

Optimal day-ahead bidding strategy with futures and bilateral contracts. Scenario generation by means of factor models

C. Corchero, F.J. Heredia, M.P. Muñoz

Group of Numerical Optimization and Modelling - GNOM

Universitat Politècnica de Catalunya - UPC, Spain

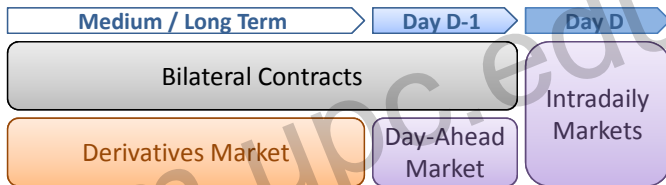
<http://gnom.upc.edu>

Project DPI2008-02154, Ministry of Science and Innovation, Spain

July 10, 2010

- 1 Introduction
 - MIBEL
 - Physical Futures and Bilateral Contracts in the MIBEL
- 2 Optimization Model and Optimal Bid Function
 - Problem definition
 - Two-stage stochastic program formulation
 - Optimal Matched Energy
 - Optimal Bid Function
- 3 Factor models
 - Day-Ahead Market price
 - Factor model estimation
 - Forecasting model
 - Factor Model Results
- 4 Results and Conclusions
 - Case Study characteristics
 - Results
 - Conclusions
 - Conclusions

Electric Energy Iberian Market: MIBEL



Derivatives Market

| Physical Futures Contracts |
|---|
| Financial and Physical Settlement. Positions are sent to OMEL's Mercado Diario for physical delivery. |
| Financial Futures Contracts |
| OMIClear cash settles the differences between the Spot Reference Price and the Final Settlement Price |

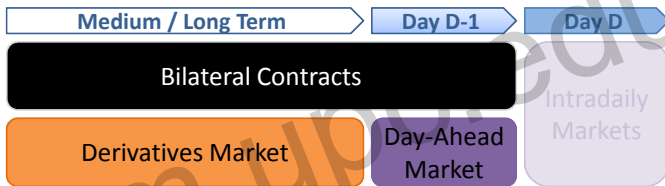
Bilateral Contracts

| Organized markets |
|---|
| - Virtual Power Plants auctions (EPE) |
| - Distribution auctions (SD) |
| - International Capacity Interconnection auctions |
| - International Capacity Interconnection nomination |
| Non organized markets |
| - National BC before the spot market |
| - International BC before the spot market |
| - National BC after the spot market |

Day-Ahead Market

| Day-Ahead Market |
|---|
| Hourly action. The matching procedure takes place 24h before the delivery period. |
| Physical futures contracts are settled through a zero price bid. |

Electric Energy Iberian Market: MIBEL



Derivatives Market

| Physical Futures Contracts |
|---|
| Financial and Physical Settlement. Positions are sent to OMEL's Mercado Diario for physical delivery. |
| Financial Futures Contracts |
| OMIClear cash settles the differences between the Spot Reference Price and the Final Settlement Price |

Bilateral Contracts

| Organized markets |
|---|
| <ul style="list-style-type: none"> - Virtual Power Plants auctions (EPE) - Distribution auctions (SD) - International Capacity Interconnection auctions - International Capacity Interconnection nomination |
| Non organized markets |
| <ul style="list-style-type: none"> - National BC before the spot market - International BC before the spot market - National BC after the spot market |

Day-Ahead Market

| Day-Ahead Market |
|---|
| Hourly action. The matching procedure takes place 24h before the delivery period. |
| Physical futures contracts are settled through a zero price bid. |

Characteristics of Physical Futures and Bilateral Contracts

Base Load Futures Contract

- A *Base Load Futures Contract* consists in a pair (L^{FC}, λ^{FC})
 - L^{FC} : amount of energy (MWh) to be procured each interval of the delivery period.
 - λ^{FC} : price of the contract (c€/MWh).

Bilateral Contracts

- A *Bilateral Contract* consists in a pair $(L_i^{BC}, \lambda_i^{BC})$ $i \in I$
 - L_i^{BC} : amount of energy (MWh) to be procured each interval i of the delivery period.
 - λ_i^{BC} : price of the contract (c€/MWh).

