{x_1, x_2 \in \mathbb{R}, z \in \mathbb{R}} 2 \cdot x_1 + z$

Constraints: ...

new $x_2 - 3 \leq z$

new $3 - x_2 \leq z$

VII Linearization of “abs” in the Constraints

Objective: $\min_{x_1, x_2 \in \mathbb{R}} 2 \cdot x_1 + x_2$

Constraints NL: $abs(3 - x_2) \leq x_1$

Objective LIN: $\min_{x_1, x_2 \in \mathbb{R}, z \in \mathbb{R}} 2 \cdot x_1 + x_2$

Constraints: $z \leq x_1$

new $x_2 - 3 \leq z$

new $3 - x_2 \leq z$

VIII Linearization of “if-then-else”

Objective NL: $\min_{x_1, x_2 \in \{0, 1, 2, \dots, M\}} 2 \cdot x_1 + (\text{if } x_2 \leq 5.5 \text{ then } a \text{ else } b)$

Constraints: . . .

Objective LIN: $\min_{x_1, x_2 \in \{0, 1, 2, \dots, M\}, z \in \{0, 1\}} 2 \cdot x_1 + b \cdot z + a \cdot (1 - z)$

Constraints: . . .

new $x_2 - 5.5 \leq M \cdot z$

new $5.5 - x_2 \leq M \cdot (1 - z)$

IX Linearization of a Product of Binary Variables

Objective: $\min_{x_1, x_2 \in \{0,1\}} 2 \cdot x_1 + x_2$

Constraints NL: including the term: $x_1 \cdot x_2$

Objective: $\min_{x_1, x_2 \in \{0,1\}, z \in \{0,1\}} 2 \cdot x_1 + x_2$

Constraints LIN: include the term z instead, where

$$z \leq x_i, \quad \text{for } i=1,2$$

$$z \geq x_1 + x_2 - 1$$

X Linearization of a Binary x Continuous Variable

Objective: $\min_{x_1 \in \{0,1\}, x_2 \in [0,M]} 2 \cdot x_1 + x_2$

Constraints NL: including the term: $x_1 \cdot x_2$

Objective: $\min_{x_1 \in \{0,1\}, x_2 \in [0,M], z \in [0,M]} 2 \cdot x_1 + x_2$

Constraints LIN: include the term z instead, where

$$z \leq M \cdot x_1, \quad \text{for } i=1,2$$

$$z \leq x_2$$

$$z \geq x_2 - (1 - x_1) \cdot M$$



Solution Tuning

Recall Example IV: Project Assignment Problem

$x_{i,j} \in \{0,1\}$ whether project i , $i=1,\dots,N$, is assigned to worker j , $j=1,\dots,N$

LP:
$$\max_{x_{i,j} \in \{0,1\}^{N \times N}} \sum_{i=1,\dots,N, j=1,\dots,N} w_{i,j} \cdot x_{i,j}$$

s.t.
$$\sum_{i=1,\dots,N} x_{i,j} = 1 \quad \text{for all } j=1,\dots,N \quad (\text{each worker gets 1 project})$$

$$\sum_{j=1,\dots,N} x_{i,j} = 1 \quad \text{for all } i=1,\dots,N \quad (\text{each project is assigned})$$

- Will the allocation always be fair?
- How “outliers” can be avoided?
- Approaches: (i) utility functions, (ii) max min, (iii) multi-objective

Approach (i): Fair Project Assignment (Non-linear)

$x_{i,j} \in \{0,1\}$ whether project $i, i=1,\dots,N$, is assigned to worker $j, j=1,\dots,N$

$$\text{NLP: } \max_{x_{i,j} \in \{0,1\}^{N \times N}} \sum_{j=1,\dots,N} u \left(\sum_{i=1,\dots,N} w_{i,j} \cdot x_{i,j} \right)$$

using, e.g., $u(z) := \ln(z)$, $u(z) := z^{0.6}$, $u(z) := -e^{-0.1 \cdot z}$

$$\text{s.t. } \sum_{i=1,\dots,N} x_{i,j} = 1 \quad \text{for all } j=1,\dots,N \quad (\text{each worker gets 1 project})$$

$$\sum_{j=1,\dots,N} x_{i,j} = 1 \quad \text{for all } i=1,\dots,N \quad (\text{each project is assigned})$$

- Idea: Avoiding low scores is better than including high scores
- **Disadvantage (i):** Non-linear solver is needed

Approach (ii): Fair Project Assignment (Linear!)

