

A structural risk-neutral model for pricing and hedging electricity derivatives

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Joint work with **René Aïd** and **Luciano Campi**

Clamart, 6th Dec. 2011



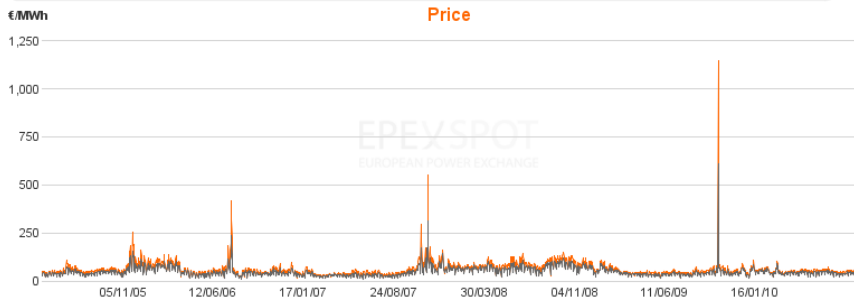
Motivation

Modelling power prices

- Joint modelling with fuel prices

Applications

- Pricing and hedging power derivatives
- Especially spread options \implies valuation of power plants



Previous Structural Model

n fuels

Variables

n	fuels, $1 \leq i \leq n$
D_t	demand (in MW)
C_t^i	capacities (in MW)
S_t^i	fuel prices
h_i	heat rates ($h_i S_t^i$ in €/MWh, \nearrow in i)

Electricity price (in €/MWh)

$$P_t = \sum_{i=1}^n h_i S_t^i \mathbf{1}_{\{\sum_{k=1}^{i-1} C_t^k \leq D_t \leq \sum_{k=1}^i C_t^k\}}$$

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Pros and cons



- Price formation mechanism
- Simple, interpretable
- Observable processes

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+

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-

- Marginal cost vs. Market price

Illustration (1/4)

Dataset

- French power market
- Time period : 13th Nov. 2006 to 30th Apr. 2010 : \sim 3.5 years
- Focus on 19th hour of each day
- 2 possible marginal fuels : coal or oil
- CO₂ price included
- Average prices (in €/MWh) : Coal \sim 47, Elec \sim 74, Oil \sim 102

Illustration (2/4)

Prices generated by the model

Spot price (in €/MWh)

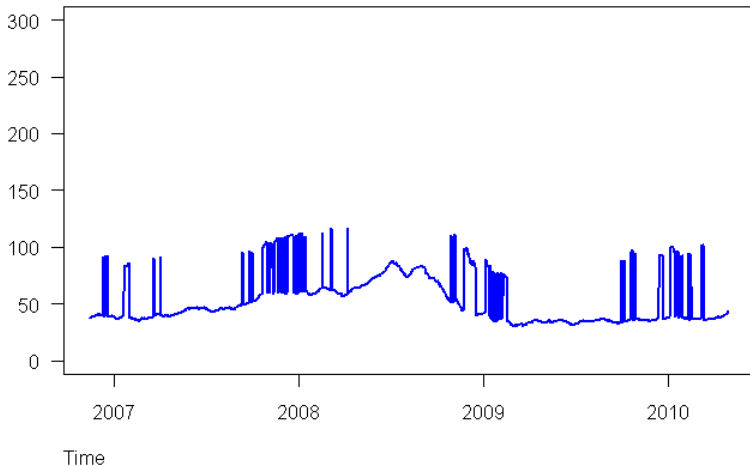


Illustration (3/4)

Comparison to realized prices

Spot price (in €/MWh)

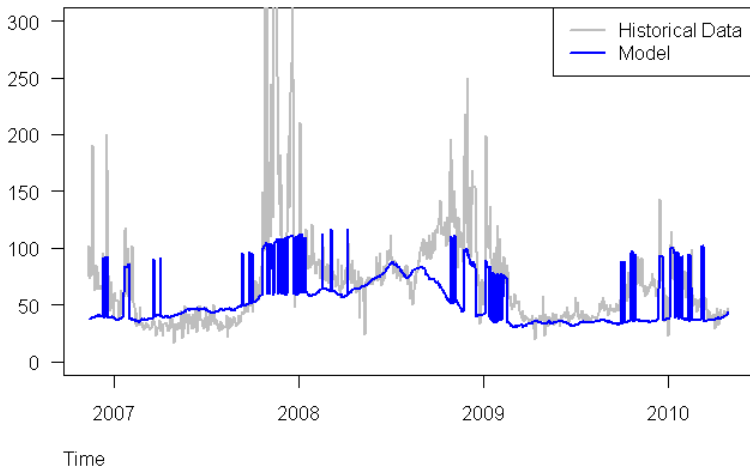
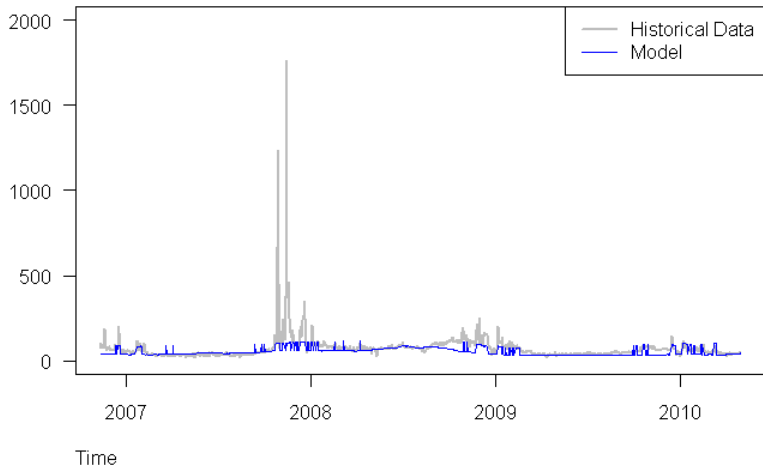