Characteristics of Physical Futures and Bilateral Contracts

Base Load Futures Contract

- A *Base Load Futures Contract* consists in a pair (L^{FC}, λ^{FC})
 - L^{FC} : amount of energy (MWh) to be procured each interval of the delivery period.
 - λ^{FC} : price of the contract (c€/MWh).

Bilateral Contracts

- A *Bilateral Contract* consists in a pair $(L_i^{BC}, \lambda_i^{BC})$ $i \in I$
 - L_i^{BC} : amount of energy (MWh) to be procured each interval i of the delivery period.
 - λ_i^{BC} : price of the contract (c€/MWh).

Integration of the futures and bilateral contracts in the day-ahead bid

The energies L^{FC} and L_i^{BC} should be integrated in the MIBEL's day-ahead bid respecting the two following rules:

- If generator t contributes with b_{it} MWh at period i to the coverage of the pool of FCs, then the energy b_{it} must be offered to the pool for sale (instrumental price bid).
- If generator t contributes with b_{it} MWh at period i to the coverage of the pool of BCs, then the energy b_{it} must be excluded from the bid to the day-ahead market. Unit t can offer its remaining production capacity $\bar{P}_t - b_{it}$ to the pool.

Integration of the futures and bilateral contracts in the day-ahead bid

The energies L^{FC} and L_i^{BC} should be integrated in the MIBEL's day-ahead bid respecting the two following rules:

- 1 If generator t contributes with f_{itj} MWh at period i to the coverage of the FC j , then the energy f_{itj} must be offered to the pool for free (**instrumental price bid**).
- 2 If generator t contributes with b_{it} MWh at period i to the coverage of the pool of BCs, then the energy b_{it} must be excluded from the bid to the day-ahead market. Unit t can offer its remaining production capacity $\bar{P}_t - b_{it}$ to the pool.

Integration of the futures and bilateral contracts in the day-ahead bid

The energies L^{FC} and L_i^{BC} should be integrated in the MIBEL's day-ahead bid respecting the two following rules:

- 1 If generator t contributes with f_{itj} MWh at period i to the coverage of the FC j , then the energy f_{itj} must be offered to the pool for free (**instrumental price bid**).
- 2 If generator t contributes with b_{it} MWh at period i to the coverage of the pool of BCs, then the energy b_{it} must be excluded from the bid to the day-ahead market. Unit t can offer its remaining production capacity $\bar{P}_t - b_{it}$ to the pool.

Day-ahead market model: definitions

Definition (Bid function)

A bid function for the thermal unit t is a non-decreasing function defined over the interval $[0, \bar{P}_t - b_{it}]$ that gives, for any feasible value of the bid energy p_{it} , the asked price per MWh from the day-ahead market:

$$\begin{aligned} \lambda_{it}^b: \quad [0, \bar{P}_t - b_{it}] &\longrightarrow \mathbb{R}^+ \cup 0 \\ p_{it} &\longmapsto \lambda_{it}^b(p_{it}) \end{aligned}$$

Day-ahead market model: definitions

Definition (Matched energy function)

The matched energy associated to the bid function λ_{it}^b is defined as the maximum bid energy with an asked price not greater than the clearing price λ_i , and is represented by the function:

$$p_{it}^M(\lambda_i) \stackrel{\text{def}}{=} \max\{p_{it} \in [0, \bar{P}_t - b_{it}] \mid \lambda_{it}^b(p_{it}) \leq \lambda_i\}$$

Day-ahead market model: assumption

Assumption

For any thermal unit i committed at period t there exists a bid function λ_{it}^b such that:

$$p_{it}^{M,S*} = p_{it}^M(\lambda_i^S) \quad \forall s \in S \quad (1)$$

with $p_{it}^{M,S}$ the optimal value of the matched energy variable $p_{it}^{M,S}$*

Problem definition

The objective of the study is to decide:

- the optimal economic dispatch of the physical futures and bilateral contract among the thermal units
- the optimal bidding at the Day-Ahead Market abiding by the MIBEL rule.

and the optimal unit commitment of the thermal units maximizing the expected Day-Ahead Market profits taking into account futures and bilateral contracts.

Problem definition

The objective of the study is to decide:

- the **optimal economic dispatch of the physical futures and bilateral contract** among the thermal units
- the **optimal bidding at Day-Ahead Market** abiding by the MIBEL rules
- the **optimal unit commitment** of the thermal units

maximizing the expected Day-Ahead Market profits taking into account futures and bilateral contracts.

