

# Connecting Discrete and Continuous Lookback and Hindsight Options Under Exponential Lévy Model

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# 1 Introduction

## 2 Continuous vs Discrete Supremum

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Let  $X$  be a real Lévy process

## Lévy-Khinchin representation

$X$  is a Lévy process with generating triplet  $(\gamma, \sigma, \nu)$  :

$$\mathbb{E}e^{iuX_s} = e^{s\varphi(u)}$$
$$\varphi(u) = i\gamma u - \frac{\sigma^2 u^2}{2} + \int_{\mathbb{R}} (e^{iux} - 1 - iux\mathbb{1}_{|x|\leq 1})\nu(dx)$$

Where  $\gamma \in \mathbb{R}$ ,  $\sigma \geq 0$  and  $\nu$  is Radon measure on  $\mathbb{R}^*$  satisfying :

$$\int_{\mathbb{R}} 1 \wedge x^2 \nu(dx) < \infty$$

## Finite activity Lévy process

It means :  $\nu(\mathbb{R}) < \infty$ . Hence

$$X_s = \gamma_0 s + \sigma B_s + \sum_{i=1}^{N_s} Y_i$$

where  $B$  is a standard Brownian motion,  $N$  is a Poisson process with parameter  $\lambda = \nu(\mathbb{R})$ ,  $(Y_i)_{i \geq 1}$  are i.i.d. r.v. with a common law  $\frac{\nu(dx)}{\nu(\mathbb{R})}$ , and

$$\gamma_0 = \gamma - \int_{|x| \leq 1} x \nu(dx)$$

## Infinite activity Lévy process

It means  $\nu(\mathbb{R}) = +\infty$ .

## Finite variation Lévy process

It means  $\sigma = 0$  and

$$\int_{|x| \leq 1} |x| \nu(dx) < \infty$$

We define

$$\begin{aligned}M_T &= \sup_{0 \leq s \leq T} X_s, & M_T^n &= \max_{0 \leq k \leq n} X_{\frac{kT}{n}} \\m_T &= \inf_{0 \leq s \leq T} X_s, & m_T^n &= \min_{0 \leq k \leq n} X_{\frac{kT}{n}}\end{aligned}$$

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Our goal is to study this expression :

$$M_T - M_T^n$$

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## Referenes

- 1 Asmussen, S., Glynn, P. et Pitman, J. (1995).
- 2 Broadie, M., Glasserman, P. et Kou, S. G.(1999).

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# Spitzer's Identity and Applications

## Proposition

Let  $X$  a real integrable Lévy process, then

$$\lim_{n \rightarrow +\infty} \sum_{k=1}^n \frac{\mathbb{E}X_{k\frac{T}{m}}^+}{k} = \int_0^T \frac{\mathbb{E}X_s^+}{s} ds$$

$$\mathbb{E}M_T^n = \sum_{k=1}^n \frac{\mathbb{E}X_{k\frac{T}{n}}^+}{k}$$

$$\mathbb{E}M_T = \int_0^T \frac{\mathbb{E}X_s^+}{s} ds$$

# Spitzer's Identity and Applications

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$$\mathbb{E}M_T^n = \sum_{k=1}^n \frac{\mathbb{E}X_{k\frac{T}{n}}^+}{k}$$

$$\mathbb{E}M_T = \int_0^T \frac{\mathbb{E}X_s^+}{s} ds$$

Hence we have

$$\mathbb{E}(M_T - M_T^n) = \int_0^T \frac{\mathbb{E}X_s^+}{s} ds - \sum_{k=1}^n \frac{\mathbb{E}X_{k\frac{T}{n}}^+}{k}$$

## Theorem

Let  $X$  be a real integrable Lévy process with generating triplet  $(\gamma, \sigma, \nu)$  satisfying  $\nu(\mathbb{R}) < \infty$ ,  $T > 0$ , and  $n \in \mathbb{N}$ .