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Comparison to realized prices

Spot price (in €/MWh)



Price spikes

- Marginal cost $MC_t := \sum_{i=1}^n h_i S_t^i \mathbf{1}_{\{\sum_{k=1}^{i-1} C_t^k \leq D_t \leq \sum_{k=1}^i C_t^k\}}$

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↔ Write $\frac{P_t}{MC_t}$ as a function of RC_t

New spot model

Estimated relation : $\frac{P_t}{MC_t} = \frac{\gamma}{(RC_t)^\nu}$, $\gamma = 6.2 \pm 0.06$, $\nu = 1.0 \pm 0.01$

Model

$$P_t = g \left(\sum_{k=1}^n C_t^k - D_t \right) \times \left(\sum_{i=1}^n h_i S_t^i \mathbf{1}_{\left\{ \sum_{k=1}^{i-1} C_t^k \leq D_t \leq \sum_{k=1}^i C_t^k \right\}} \right)$$

Scarcity function

$$g(x) := \min \left(\frac{\gamma}{x^\nu}, M \right) \mathbf{1}_{\{x > 0\}} + M \mathbf{1}_{\{x \leq 0\}}$$

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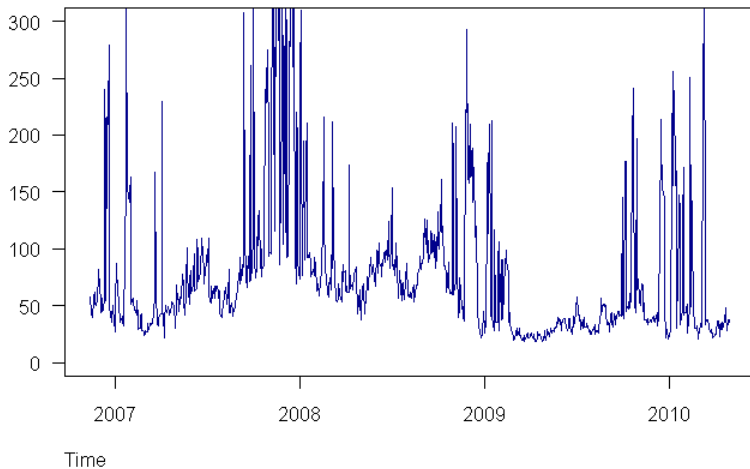
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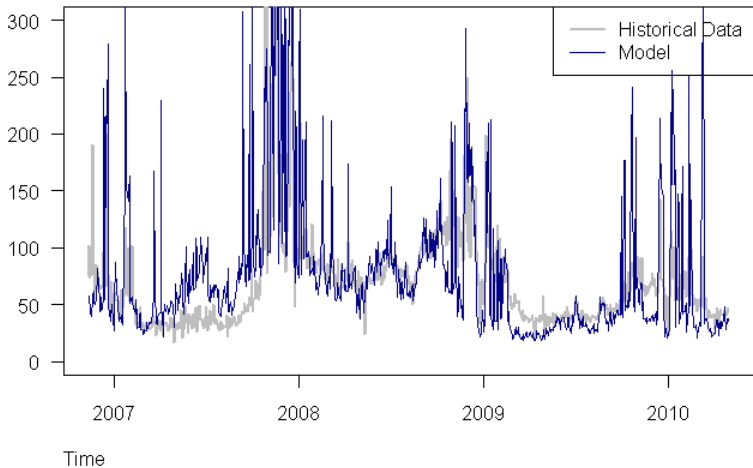
Spot price (in €/MWh)



Illustration

Comparison to realized prices

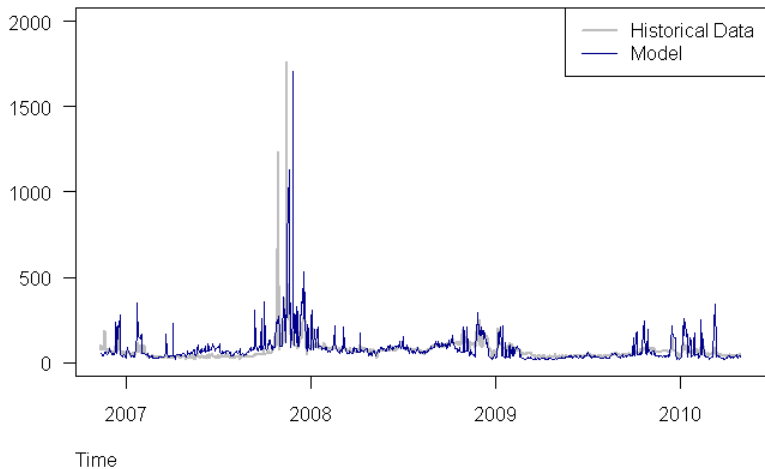
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Illustration

Comparison to realized prices

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Hedging criterion

Incomplete market

- Choice of a hedging criterion
- Ex : super-replication, utility indifference, mean-variance,...
- Our choice : **Local Risk Minimization**