Problem definition

The objective of the study is to decide:

- the **optimal economic dispatch of the physical futures and bilateral contract** among the thermal units
- the **optimal bidding at Day-Ahead Market** abiding by the MIBEL rules
- the **optimal unit commitment** of the thermal units

maximizing the expected Day-Ahead Market profits taking into account futures and bilateral contracts.

Problem definition

The objective of the study is to decide:

- the **optimal economic dispatch of the physical futures and bilateral contract** among the thermal units
- the **optimal bidding at Day-Ahead Market** abiding by the MIBEL rules
- the **optimal unit commitment** of the thermal units

maximizing the expected Day-Ahead Market profits taking into account futures and bilateral contracts.

Problem definition

The objective of the study is to decide:

- the **optimal economic dispatch of the physical futures and bilateral contract** among the thermal units
- the **optimal bidding at Day-Ahead Market** abiding by the MIBEL rules
- the **optimal unit commitment** of the thermal units

maximizing the expected Day-Ahead Market profits taking into account futures and bilateral contracts.

Problem definition

Model characteristics

- Stochastic mixed integer quadratic programming model
- *Price-taker* generation company
- Set of thermal generation units, T
- Optimization horizon of 24h, I
- Set of physical futures contracts, F , of energy L_j^{FC} $j \in F$.
- A pool of bilateral contracts of energy L^{BC} .
- Set of day-ahead market price scenarios, λ^s , $s \in \mathcal{S}$

Variables

First stage variables: $\forall t \in T, \forall i \in I$

- Unit commitment: $u_i^t, c_t^u, c_d^t \in \{0, 1\}$
- Instrumental price offer bid : q_i^t
- Scheduled energy for futures contract j : $f_{itj} \quad \forall j \in F$
- Scheduled energy for bilaterals contract: b_{it}

Second stage variables $\forall t \in T, \forall i \in I, \forall s \in S$

- Matched energy: $p_{it}^{M,s}$
- Total generation: p_{it}^S

Variables

First stage variables: $\forall t \in T, \forall i \in I$

- Unit commitment: $u_i^t, c_t^u, c_d^t \in \{0, 1\}$
- Instrumental price offer bid : q_i^t
- Scheduled energy for futures contract j : $f_{itj} \quad \forall j \in F$
- Scheduled energy for bilaterals contract: b_{it}

Second stage variables $\forall t \in T, \forall i \in I, \forall s \in S$

- Matched energy: $p_{it}^{M,s}$
- Total generation: p_{it}^S

Physical Future and Bilateral Contracts model

Physical future contract coverage:

$$\sum_{t \in T} f_{itj} = L_j^{FC}, \forall j \in F, \forall i \in I$$
$$f_{itj} \geq 0, \forall j \in F, \forall t \in T, \forall i \in I$$

Bilateral contract coverage:

$$\sum_{t \in T} b_{it} = L_i^{BC}, \forall i \in I$$
$$0 \leq b_{it} \leq \bar{P}_t, \forall t \in T, \forall i \in I$$

Physical Future and Bilateral Contracts model

Physical future contract coverage:

$$\sum_{t \in T} f_{itj} = L_j^{FC}, \forall j \in F, \forall i \in I$$
$$f_{itj} \geq 0, \forall j \in F, \forall t \in T, \forall i \in I$$

Bilateral contract coverage:

$$\sum_{t \in T} b_{it} = L_i^{BC}, \forall i \in I$$
$$0 \leq b_{it} \leq \bar{P}_t, \forall t \in T, \forall i \in I$$

Day-ahead market model: constraints

Matched energy:

$$p_{it}^{M,S} \leq \bar{P}_t - b_{it}, \forall t \in T, \forall i \in I, \forall s \in S$$

$$p_{it}^{M,S} \geq q_{it}, \forall t \in T, \forall i \in I, \forall s \in S$$

Instrumental price bid:

$$q_{it} \geq \underline{P}_t - b_{it}, \forall t \in T, \forall i \in I$$

$$q_{it} \geq 0, \forall t \in T, \forall i \in I$$

$$q_{it} \geq \sum_{j \in F_i} f_{itj}, \forall t \in T, \forall i \in I$$

Total energy generation:

$$p_{it}^S = b_{it} + p_{it}^{M,S}, \forall t \in T, \forall i \in I, \forall s \in S$$

