



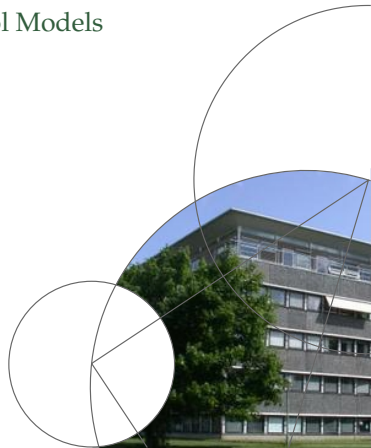
Faculty of Science



Valuation and Operation of Three Types of Power Plants using Continuous Time Stochastic Control Models

Rune Ramsdal Ernsten
Department of Mathematical Sciences

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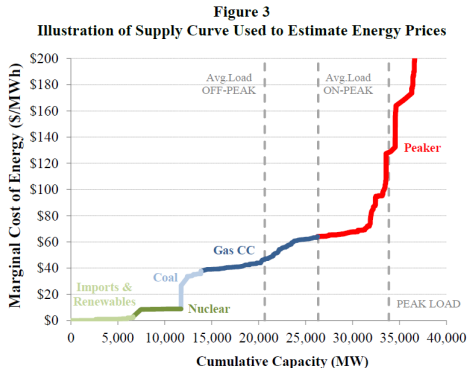
Overview

- ① Motivation
- ② Power plant models
 - Wind power plant
 - Gas power plant
 - Hydro power plant
- ③ Stochastic Optimal Control
 - Heuristics of HJB equation
 - Applying the equation
- ④ Price model
 - Calibration with EM algorithm



Motivation

Energy prices based on supply and demand.



Increasing the amount of renewables will change price dynamics in the energy market.





Motivation

Investigate impact of changes in dynamics on:

- ① Wind power plant
 - Non-controllable
 - Stochastic generation
 - No variable cost

- ② Gas power plant
 - Controllable generation
 - Variable cost

- ③ Hydro plant
 - Controllable output/input levels
 - Stochastic input





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Wind power plant

$$V_{wind}(t, P, W) = E \left[\int_t^T e^{-\rho(s-t)} P_s h(W_s) ds \mid P_t = P, W_t = W \right]$$
$$st. \begin{aligned} dP_t &= \mu_P(P_t, t) dt + \sigma_P(P_t, t) dZ_t^P + \gamma(P_t, t) dJ_t \\ dW_t &= \mu_W(W_t, t) dt + \sigma_W(W_t, t) dZ_t^W \end{aligned}$$

Z_t is a Brownian Motion and J_t is a pure jump process.

V_{wind} : Value of wind power plant given wind and price.