Local Risk Minimization

- Explicit formulae
- Split contingent claims between the **hedgeable part** (fuels) and the **non-hedgeable part** (demand, capacities)
- Pricing : expected discounted payoff under $\hat{\mathbb{Q}}$

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Futures

Example

Pricing

$$F_t^e(T) = \sum_{i=1}^n h_i G_i^T(t, C_t, D_t) F_t^i(T)$$

Futures

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Pricing

$$F_t^e(T) = \sum_{i=1}^n h_i G_i^T(t, C_t, D_t) F_t^i(T)$$

$$G_i^T(t, C_t, D_t) = \mathbb{E}_t \left[g \left(\sum_{k=1}^n C_T^k - D_T \right) \mathbf{1}_{\left\{ \sum_{k=1}^{i-1} C_T^k \leq D_T \leq \sum_{k=1}^i C_T^k \right\}} \right]$$

Modelling the inputs

$$D_t, C_t^i, i = 1 \dots n$$

deterministic function + Ornstein-Uhlenbeck

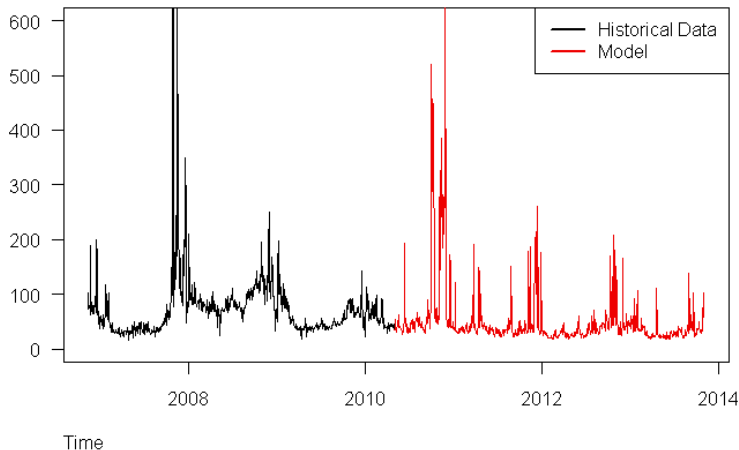
$$S_t^i, i = 1 \dots n$$

spread $Y_t^i := h_i S_t^i - h_{i-1} S_t^{i-1}$

Y_t^i : geometric Brownian motion

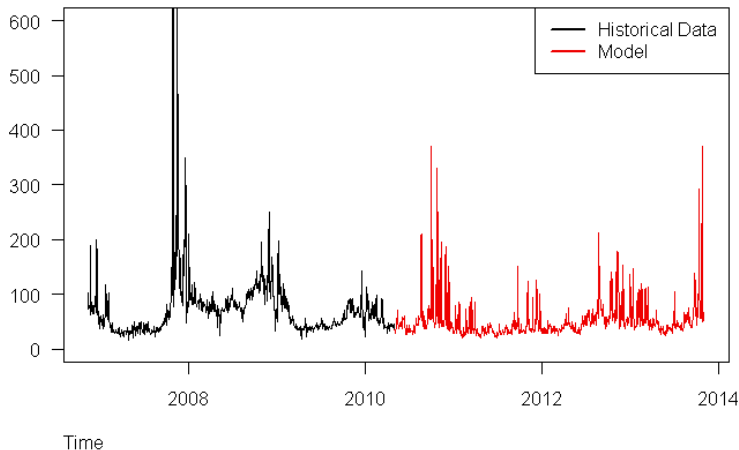
Spot Trajectories (1/3)

Spot price (in €/MWh)



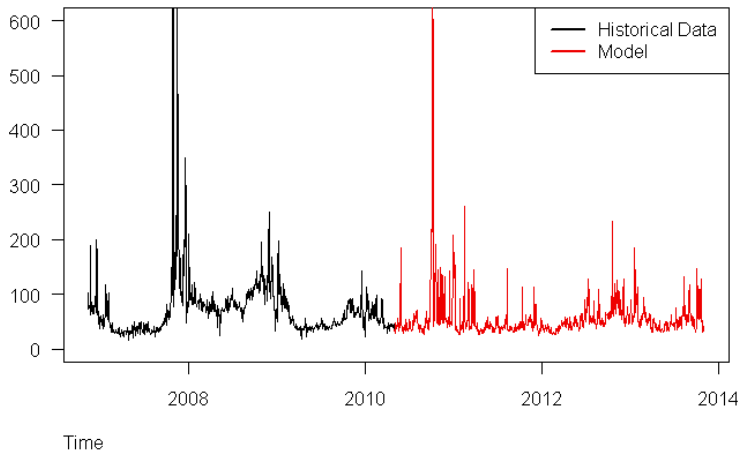
Spot Trajectories (2/3)

Spot price (in €/MWh)



Spot Trajectories (3/3)

Spot price (in €/MWh)



Simple numerical test

Partial hedging of Power Futures (1/2)

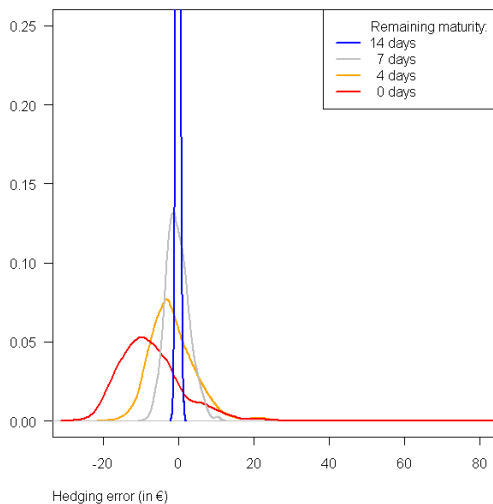
- Hedging a (fictitious) power futures with a delivery period of 1 hour