Day-ahead market model: constraints

Matched energy:

$$p_{it}^{M,s} \leq \bar{P}_t - b_{it}, \forall t \in T, \forall i \in I, \forall s \in S$$

$$p_{it}^{M,s} \geq q_{it}, \forall t \in T, \forall i \in I, \forall s \in S$$

Instrumental price bid:

$$q_{it} \geq \underline{P}_t - b_{it}, \forall t \in T, \forall i \in I$$

$$q_{it} \geq 0, \forall t \in T, \forall i \in I$$

$$q_{it} \geq \sum_{j \in F_i} f_{itj}, \forall t \in T, \forall i \in I$$

Total energy generation:

$$p_{it}^s = b_{it} + p_{it}^{M,s}, \forall t \in T, \forall i \in I, \forall s \in S$$

Day-ahead market model: constraints

Matched energy:

$$p_{it}^{M,s} \leq \bar{P}_t - b_{it}, \forall t \in T, \forall i \in I, \forall s \in S$$

$$p_{it}^{M,s} \geq q_{it}, \forall t \in T, \forall i \in I, \forall s \in S$$

Instrumental price bid:

$$q_{it} \geq \underline{P}_t - b_{it}, \forall t \in T, \forall i \in I$$

$$q_{it} \geq 0, \forall t \in T, \forall i \in I$$

$$q_{it} \geq \sum_{j \in F_i} f_{itj}, \forall t \in T, \forall i \in I$$

Total energy generation:

$$p_{it}^s = b_{it} + p_{it}^{M,s}, \forall t \in T, \forall i \in I, \forall s \in S$$

Day-ahead market model: constraints

Other set of constraints:

- Unit commitment constraints: including the start-up and shut-down costs and the minimum operation and idle time control taking into account the initial state of the units.
- Operational limits for the total generation.

Objective function

Maximization of the day-ahead market clearing's benefits

$$\max_{p,q,f,b} \sum_{t \in T} \sum_{i \in I} \left(-c_{it}^u - c_{it}^d - c_{it}^b u_{it} + \right. \\ \left. + \sum_{s \in S} P^s \left[\lambda_t^{D_s} p_{it}^{M,s} - (c_i^l p_{it}^s + c_i^q (p_{it}^s)^2) \right] \right)$$

Incomes from Futures and bilateral contracts:

- **Futures contracts:** $\sum_{t \in T} \sum_{j \in J} (\lambda_j^{FC} - \lambda_t) L_t^{FC}$
- **Bilateral contracts:** $\sum_{t \in T} \lambda_t^{BC} L_t^{BC}$
- They don't depend on the decision variables.

Objective function

Maximization of the day-ahead market clearing's benefits

$$\max_{p,q,f,b} \sum_{t \in T} \sum_{i \in I} \left(-c_{it}^u - c_{it}^d - c_{it}^b u_{it} + \right. \\ \left. + \sum_{s \in S} P^s \left[\lambda_t^{D_s} p_{it}^{M,s} - (c_i^l p_{it}^s + c_i^q (p_{it}^s)^2) \right] \right)$$

Incomes from Futures and bilateral contracts:

- **Futures contracts:** $\sum_{t \in T} \sum_{j \in J} \left(\lambda_j^{FC} - \lambda_t \right) L_t^{FC}$
- **Bilateral contracts:** $\sum_{t \in T} \lambda_t^{BC} L_t^{BC}$
- They don't depend on the decision variables.

Summary of the model

Problem OBIFUC

(Optimal bid with **B**ilateral and **F**utures **C**ontracts)

Max Day-ahead market clearing's benefits

s.t: Physical future contract coverage

Bilateral contract coverage

Matched energy

Instrumental price bid

Total energy generation

Unit commitment

Optimal Matched Energy

Lemma

Let $x^{*'} = [p^*, p^{M,*}, q^*, f^*, b^*]'$ be an optimal solution of problem (OBIFUC). Then for any thermal unit i the optimal value of the matched energy $p_{it}^{M,S*}$ can be expressed as:

$$p_{it}^{M,S*} = \max\{q_{it}^*, \rho_{it}^s(b_{it}^*)\} \quad (2)$$

Bid's functions Optimality Conditions

Definition (Bid functions's optimality conditions)

