

A General Approach for Drawdown (Drawup) Risks of Time-Homogeneous Markov Processes^[1]

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¹Based on a joint work with David Landriault (University of Waterloo) and Hongzhong Zhang (Columbia University)

- Introduction of drawdown and its applications
- Main results

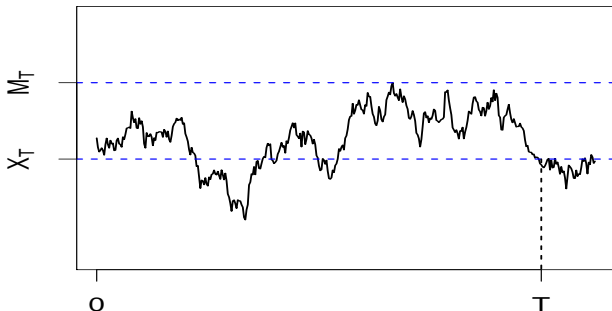


Definition

- **Drawdown** measures the **decline** in value **from the historical peak** for an investment, fund or commodity.
- Consider a stochastic process $X = \{X_t : t \geq 0\}$, the **magnitude of drawdown at time T** is defined by

$$Y_T := M_T - X_T,$$

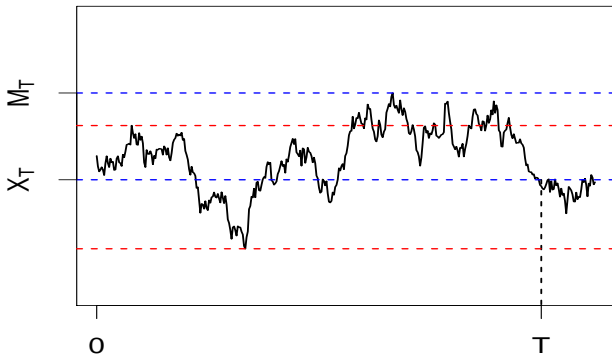
where $M_T = \sup_{0 \leq t \leq T} X_t$



Maximum drawdown

- The **maximum drawdown** up to time T is defined by

$$\sup_{0 \leq t \leq T} Y_t = \sup_{0 \leq t \leq T} \{M_t - X_t\}$$

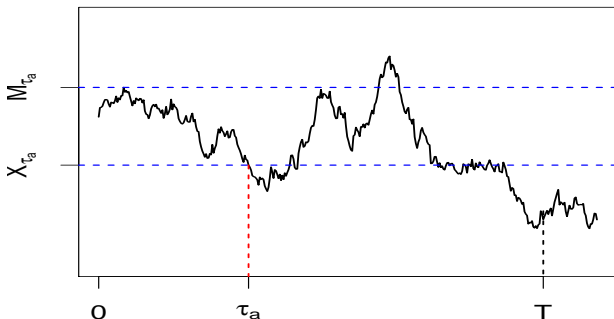


Maximum drawdown (cont.)

- The first time the **magnitude** of drawdown exceeds a pre-specified level $a > 0$ is denoted by

$$\tau_a := \inf\{t > 0 : Y_t \geq a\}.$$

- (Maximum drawdown before time T exceeds a) $\iff (\tau_a \leq T)$



Who cares about drawdown?

- mutual funds managers
- financial mathematicians
- statisticians
- probabilists
- ...
- myself :)



Applications: mutual funds

*“The mutual fund industry and many investment professionals have a well-guarded **secret** they do not want the investing public to know about: **drawdowns**. Drawdowns are, in our opinion, the **single most important determinant** of investing success or failure for most investors.*

*One of the worst characteristics of drawdowns is that they **frequently strike** like tornados. They hit quickly, without warning, and cause immense damage. It’s often difficult to realize their devastation until after they have struck.”*

Greg Miller, CPA, CEO
Wellesley Investment Advisors, Inc.
Wellesley, Massachusetts
April 2006 Issue of Investment Advisor



Applications: mutual funds (cont.)

- Frequently quoted **performance measures** (Schuhmacher and Eling, 2011)
 - Calmar ratio = $\frac{\text{annual rate of return}}{\text{maximum drawdown}}$
 - Sterling ratio = $\frac{\text{annual rate of return}}{\text{average of maximum drawdowns}}$
 - Burke ratio, Martin ratio, Pain ratio, etc.
- Drawdown is an alternative measurement for **volatility**.