- Using a (daily rebalanced) portfolio of futures on fuels

Result

Partial hedging of Power Futures (2/2)

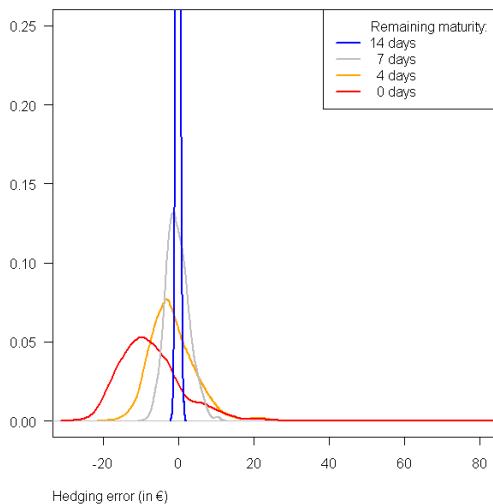
Distribution of hedging error: Time evolution



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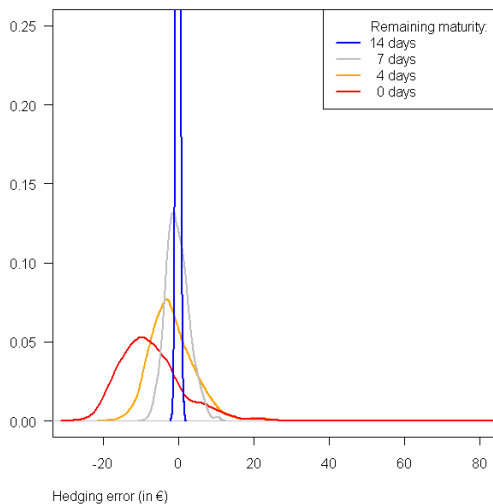


- Far from maturity : perfect hedging, basket of fuels

Result

Partial hedging of Power Futures (2/2)

Distribution of hedging error: Time evolution



- Far from maturity : perfect hedging, basket of fuels
- Close to maturity : inefficient partial hedging

Other examples

- Spread options, options on futures, . . .
- Semi-explicit pricing : numerical integration
- Partial hedging by means of fuel futures and power futures

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Investments in power plants

Spot model \implies Spread options \implies Power plants \implies Investments

Stochastic control problem

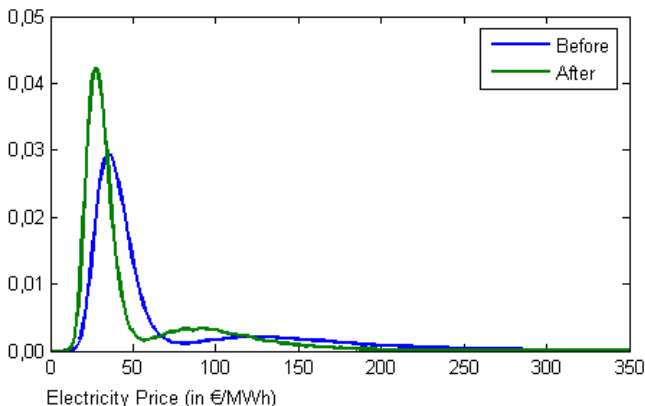
$$\sup_{I \in \mathcal{A}} \mathbb{E} \left[\int_0^{\infty} e^{-\rho s} I_s \cdot [(P_s - h \times S_s)^+ - \kappa] ds - \sum_{s \geq 0} e^{-\rho s} K(I_s - I_{s-}) \right]$$

Feedback effect

Before : price density at time $T = 1$ year

After : add 2GW of new peak-load assets

Price densities



Conclusion

- **New spot model : marginal cost + scarcity effect**
- Derivatives pricing : futures, spread options, options on futures, . . .
- Partial hedging using fuel futures and power futures
- Still extensions and improvements
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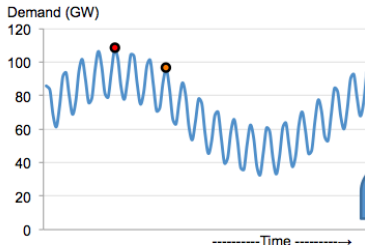
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Questions

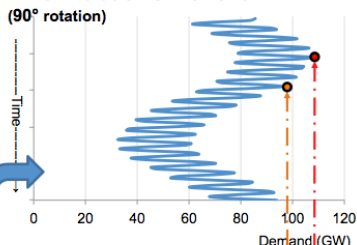


Time evolution of marginal costs

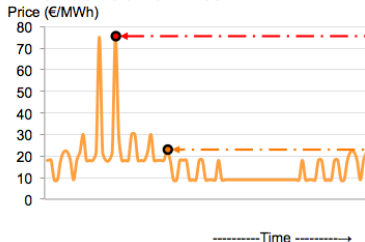
Time Evolution of Demand



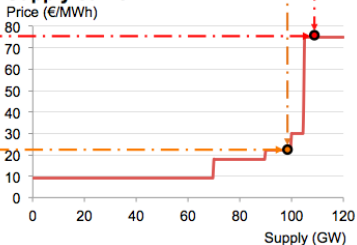
Time Evolution of Demand (90° rotation)