P_t : Spot price

W_t : Wind speed

ρ : Discount factor

h : Power curve



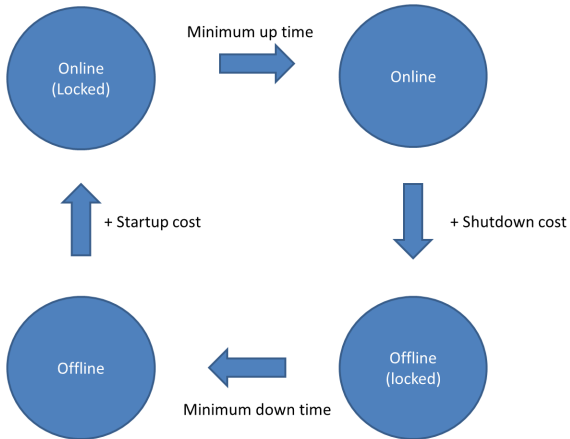


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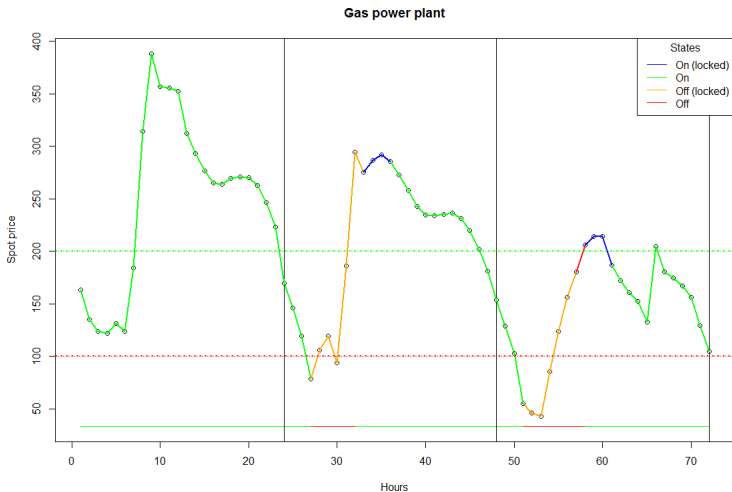
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Gas power plant



Gas power plant



Gas power plant

$$V_{gas}(t, P, t_0, 1) = \max_{\tau, Q_s} E \left[\int_t^{\tau} e^{-\rho(s-t)} \Pi(Q_s, P_s, 1) ds \quad \text{[On locked]} \right. \\ \left. + e^{-\rho(\tau-t)} V_{gas}(t + \tau, P_{t+\tau}, 0, 2) \middle| P_t = P \right]$$

$$V_{gas}(t, P, t_0, 2) = \max_{\tau, Q_s} E \left[\int_t^{\tau} e^{-\rho(s-t)} \Pi(Q_s, P_s, 2) ds \quad \text{[On]} \right. \\ \left. + e^{-\rho(\tau-t)} (C_{off} + V_{gas}(t + \tau, P_{t+\tau}, 0, 3)) \middle| P_t = P \right]$$

$$V_{gas}(t, P, t_0, 3) = \max_{\tau, Q_s} E \left[\int_t^{\tau} e^{-\rho(s-t)} \Pi(Q_s, P_s, 3) ds \quad \text{[Off locked]} \right. \\ \left. + e^{-\rho(\tau-t)} V_{gas}(t + \tau, P_{t+\tau}, 0, 4) \middle| P_t = P \right]$$

$$V_{gas}(t, P, t_0, 4) = \max_{\tau, Q_s} E \left[\int_t^{\tau} e^{-\rho(s-t)} \Pi(Q_s, P_s, 4) ds \quad \text{[Off]} \right. \\ \left. + e^{-\rho(\tau-t)} (C_{on} + V_{gas}(t + \tau, P_{t+\tau}, 0, 1)) \middle| P_t = P \right]$$





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Hydro Power Plant

$$V_{hydro}(t, P, I, L) = \max_{(v_s)_{s \in [t, T]}} E \left[\int_t^T e^{-\rho(s-t)} P_s H(v_s, L_s) ds \mid P_t = P, I_t = I \right]$$

$$st. dP_t = \mu_P(P_t, t) dt + \sigma_P(P_t, t) dZ_t^P + \gamma(P_t) dJ_t$$

$$dI_t = \mu_I(I_t, t) dt + \sigma_I(I_t, t) dZ_t^I$$

$$dL_t = (I_t - v_t) dt$$

$$(I_t - v_t) 1_{(L_t > L_{min})} \leq I_t - v_t \text{ Positive inflow at lower bound}$$

$$(I_t - v_t) 1_{(L_t < L_{max})} \geq I_t - v_t \text{ Negative inflow at upper bound}$$

$$v_{min} \leq v_t \leq v_{max}$$

P_t : Spot price

I_t : Inflow rate to storage facility (melt water and rainfall etc.)

L_t : Water level in hydro power plant

v_t : outflow rate

H : Power produced

$-v_{min}$: Max inflow rate

v_{max} : Max outflow rate





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Stochastic Optimal Control

Let $u(t)$ be the control at time t .

Minimize

$$J_t = E_{tx} \left[\int_t^T L(s, X_s, u(s)) ds \right]$$

Define value function

$$V(t, x) = \inf_{u \in C} J_t$$

C is set of admissible controls.