\*If  $\sigma > 0$

$$\mathbb{E}(M_T - M_T^n) = -\frac{\sigma\sqrt{T}\zeta\left(\frac{1}{2}\right)}{\sqrt{2\pi n}} + \frac{\theta_+}{2n} + O\left(\frac{1}{n^{\frac{3}{2}}}\right)$$

Where  $\zeta$  is the Riemann's zeta function and

$$\begin{aligned}\theta_+ &= \frac{\gamma_0 t}{2} + \lambda T \mathbb{E} Y_1^+ - \sigma\sqrt{T} \mathbb{E} \phi\left(\frac{\gamma_0}{\sigma}\sqrt{T} + \frac{\sum_{i=1}^{N_T} Y_i}{\sigma\sqrt{T}}\right) \\ &\quad - \mathbb{E}\left(\gamma_0 T + \sum_{i=1}^{N_T} Y_i\right) \Phi\left(\frac{\gamma_0}{\sigma}\sqrt{T} + \frac{\sum_{i=1}^{N_T} Y_i}{\sigma\sqrt{T}}\right)\end{aligned}$$

## Theorem (Next)

\*If  $\sigma = 0$

$$\mathbb{E}(M_T - M_T^n) = \frac{1}{2n} (\gamma_0^+ T + \lambda T \mathbb{E}Y_1^+ - \mathbb{E}X_T^+) + o\left(\frac{1}{n}\right)$$

\*If  $\sigma = 0$  and  $Y_1$  has a density function  $f_{Y_1} \in C(\mathbb{R})$

$$\mathbb{E}(M_T - M_T^n) = \frac{1}{2n} (\gamma_0^+ T + \lambda T \mathbb{E}Y_1^+ - \mathbb{E}X_T^+) + O\left(\frac{1}{n^2}\right)$$

\*If  $\sigma = 0$  and  $\gamma_0 = 0$

$$\mathbb{E}(M_T - M_T^n) = \frac{1}{2n} (\lambda T \mathbb{E}Y_1^+ - \mathbb{E}X_T^+) + O\left(\frac{1}{n^2}\right)$$

## Theorem

Let  $X$  be a real Lévy process integrable with generating triplet  $(\gamma, 0, \nu)$ . Assume that  $\int_{|x| \leq 1} |x| |\log(|x|)| \nu(dx) < \infty$ , then

$$\mathbb{E}(M_T - M_T^n) = \frac{1}{2n} \left( \left( \gamma_0^+ + \int_{\mathbb{R}} x^+ \nu(dx) \right) T - \mathbb{E}X_T^+ \right) + o\left(\frac{1}{n}\right)$$

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Define  $\delta = \frac{T}{n}$ , we have

$$\mathbb{E}(M_T - M_T^n) = \int_0^\delta \left( \frac{\mathbb{E}X_s^+}{s} - \frac{\mathbb{E}X_\delta^+}{\delta} \right) ds + \sum_{k=2}^n \int_{(k-1)\delta}^{k\delta} \left( \frac{\mathbb{E}X_s^+}{s} - \frac{\mathbb{E}X_{k\delta}^+}{k\delta} \right) ds$$

## Theorem

Let  $X$  be a real Lévy process integrable with generating triplet  $(\gamma, \sigma, \nu)$ .

- ① If  $\sigma > 0$

$$\mathbb{E}(M_T - M_T^n) = O\left(\frac{1}{\sqrt{n}}\right)$$

- ② If  $\sigma = 0$

$$\mathbb{E}(M_T - M_T^n) = o\left(\frac{1}{\sqrt{n}}\right)$$

- ③ If  $\sigma = 0$  and  $\int_{|x| \leq 1} |x| \nu(dx) < \infty$

$$\mathbb{E}(M_T - M_T^n) = O\left(\frac{\log(n)}{n}\right)$$

# Compound Poisson Process Generalization

Let  $X$  a real finite activity Lévy process with generating triplet  $(\gamma, 0, \nu)$ .

$$X_s = \gamma_0 s + \sum_{i=1}^{N_s} Y_i$$

## Theorem

Let  $f \in C_b^1$ , then

$$\mathbb{E}(f(M_T) - f(M_T^n)) = \frac{T}{2n} \left( |\gamma_0| \mathbb{E}f'(M_T)Z + C \right) + o\left(\frac{1}{n}\right)$$

Where the r.v.  $Z$  and the constant  $C$  can be derived explicitly.