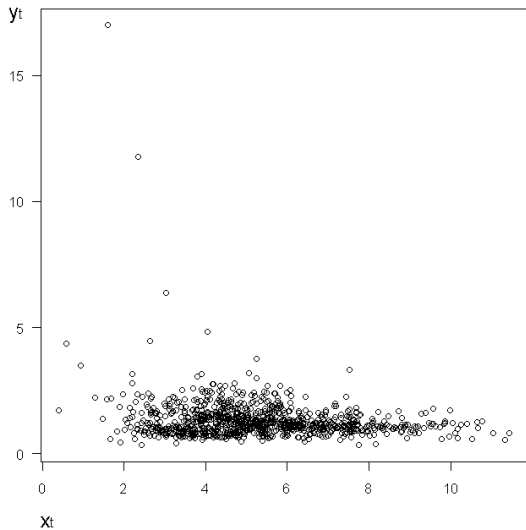
Time Evolution of Price



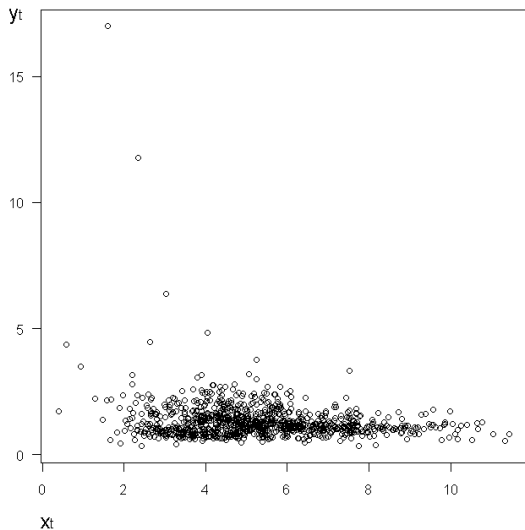
Supply curve



Estimation (1/8)

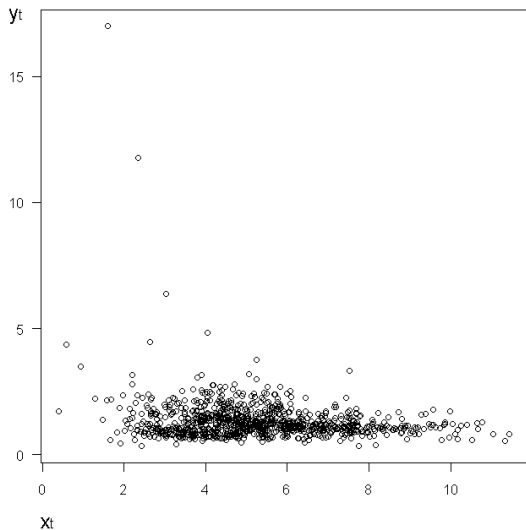


Estimation (1/8)



- Decreasing relation

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- Tricky estimation

Estimation (2/8)

Quantiles

$$\mathbb{P}(X \leq q_X(p)) = p$$

$$\mathbb{P}(Y \leq q_Y(p)) = p$$

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Idea

If $Y = h(X)$, $h \searrow$

Then $q_Y(1-p) = h(q_X(p))$, $0 \leq p \leq 1$

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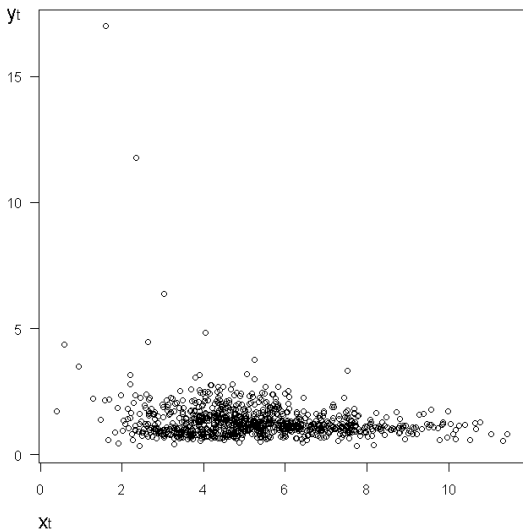
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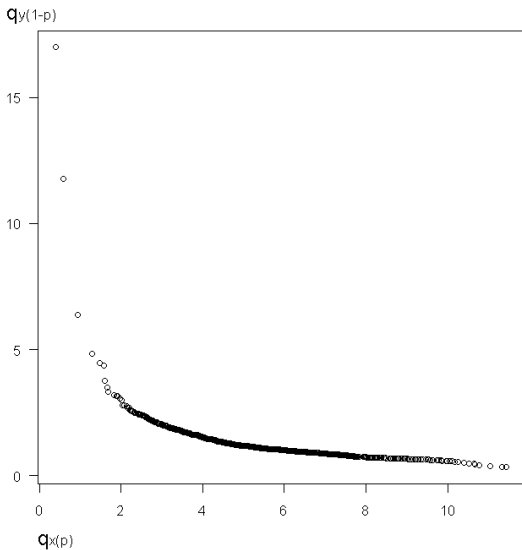
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⇒ Estimation on the quantiles

