

Etude d'un problème de tarification en télécommunication

Luce Brotcorne

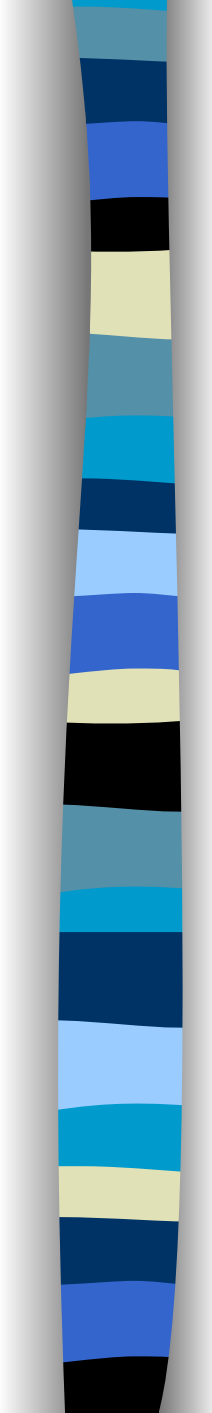


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6. Focus on the F. T. Context Joint Design and Pricing

Companies must decide on **prices** for services

⇒ Maximize revenue

Design choices capture many of the most important features of companies

⇒ Cost of providing service and attractiveness

At the present time, design and pricing issues are not treated jointly

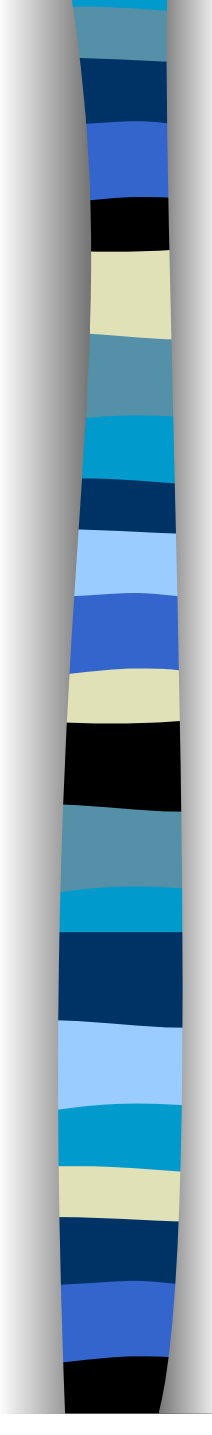
Tradeoff between

the **cost** of service

the **revenue** generated

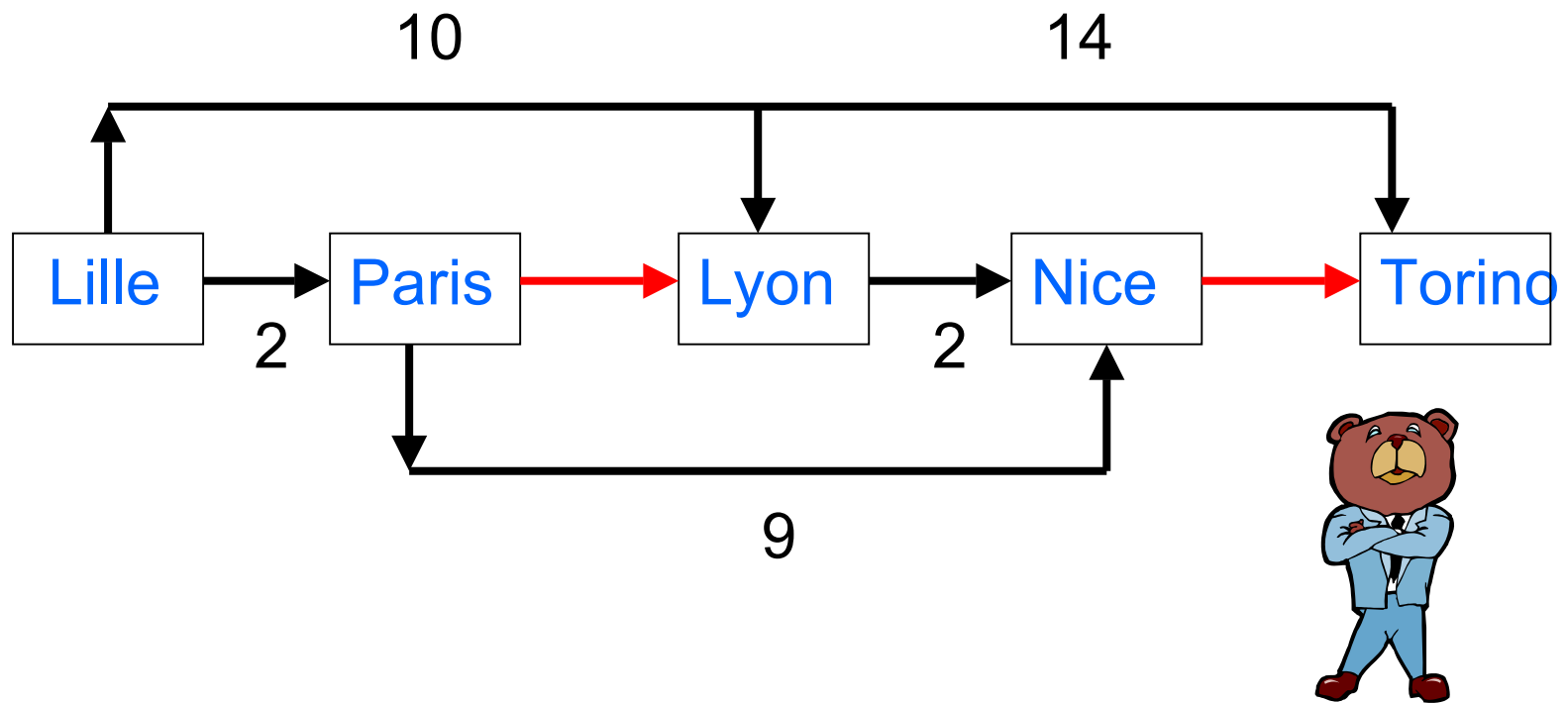
when choosing its system **design** and **prices**





