

DAGSTUHL SEMINAR

Bilevel optimization for the retail electricity market

Quentin Jacquet^{a,b}
Wim Van Ackooij^a, Clémence Alasseur^a, Stéphane Gaubert^b

^aOSIRIS, EDF Lab, 91120 Palaiseau, France

^bINRIA, CMAP, Ecole Polytechnique, 91120, Palaiseau, France



TABLE OF CONTENTS

- 1 INTRODUCTION

- 2 LEADER-FOLLOWER
 - Deterministic model
 - Logit regularization
 - Quadratic regularization

- 3 DEMAND ELASTICITY
 - Elasticity of the demand
 - Distortion of the Polyhedral Complex
 - Linear Reformulation

- 4 DYNAMIC EXTENSION
 - Switching costs
 - Ergodic control

- 5 PERSPECTIVES

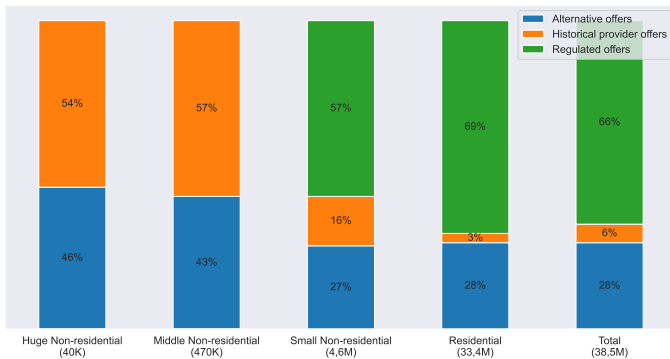
INTRODUCTION

A WIDE VARIETY OF OFFERS...

- ◇ Since 2007, French electricity market is open to competition :

Market Offers
Company freely determines the prices

Regulated Offers
Fixed prices



A WIDE VARIETY OF OFFERS...

- ◇ Since 2007, French electricity market is open to competition :

Market Offers
 Company freely determines the prices

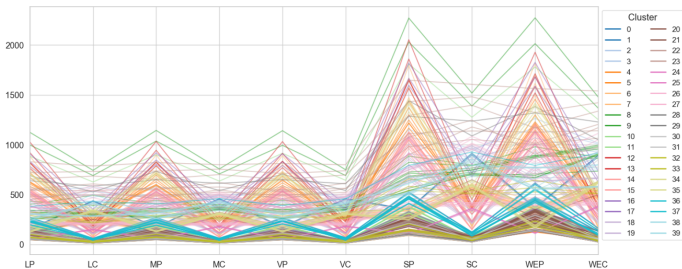
Regulated Offers
 Fixed prices

- ◇ Contracts structure:

	$\overbrace{\hspace{15em}}^{W \text{ contracts}}$		
	Baseload version	Peak/Off-peak version	
$\left. \begin{array}{l} \text{Variable portion} \\ (\text{€}/\text{kWh}) \end{array} \right\} H \text{ attributes (2 or 3)}$	unique price	peak price	
		off-peak price	
Fixed portion (€)	power	power	

... AND A WIDE VARIETY OF CUSTOMERS

- ◇ Load curves¹ of customers reflect different consumption behaviors:



- ◇ Consumption preferences change
Digital ("Digiwatt"), Green ("Vert Electrique")
Self-consumption (solar panel, batteries, ...)

Assumption: The population can be aggregated into S customers segments.

¹We use simulated load curves, from SMACH.

CHALLENGE



Issue

How to determine fair prices to attract/keep customers while secure a sufficient profit ?

Leader-follower game:

- ◇ First player (*leader*) decides
- ◇ Second player (*follower*) reacts



SKETCH OF THE MODELIZATION

Multi-leader-common-followers game [LM10]

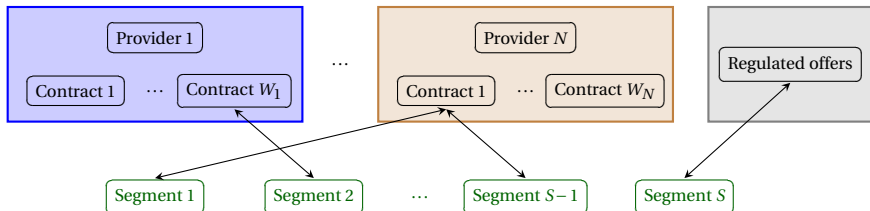


Figure: Representation of the problem

- ◇ Nash equilibrium at upper level
- ◇ *Envy-Free* : no limitation on the maximum number of customers able to purchase the same contract

FOCUS ON STATIC COMPETITION

Leader-follower game (Stackelberg)