Bellman's principle of dynamic programming:

$$V(t, x) = \inf_{u \in C} E_{tx} \left[\int_t^{t+h} L(s, X_s, u(s)) ds + V(t+h, X_{t+h}) \right]$$



Stochastic Optimal Control

Minimizing

$$J_t = E_{tx} \left\{ \int_t^T L(s, X_s, u(s)) ds \right\}$$

with

$$V(t, x) = \inf_{u \in C} J_t$$

becomes equivalent to, under some regularity conditions, to determining $V(t, x)$ through the HJB equation

$$\inf_{u \in C} \{A^u V(t, x) + L(t, x, v)\} = 0$$

with $V(T, x) = 0$.

A^u is the "derivative of the expected value" defined as

$$A^u V(t, x) = \lim_{h \rightarrow 0^+} \frac{1}{h} [E_{tx} V(t+h, X_{t+h}) - V(t, x)]$$





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$$V(t, P, W) = E \left[\int_t^T e^{-\rho(s-t)} P_s h(W_s) ds \mid P_t = P, W_t = W \right]$$

$$st. \begin{aligned} dP_t &= \mu_P(P_t, t) dt + \sigma_P(P_t, t) dZ_t^P + \gamma(P_t) dJ_t \\ dW_t &= \mu_W(W_t, t) dt + \sigma_W(W_t, t) dZ_t^W \end{aligned}$$

V solves the Hamilton-Jacobi-Bellman equation

$$\begin{aligned} V_t + \mu_P V_P + \mu_W V_W + \frac{1}{2} \sigma_P^2 V_{PP} + \frac{1}{2} \sigma_W^2 V_{WW} + \rho_{PW} \sigma_P \sigma_W V_{PW} \\ + \lambda E \left[V^+ - V \mid P_t = P, W_t = W \right] + Ph(W) - \rho V = 0 \\ V(T, P, W) = 0 \end{aligned}$$



Gas power plant

The Hamilton-Jacobi-Bellman PIDEs are

$$A(V_{gas}(t, P, t_0, 1)) + \max_Q \{\Pi(P, Q, 1)\} - \rho V_{gas}(t, P, t_0, 1) = 0,$$

$$A(V_{gas}(t, P, t_0, 2)) + \max_Q \{\Pi(P, Q, 2)\} - \rho V_{gas}(t, P, t_0, 2) = 0,$$

$$A(V_{gas}(t, P, t_0, 3)) + \max_Q \{\Pi(P, Q, 3)\} - \rho V_{gas}(t, P, t_0, 3) = 0,$$

$$A(V_{gas}(t, P, t_0, 4)) + \max_Q \{\Pi(P, Q, 4)\} - \rho V_{gas}(t, P, t_0, 4) = 0,$$

with boundary conditions $V_{gas}(T, P, t_0, i) = 0$, $i = 1, 2, 3, 4$, and

$$V_{gas}(t, P, T_{up}, 1) = V_{gas}(t, P, 0, 2),$$

$$V_{gas}(t, \mathcal{P}_{off}^*(t), t_0, 2) = V_{gas}(t, \mathcal{P}_{off}^*(t), 0, 3) - C_{shutdown},$$

$$V_{gas}(t, P, T_{down}, 3) = V_{gas}(t, P, 0, 4),$$

$$V_{gas}(t, \mathcal{P}_{on}^*(t), t_0, 4) = V_{gas}(t, \mathcal{P}_{on}^*(t), 0, 1) - C_{startup}.$$



Storage facility (Hydro Power Plant)

$$V(t, P, I, L) = \max_{(v_s)_{s \in [t, T]}} E \left[\int_t^T e^{-\rho(s-t)} P_s H(v_s, L_s) ds \mid P_t = P, I_t = I \right]$$

$$st. dP_t = \mu_P(P_t, t) dt + \sigma_P(P_t, t) dZ_t^P + \gamma(P_t) dJ_t$$

$$dI_t = \mu_I(I_t, t) dt + \sigma_I(I_t, t) dZ_t^I$$

$$dL_t = (I_t - v_t) dt$$

$$(I_t - v_t) 1_{(L_t > L_{min})} \leq I_t - v_t$$

$$(I_t - v_t) 1_{(L_t < L_{max})} \geq I_t - v_t$$

$$v_{min} \leq v_t \leq v_{max}$$

HJB equation:

$$\begin{aligned} V_t + \mu_P V_P + \frac{1}{2} \sigma_P^2 V_{PP} + \mu_I V_I + \frac{1}{2} \sigma_I^2 V_{II} + \rho_{PI} \sigma_I \sigma_P V_{PI} \\ + \max_v \{ PH(v, L) + (I - v) V_L \} \\ + \lambda E \left[V^+ - V \mid P_t = P, I_t = I, L_t = L \right] - \rho V = 0 \end{aligned}$$

$$V(T, P, W) = 0$$





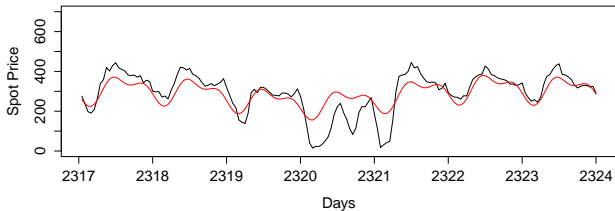
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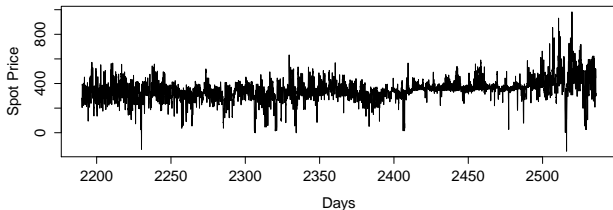
Price model

Spot prices (Week)



- Seasonality
- Mean reversion

Spot prices (Year)



- Jumps
- Negative prices



Price model

Spot priced modelled as

$$P_t = e^{\theta_t} e^{X_t} - M$$

where

$$dX_t = \kappa(\alpha - X_t) dt + \sigma dZ_t^P + dJ_t$$

θ_t : Seasonal component

M : Constant to allow for negative prices

Explicit solution

$$P_t = e^{\theta_t} \exp \left\{ X_s e^{-\kappa(t-s)} + \alpha(1 - e^{-\kappa(t-s)}) + \sigma \int_s^t e^{-\kappa(t-u)} dZ_u^P \right. \\ \left. + \sum_{n=N_s+1}^{N_t} e^{-\kappa(t-T_n)} Y_n \right\} - M$$





Price model

Calibration gives

- Expected price 297.10
- Hourly volatility ~ 17 (95% in $(-33.3, 33.3)$)
- Mean reversion level 279.57
- Mean reversion (~ 3.5 hours to get halfway to mean)

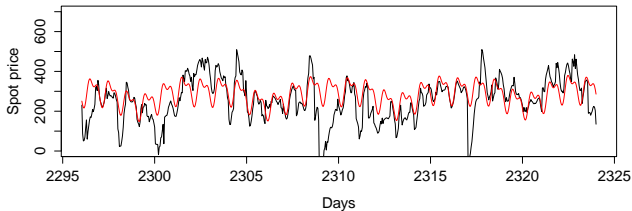
Jumps modelled as medium and small jumps

- Small jumps (~ 3.5 hours between jumps, 95% in $(-100, 100)$)
- Medium jumps (~ 38 hours between jumps, 95% in $(-360, 360)$)

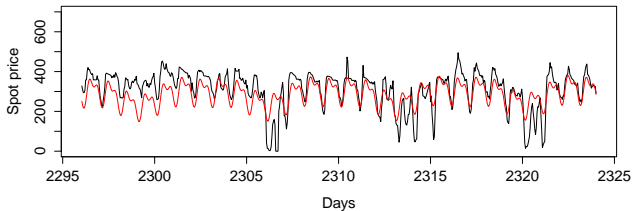


Price model

Simulated price (Month)



Actual price (Month)





Conclusion

- Possible to determine value and control of power plants over long time horizons
- Problem is solved locally
- Running time increases linearly with T
- Incorporates non-linear payoff functions
- Capture specific marginal distributions
- Approximate discrete problem with continuous time problem





References

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Questions?