Problem

- An **operator** strives **to maximize its revenue**
 - Open connections (interconnection or connection building)
 - Determine the associated tariffs
- **Customers** are set to send prescribed amounts of flows from origins to destinations at **minimum cost**.
Choice among several operators



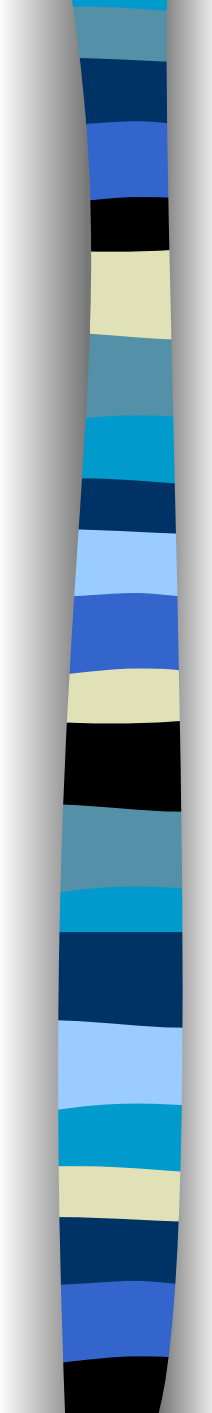
No fixed opening cost: Fixed opening cost (1):

$$T_{PL} = 7$$

$$T_{NT} = 12$$

$$0 \leq f_{PL} \leq 7 \Rightarrow T_{PL} = 7$$

$$0 \leq f_{NT} \leq 12 \Rightarrow T_{NT} = 12$$



Hierarchical process with opposite interests

Bilevel Programming

7. Bilevel Programming

Leader



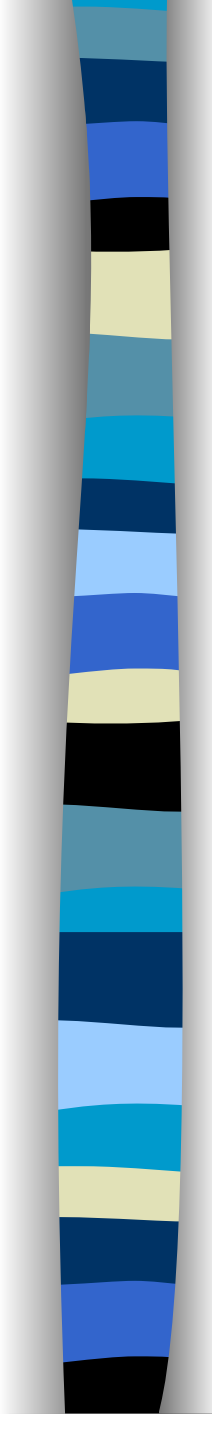
Follower

$$\min_x f(x, y)$$

$$\text{s.c. } (x, y) \in X,$$

$$y \text{ solution de } \min_y g(x, y),$$

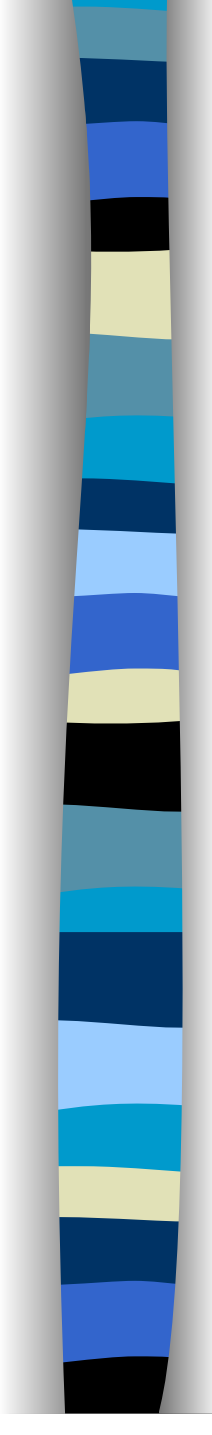
$$\text{s.c. } (x, y) \in Y.$$



For fixed x , the second level problem is defined as:

$$y \in P(\tilde{x}) = \operatorname{argmin}_y g(\tilde{x}, y) \\ \text{s.c. } (\tilde{x}, y) \in Y.$$

Problem parametrized in x .