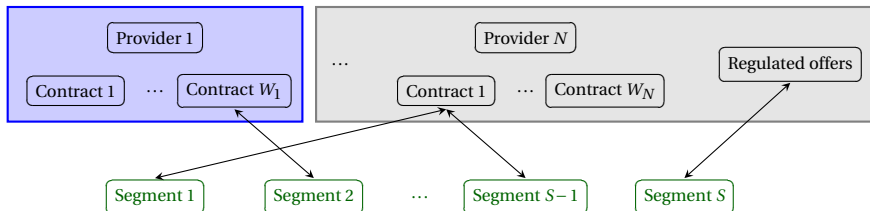


Figure: Representation of the problem

- ◇ *Nash equilibrium at upper level* → static competition
- ◇ *Envy-Free*: no limitation on the maximum number of customers able to purchase the same contract

FOCUS ON STATIC COMPETITION

Leader-follower game (Stackelberg)

General formulation of *Bilevel problems* [Dem+15]

$$\left\{ \begin{array}{ll} \text{“min”} & F(x, y) \\ x & \\ \text{s.t.} & x \in X \\ & y \in \Psi(x) := \underset{y}{\text{Argmin}} \{f(x, y); g(x, y) \leq 0\} \end{array} \right. \begin{array}{l} \leftarrow \text{“Upper level”} \\ \\ \\ \leftarrow \text{“Lower level”} \end{array}$$

- ◇ x is called “Upper variable”, controlled by the leader
- ◇ y is called “Lower variable”, controlled by the follower

Complexity results

Linear Bilevel problems are NP-Hard [Jer85].

LEADER-FOLLOWER GAMES

[Jac+21]

DETERMINISTIC MODEL

Notations:

- ◇ $[S] := \{1 \dots S\}$ customers segments,
- ◇ $[W]$ contracts of the leader,
- ◇ $[H]$ attributes per contract

Variables:

- ◇ x_w^h price of attribute h and contract w ,
- ◇ $\mu_{sw} = \begin{cases} 1 & \text{if segment } s \text{ chooses } w, \\ 0 & \text{otherwise.} \end{cases}$

Data:

- ◇ C_{sw} cost to supply s if he chooses w ,
- ◇ R_{sw} reservation price of s for contract w ,
- ◇ Customer invoice is a linear form of the prices

$$\theta_{sw}(x) := \langle E_{sw}, x_w \rangle_H$$

Deterministic bilevel problem

$$\left\{ \begin{array}{l} \max_{x \in X, \mu^*} \sum_{s \in [S]} \rho_s \langle \theta_s(x) - C_s, \mu_s^* \rangle_W \rightarrow \text{leader pb} \\ \text{s. t. } \mu^* \in \operatorname{argmin}_{\mu \in (\Delta_{W+1})^S} \left\{ \sum_{s \in [S]} \langle \theta_s(x) - R_s, \mu_s \rangle_W \right\} \\ \quad \leftarrow \text{follower pb} \end{array} \right.$$

Profit function

$$\pi(x) := \sum_{s \in [S]} \rho_s \langle \theta_s(x) - C_s, \mu_s^*(x) \rangle_W ,$$

where $\mu^*(\cdot)$ is the optimal follower response.

KKT TRANSFORMATION

The **follower problem** is linear, and can be replaced by KKT conditions:

$$\begin{aligned} \max_{x \in X, \mu, \eta} \quad & \sum_{s \in [S]} \rho_s \eta_s + \rho_s \langle R_s - C_s, \mu_s \rangle_W \\ \text{s. t.} \quad & 0 \leq \mu_{sw} \perp \theta_{sw}(x) - R_{sw} - \eta_s \geq 0, \forall s, w \\ & 0 \leq \mu_{s0} \perp \eta_s \leq 0, \forall s \\ & \mu_s \in \Delta_{W+1}, \forall s \end{aligned}$$

This leads to a *Linear Program under Complementarity Constraints* (LPCC).

Usually, we replace the complementarity constraints by Big- M constraints
 \rightsquigarrow MILP formulation, generalization of [STM11; Fer+16].

PRICE COMPLEX AND INSTABILITY

One customer ($S = 1$), 2 contracts ($W = 2$)

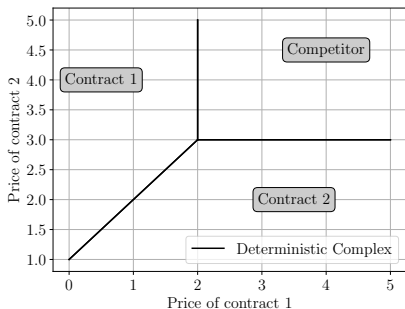


Figure: Response of follower in the space of prices

(developed in [BK19; Eyt18])

Five customers ($S = 5$), 1 contract ($W = 1$)

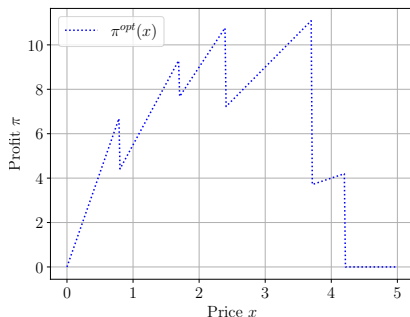


Figure: Instability in the profit function

(developed in [GMS15])

Proposition [Jac+21]

In the general case, the optimal profit is achieved at a discontinuity.