Let $x^{*'} = [u^*, c^{*u}, c^{*d}, p^{M,*}, p^*, q^*, f^*, b^*]'$ be an optimal solution of the (OBIFUC) problem. The bid function λ_{it}^{b*} of a thermal unit i committed at period t (i.e. $i \in U_t$) is said to be optimal w.r.t. the (OBIFUC) problem and solution x^* if the value of the matched energy function associated to any scenario's clearing price λ_t^s , $p_{it}^M(\lambda_t^s)$, coincides with the optimal matched energy $p_{it}^{M,s*}$, that is:

$$p_{it}^M(\lambda_t^s) = p_{it}^{M,s*} = \max\{q_{it}^*, \rho_{it}^s(b_{it}^*)\}$$

OBIFUC's optimal bid function

Lemma (Optimal bid function)

Let $x^{*'} = [u^*, c^{*u}, c^{*d}, p^{M,*}, p^*, q^*, f^*, b^*]'$ be an optimal solution of the (OBIFUC) problem and i any thermal unit committed on period t at the optimal solution (i.e. $i \in U_t$). Then the bid function:

$$\lambda_{it}^*(p_{it}; b_{it}^*) = \begin{cases} 0 & \text{if } p_{it} \leq q_{it}^* \\ 2c_i^q (p_{it} + b_{it}^*) + c_i^f & \text{if } q_{it}^* < p_{it} \leq (\bar{P}_t - b_{it}^*) \end{cases} \quad (3)$$

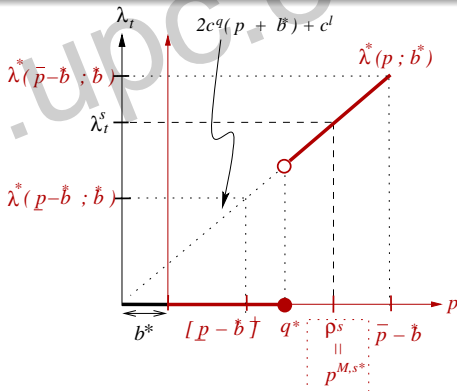
is optimal w.r.t. the (OBIFUC) problem and the optimum x^* .

OBIFUC's optimal bid function graphical representation

Matched energy at scenario s :

$$p_{it}^M(\lambda_t^s) = p_{it}^{M,s*} = \max\{q_{it}^*, \rho_{it}^s(b_{it}^*)\} \quad (4)$$

$$q_{it}^* \leq \rho_{it}^s(b_{it}^*) \implies p_{it}^{M,s*} = \rho_{it}^s(b_{it}^*)$$

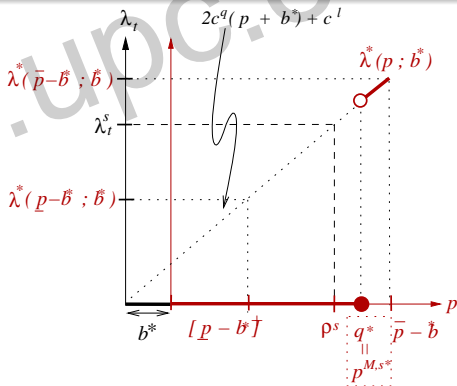


OBIFUC's optimal bid function graphical representation

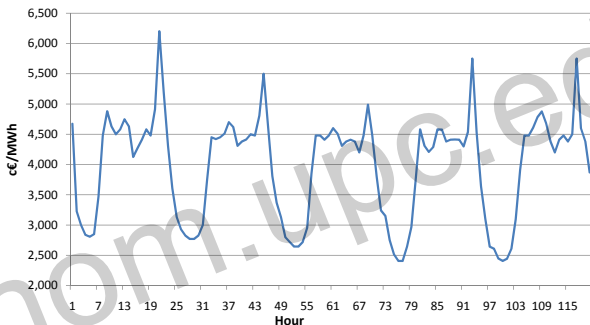
Matched energy at scenario s :

$$p_{it}^M(\lambda_t^s) = p_{it}^{M,s*} = \max\{q_{it}^*, \rho_{it}^s(b_{it}^*)\} \quad (4)$$

$$q_{it}^* \geq \rho_{it}^s(b_{it}^*) \implies p_{it}^{M,s*} = q_{it}^*$$



Price characteristics



Electricity spot prices exhibit:

- Non-constant mean and variance
- Calendar effects
- Daily and weekly seasonality
- High volatility and presence of outliers

Factor Model Approach

To apply the methodology of factor models in the next way:

- The spot price is interpreted not as a single time series but a set of 24 time series, one for each hour.
- The factor model allows to identify common unobserved factors which represent the relationship between the hours of a day.
- The forecasting model provide suitable scenarios for the optimization model.