$x_{i,j} \in \{0,1\}$ whether project $i, i=1,\dots,N$, is assigned to worker $j, j=1,\dots,N$

LP:
$$\max_{x_{i,j} \in \{0,1\}^{N \times N}, z \in \mathbb{R}} z \quad \text{s.t.} \quad z \leq \sum_{i=1,\dots,N} w_{i,j} \cdot x_{i,j} \quad \text{for all } j=1,\dots,N$$

$$\sum_{i=1,\dots,N} x_{i,j} = 1 \quad \text{for all } j=1,\dots,N \quad (\text{each worker gets 1 project})$$

$$\sum_{j=1,\dots,N} x_{i,j} = 1 \quad \text{for all } i=1,\dots,N \quad (\text{each project is assigned})$$

- Idea: Optimize the lowest willingness (cf. worst case criteria)
- **Disadvantage (ii):** Total willingness score can be low

Approach (iii): Fair Project Assignment (Linear!)

$x_{i,j} \in \{0,1\}$ whether project $i, i=1,\dots,N$, is assigned to worker $j, j=1,\dots,N$

LP:
$$\max_{x_{i,j} \in \{0,1\}^{N \times N}, z \in \mathbb{R}} \sum_{i=1,\dots,N, j=1,\dots,N} w_{i,j} \cdot x_{i,j} + \alpha \cdot z, \quad \text{with parameter } \alpha \geq 0$$

s.t.
$$z \leq \sum_{i=1,\dots,N} w_{i,j} \cdot x_{i,j} \quad \forall j$$

$$\sum_{i=1,\dots,N} x_{i,j} = 1 \quad \text{for all } j=1,\dots,N \quad (\text{each worker gets 1 project})$$

$$\sum_{j=1,\dots,N} x_{i,j} = 1 \quad \text{for all } i=1,\dots,N \quad (\text{each project is assigned})$$

- Idea: Combine both objectives as a weighted sum
- **Disadvantage (iii):** Suitable weighting factor α has to be determined



Penalty Approaches & Efficient Frontiers

Penalty Formulations for Pareto-Optimal Relaxations

Objective: $\max_{x_1, \dots, x_N \in \{0,1\}} \sum_{i=1, \dots, N} u_i \cdot x_i$ Knapsack example

Constraints: $\sum_{i=1, \dots, N} s_i \cdot x_i \leq C$ (One) Hard Constraint

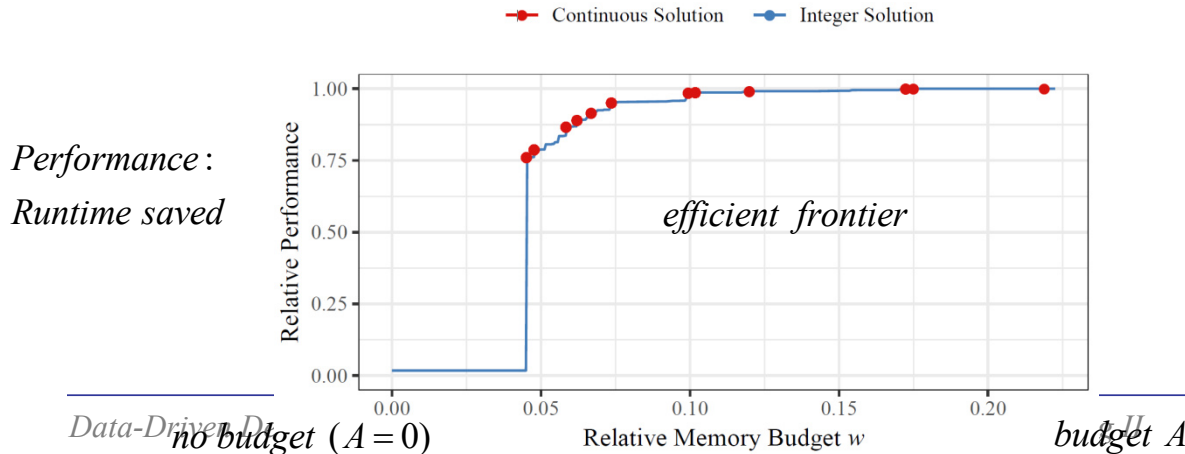
Penalty-Objective: $\max_{x_1, \dots, x_N \in \{0,1\}} \sum_{i=1, \dots, N} u_i \cdot x_i - \alpha \cdot \sum_{i=1, \dots, N} s_i \cdot x_i$ (Soft Constraint)