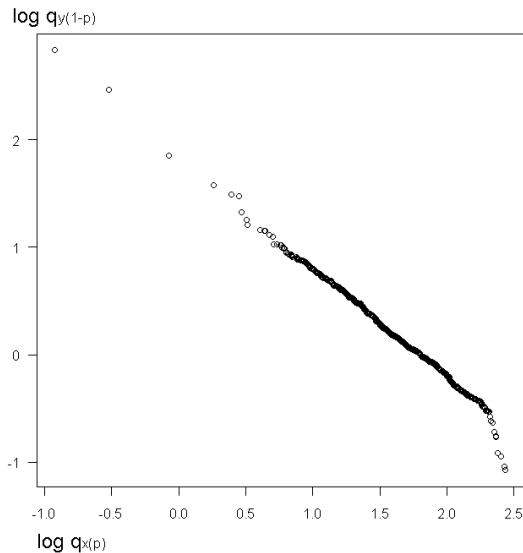
Estimation (3/8)



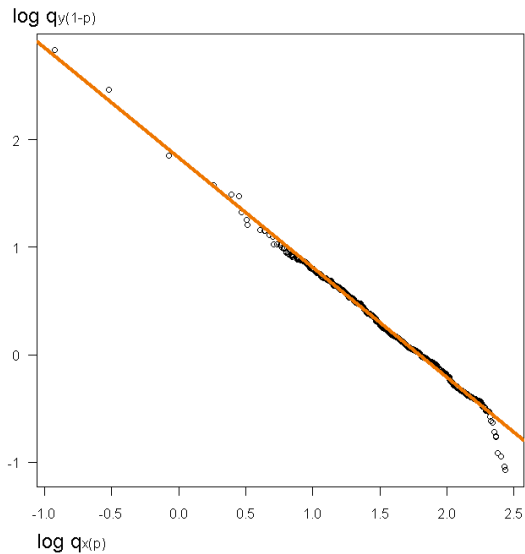
Estimation (4/8)



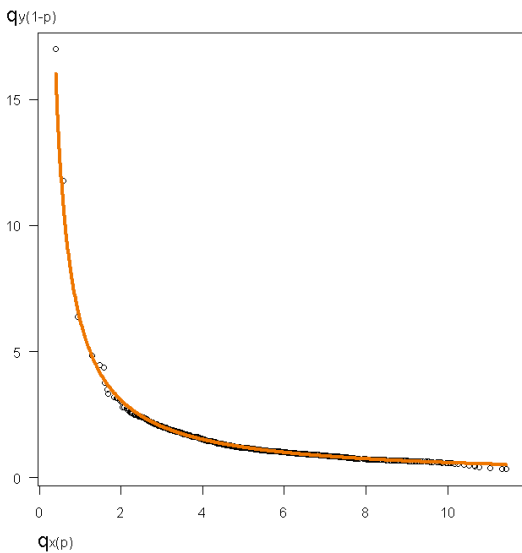
Estimation (5/8)



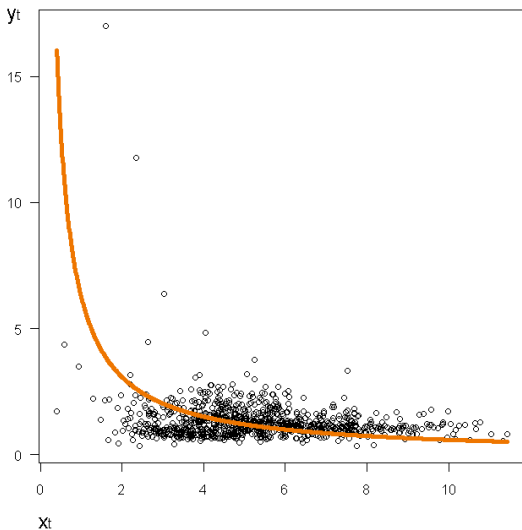
Estimation (6/8)



Estimation (7/8)

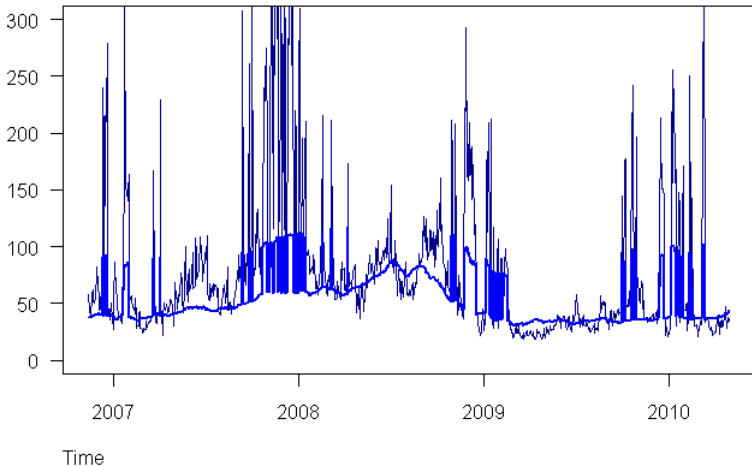


Estimation (8/8)



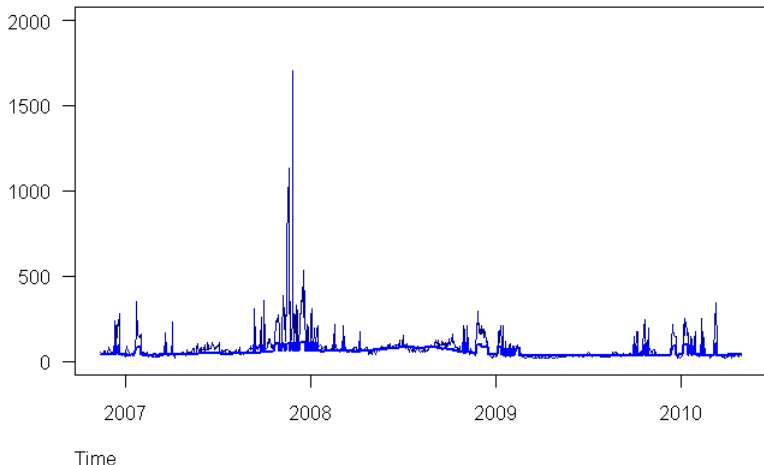
Comparison of the two models (1/2)