LOGIT REGULARIZATION

Multinomial Logit model

$$\left\{ \begin{array}{l} \max_{x \in X, \mu^*} \sum_{s \in [S]} \rho_s \langle \theta_s(x) - C_s, \mu_s^* \rangle_W \\ \text{s.t. } \mu^* \in \operatorname{argmin}_{\mu \in (\Delta_{W+1})^S} \left\{ \begin{array}{l} \sum_{s \in [S]} \langle \theta_s(x) - R_s, \mu_s \rangle_W \\ + \frac{1}{\beta} \langle \log(\mu_s), \mu_s \rangle_{W+1} \end{array} \right\} \end{array} \right.$$

$$\rightsquigarrow \mu_{sw}^*(x) = \frac{e^{-\beta(\theta_{sw}(x) - R_{sw})}}{1 + \sum_{w' \in [W]} e^{-\beta(\theta_{sw'}(x) - R_{sw'})}}$$

$\Rightarrow \mu_s^* \in \operatorname{Int} \Delta_{W+1}$, no polyhedral complex

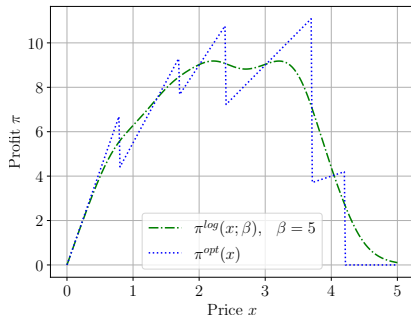
Five customers ($S = 5$), 1 contract ($W = 1$)

Figure: Logit regularization

(developed in [GMS15; Li+19])

Proposition [Li+19]

For a heterogeneous population ($S > 1$), π is in general non-concave.

QUADRATIC REGULARIZATION (1)

Multinomial Logit model

$$\left\{ \begin{array}{l} \max_{x \in X, \mu^*} \sum_{s \in [S]} \rho_s \langle \theta_s(x) - C_s, \mu_s^* \rangle_W \\ \text{s. t. } \mu^* \in \underset{\mu \in (\Delta_{W+1})^S}{\text{argmin}} \left\{ \begin{array}{l} \sum_{s \in [S]} \langle \theta_s(x) - R_s, \mu_s \rangle_W \\ + \frac{1}{\beta} \langle \log(\mu_s), \mu_s \rangle_{W+1} \end{array} \right\} \end{array} \right\}$$

$$\rightsquigarrow \mu_{sw}^*(x) = \frac{e^{-\beta(\theta_{sw}(x) - R_{sw})}}{1 + \sum_{w' \in [W]} e^{-\beta(\theta_{sw'}(x) - R_{sw'})}}$$

Quadratic model

$$\left\{ \begin{array}{l} \max_{x \in X, \mu} \sum_{s \in [S]} \rho_s \langle \theta_s(x) - C_s, \mu_s^* \rangle_W \\ \text{s. t. } \mu^* \in \underset{\mu \in (\Delta_{W+1})^S}{\text{argmin}} \left\{ \begin{array}{l} \sum_{s \in [S]} \langle \theta_s(x) - R_s, \mu_s \rangle_W \\ + \frac{1}{\beta} \langle \mu_s - 1, \mu_s \rangle_{W+1} \end{array} \right\} \end{array} \right\}$$

$$\rightsquigarrow \mu_s^*(x) = \text{Proj}_{\Delta_{W+1}} \left(\frac{\beta}{2} (R_s - \theta_s(x)) \right)$$

↪ powerful algorithm for projection on the simplex, see [Con16]

QUADRATIC REGULARIZATION (2)

One customer ($S = 1$), 2 contracts ($W = 2$)

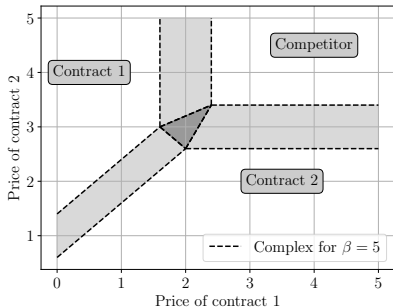


Figure: Response of follower in the space of prices

Five customers ($S = 5$), 1 contract ($W = 1$)

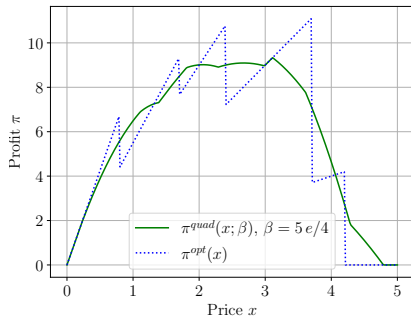


Figure: Quadratic regularization

Theorem [Jac+21]

The profit π is *continuous*. Moreover, it is *concave* on each cell of the polyhedral complex.

