

Research Article

Optimal Control of Investment-Reinsurance Problem for an Insurer with Jump-Diffusion Risk Process: Independence of Brownian Motions

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This paper investigates the excess-of-loss reinsurance and investment problem for a compound Poisson jump-diffusion risk process, with the risk asset price modeled by a constant elasticity of variance (CEV) model. It aims at obtaining the explicit optimal control strategy and the optimal value function. Applying stochastic control technique of jump diffusion, a Hamilton-Jacobi-Bellman (HJB) equation is established. Moreover, we show that a closed-form solution for the HJB equation can be found by maximizing the insurer's exponential utility of terminal wealth with the independence of two Brownian motions $W(t)$ and $W_1(t)$. A verification theorem is also proved to verify that the solution of HJB equation is indeed a solution of this optimal control problem. Then, we quantitatively analyze the effect of different parameter impacts on optimal control strategy and the optimal value function, which show that optimal control strategy is decreasing with the initial wealth x and decreasing with the volatility rate of risk asset price. However, the optimal value function $V(t; x; s)$ is increasing with the appreciation rate μ of risk asset.

1. Introduction

By means of investment and reinsurance, insurers can protect themselves against potentially large losses or ensure their earnings remain relatively stable. Therefore, many optimal investment and reinsurance problems have arisen in insurance risk management and have been extensively studied in the literature.

In the older forms, reinsurance was often referred to as “proportional” reinsurance; few studies pay attention to reinsurance. Since Borch [1] studied the safety loading of reinsurance premiums, a vast amount of literature is particularly concerned about reinsurance. However, the excess-of-loss reinsurance is a tool commonly employed in risk management in the recent thirty years. Tapiero and Zuckerman [2] gave the optimum excess-loss reinsurance under a dynamic framework. Taylor [3] studied reserving consecutive layers of inwards excess-of-loss reinsurance. Cao and Xu [4] investigated both proportional and excess-of-loss reinsurance under investment gains. Gu et al. [5] investigated

optimal control of excess-of-loss reinsurance and investment for insurers under a constant elasticity of variance model but without compound Poisson jump in their research. Zhao et al. [6] studied optimal excess-of-loss reinsurance and investment problem for an insurer with jump-diffusion risk process under the Heston model.

It is well known that the compound Poisson process is the most popular and useful model to describe claims process ever since the classical Cramér-Lundberg model in risk theory. However, as a compound Poisson process perturbed by a standard Brownian motion, jump diffusion has been researched extensively in the recent ten years. Jump diffusion can give more practical description of claims than continuous models, which widely used to describe dynamics of surplus process. Yang and Zhang [7] investigated the problem of optimal investment for insurer with jump-diffusion risk process. Li et al. [8] studied the threshold dividend strategy for renewal jump-diffusion process. Ruan et al. [9] studied the optimal portfolio and consumption with habit formation in a jump-diffusion market.

Not only does the insurer cede part of premiums for reinsurance, but the insurer also invests in a financial market consisting of one risk-free asset and a risk asset. The constant elasticity of variance (CEV) model describes the volatility of risk asset that received widespread interests after being proposed by Cox and Ross [10] for European option pricing, partly because the CEV model can capture the implied volatility skew and can explain the volatility smile as well. Hsu et al. [11] gave the integration and detailed derivation for constant elasticity of variance (CEV) option pricing model. Campi et al. [12] investigated systematic equity-based credit risk for a CEV model with jump to default. Applying Legendre transform and dual theory, Xiao et al. [13] investigated analytical strategies for annuity contracts under constant elasticity of variance model. The CEV model was applied to the optimal investment and reinsurance problems in Gu et al. [14]. By maximizing the expected exponential utility of terminal wealth, Lin and Li [15] studied the optimal reinsurance and investment for a jump-diffusion risk process under the CEV model.

The goal of this paper is therefore to investigate the optimal control strategy of excess-of-loss reinsurance and investment problem for an insurer with compound Poisson jump-diffusion risk process from the stochastic control point of view. Under the hypothesis that two Brownian motions $W(t)$ and $W_1(t)$ are mutually independent, we present three main results. We first give the optimal excess-of-loss reinsurance strategy and investment strategy, respectively, according to the first-order necessary condition of maximum. We also obtain the optimal value function by plugging a reasonable conjecture into the HJB equation and solving linear second-order partial differential equations. In the third part, we specialize in a verification theorem. We also prove the verification theorem in detail inspired by the results of Taksar and Zeng [16], Gu et al. [5], and Zhao et al. [6]. At last, we present some numerical examples to illustrate the impacts of model parameters.