- **Portfolio selection and optimization:** Grossman and Zhou (1993), Pospisil and Vecer (2010), Cherny and Obloj (2013), Sekine (2013)
- **Option pricing:** Russian option

$$\begin{aligned}\text{Payoff} &= \sup_{\tau} e^{-a\tau} \{K, M_{\tau}\} \\ \tau^* &= \inf \{t \geq 0 : Y_t \geq k^*\} \text{ for some } k^*\end{aligned}$$

- **Insurance:** Carr et al. (2011), De Finetti dividend problem



- **Change point detection**: Hadjiliadis and Moustakides (2006), Khan (2008)
- **Queueing**: Asmussen (1989), Borovkov (1976), Prabhu (1997)



- Drawdown and drawup processes are usually referred as **reflected processes**

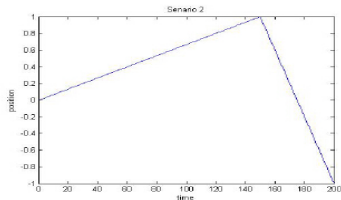
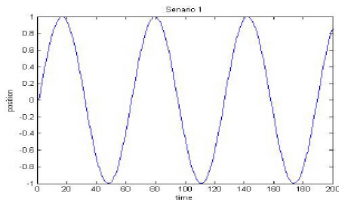
$$Y_t = M_t - X_t \text{ and } \hat{Y}_t = X_t - m_t$$

- **Time-homogeneous diffusion processes**: Taylor (1975), Lehoczky (1977), Magdon et al. (2004), Pospisil et al. (2009)
- **Lévy processes**: Asmussen et al. (2004), Pistorius (2004), Mijatovic and Pistorius (2012), Ivanovs and Palmowski (2012)



Frequency of drawdowns

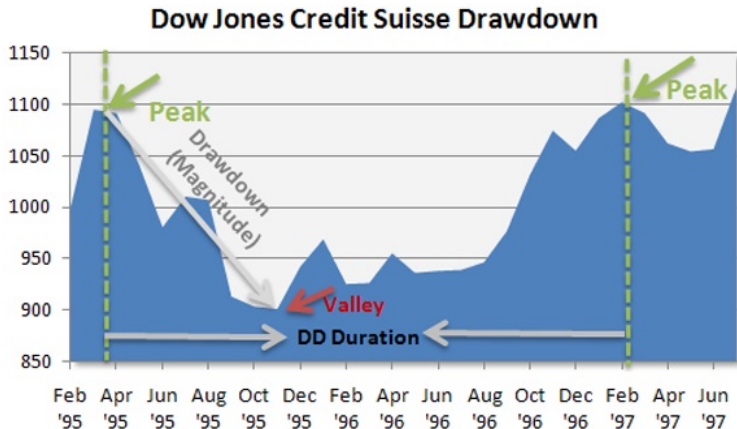
- In addition to the **magnitude**, there are other two aspects of drawdowns: **frequency** and **duration**.



- **Frequency** of drawdowns: Landriault, L., Zhang (2015)



Duration of drawdowns



- Duration of drawdowns: Landriault, L., Zhang (2014)



- Introduction of drawdown and its applications
- Main results



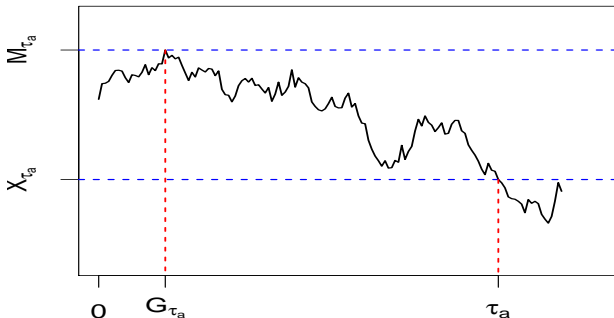
Objectives

- Interested in the joint law of $(\tau_a, M_{\tau_a}, Y_{\tau_a}, G_{\tau_a})$, where

$$G_{\tau_a} = \sup \{0 \leq t < \tau_a : M_t = X_t\}$$

is the last time at maximum

- G_{τ_a} is the turning point from rising to crashing



- Previous approaches (spectrally negative Lévy models and time-homogeneous diffusion models)
 - Ito excursion theory: Avram et al. (2004), Pistorius (2004), Mijatovic and Pistorius (2012)
 - Martingale theory: Taylor (1975), Asmussen et al. (2004), Nguyen-Ngoc and Yor (2005)
 - Approximation approach: Lehoczky (1977), L. et al. (2013), Zhang (2015)