CUSTOMERS' RESPONSE AS A POLYHEDRAL COMPLEX

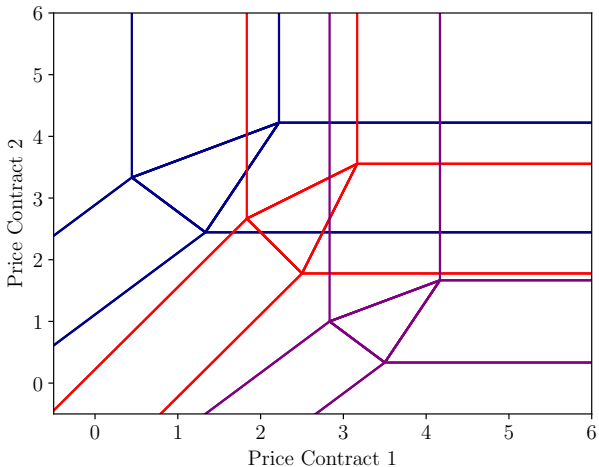


Figure: Example with $S = 3$ segments and $W = 2$ contracts

CUSTOMERS' RESPONSE AS A POLYHEDRAL COMPLEX

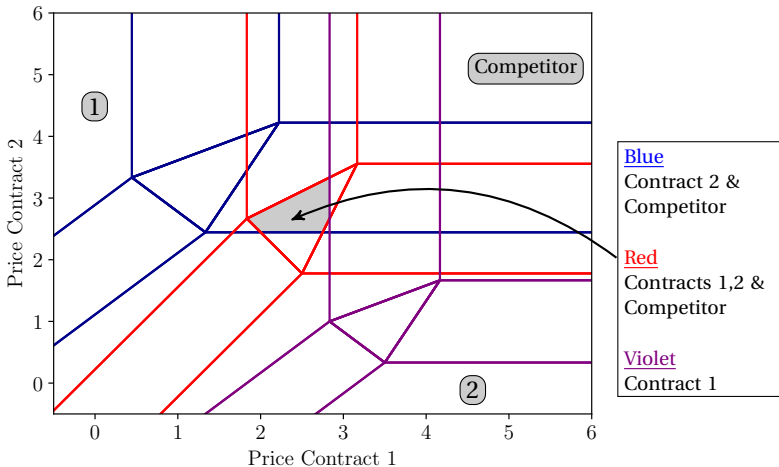


Figure: Example with $S = 3$ segments and $W = 2$ contracts

DESIGN OF A HEURISTIC – ON AN EXAMPLE

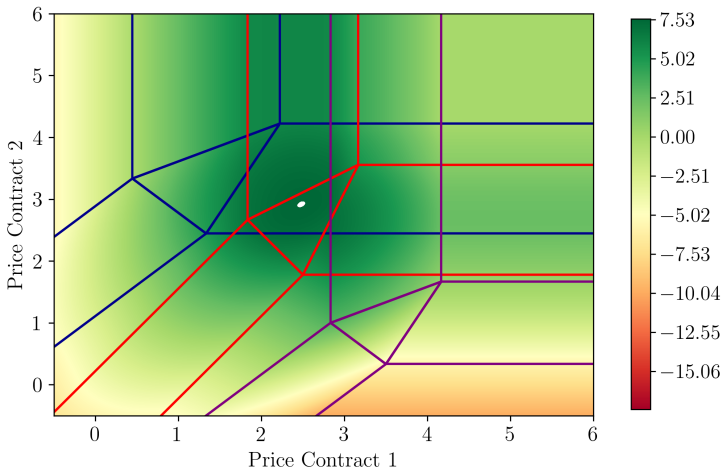


Figure: Example with $S = 3$ segments and $W = 2$ contracts

DESIGN OF A HEURISTIC – ON AN EXAMPLE

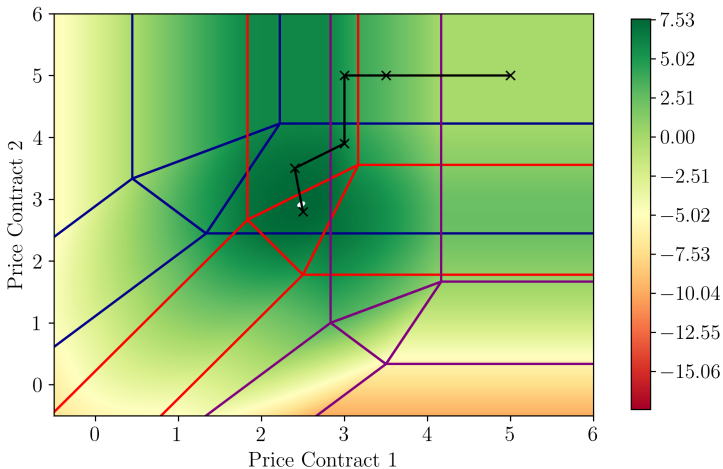
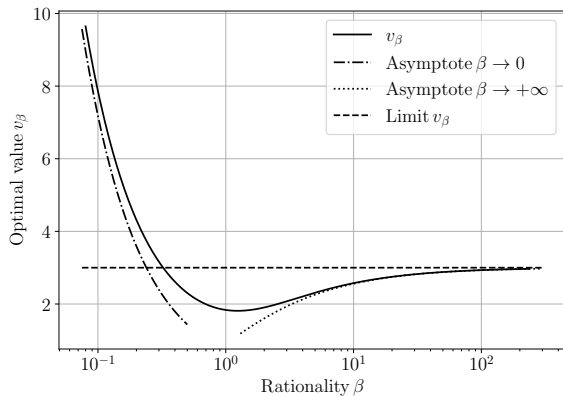
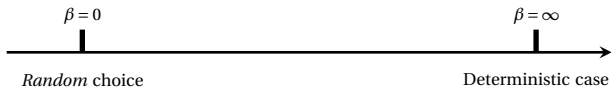


Figure: Example with $S = 3$ segments and $W = 2$ contracts

IMPACT OF THE REGULARIZATION INTENSITY

Figure: Optimal value as a function of β

DEMAND ELASTICITY

(Ongoing research)

AND IF CONSUMERS *adapt* THEIR CONSUMPTION TO PRICES ?

Intuition

“If the electricity is too costly, I will reduce my consumption.”

↔ Isoelastic utility function of the electricity demand (CRRA):

$$U_s : E \in \mathbb{R}^H \mapsto \sum_{h \in [H]} \alpha_s^h \frac{(E^h)^\eta}{\eta}, \quad \eta \in \underbrace{]-\infty, 0[}_{\text{residential}} \cup \underbrace{]0, 1[}_{\text{industrial}} .$$

AND IF CONSUMERS *adapt* THEIR CONSUMPTION TO PRICES ?

Intuition

“If the electricity is too costly, I will reduce my consumption.”

↪ Isoelastic utility function of the electricity demand (CRRA):