Non unicity of follower solution

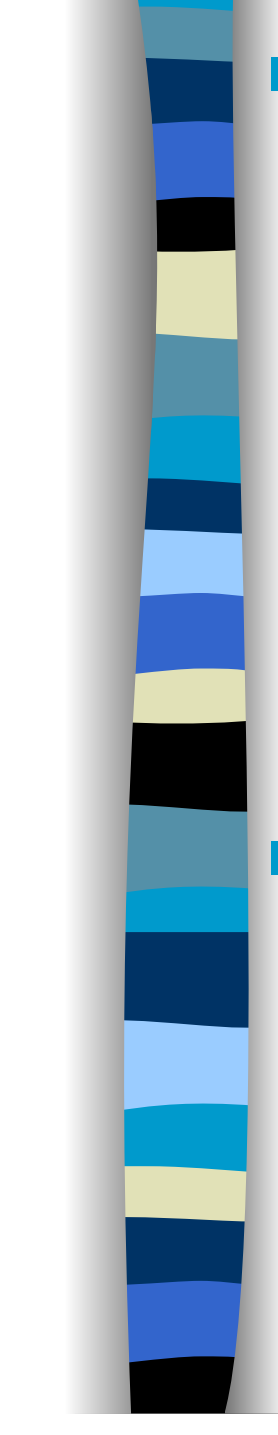
1) Non Cooperation:

$$\begin{aligned} \min_x \quad & \max_y f(x, y) \\ \text{s.c.} \quad & (x, y) \in X, \\ & y \in P(x). \end{aligned}$$

2) Cooperation:

$$\begin{aligned} \min_{x,y} \quad & f(x, y) \\ \text{s.c.} \quad & (x, y) \in X, \\ & y \in P(x). \end{aligned}$$

$$\begin{aligned} \min_{x,y} \quad & f(x, y) \\ \text{s.c.} \quad & (x, y) \in X, \\ \min_y \quad & g(x, y) \\ \text{s.c.} \quad & (x, y) \in Y. \end{aligned}$$



■ Bilevel Programs

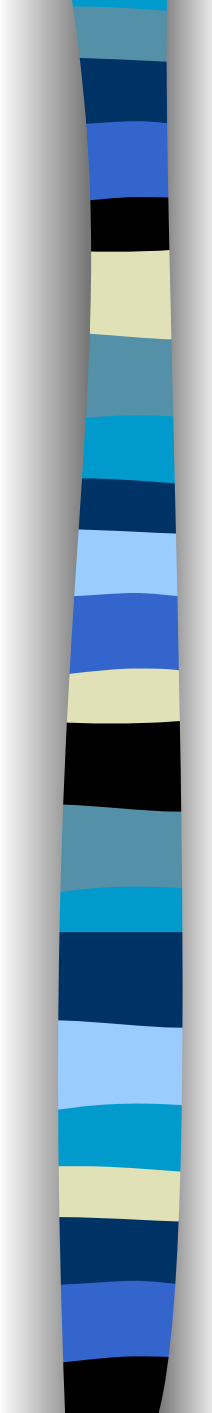
$$\begin{aligned} & \min_{x,y} f(x, y) \\ & \text{s.c. } (x, y) \in X, \\ & \min_y g(x, y) \\ & \text{s.c. } (x, y) \in Y. \end{aligned}$$

■ Difficulties

- Leader's domain is generally non convex and discontinuous

$$IR = \{(x, y) : (x, y) \in X, y \in P(x)\}$$

- $y(x)$ is generally non differentiable

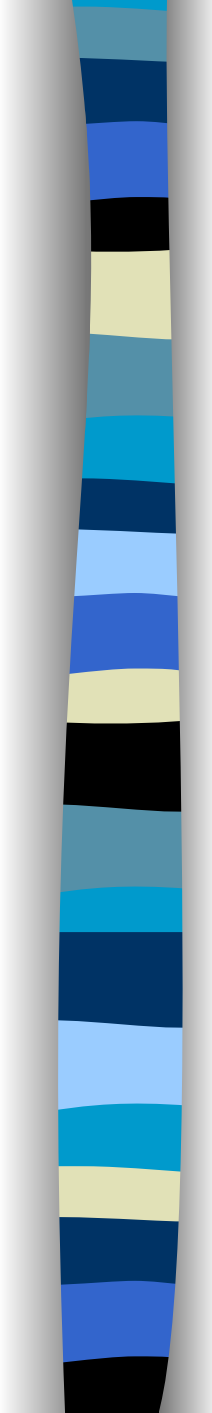


■ Properties

- Non convex problems
- Optimal solution non pareto-optimal
- Linear Linear bilevel programs:
 - Sharply NP-Difficult
 - Prove optimality is NP-Difficult

■ Algorithms

- General Case: Penalties methods, Descent methods,...
- Linear Linear bilevel programs: Branch and Bound, Extremal Points, Descent methods, Penalties methods,...



■ Applications

- Tarification
- Network Design
- Planning
- ...

■ References

- Shimizu K., Ishizuka Y. and Bard J.F., *Non Differentiable and two-level mathematical programming*, Kluwer Academic Publisher, 1997.
- Vicente N. and Calamai P.H., Bilevel and multilevel programming: a bibliography, *Journal of Global Optimization*, 5, PP. 291-306, 1994

8. Bilevel Model for JDP

$k \in K$: Set of origin destination pairs $((o(k), d(k)))$
 n^k : Demand for a commodity k