# Asmussen et al. Theorem's Extension

Let  $X$  be a Lévy process with generating triplet  $(\gamma, \sigma, \nu)$  satisfying  $\sigma > 0$ , and we define

$$W = \min_{n \in \mathbb{Z}} R(U + n)$$

where  $U$  is uniform on  $(0, 1)$  and  $R$  is a two-sided version three-dimensional Bessel process.

## Theorem (Asmussen, Glynn, Pitman (1995))

If  $\nu = 0$ , the sequence  $(\sqrt{n}(M_T - M_T^n), X)$  converges in distribution to  $(\sigma\sqrt{T}W, X)$ , where  $W$  and  $X$  are independent.

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## Theorem

If  $X$  is a finite activity Lévy process, the sequence  $(\sqrt{n}(M_T - M_T^n), M_T)$  converges in distribution to  $(\sigma\sqrt{T}W, M_T)$ , where  $W$  and  $M_T$  are independent.

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# The Exponential Lévy Model

## The dynamic

The behaviour of the underlying asset is modeled as :

$$S_T = S_0 e^{X_T}$$

Where  $X$  is a Lévy process. And we assume that

$$\left( e^{-(r-\delta)\tau} S_T \right)_{\tau \in [0, T]}$$

is a martingale. With

- $r$  : the continuously compounded interest rate.
- $\delta$  : the continuously compounded dividend rate.

# Lookback Option

Let  $\Delta = \frac{T}{n}$ .

## Continuous prices

The continuous prices at date  $\tau \in [0, T]$  :

$$V(S_-) = e^{-r(T-\tau)} \mathbb{E} \left( S_0 e^{X_T} - \min(S_-, S_0 e^{m\tau}) \right), \text{ call}$$

$$V(S_+) = e^{-r(T-\tau)} \mathbb{E} \left( \max(S_+, S_0 e^{M_T}) - S_0 e^{X_T} \right), \text{ put}$$

## Discrete prices

The discrete prices at date  $k\Delta$  with  $k \in \{0, \dots, n\}$  :

$$V^n(S_-) = e^{-r\Delta(n-k)} \mathbb{E} \left( S_0 e^{X_{n\Delta}} - \min(S_-, S_0 e^{m_T^n}) \right), \text{ call}$$

$$V^n(S_+) = e^{-r\Delta(n-k)} \mathbb{E} \left( \max(S_+, S_0 e^{M_T^n}) - S_0 e^{X_{n\Delta}} \right), \text{ put}$$

# Hindsight Option

## Continuous prices

The continuous prices at date  $\tau \in [0, T]$  :

$$V(S_+, K) = e^{-r(T-\tau)} \mathbb{E} \left( \max \left( S_+, S_0 e^{M_\tau} \right) - K \right)^+, \text{ call}$$

$$V(S_-, K) = e^{-r(T-\tau)} \mathbb{E} \left( K - \min \left( S_-, S_0 e^{m_\tau} \right) \right)^+, \text{ put}$$

## Discrete prices

The discrete prices at date  $k\Delta$  avec  $k \in \{0, \dots, n\}$  :

$$V^n(S_+, K) = e^{-r\Delta(n-k)} \mathbb{E} \left( \max \left( S_+, S_0 e^{M_\tau^n} \right) - K \right)^+, \text{ call}$$

$$V^n(S_-, K) = e^{-r\Delta(n-k)} \mathbb{E} \left( K - \min \left( S_-, S_0 e^{m_\tau^n} \right) \right)^+, \text{ put}$$

- (H1)  $X$  is a finite activity Lévy process, integrable, satisfying  $\sigma > 0$  et  $\exists \alpha > 0 / \mathbb{E}e^{(1+\alpha)M_T} < \infty$ .
- (H2)  $X$  is a finite activity Lévy process, integrable, satisfying  $\sigma > 0$ .
- (H3)  $X$  is a compound Poisson process (with a drift) integrable and satisfying  $\gamma_0 = 0$  or  $Y_1$  have a bounded density.
- (H4)  $X$  is a integrable infinite activity Lévy process.