Constraints: none

Results: Pareto-optimal combinations of “Utility” and “Space”

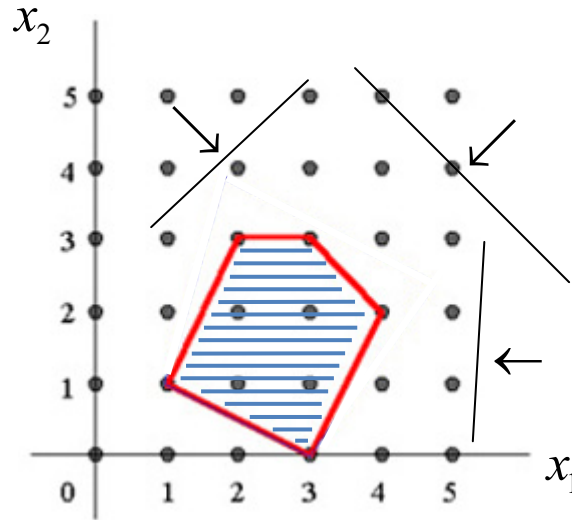
Pareto-Optimal Relaxations (int. vs. cont. solutions)

- (i) Optimal integer solution (blue): $\min_{\bar{x} \in \{0,1\}^N} F(\bar{x}) \text{ s.t. } M(\bar{x}) \leq A \Rightarrow \bar{x}^*(A) \text{ optimal}$
- (ii) Continuous relaxation: $\rightarrow \min_{\bar{x} \in [0,1]^N} F(\bar{x}) \text{ s.t. } M(\bar{x}) \leq A \Rightarrow \bar{x}^*(A) \in \{0,1\}^N ?$
- (iii) Penalty formulation (red): $\min_{\bar{x} \in [0,1]^N} F(\bar{x}) + \alpha \cdot M(\bar{x}) \Rightarrow \bar{x}^*(\alpha) \in \{0,1\}^N \text{ and}$
 \uparrow *Pareto-optimal!*



When do Integer & Continuous Solutions Coincide?

maximize $a \cdot x_1 + b \cdot x_2$ s.t. ... with $x_1, x_2 \in \mathbb{R}$ vs. $x_1, x_2 \in \mathbb{N}$



- **Answer:** The corners of the polygon have to be “integers”!



HPI Master Project Assignment Problem

Description HPI Master Project Assignment Problem

- Each worker gets 1 project
- 0, 3, 4, 5, or 6 students per project
- Each student has a “first/second/third choice” project
- Can a student exclude one/two projects? -> no
- Average number of students/projects?
- Maximize the number of first choices?
- Minimize the number of unfulfilled dreams?
- Weighted sum?



Next Week

Homework: Try to apply/implement the Linearizations I-X!

Formulate the HPI Master Project Assignment Problem

Outlook:

- Nonlinear Programming and Suitable Solvers
- Linear Regression
- Logistic Regression
- Probabilities & Random Variables

Overview

2	April 25	Linear Programming I
3	April 29	Linear Programming II
4	May 2	Linear/Logistic Regression + Homework (2 weeks time)
5	May ?	Exercise Implementations (postponed)
6	May 16	Dynamic Programming I
7	May 20	Dynamic Programming II
8	May 23	Response Strategies / Game Theory
9	May 27	Project Assignments
10	June 3	Robust Optimization
11	June 13	Workshop / Group Meetings
12	June 20	Presentations (First Results)
13/14	June 24/27	Workshop / Group Meetings
15/16	July 1/4	Workshop / Group Meetings
17	July 11	Presentations (Final Results), Feedback, Documentation (Aug 31)