C_a : Operating cost per unit on $a \in A_1$

T_a : **Tariff** per unit on $a \in A_1$

f_a : fixed opening cost on $a \in A_1$

v_a : **Binary variable** for $a \in A_1$

d_a : Transshipment cost per unit on $a \in A_2$

x_a^k : Commodity k **flow** on $a \in A_1$

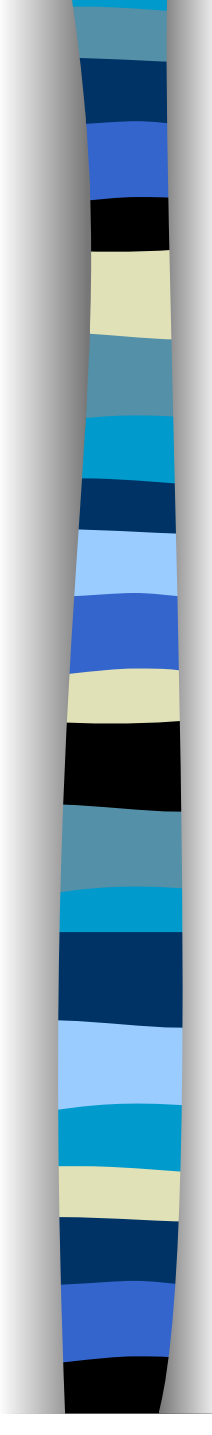
y_a^k : Commodity k **flow** on $a \in A_2$

Bilevel Model

$$\begin{aligned}
 & \max_{T, v} \quad \boxed{\sum_{k \in \mathcal{K}} \sum_{a \in \mathcal{A}_1} n^k \underline{T_a x_a^k} - \sum_{a \in \mathcal{A}_1} f_a \underline{v_a} - \sum_{k \in \mathcal{K}} \sum_{a \in \mathcal{A}_1} n^k \underline{c_a x_a^k}} \quad \text{Revenue} \\
 & \text{s.t.} \quad v_a \in \{0, 1\} \quad \forall a \in \mathcal{A}_1, \\
 & \min_{x, y} \quad \boxed{\sum_{k \in \mathcal{K}} n^k \left(\sum_{a \in \mathcal{A}_1} \underline{T_a x_a^k} + \sum_{a \in \mathcal{A}_2} \underline{d_a y_a^k} \right)} \quad \text{Customer cost} \\
 & \quad \underline{A_1 x^k} + \underline{A_2 y^k} = e^k \quad \forall k \in \mathcal{K}, \\
 & \quad \underline{x_a^k} \leq \underline{v_a} \quad \forall k \in \mathcal{K} \quad \forall a \in \mathcal{A}_1, \\
 & \quad x^k, y^k \geq 0 \quad \forall k \in \mathcal{K}.
 \end{aligned}$$

The operator (leader) $\Rightarrow$ maximizes the net revenue

The customers (followers) $\Rightarrow$ minimize the cost

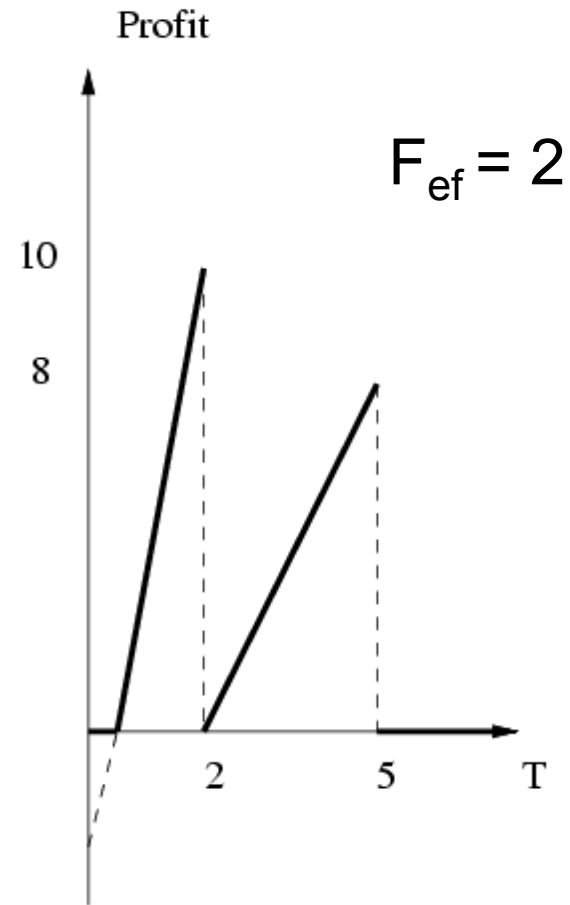
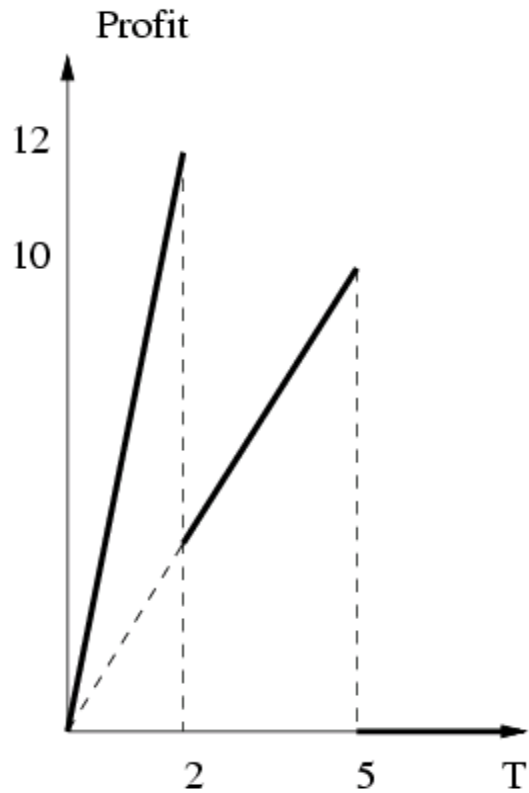
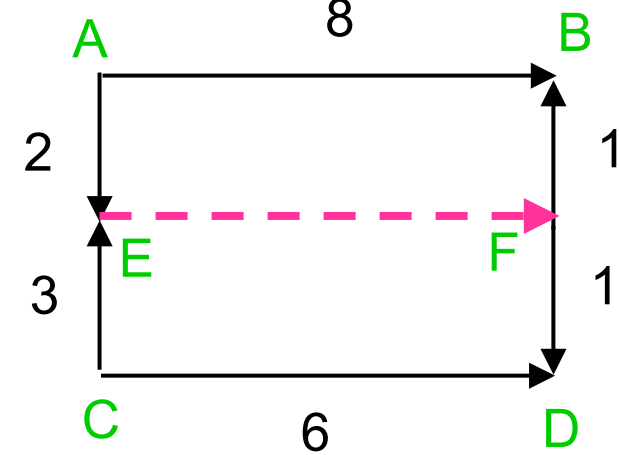