## Proposition

The price of a discrete lookback option at the  $k^{\text{th}}$  fixing date and its continuous version at time  $\tau = k\Delta t$  satisfy

$$V_n(S_{\pm}) = e^{\mp\beta_1\sigma\sqrt{\frac{T}{n}}} V\left(S_{\pm}e^{\pm\beta_1\sigma\sqrt{\frac{T}{n}}}\right) \pm \left(e^{\mp\beta_1\sigma\sqrt{\frac{T}{n}}} - 1\right) e^{-\delta(T-\tau)} S_{\tau} + o\left(\frac{1}{\sqrt{n}}\right)$$
$$V(S_{\pm}) = e^{\pm\beta_1\sigma\sqrt{\frac{T}{n}}} V_n\left(S_{\pm}e^{\mp\beta_1\sigma\sqrt{\frac{T}{n}}}\right) \pm \left(e^{\pm\beta_1\sigma\sqrt{\frac{T}{n}}} - 1\right) e^{-\delta(T-\tau)} S_{\tau} + o\left(\frac{1}{\sqrt{n}}\right)$$

where in  $\pm$  and  $\mp$ , the top case applies to the put and the bottom to the call,

$$\beta_1 = -\frac{\zeta\left(\frac{1}{2}\right)}{\sqrt{2\pi}}$$

and  $\zeta$  is the Riemann-zeta function. The put relations are true under  $H1$ , and those for the call under  $H2$ .

## Proposition

The price of a discrete hindsight option at the  $k^{\text{th}}$  fixing date and its continuous version at time  $\tau = k\Delta t$  satisfy

$$V_n(S_{\pm}, K) = e^{\mp\beta_1\sigma\sqrt{\frac{\tau}{n}}} V\left(S_{\pm}e^{\pm\beta_1\sigma\sqrt{\frac{\tau}{n}}}, Ke^{\pm\beta_1\sigma\sqrt{\frac{\tau}{n}}}\right) + o\left(\frac{1}{\sqrt{n}}\right)$$

$$V(S_{\pm}, K) = e^{\pm\beta_1\sigma\sqrt{\frac{\tau}{n}}} V_n\left(S_{\pm}e^{\mp\beta_1\sigma\sqrt{\frac{\tau}{n}}}, Ke^{\mp\beta_1\sigma\sqrt{\frac{\tau}{n}}}\right) + o\left(\frac{1}{\sqrt{n}}\right)$$

where in  $\pm$  and  $\mp$ , the top case applies to the call and the bottom to the put. The call relations are true under  $H1$ , and those for the put under  $H2$ .

## Proposition

Under  $H3$ , the price of a discrete lookback option at the  $k^{\text{th}}$  fixing date and its continuous version at time  $\tau = k\Delta t$  satisfy

- 1 for the call

$$V_n(S_-) = V(S_-) + \frac{\alpha}{n} + o\left(\frac{1}{n}\right)$$

where  $\alpha$  is a constant which we can derive explicitly.

- 2 for the put, if it exists  $\beta > 1$  such that  $\mathbb{E}e^{\beta M_\tau} < \infty$ , then

$$V_n(S_+) = V(S_+) + O\left(\frac{1}{n^{\frac{\beta-1}{\beta}}}\right)$$

## Proposition

Under  $H4$ , the price of a discrete call lookback option at the  $k^{\text{th}}$  fixing date and its continuous version at time  $\tau = k\Delta t$  satisfy