Schema

Hourly electricity prices



Factor analysis



One-step-ahead forecast price and confidence interval



Scenario generation



Optimization model

Time Series Factor Analysis

Time Series Factor Analysis

Time Series Factor Analysis ^a (*TSFA*) estimates measurement model for time series data with as few assumptions as possible about the dynamic process governing the factors. It estimates parameters and predicts factor scores.

^aGilbert P.D., Meijer E. (2005). Time Series Factor Analysis with an Application to Measuring Money

Factor Model Estimation

Let y_t be a M -vector of observed time series of length T and k unobserved factors ($k \ll M$) collected in the K -vector ξ . The relationship between the observed time series and the factors is assumed to be linear and described by equation:

$$y_t = \alpha_t + B\xi_t + \epsilon_t$$

where α_t is an M -vector of intercept parameters, B is an $M \times k$ matrix parameter of loadings, assumed time-invariant, and ϵ is a random M -vector of measurement errors.

Parameters are estimated by **maximum likelihood**.

Forecasting model

The factors obtained have to be implemented into a forecasting model in order to obtain the price forecasts.

A **one-step-ahead forecasting model** is specified and estimated as a **linear multiple regression model** with the **factors as predictors**¹:

$$y_{t+1} = \beta \hat{\xi}_t + \alpha(L)y_t + \varepsilon_{t+1}$$

The out of the sample forecast for $y_{T+1|T}$ is given by the conditional expectation

$$y_{T+1|T} = \hat{\beta} \hat{\xi}_T + \hat{\alpha}(L)y_T$$

¹Stock J., Watson M.W. (2002). Forecasting Using Principal Components From a Large Number of Predictors

Data analysis

- Random variable: Iberian Day-Ahead Market electricity prices
- Data set: work days from January 1^{sts}, 2007 to March 30th, 2008.
- 3 significant factors, based on eigenvalues of the sample correlation matrix.
- The data has been analyzed using R (version 2.7.0) with the library TSFA available at CRAN (www.cran.r-project.org).

Iberian Day-Ahead Electricity Market price

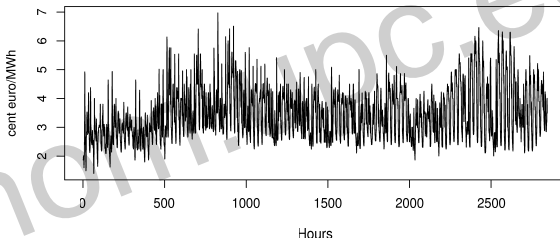


Figure 1: Iberian Day-Ahead Electricity Market price (January 1st, 2007 - March 30th, 2008)

24 Time Series

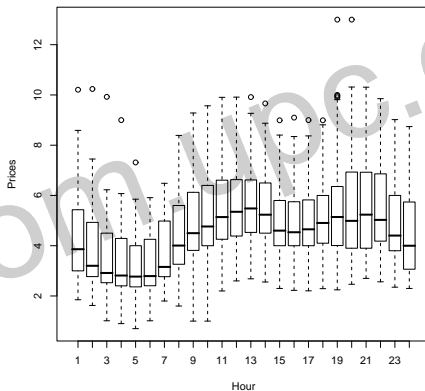


Figure 3: Iberian Day-Ahead Electricity Market price for each hour(January 1st, 2007 - March 30th, 2008)

Factor model results

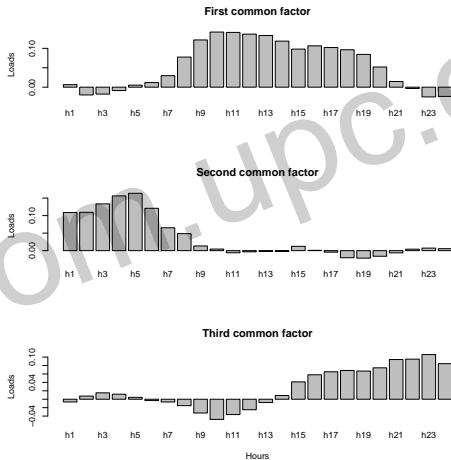


Figure 6: Loads of the common factors

Out of sample forecasting results (I/II)