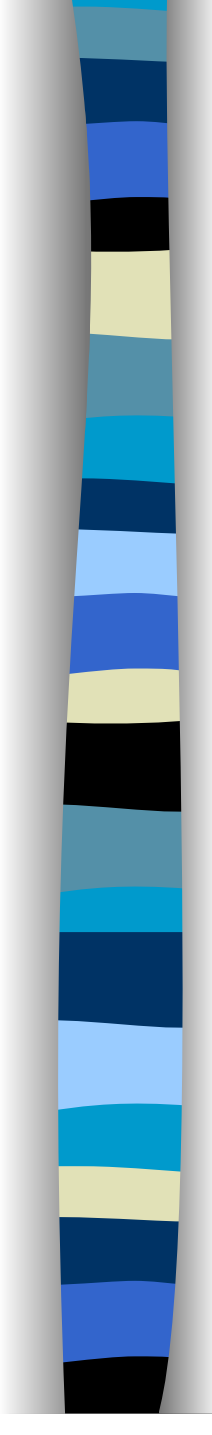
Assumptions

1. If the set of solutions of (FP) not single valued
⇒ The customers select minimal cost paths
generating the highest profit
2. $\exists$ one untariff path for each O-D pair
3. $\exists$ a tariff setting scheme generating profit and
/creating negative cost cycles

Profit Function

$$\begin{aligned} 2: 1 &\rightarrow 2 \\ 4: 3 &\rightarrow 4 \end{aligned}$$

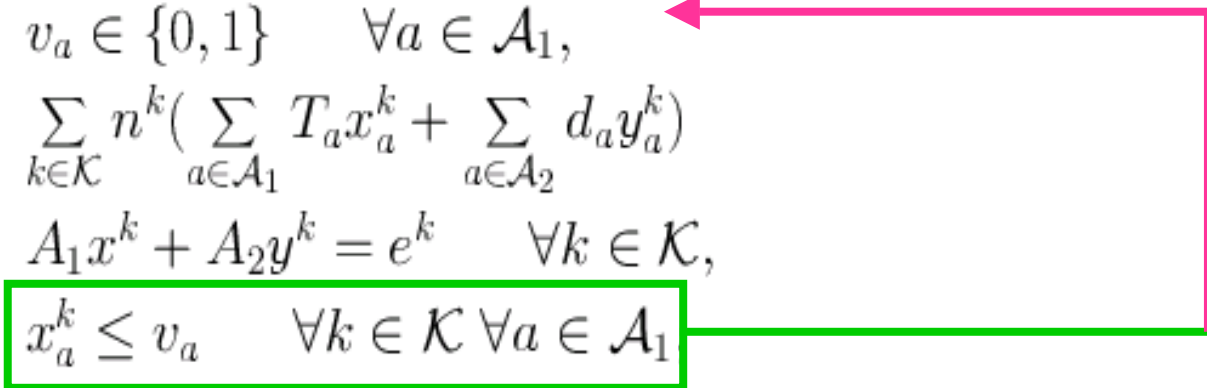




Properties

1. One commodity
 $\Rightarrow$ JDP is a toll setting problem
2. Upper bound on the profit
3. Moving constraints from the lower level to the upper level

Moving constraints from second to first level

$$\begin{aligned} \max_{T,v} \quad & \sum_{k \in \mathcal{K}} \sum_{a \in \mathcal{A}_1} n^k T_a x_a^k - \sum_{a \in \mathcal{A}_1} f_a v_a - \sum_{k \in \mathcal{K}} \sum_{a \in \mathcal{A}_1} n^k c_a x_a^k \\ \text{s.t.} \quad & v_a \in \{0, 1\} \quad \forall a \in \mathcal{A}_1, \\ \min_{x,y} \quad & \sum_{k \in \mathcal{K}} n^k \left(\sum_{a \in \mathcal{A}_1} T_a x_a^k + \sum_{a \in \mathcal{A}_2} d_a y_a^k \right) \\ & A_1 x^k + A_2 y^k = e^k \quad \forall k \in \mathcal{K}, \\ & x_a^k \leq v_a \quad \forall k \in \mathcal{K} \forall a \in \mathcal{A}_1 \\ & x^k, y^k \geq 0 \quad \forall k \in \mathcal{K}. \end{aligned}$$