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- Our approach (bound approach)
 - Two steps: constructing tight bounds + taking limits (some regularity assumptions)
 - Advantages: simple, strict, general
 - Building direct connections between drawdown problems and exit problems



Notation of exit times

- Consider a **time-homogeneous Markov process** X
- Define the first passage times

$$T_x^+ = \inf \{t \geq 0 : X_t > x\} \text{ and } T_x^- = \inf \{t \geq 0 : X_t < x\}$$

- For $u \leq x \leq v$, define two functions on two-sided exits

$$B^{(q,s)}(x; u, v) = \mathbb{E}_x[e^{-qT_v^+ - s(X_{T_v^+} - v)} 1_{\{T_v^+ < T_u^-\}}]$$

$$C^{(q,s)}(x; u, v) = \mathbb{E}_x[e^{-qT_u^- - s(u - X_{T_u^-})} 1_{\{T_u^- < T_v^+\}}]$$

- Analytically tractable models for B and C : one-sided Lévy, Kou's double-exponential jump diffusion, meromorphic Lévy, time-homogeneous diffusion, etc.



Lemma

For any $\varepsilon \in (0, a)$, we have

$$\mathbb{E}_x[e^{-qT_{x+\varepsilon}^+} 1_{\{T_{x+\varepsilon}^+ < \tau_a\}}] \leq B^{(q,0)}(x; x-a, x+\varepsilon)$$

$$\mathbb{E}_x[e^{-qT_{x+\varepsilon}^+} 1_{\{T_{x+\varepsilon}^+ < \tau_a\}}] \geq B^{(q,0)}(x; x+\varepsilon-a, x+\varepsilon)$$

and for $\delta \geq 0$,

$$\mathbb{E}_x[e^{-q\tau_a - s(Y_{\tau_a} - a) - \delta(M_{\tau_a} - x)} 1_{\{\tau_a < T_{x+\varepsilon}^+\}}] \leq e^{s\varepsilon} C^{(q,s)}(x; x+\varepsilon-a, x+\varepsilon)$$

$$\mathbb{E}_x[e^{-q\tau_a - s(Y_{\tau_a} - a) - \delta(M_{\tau_a} - x)} 1_{\{\tau_a < T_{x+\varepsilon}^+\}}] \geq e^{-(s+\delta)\varepsilon} C^{(q,s)}(x; x-a, x+\varepsilon)$$



Main result 1

Assumption (1)

The following limits exist and are equal

$$K_a^{(q,s)} := \lim_{\varepsilon \downarrow 0} \frac{C^{(q,s)}(0; \varepsilon - a, \varepsilon)}{1 - B^{(q,0)}(0; \varepsilon - a, \varepsilon)} = \lim_{\varepsilon \downarrow 0} \frac{C^{(q,s)}(0; -a, \varepsilon)}{1 - B^{(q,0)}(0; -a, \varepsilon)}.$$

Theorem

Consider a *general Lévy process* X satisfying Assumption (1). Then

$$\mathbb{E}[e^{-q\tau_a - s(Y_{\tau_a} - a)}] = K_a^{(q,s)}.$$

- In particular, if X is *spectrally negative Lévy*, we recover Theorem 1 of Avram et al. (2004) and Proposition 2 of Pistorius (2004).