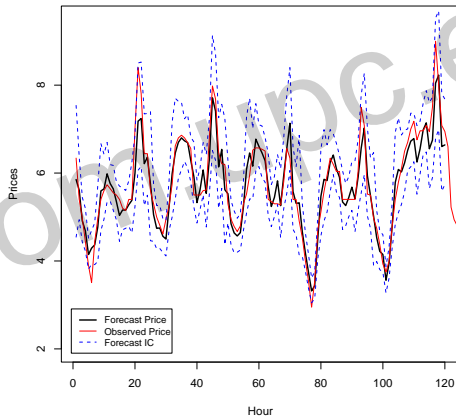


Figure 7: One-step-ahead forecast prices

Out of sample forecasting results (II/II)

| | | | | | | |
|-------|-------|-------|-------|-------|-------|-------|
| Hour | 1 | 2 | 3 | 4 | 5 | 6 |
| R^2 | 99.1 | 95.3 | 97.1 | 99.8 | 99.8 | 97.6 |
| MSE | 0.017 | 0.004 | 0.003 | 0.003 | 0.002 | 0.002 |
| Hour | 7 | 8 | 9 | 10 | 11 | 12 |
| R^2 | 96.0 | 99.6 | 99.7 | 99.8 | 96.3 | 98.3 |
| MSE | 0.003 | 0.008 | 0.008 | 0.004 | 0.003 | 0.001 |
| Hour | 13 | 14 | 15 | 16 | 17 | 18 |
| R^2 | 99.9 | 97.7 | 99.8 | 99.9 | 99.9 | 97.1 |
| MSE | 0.002 | 0.002 | 0.004 | 0.002 | 0.002 | 0.002 |
| Hour | 19 | 20 | 21 | 22 | 23 | 24 |
| R^2 | 99.7 | 96.6 | 94.2 | 99.7 | 99.7 | 95.1 |
| MSE | 0.006 | 0.005 | 0.007 | 0.007 | 0.007 | 0.005 |

Table 1: Summary of the forecast models for each hour

Case Study characteristics

- Real data from the Spanish Market about the generation company and the market prices.
- 9 thermal generation units (6 coal, 3 fuel) from a Spanish generation company with daily bidding in the MIBEL

| | | | | | |
|----------------------------------|---------|---------|---------|---------|-------|
| $[\bar{P} - \underline{P}]$ (MW) | 160-243 | 250-550 | 80-260 | 160-340 | 30-70 |
| $min_{on/off}$ (h) | 3 | 3 | 3 | 4 | 4 |
| $[\bar{P} - \underline{P}]$ (MW) | 60-140 | 160-340 | 110-157 | 110-157 | |
| $min_{on/off}$ (h) | 3 | 3 | 4 | 4 | |

- Model implemented and solved with AMPL/CPLEX 11.0.

Case Study characteristics

- Real data from the Spanish Market about the generation company and the market prices.
- 9 thermal generation units (6 coal, 3 fuel) from a Spanish generation company with daily bidding in the MIBEL

| | | | | | |
|----------------------------------|---------|---------|---------|---------|-------|
| $[\bar{P} - \underline{P}]$ (MW) | 160-243 | 250-550 | 80-260 | 160-340 | 30-70 |
| $min_{on/off}$ (h) | 3 | 3 | 3 | 4 | 4 |
| $[\bar{P} - \underline{P}]$ (MW) | 60-140 | 160-340 | 110-157 | 110-157 | |
| $min_{on/off}$ (h) | 3 | 3 | 4 | 4 | |

- Model implemented and solved with AMPL/CPLEX 11.0.

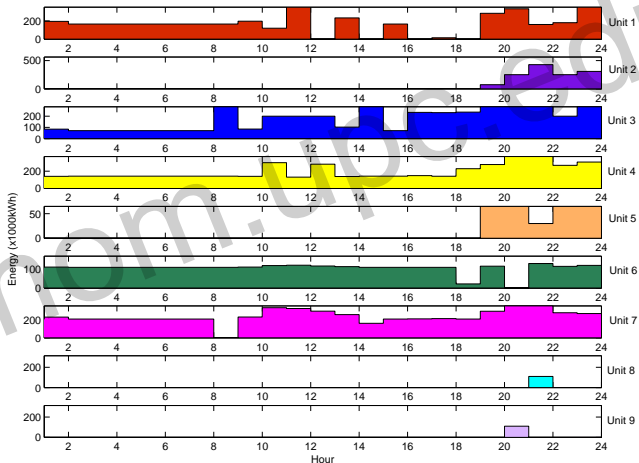
Case Study characteristics

- Real data from the Spanish Market about the generation company and the market prices.
- 9 thermal generation units (6 coal, 3 fuel) from a Spanish generation company with daily bidding in the MIBEL