Solution Procedure for JDP

Lagrangian Relaxation

$$x_a^k \leq v_a \quad \forall k \in \mathcal{K}, \forall a \in \mathcal{A}_1$$

$$u = (u_a^k) \geq 0,$$

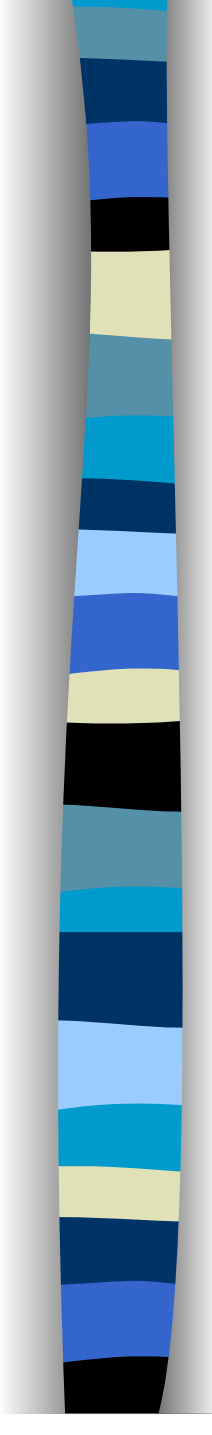
$$\begin{aligned} (LSP(u)) \quad ZL(u) = & \max_{T, v, x, y, \lambda} \sum_{a \in \mathcal{A}_1} \sum_{k \in \mathcal{K}} n^k T_a x_a^k - \sum_{a \in \mathcal{A}_1} f_a v_a \\ & - \sum_{a \in \mathcal{A}_1} \sum_{k \in \mathcal{K}} n^k c_a x_a^k + \sum_{a \in \mathcal{A}_1} \sum_{k \in \mathcal{K}} u_a^k (v_a - x_a^k) \\ \text{s.t.} \quad & v_a \in \{0, 1\} \quad \forall a \in \mathcal{A}_1, \\ \min_{x, y} \quad & \sum_{k \in \mathcal{K}} n^k \left(\sum_{a \in \mathcal{A}_1} T_a x_a^k + \sum_{a \in \mathcal{A}_2} d_a y_a^k \right) \\ \text{s.t.} \quad & Ax^k + By^k = e^k \quad \forall k \in \mathcal{K}, \\ & x^k, y^k \geq 0 \quad \forall k \in \mathcal{K}, \end{aligned}$$

Dual Lagrangian Problem

$$\min_{u \geq 0} z(u)$$

Primal Dual Algorithm for Lagrangian Relaxation

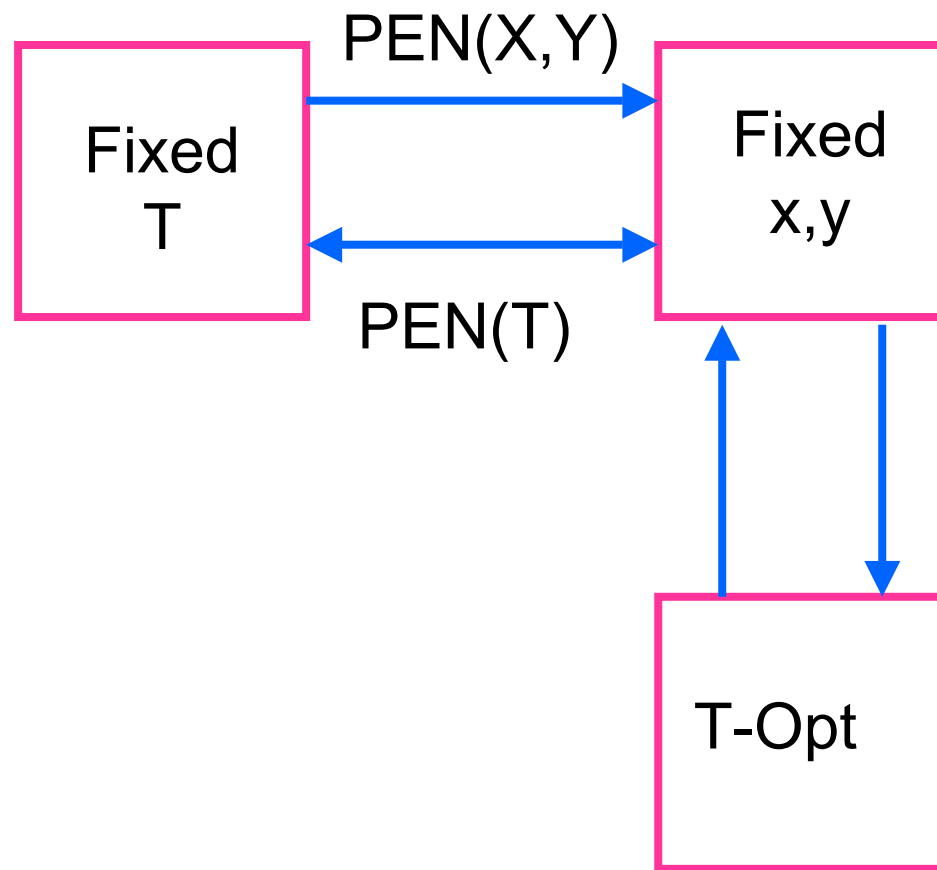
$$\begin{aligned}
 (PEN) \quad & \max_{T,v,x,y,\lambda} \quad \sum_{k \in \mathcal{K}} \sum_{a \in \mathcal{A}_1} n^k T_a x_a^k - \sum_{a \in \mathcal{A}_1} f_a v_a - \sum_{k \in \mathcal{K}} \sum_{a \in \mathcal{A}_1} n^k c_a x_a^k \\
 & + \sum_{k \in \mathcal{K}} \sum_{a \in \mathcal{A}_1} u_a^k (v_a - x_a^k) - M_1 \sum_{k \in \mathcal{K}} n^k (T x^k + d y^k - \lambda^k e^k) \\
 \text{s.t.} \quad & A_1 x^k + A_2 y^k = e^k \quad \forall k \in \mathcal{K}, \\
 & \lambda^k A_1 \leq T \quad \forall k \in \mathcal{K}, \\
 & \lambda^k A_2 \leq d \quad \forall k \in \mathcal{K}, \\
 & v_a \in \{0, 1\} \quad \forall a \in \mathcal{A}_1, \\
 & x_a^k \geq 0 \quad \forall k \in \mathcal{K}, \forall a \in \mathcal{A}_1, \\
 & y_a^k \geq 0 \quad \forall k \in \mathcal{K}, \forall a \in \mathcal{A}_2.
 \end{aligned}$$