Main result 2

Assumption (2)

The following limits exist and satisfy

$$\begin{aligned}b_a^{(q,0)}(x) &:= \lim_{\varepsilon \downarrow 0} \frac{1 - B^{(q,0)}(x; x - a, x + \varepsilon)}{\varepsilon} \\&= \lim_{\varepsilon \downarrow 0} \frac{1 - B^{(q,0)}(x; x + \varepsilon - a, x + \varepsilon)}{\varepsilon} \\&= \lim_{\varepsilon \downarrow 0} \frac{1 - B^{(q,0)}(x - \varepsilon; x - a, x)}{\varepsilon} \\&= \lim_{\varepsilon \downarrow 0} \frac{1 - B^{(q,0)}(x - \varepsilon; x - \varepsilon - a, x)}{\varepsilon}\end{aligned}$$

$$\begin{aligned}c_a^{(q,s)}(x) &:= \lim_{\varepsilon \downarrow 0} \frac{C^{(q,s)}(x; x - a, x + \varepsilon)}{\varepsilon} = \lim_{\varepsilon \downarrow 0} \frac{C^{(q,s)}(x; x + \varepsilon - a, x + \varepsilon)}{\varepsilon} \\&= \lim_{\varepsilon \downarrow 0} \frac{C^{(q,s)}(x - \varepsilon; x - a, x)}{\varepsilon} = \lim_{\varepsilon \downarrow 0} \frac{C^{(q,s)}(x - \varepsilon; x - \varepsilon - a, x)}{\varepsilon}\end{aligned}$$



Main result 2 (cont.)

Theorem

Consider a *spectrally negative time-homogeneous Markov process* X satisfying Assumption (2). Then,

$$\mathbb{E}_x \left[e^{-q\tau_a - s(Y_{\tau_a} - a) - \delta(M_{\tau_a} - x)} \right] = \int_x^\infty e^{-\int_x^y (\delta + b_a^{(q,0)}(z)) dz} c_a^{(q,s)}(y) dy$$

- In particular, if X is a *time-homogeneous diffusion*, we recover results of Lehoczky (1977) by allowing downside jumps and incorporating the law of Y_{τ_a} .



Consider a Lévy process with two-sided jumps

$$X_t = \tilde{X}_t + \sum_{i=1}^{N_t^+} J_i^+$$

- \tilde{X} is a spectrally negative Lévy process with Gaussian coefficient $\sigma > 0$
- The Lévy measure Π satisfying $\int_{(-1,0)} |x| \Pi(dx) < \infty$
- N^+ is a Poisson process with arrival rate



Main result 3 (cont.)

Theorem

Consider a Lévy process X of the above form. Then,

$$\mathbb{E} \left[e^{-q\tau_a - rG_{\tau_a} - s(Y_{\tau_a} - a) - \delta M_{\tau_a}} \right] = \frac{\lim_{\varepsilon \downarrow 0} \frac{C^{(q,s)}(a-\varepsilon; 0, a)}{\varepsilon}}{\delta + \lim_{\varepsilon \downarrow 0} \frac{1 - B^{(q+r, \delta)}(a-\varepsilon; 0, a)}{\varepsilon}}$$

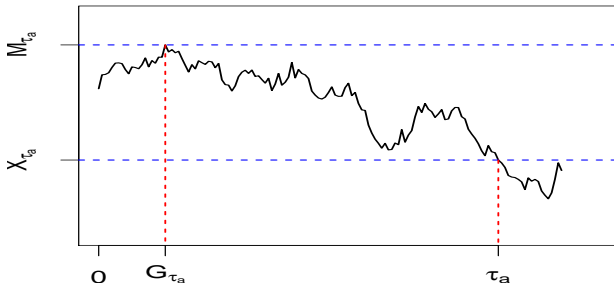
Corollary

(G_{τ_a}, M_{τ_a}) is *independent* of $(\tau_a - G_{\tau_a}, Y_{\tau_a})$



Independence of drawdown estimates

- We showed (G_{τ_a}, M_{τ_a}) is independent of $(\tau_a - G_{\tau_a}, Y_{\tau_a})$ for this particular Lévy model with two-sided jumps
- Landriault, L., Zhang (2014) shows that (G_{τ_a}, M_{τ_a}) is independent of $(\tau_a - G_{\tau_a}, Y_{\tau_a}, Y_{\tau_a-})$ for spectrally negative Lévy processes
- Rising part and crashing part of drawdowns are independent in both time and level scale!!!



Conjecture

(G_{τ_a}, M_{τ_a}) is independent of $(\tau_a - G_{\tau_a}, Y_{\tau_a}, Y_{\tau_a-})$ for *general Lévy processes*



Thank you for your attention!