$$U_s : E \in \mathbb{R}^H \mapsto \sum_{h \in [H]} \alpha_s^h \frac{(E^h)^\eta}{\eta}, \quad \eta \in \underbrace{]-\infty, 0[}_{\text{residential}} \cup \underbrace{]0, 1[}_{\text{industrial}} .$$

↪ The customer *not only* decides the contract, but also maximizes

$$U_s^* : x \in \mathbb{R}^H \mapsto \max_{E \in \mathbb{R}^H} \{U_s(E) - \langle x, E \rangle_H\} .$$

The optimal energy consumption is $\mathcal{E}_s^h(x^h) = \left(\frac{\alpha_s^h}{x^h}\right)^{\frac{1}{1-\eta}}$.

AND IF CONSUMERS *adapt* THEIR CONSUMPTION TO PRICES ?

Intuition

“If the electricity is too costly, I will reduce my consumption.”

↪ Isoelastic utility function of the electricity demand (CRRA):

$$U_s : E \in \mathbb{R}^H \mapsto \sum_{h \in [H]} \alpha_s^h \frac{(E^h)^\eta}{\eta}, \quad \eta \in \underbrace{]-\infty, 0[}_{\text{residential}} \cup \underbrace{]0, 1[}_{\text{industrial}} .$$

↪ The customer *not only* decides the contract, but also maximizes

$$U_s^* : x \in \mathbb{R}^H \mapsto \max_{E \in \mathbb{R}^H} \{U_s(E) - \langle x, E \rangle_H\} .$$

The optimal energy consumption is $\mathcal{E}_s^h(x^h) = \left(\frac{\alpha_s^h}{x^h}\right)^{\frac{1}{1-\eta}}$.

↪ The invoice is now a *nonlinear* function:

$$\theta_{sw}(x_w) := \langle \tilde{E}_{sw}, x_w \rangle_H \rightarrow \Theta_s(x_w) := \langle \mathcal{E}_s(x_w), x_w \rangle_H$$

Remark: We recover the inelastic case for $\eta \rightarrow -\infty$.

