Risk-neutral valuation of Swing options in presence of jumps

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Goal of this presentation

 The main goal of this presentation is to show the basic framework for the risk-neutral pricing of swing options valuation where the logarithm of the spot price is the sum of a deterministic seasonal trend and the Ornstein-Uhlenbeck process driven by a Lévy process (jump-diffusion).

 This presentation has been designed for quants analyst, and the content of this presentation can hurt your feelings if you do not used to work with stochastic differential equations, PDE-PIDE and some measure theory.

Questions addressed in this presentation:

- Which is the stochastic differential equation that better fits the "most correct" dynamics of the commodities involved in a swing option?
- Can we build a risk-neutral model that allow us to model the forward prices?
- Could we build an analytical pricing framework that allows us to estimate not only the prices but also the Greeks for the swing option hedging?
- How can we obtain the most robust parameters that we have to introduce in the model such that guarantee the most stable possible pricing and hedging?

1. Introduction

Introduction

Definition

A **Swing option** is a financial contract with the following payoff characteristics:

- **1** Maturity contract: runs over [0, T]
- **2 Strike**: fixed price *K* Eur/MWh
- **Swing action times**: finite set of dates $\{T_n\}_{n=1}^N$ with $0 \le T_1 < T_2 < ... < T_N < T$
- **Swing action**: At each swing action date T_n the holder decides on the **amount of energy** B_n^d MWh to be bought at fixed price K Eur/MWh over each of the D periods $\left(T_n^d, T_n^{d+1}\right]$, $1 \le d \le D$.
- **Total and partial boundaries**: assume that $B_n^d \in \mathcal{O} \subseteq [0, \infty)$ where \mathcal{O} is either a closed interval $\mathcal{O} = [\underline{\mathsf{B}}, \overline{B}]$ or a discrete set. Additionally, the holder must buy at least $\underline{\mathsf{M}}$ MWh, and the most $\overline{\mathsf{M}}$ MWh in total
- **Settlement**: All swing options are financially settled.

Introduction

The **outline** of this presentation (excluding this introduction) is the following:

2. Section 1: Deng (2000) spot price model

 This section introduce the model developed by Deng (2000) where the logarithm of the spot price is the sum of a seasonality term and a Ornstein-Uhlenbeck process driven by a jump diffusion.

3 Section 2: Model calibration

 Section 2 is devoted to the calibration or estimation of the parameters of the model, using a combination between a least-squares method (Lucia and Schwartz (2002)) and the Fourier transform based in the maximum likelihood approach by Singleton (2001).

4. Section 3: Swing option price estimation and numerical algorithm

- In the last section, we proceed to develop the swing option pricing algorithm based in the discretization of a PIDE using a finite difference method, similar as **Cont and Tankov (2003)**.

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2. Section 1: Deng (2000) spot price model

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- 2. Section 1: Deng (2000) spot price model

Section 1: Deng (2000) spot price model

• We consider that power spot prices $S = \{S_t : 0 \le t \le T^*\}$ lives in a **continuous-time trading economy** in $[0, T^*]$ driven by a compensated **Lévy** process L on a 'sufficiently rich' stochastic basis $(\Omega, \mathcal{L}, \mathbb{L}, \mathbb{P})$.

Definition

Assume that the **power-spot prices follows the** \mathbb{P} -dynamics:

$$\begin{cases}
S_t = \exp(f(t) + X_t) \\
dX_t = -\alpha X_{t-} dt + dL_t
\end{cases}$$
(1)

where $\alpha>0$ is fixed, $f\left(t\right)$ is a **deterministic seasonal trend** and L_{t} is a **compensated Lévy process** that accept the canonical **Lévy-Ito decomposition** with triplet $(\sigma,v,0)$ such that

$$L(t) = \int_0^t \sigma W(ds) + \int_0^t \int_{|x| > \varepsilon} x \cdot (J_X - v) (ds, dx)$$

= $\sigma W_t + U_t^x - v \mathbb{E}_P[x] t$ (2)