Spot price (in €/MWh)



Comparison of the two models (2/2)

Spot price (in €/MWh)



C_t and D_t

Building blocks

$$D_t = f_D(t) + Z_D(t)$$

$$C_t^i = f_i(t) + Z_i(t)$$

Stochastic part

$$dZ_D(t) = -\alpha_D Z_D(t) dt + b dW_t^D$$

$$dZ_i(t) = -\alpha_i Z_i(t) dt + \beta_i dW_t^i$$

Deterministic part

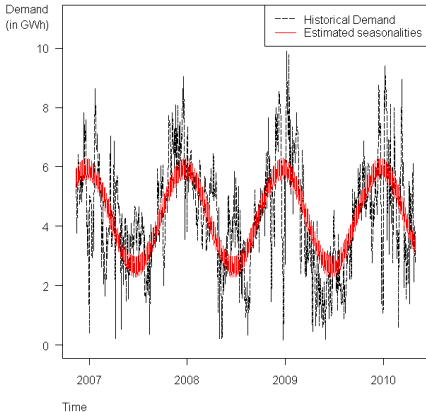
$$f_D(t) = d_1 + d_2 \cos\left(2\pi \frac{t - d_3}{l_1}\right) + d_4 \cos\left(2\pi \frac{t - d_5}{l_2}\right)$$

$$f_i(t) = c_1^i + c_2^i \cos\left(2\pi \frac{t - c_3^i}{l_1}\right) + f_i^{\text{evo}}(t)$$

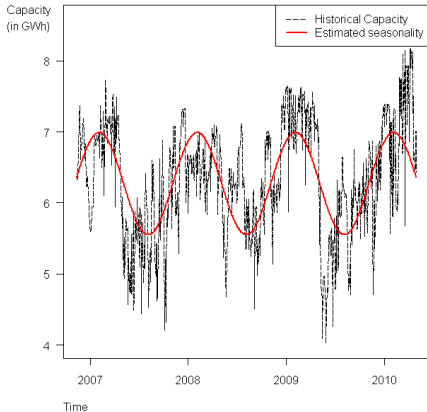
C_t and D_t

Historical data

Demand seasonalities



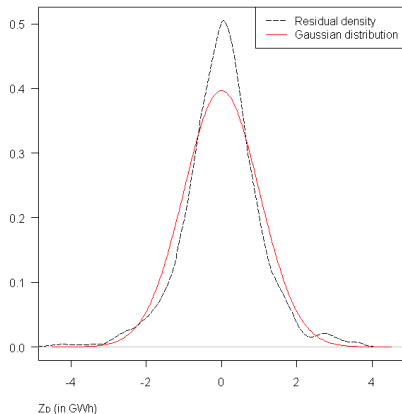
Coal capacity seasonality



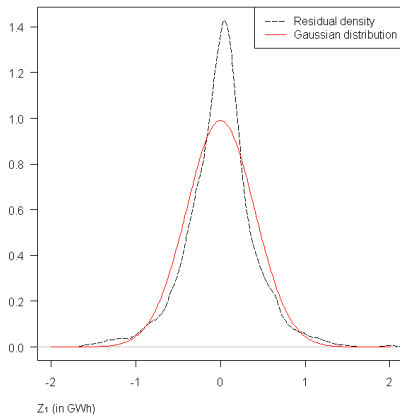
C_t and D_t

Residuals

Demand: Residual density vs. Gaussianity



Coal capacity: Residual density vs. Gaussianity



Hedging strategy

Payoff H , price $V_t^H := \widehat{\mathbb{E}}[H|\mathcal{F}_t] = \phi(t, F_t(T^*), C_t, D_t)$

Hedging strategy

$$\xi_t^e = \frac{1}{\|\theta_t^C, \theta_t^D\|^2} \left\{ \sum_{i=1}^n \theta_t^{C,i} \frac{\partial \phi}{\partial c_i} \beta_i + \theta_t^D \frac{\partial \phi}{\partial z} b \right\}$$

$$\xi_t^i = \frac{\partial \phi}{\partial y_i} + \frac{h_i G_i^{T^*}(t, C_t, D_t)}{\|\theta_t^C, \theta_t^D\|^2} \left\{ \sum_{i=1}^n \theta_t^{C,i} \frac{\partial \phi}{\partial c_i} \beta_i + \theta_t^D \frac{\partial \phi}{\partial z} b \right\}$$

where $dF_t^e := \theta_t^S \cdot d\widehat{W}_t + \theta_t^C \cdot dW_t^C + \theta_t^D dW_t^D$