Binaries Variables : v

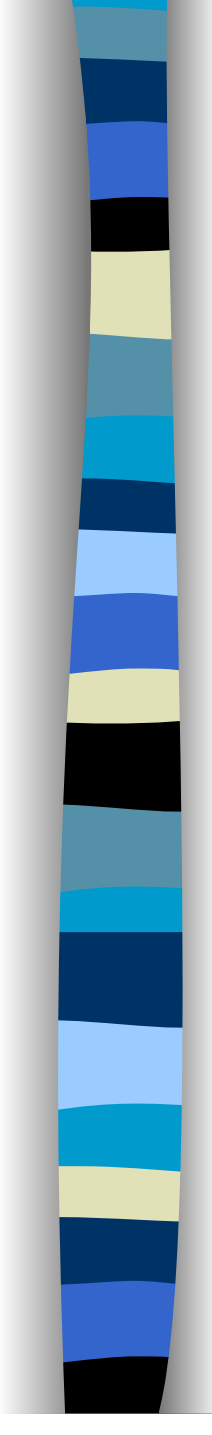
$$\begin{aligned} (PEN1(v)) \quad & \max_v \quad \sum_{a \in \mathcal{A}_1} \sum_{k \in \mathcal{K}} (u_a^k - f_a) v_a \\ & \text{s.t.} \quad v_a \in \{0, 1\} \quad \forall a \in \mathcal{A}_1. \end{aligned}$$

Gauss-Seidel method for PEN



Inverse Optimization Program

$$\begin{aligned} (T - OPT) \quad & \max_{T, \lambda} \quad \sum_{a \in \mathcal{A}_1} \sum_{k \in \mathcal{K}} n^k T_a x_a^k - \sum_{a \in \mathcal{A}_1} \sum_{k \in \mathcal{K}} n^k c_a x_a^k \\ \text{s.t.} \quad & \lambda^k A \leq T \quad \forall k \in \mathcal{K}, \\ & \lambda^k B \leq d \quad \forall k \in \mathcal{K}, \\ & n^k (T x^k + d y^k - \lambda^k e^k) = 0 \quad \forall k \in \mathcal{K}. \end{aligned}$$



Numerical Results

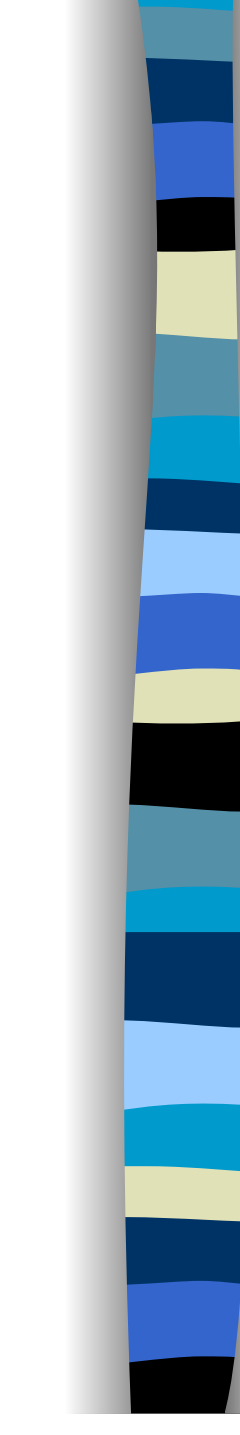
- Grid Networks : 60 nodes- 208 arcs
- Commodities: 10 – 20
- 5% to 20% of tariff arcs
- Heuristic / MIP formulation solved with Cplex