Section 1: Deng (2000) spot price model

• The solution S_t to the SDE (1) started in t_0 is given by

$$S_{t} = S_{t_{0}}e^{-\alpha(t-t_{0})} - \frac{v\mathbb{E}_{P}[x]}{\alpha} \left(1 - e^{-\alpha(t-t_{0})}\right) + \sigma \int_{t_{0}}^{t} e^{-\alpha(t-t_{0})}dW_{s} + \sum_{i=N_{t_{0}}}^{N_{t}} e^{-\alpha(t-t_{i})}X_{i}$$
(3)

ullet The density of S_t is generally not known explicitly, but using the FT for any $t \in \mathbb{R}^+$

$$\hat{\mu}_t(z) = e^{(t-t_0)\psi(z)} \text{ with } z \in \mathbb{R}^d$$
(4)

where $\psi_{j}\left(z\right)$ is the **Lévy-Khitchine exponent** with the following representation

$$\psi(z) = izxe^{-\alpha(t-t_0)} - \frac{\sigma^2 z^2}{4\alpha} \left(1 - e^{-2\alpha(t-t_0)} \right)$$

$$+ \int_{\mathbb{R}^d} \left(e^{izx} - 1 - izxe^{-\alpha(t-t_0)} \right) v(dx)$$
(5)

Section 1: Deng (2000) spot price model

Fact

Forward prices, under a 'diffusive' risk-neutral measure, has the following expression:

$$F(t,T) = \exp\left(f(T) + (\log S_t - f(t)) e^{-\alpha(T-t)}\right)$$

$$\times \exp\left(\frac{\sigma\lambda}{\alpha} \left(1 - e^{-\alpha(T-t)}\right) + \frac{\sigma^2}{4\alpha} \left(1 - e^{-2\alpha(T-t)}\right)\right)$$

$$\times \exp\left(\int_{\mathbb{R}^d} \left(e^x - 1 - xe^{-\alpha(T-t)}\right) v(dx)\right)$$

$$\equiv F_{season} \times F_{diffusion} \times F_{jump}$$
(6)

Proof.

That can be obtained following **Lucia and Schwartz**(2002)



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Section 2: Model Calibration

Q Let us define a **periodic seasonal trend** $f(t|\Theta)$ with $\Theta = (A_0, A_n, B_n)_{n=1}^N$ such that

$$f(t|\Theta) = A_0 + \sum_{n=1}^{N} A_n \cos(2\pi f_n t + B_n)$$
 (7)

where the parameter vector Θ is unknown and is to be estimated from data.

- ② Notice that the **stationary covariance function** of the $\ln S_t$ is given by $Cov\left(\ln S_{t+\Delta}, \ln S_t\right) = e^{-\alpha \Delta} Var\left(\ln S_t\right)$ so α is estimated from sample covariances and variances.
- Next, we subtract the seasonal component $f(t|\Theta)$ from the $\ln S_t$, and we estimate the remaining parameters with the **maximum likelihood method**. Notice that if we unknown the density function we may form the \log **likelihood function** using the inverse **Fourier transformation** of the characteristic function, following **Deng**(2000) or **Singleton**(2001).
- Finally, to estimate the **market price of risk** λ , we minimized the distance between market and model prices, such that

$$\hat{\lambda} = \arg\min \sum_{k=1}^{K} \left| \hat{F}(t, T) - F(t, T) \right|^{2} \tag{8}$$

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Section 3: Swing option price estimation

- Let us remind that at time T_n , the swing option holder decide to buy B_n^d MWh at K EUR/MWh.
- A swing action is described as a D-vector $\left\{B_n^1...B_n^D\right\}$, and defined as $\Delta_n = \sum_{d=1}^D B_n^d$ such that $\Delta_n \in S = [0, D\bar{B}]$. A swing action means choosing Δ_n without violating the contract constraints, and receiving the amount $g\left(T_n, s, \Delta_n\right)$.
- ullet Finally, let us define $Z_t = \sum_{n=1}^j \Delta_n$ for $t \in \left(T_j, T_{j+1}\right]$.