$$V_n(S_-) = V(S_-) + o\left(\frac{1}{\sqrt{n}}\right)$$

If  $\int_{|x| \leq 1} |x| \nu(dx) < \infty$

$$V_n(S_-) = V(S_-) + O\left(\frac{\log(n)}{n}\right)$$

If  $\int_{|x| \leq 1} |x| |\log(|x|)| \nu(dx) < \infty$

$$V_n(S_-) = V(S_-) + O\left(\frac{1}{n}\right)$$

## Proposition

We assume *H4* and that exists  $\beta > 1$  such that  $\mathbb{E}e^{\beta M_T} < \infty$ . The price of a discrete put lookback option at the  $k^{\text{th}}$  fixing date and its continuous version at time  $\tau = k\Delta t$  satisfy,  $\forall \epsilon > 0$

$$V_n(S_+) = V(S_+) + o\left(\frac{1}{n^{\frac{\beta-1}{2\beta}-\frac{\epsilon}{2}}}\right)$$

If  $\int_{|x|\leq 1} |x| \nu(dx) < \infty$

$$V_n(S_+) = V(S_+) + O\left(\left(\frac{\log(n)}{n}\right)^{\frac{\beta-1}{\beta}-\epsilon}\right)$$

If  $\int_{|x|\leq 1} |x| |\log(|x|)| \nu(dx) < \infty$

$$V_n(S_+) = V(S_+) + O\left(\frac{1}{n^{\frac{\beta-1}{\beta}-\epsilon}}\right)$$

## Corollary

The results found for lookback under  $H3$   $H4$  and  $H5$  are true for hindsight, by interchanging "call" and "put".

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# Double Exponential Jump Diffusion Model

We have

$$X_s = \gamma_0 s + \sigma B_s + \sum_{i=1}^{N_s} Y_i$$

$N$  is a poisson process with parameter  $\lambda$ , and  $Y_1$  follows an asymmetric double exponential distribution of density :

$$f_Y(y) = p\eta_1 e^{-\eta_1 y} \mathbb{1}_{\{y \geq 0\}} + q\eta_2 e^{\eta_2 y} \mathbb{1}_{\{y < 0\}}$$

where  $p, q \geq 0$  are constants,  $p + q = 1$ ,  $\eta_1 > 1$  and  $\eta_2 > 0$ .

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where  $p, q \geq 0$  are constants,  $p + q = 1$ ,  $\eta_1 > 1$  and  $\eta_2 > 0$ .

## Parameters of simulations

The parameters are :  $\sigma = 0.3$ ,  $p = 0.6$ ,  $\lambda = 7$ ,  $\eta_1 = 50$  and  $\eta_2 = 25$ .

# Call Case

n	approx. relative error	discrete relative error
1	-1.56%	-57.88%
2	-0.95%	-39.98%
3	-0.66%	-32.27%
4	-0.60%	-27.72%
5	-0.52%	-24.65%
6	-0.48%	-22.39%
7	-0.42%	-20.65%
8	-0.40%	-19.26%
9	-0.38%	-18.13%
10	-0.38%	-17.18%

**Table:** Performance of the propositions of approximation in the double exponential jump diffusion model. The parameters are :  $S_0 = 100$ ,  $r = 0.05$ ,  $\delta = 0$ ,  $T = 1$ ,  $S_- = 90$ . The true continuous call price is 26.1436 and the relative Monte Carlo error is 0.01%.

# Put Case

n	approx. relative error	discrete relative error
1	9.75%	-55.64%
2	2.97%	-42.93%
3	0.82%	-36.49%
4	0.37%	-32.35%
5	0.34%	-29.40%
6	0.29%	-27.18%
7	0.23%	-25.42%
8	0.17%	-24.00%
9	0.12%	-22.80%
10	0.06%	-21.80%

**Table:** Performance of the propositions of approximation in the double exponential jump diffusion model. The parameters are :  $S_0 = 100$ ,  $r = 0.05$ ,  $\delta = 0$ ,  $T = 1$ ,  $S_+ = 110$ . The true continuous put price is 26.0449 and the relative Monte Carlo error is 0.01%.

Thank you for your attention!