Heuristic

Cplex

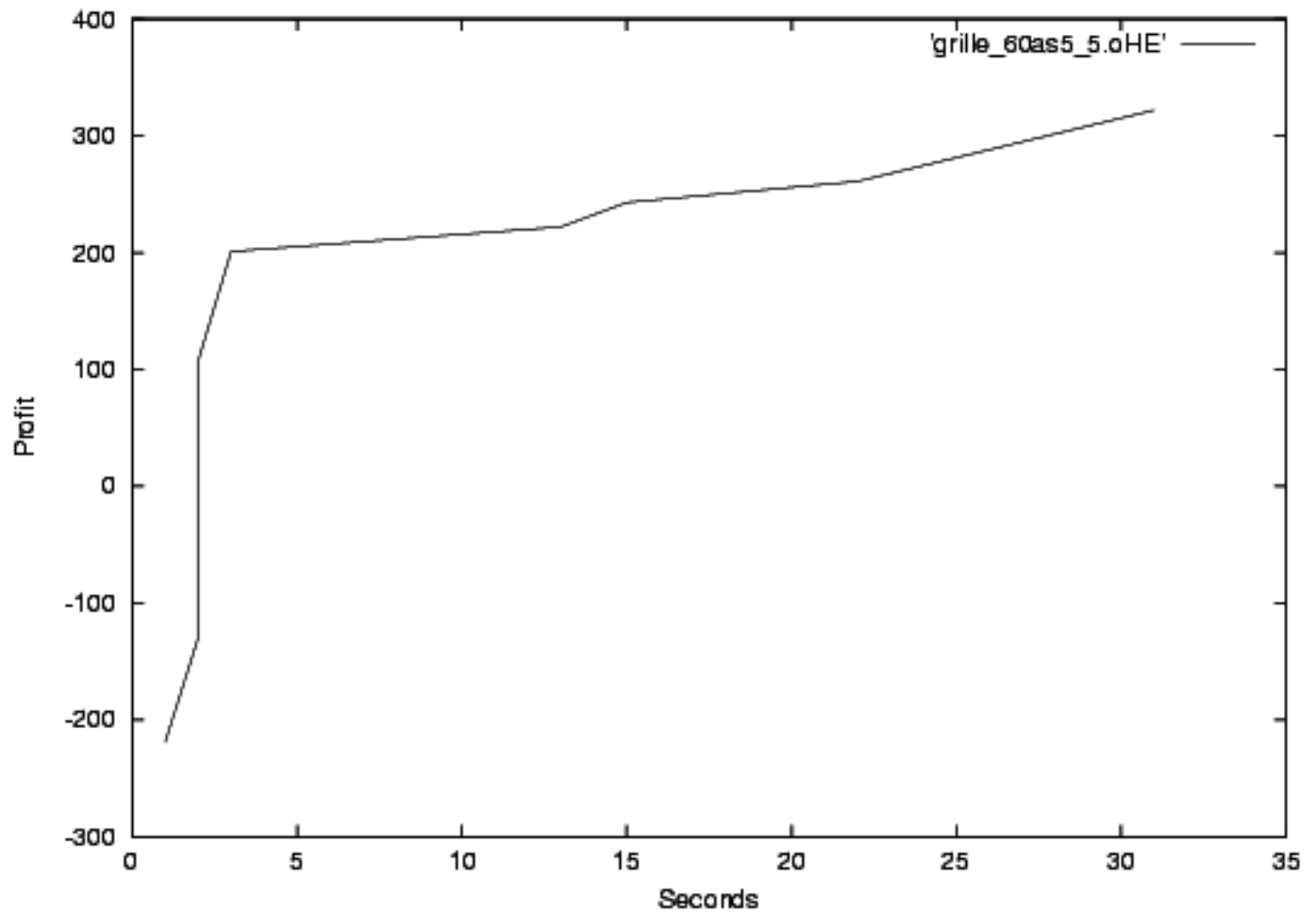
%TA	#TA	#BA	%OPT	CPU	GAP	CPU
5	4.8	159.6	0.98	62.0	73.62	433.0
10	5.4	187.4	0.97	84.8	58.20	781.0
15	9.8	196.0	1.08	118.4	46.95	10371.4
20	12.6	348.2	1.22	268.6	51.28	9223.8

Grid networks (60,208) - 20 Commodities

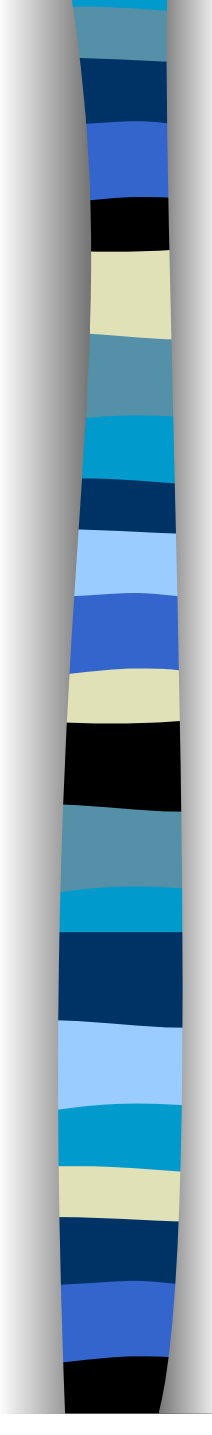


Heuristic				Cplex		
f_a	%TA	NOPT	%OPT	CPU	GAP	CPU
0	5	5	0.99	28.4	30.29	1546.
	15	0	1.09	53.4	28.61	4958.
30	5	5	0.98	6.02	73.62	433.0
	15	4	1.08	118.4	46.95	10371
60	5	5	1.00	67.8	111.81	35.0
	15	5	1.00	140.8	44.92	2333.

Grid networks (60,208) - 20 Commodities
Fixed cost Sensitivity



Grid networks (60,208) - 20 Commodities
Upper level envelopes on profit function



9. Conclusion

Algorithm based on a novel application of Lagrangian relaxation within a bilevel programming framework

Extensions :

capacities on the link of the network

Yield management (determine the good price and the good services availability at the right time)