Definition

We can define the **value of a swing option** at moment t, when $s=S_t$ and $z=Z_t$, as $V\left(t,s,z\right)$ given by

$$V(t,s,z) = \begin{cases} \sup \sum_{n=j}^{N} e^{-r(T_n - T_j)} \mathbb{E}\left[g\left(T_n, S_{T_n}, \Delta_n\right) \middle| \mathcal{F}_{T_j}\right] & t = T_j \\ e^{-r(T_n - T_j)} \mathbb{E}\left[V\left(T_j, S_{T_j}, z\right) \middle| \mathcal{F}_{T_j}\right] & T_j < t < T_j, \ j > 1 \\ t < T_1 & (9) \end{cases}$$

Section 3: Swing option price estimation

Theorem

There exists at least one optimal swing action strategy $\{\Delta_n^*\}_{n=j}^N$ such that the supremum is attainable.

According with the **Feynman-Kac theorem**, the swing option price $V\left(t,s,z\right)$ is the unique solution to the following parabolic partial integro-differential equation (PIDE)

$$\frac{\partial V}{\partial t} + \mathcal{L}_x V - rV = 0 \tag{10}$$

where the operator \mathcal{L}_x ca be splitted into two parts (integral and differential) such that $\mathcal{L}_x = \mathcal{D}_x + v\mathcal{I}_x$ with

$$\mathcal{D}_{x}V\left(t,x,z\right) = \frac{\sigma^{2}}{2}\frac{\partial^{2}V}{\partial x^{2}} - \left(\sigma\lambda + \alpha x\right)\frac{\partial V}{\partial x}$$

$$\mathcal{I}_{x}V\left(t,x,z\right) = \int_{\mathbb{R}}\left[V\left(t,x_{l}+y,z\right) - V\left(t,x_{l},z\right) - y\frac{\partial V}{\partial x}\right]f(y)dy$$
(11)

that can be solved numerically splitting the diffusion or continuous part, from the integral or discontinuous part.

Section 3: Swing option price estimation

• The space derivatives are discretized using finite differences:

$$\left(\frac{\partial V}{\partial x}\right)_{n} \approx \begin{cases}
\frac{V_{n+1} - V_{n}}{\Delta x} & \text{if } \sigma\lambda + \alpha x < 0 \\
\frac{V_{n} - V_{n-1}}{\Delta x} & \text{if } \sigma\lambda + \alpha x \ge 0
\end{cases}$$

$$\left(\frac{\partial^{2} V}{\partial x^{2}}\right)_{n} \approx \frac{V_{n+1} - 2V_{n} - V_{n-1}}{(\Delta x)^{2}}$$

• In order to approximate the **integral terms** one can use the **trapezoidal quadrature rule** with the same grid resolution Δx . More specifically, if we define:

$$V_n = \int_{\Delta x \left(n - \frac{1}{2}\right)}^{\Delta x \left(n + \frac{1}{2}\right)} f(y) dy$$

then $\mathcal{I}_{x}V\left(t,x,z\right)$ may approximated as:

$$\mathcal{I}_{x}V\left(t_{k+1},x_{l},z\right) \approx \lambda \sum_{n=-N}^{N} \left[V_{k+1}^{l+n} - V_{k+1}^{l} - \frac{n}{2}\left(V_{k+1}^{l+1} - V_{k+1}^{l-1}\right)\right] \nu_{n}$$

Answers:

- 1. We have introduced a **stochastic process** L_t , that has "some good" properties for the commodities price modelling.
- We have built the risk-neutral dynamics for forward prices where we have included seasonality, mean-reversion, Brownian motion and jumps.
- 3. We have proposed a calibration methodology, that include the market price of risk for the diffusion part.
- 4. We have defined a pricing framework for swing option based in a PIDE. The solution of the PIDE can be found with a numerical pricing algorithm based in finite differences (using a combination of explicit-implicit schemes).

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