Dynamics of power futures



$$\begin{aligned}
 dF_t^e &= \theta_t^S \cdot d\widehat{W}_t + \theta_t^C \cdot dW_t^C + \theta_t^D \cdot dW_t^D \\
 &= e^{r(T^*-t)} \sum_{i=1}^n \left(\sum_{k=i}^n G_k^{T^*}(t, C_t, D_t) \right) \sigma_i Y_t^i d\widehat{W}_t^i \\
 &\quad + \sum_{i=1}^n h_i F_t^i(T^*) \frac{\partial G_i^{T^*}}{\partial d}(t, C_t, D_t) b(t, D_t) dW_t^D \\
 &\quad + \sum_{i=1}^n h_i F_t^i(T^*) \sum_{k=1}^n \frac{\partial G_i^{T^*}}{\partial C_k}(t, C_t, D_t) \beta_k(t, C_t^k) dW_t^{C,k}
 \end{aligned}$$

Computing G_i^T

and its partial derivatives

- Explicit, as a function of the *extended incomplete Goodwin-Staton integral* :

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$$\Gamma(\alpha, x) = \text{incomplete Gamma function} = \int_x^\infty t^{\alpha-1} e^{-t} dt$$

Spread option

$$\text{Payoff } H = (P_T - h_1 S_T^1 - K)^+$$

Pricing

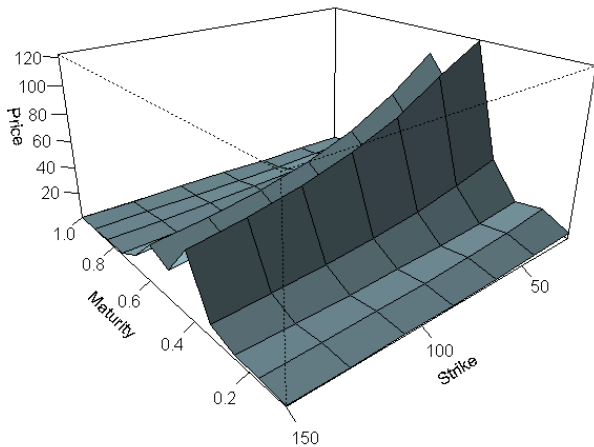
$$\pi_0 = \int_{\mathbb{R}^2} f_{C_T^1 - D_T}(z) f_{C_T^2}(c) \{ \phi_1(c, z) \mathbf{1}_{\{z > 0\}} + \phi_2(c, z) \mathbf{1}_{\{z \leq 0\}} \} dc dz,$$

$$\phi_1 = (g-1) BS_0 \left(\sigma_1, \frac{K}{g-1} \right) \mathbf{1}_{\{g > 1\}} \quad g = g(c+z)$$

$$\begin{aligned} \phi_2 = & g \int_0^\infty \hat{f}_{Y_T^1}(y) BS_0 \left(\sigma_2, \frac{K + (1-g)y}{g} \right) \left(\mathbf{1}_{\{g \leq 1\}} + \mathbf{1}_{\{g > 1\}} \mathbf{1}_{\{y < \frac{K}{g-1}\}} \right) dy \\ & + \left(g Y_0^2 \mathcal{N} \left(\frac{\left(r - \frac{\sigma_1^2}{2} \right) T - \ln \left(\frac{K}{(g-1) Y_0^1} \right)}{\sigma_1 \sqrt{T}} \right) + (g-1) BS_0 \left(\sigma_1, \frac{K}{g-1} \right) \right) \mathbf{1}_{\{g > 1\}} \end{aligned}$$

Spread option

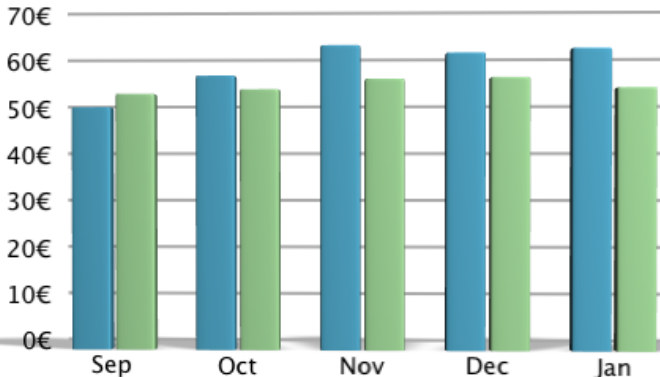
Numerical example



Market vs. model (1/2)

Risk premium vs. model error

Power Futures



Traded

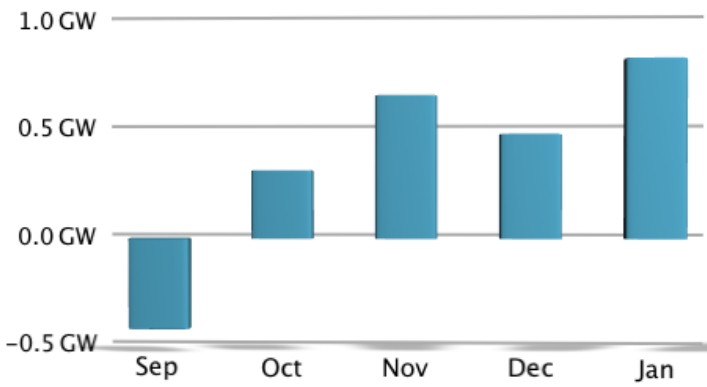
Model

08/28/2010

Market vs. model (2/2)

Risk premium vs. model error

Implied Load Spread



08/28/2010

Extensions

Done

- Extend the numerical application to every single hour of the day (3 marginal fuels)
- Extension to payoffs with delivery periods $[T_1, T_2]$
- Compute the residual market price of risk

To do

- Better models for fuels (spot and futures) : convenience yield, interest rates, cointegration...
- Better volatility models
- Compare different hedging criteria
- Application to investment opportunities