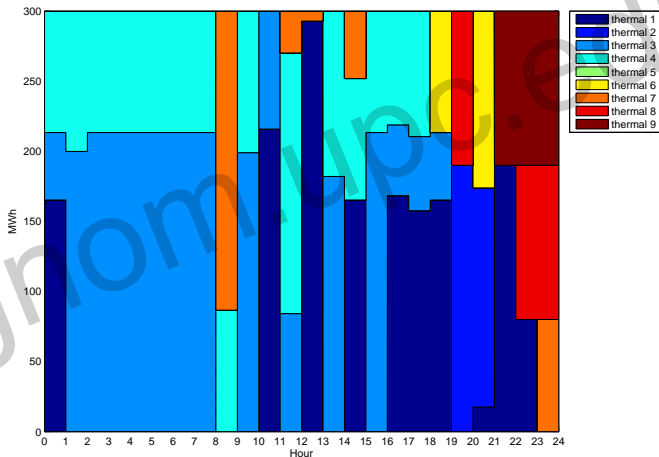
| | | | | | |
|----------------------------------|---------|---------|---------|---------|-------|
| $[\bar{P} - \underline{P}]$ (MW) | 160-243 | 250-550 | 80-260 | 160-340 | 30-70 |
| $min_{on/off}$ (h) | 3 | 3 | 3 | 4 | 4 |
| $[\bar{P} - \underline{P}]$ (MW) | 60-140 | 160-340 | 110-157 | 110-157 | |
| $min_{on/off}$ (h) | 3 | 3 | 4 | 4 | |

- Model implemented and solved with AMPL/CPLEX 11.0.

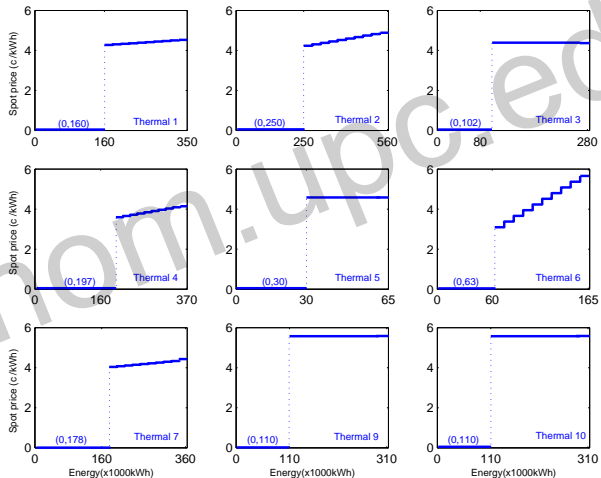
Results: unit commitment and zero price bid



Results: procurement of bilateral contracts



Results: optimal bidding curves



Conclusions

- The forecast procedure based on factor models gives suitable results, equivalent to the ones obtained through an ARIMA model.
- The advantage of the procedure presented lies in its simplicity, easy to implement and to present.
- The improved forecasts have been used to successfully generate a set of scenarios to feed the stochastic optimization model.

Conclusions

- It has been built an Optimal Bidding Model for a price-taker generation company operating both in the MIBEL Derivatives and Day-Ahead Electricity Market.
- The model developed gives the producer:
 - Optimal bid for the spot market: quantity at 0€/MWh and the rest of the power capacity at the unit's marginal cost
 - Unit commitment
 - Optimal allocation of the physical futures contracts among the thermal units

following in detail the MIBEL rules.

Optimal day-ahead bidding strategy with futures and bilateral contracts. Scenario generation by means of factor models

C. Corchero, F.J. Heredia, M.P. Muñoz

Group of Numerical Optimization and Modelling - GNOM

Universitat Politècnica de Catalunya - UPC, Spain

<http://gnom.upc.edu>

Project DPI2008-02154, Ministry of Science and Innovation, Spain

July 10, 2010