DISTORTION OF THE POLYHEDRAL COMPLEX

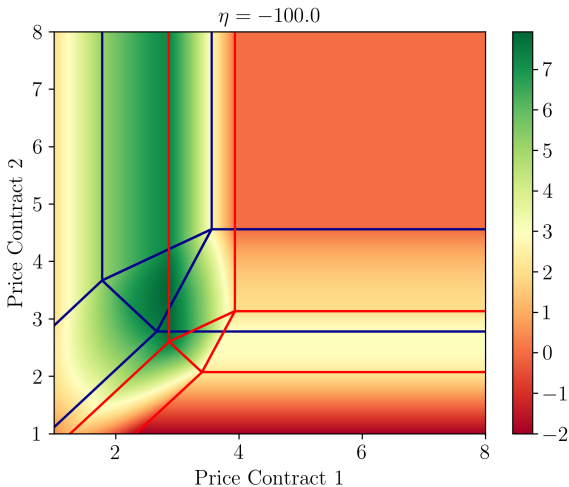


Figure: Example with $S = 2$ segments and $W = 2$ contracts

DISTORTION OF THE POLYHEDRAL COMPLEX

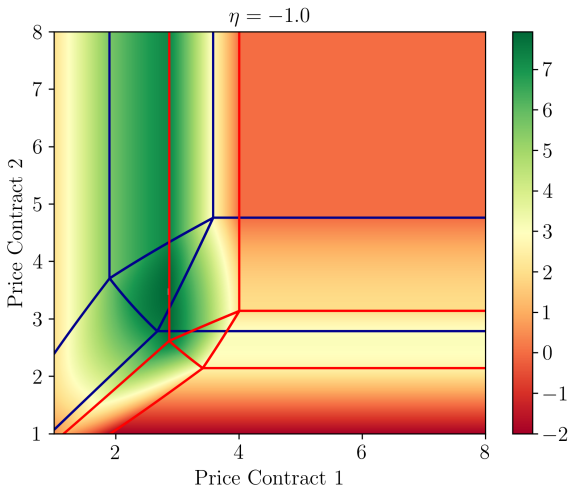
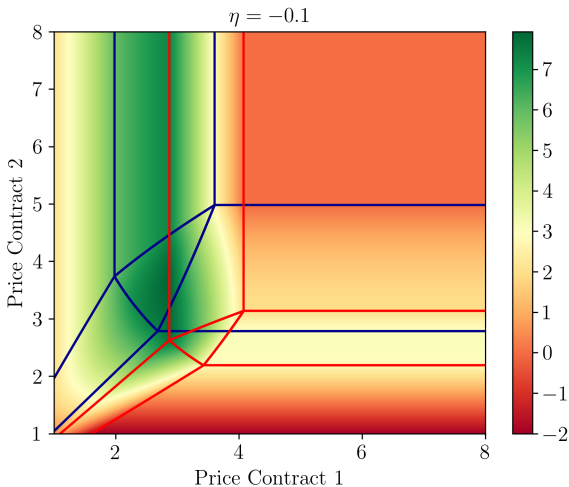


Figure: Example with $S = 2$ segments and $W = 2$ contracts

DISTORTION OF THE POLYHEDRAL COMPLEX

Figure: Example with $S = 2$ segments and $W = 2$ contracts

DISTORTION OF THE POLYHEDRAL COMPLEX

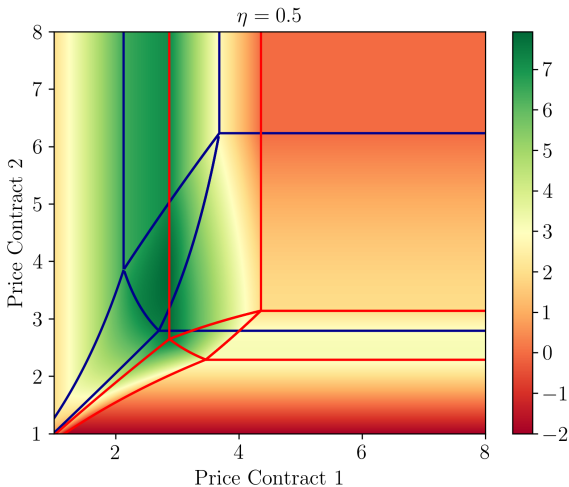


Figure: Example with $S = 2$ segments and $W = 2$ contracts

RETRIEVING A POLYHEDRAL COMPLEX (FIRST ORDER COST)

Proposition

Suppose that the price constraints are of the following form

$$X = \mathcal{O}_{(\underline{x}, \bar{x}, \kappa)}(P) := \left\{ x_w^h \in [\underline{x}_w^h, \bar{x}_w^h] \mid x_w^h \leq \kappa_w^h x_w^{h'} \text{ for } (w, h) \leq_P (w', h') \right\},$$

where P is a partial order set. Then, the bilevel problem

$$\begin{aligned} \max_{x \in X, \mu^*} \quad & \sum_{s \in [S]} \rho_s \langle \Theta_s(x) - C_s, \mu_s \rangle_W \\ \text{s. t. } \mu^* \in \quad & \operatorname{argmin}_{\mu \in (\Delta_{W+1})^S} \left\{ \sum_{s \in [S]} \langle \Theta_s(x_w) - R_s, \mu_s \rangle_W \right\}, \end{aligned}$$

can be equivalently defined using variables $z_w^h := (x_w^h)^{-\frac{\eta}{1-\eta}}$



CONSUMPTION-DEPENDENT COSTS

The retailer cost is not constant anymore, but depends on the total consumption:

$$\cancel{\max_{x \in X, \mu^*} \sum_{s \in [S]} \rho_s \langle \Theta_s(x) - C_s, \mu_s^* \rangle_W} \rightarrow \max_{x \in X, \mu^*} \sum_{s \in [S]} \rho_s \langle \Theta_s(x), \mu_s^* \rangle_W - C \left(\underbrace{\sum_{s \in [S]} \rho_s \sum_{w \in [W]} \mathcal{E}_s(x_w) \mu_{sw}^*}_{:= \mathcal{E}^{\text{tot}}(x, \mu^*) \text{ (total consumption)}} \right)$$

with $C(\cdot)$ a convex nondecreasing function.

Proposition

In the Z space,

- ◇ the energy consumption $z_w \mapsto \mathcal{E}_s(z_w)$ is always *convex*,
- ◇ the total energy $z \mapsto \mathcal{E}^{\text{tot}}(z, \mu^*(z))$ is *convex* on each cell for a sufficiently large regularization intensity β^{-1} .

DYNAMIC EXTENSION

[Jac+22]

AND IF CONSUMERS *do not immediately* REACT ?

Intuition

“I switch to a new contract if there is a *sufficient* difference with my current offer.”

This notion is known in Economics:

↪ Customers have *switching costs* (imperfect market), see e.g. [DHR10; HP10]



MARKOVIAN DECISION PROCESS

Modelization as a *Markovian Decision Process* (MDP)

$$\mu_{t+1} = \mu_t P(x_t),$$

where $P(x_t)$ is the transition matrix,
obtained by solving the *lower* problem knowing the *upper* decision x_t at time t .

We choose a *logit* transition

$$P(x_t) = \text{diag}(\{P(x_t)_s\}_{s \in [S]}), \quad [P_s(x_t)]_{(v,w)} = \frac{e^{\beta(R_{sw} - \theta_{sw}(x_t)) + \gamma_{sv} \mathbb{1}_{(w=v)}}}{1 + \sum_{w' \in [W]} e^{\beta(R_{sw'} - \theta_{sw'}(x_t)) + \gamma_{sv} \mathbb{1}_{(w'=v)}}$$

$\rightsquigarrow \gamma_{sw}$ is the switching cost the customer s would pay if he switches to another offer.

- ◇ The previous (static) model is recover when $\gamma \equiv 0$.
- ◇ $P(x) \gg 0$ for all x , and we can define \mathcal{D} such that

$$\mu_t \in \mathcal{D} \subset \text{relint}\left(\Delta_W^S\right), t \geq 1 .$$

ERGODIC CONTROL

For a policy $\pi = \{\pi_t\}_{t \geq 1}$, $\mathbf{x}_t = \pi_t(\boldsymbol{\mu}_t)$ is the action taken by the controller at t .
Now, we aim to maximize the *average long-term reward*, i.e.,

$$g^* = \sup_{\pi \in \Pi} \liminf_{T \rightarrow \infty} \frac{1}{T} \sum_{t=1}^T r(\pi_t(\boldsymbol{\mu}_t), \boldsymbol{\mu}_t) , \quad (1)$$

where $r(\cdot, \cdot)$ is the objective defined in the static model.

For any function $v: \Delta_W^S \rightarrow \mathbb{R}$, the Bellman operator \mathcal{B} is defined as

$$\mathcal{B} v(\boldsymbol{\mu}) = \max_{\mathbf{x} \in X} \{r(\mathbf{x}, \boldsymbol{\mu}) + v(\boldsymbol{\mu}P(\mathbf{x}))\} .$$

Theorem [Jac+22]

Assume that $\mathbf{x} \mapsto P_s(\mathbf{x})$ is continuous and $P_s(\mathbf{x}) \gg 0$ for all \mathbf{x} and $s \in [S]$. Then, the ergodic eigenproblem

$$g \mathbb{1}_{\mathcal{D}} + h = \mathcal{B} h$$

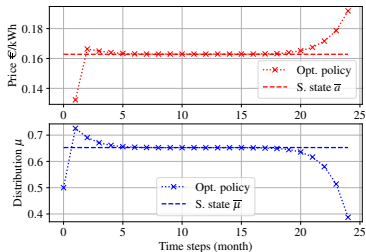
admits a solution $g^* \in \mathbb{R}$ and h^* Lipschitz and convex on \mathcal{D} .

Moreover, g^* satisfies (1), and a maximizer $\mathbf{x}^*(\cdot) \in \arg \max \mathcal{B} h^*$ defines an optimal policy for the average gain problem.

IMPACT OF SWITCHING COSTS ON TOY MODEL

low γ

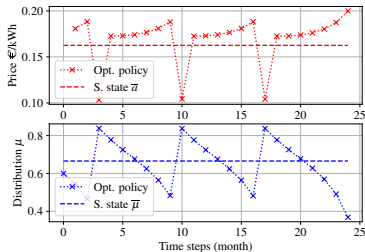
“Turnpike” like strategy:
Attraction to a steady-state



(a) Optimal finite horizon trajectory (provider action and customer distribution) for *low* switching cost.

high γ

Cyclic strategy:
A promotion is periodically applied



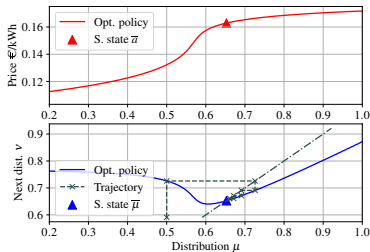
(b) Optimal finite horizon trajectory (provider action and customer distribution) for *high* switching cost.

↪ Phenomenon already mentioned in Economics, see e.g. [HP10].

IMPACT OF SWITCHING COSTS ON TOY MODEL

low γ

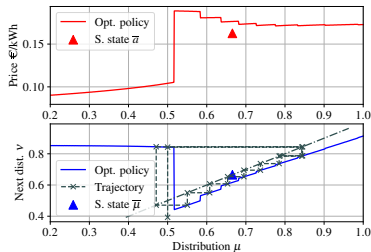
“Turnpike” like strategy:
Attraction to a steady-state



(a) Optimal decision for the long-run average reward (provider action and next customer distribution)

high γ

Cyclic strategy:
A promotion is periodically applied



(b) Optimal decision for the long-run average reward (provider action and next customer distribution)

↪ Phenomenon already mentioned in Economics, see e.g. [HP10].

PERSPECTIVES

Future works

- ◇ Analyze of turnpike property for the dynamic extension
- ◇ Definition of continuous-time model
- ◇ Competition at the upper level (between leaders)

REFERENCES I

- [Jer85] Robert G. Jeroslow. “The polynomial hierarchy and a simple model for competitive analysis”. In: [Mathematical Programming](#) 32.2 (June 1985), pp. 146–164.
- [DHR10] Jean-Pierre Dubé, Günter J. Hitsch, and Peter E. Rossi. “State dependence and alternative explanations for consumer inertia”. In: [The RAND Journal of Economics](#) 41.3 (Aug. 2010), pp. 417–445.
- [HP10] Dan Horsky and Polykarpos Pavlidis. “Brand Loyalty Induced Price Promotions: An Empirical Investigation”. In: [SSRN Electronic Journal](#) (2010).
- [LM10] Sven Leyffer and Todd Munson. “Solving multi-leader-common-follower games”. In: [Optimization Methods and Software](#) 25.4 (Aug. 2010), pp. 601–623.
- [STM11] R. Shioda, L. Tunçel, and T. Myklebust. “Maximum utility product pricing models and algorithms based on reservation price”. In: [Computational Optimization and Applications](#) 48 (Mar. 2011), pp. 157–198.
- [Dem+15] Stephan Dempe, Vyacheslav Kalashnikov, Gerardo A. Pérez-Valdés, and Nataliya Kalashnykova. [Bilevel Programming Problems](#). Springer Berlin Heidelberg, 2015.
- [GMS15] François Gilbert, Patrice Marcotte, and Gilles Savard. “A Numerical Study of the Logit Network Pricing Problem”. In: [Transportation Science](#) 49 (Jan. 2015), p. 150105061815001.
- [Con16] Laurent Condat. “Fast Projection onto the Simplex and the l1 Ball”. In: [Mathematical Programming, Series A](#) 158.1 (July 2016), pp. 575–585.
- [Fer+16] Cristina G Fernandes, Carlos E Ferreira, Alvaro JP Franco, and Rafael CS Schouery. “The envy-free pricing problem, unit-demand markets and connections with the network pricing problem”. In: [Discrete Optimization](#) 22 (2016), pp. 141–161.

REFERENCES II

- [Eyt18] Jean-Bernard Eytard. “A tropical geometry and discrete convexity approach to bilevel programming: application to smart data pricing in mobile telecommunication networks”. PhD thesis. Université Paris-Saclay (ComUE), 2018.
- [BK19] Elizabeth Baldwin and Paul Klemperer. “Understanding preferences: “demand types”, and the existence of equilibrium with indivisibilities”. In: *Econometrica* 87.3 (2019), pp. 867–932.
- [Li+19] Hongmin Li, Scott Webster, N. Mason, and K. Kempf. “Product-Line Pricing Under Discrete Mixed Multinomial Logit Demand”. In: *Manuf. Serv. Oper. Manag.* 21 (2019), pp. 14–28.
- [Jac+21] Quentin Jacquet, Wim van Ackooij, Clémence Alasseur, and Stéphane Gaubert. A Quadratic Regularization for the Multi-Attribute Unit-Demand Envy-Free Pricing Problem. 2021.
- [Jac+22] Quentin Jacquet, Wim van Ackooij, Clémence Alasseur, and Stéphane Gaubert. Ergodic control of a heterogeneous population and application to electricity pricing. 2022.

THANK YOU FOR YOUR ATTENTION !