The paper is organized as follows. Section 2 introduces the risk asset price model under the CEV model and the wealth process with jump diffusion. In Section 3, we consider the optimal excess-of-loss reinsurance and investment strategy and a verification theorem as well. Section 4 contains some numerical simulations and theoretical results. Section 5 brings this paper to an end.

2. The Model

Throughout this paper, $(\Omega, \mathcal{F}, P, \{\mathcal{F}_t\}_{0 \leq t \leq T})$ denotes a complete probability space satisfying the usual condition, where a finite constant $T > 0$ represents the investment time horizon; two standard Brownian motions $W_1(t)$ and $W(t)$ are independent of each other; γ is a Poisson random measure; $\mathcal{F}_t := \mathcal{F}_t^W \vee \mathcal{F}_t^{W_1} \vee \mathcal{F}_t^\gamma$ represents the minimal σ -field generated by \mathcal{F}_t^W , $\mathcal{F}_t^{W_1}$, and \mathcal{F}_t^γ , which contains all the information available until time t . All stochastic processes involved in this paper are supposed to be \mathcal{F}_t adapted. In addition, let $x \wedge y = \min\{x, y\}$ below.

Following the same formulation of Zhao et al. [6], the surplus process of an insurer is described by the following jump-diffusion model:

$$dR(t) = cdt + \sigma dW(t) - dC(t), \tag{1}$$

where c and σ are positive constants, $W(t)$ is a standard Brownian motion, and $\sigma W(t)$ describes the uncertainty associated with the surplus of the insurer at time t . Assume $C(t) = \sum_{i=1}^{N(t)} Z_i$ which represents the cumulative claims until time t , where $N(t)$ is a homogeneous Poisson process with intensity λ and the claim sizes $\{Z_i, i \geq 1\}$ are independent and identically distributed positive random variables with common distribution $F(z)$. Denote the mean value $E[Z_i] = \mu_\infty$ and

$$D := \sup \{z : F(z) \leq 1\} < +\infty. \tag{2}$$

Suppose that $F(0) = 0, 0 < F(z) < 1$ if $z \in (0, D)$ and $F(z) = 1$ if $z \geq D$. The premium according to the expected value principle is $c = (1 + \eta)\lambda\mu_\infty$, where $\eta > 0$ is the safety loading of the insurer.

According to the theory of Poisson random measure (referring to Øksendal and Sulem [17]), we can rewrite the compound Poisson process $C(t) = \sum_{i=1}^{N(t)} Z_i$. γ denotes the Poisson random measure; then

$$\begin{aligned} \sum_{i=1}^{N(t)} Z_i &= \int_0^t \int_{\mathbb{R}^+} z \gamma(dz, du), \\ C^a(t) &= \sum_{i=1}^{N(t)} Z_i = \int_0^t \int_{\mathbb{R}^+} (z \wedge a) \gamma(dz, du). \end{aligned} \tag{3}$$

The compensator ν of the random measure γ is

$$\nu(dz, dt) = \lambda dF(z) dt, \tag{4}$$

so the compensated Poisson random measure of γ is

$$\tilde{\gamma}(dz, dt) = \gamma(dz, dt) - \lambda dF(z) dt. \tag{5}$$

Insurance company can purchase excess-of-loss reinsurance to reduce the risk. Supposing the insurer's (fixed) retention level denoted by a , the corresponding reserve process is

$$dR(t) = c^{(a)} dt + \sigma dW(t) - dC^{(a)}(t) \tag{6}$$

and the reinsurer's premium rate is

$$\begin{aligned} c^{(a)} &= (1 + \eta)\lambda\mu_\infty - (1 + \theta)\lambda\{\mu_\infty - E[Z_i \wedge a]\} \\ &= (\eta - \theta)\lambda\mu_\infty + (1 + \theta)\lambda E[Z_i \wedge a], \\ E[Z_i \wedge a] &= \int_0^a z dF(z) + \int_a^\infty a dF(z) \\ &= \int_0^a (1 - F(z)) dz \triangleq \int_0^a \overline{F}(z) dz \end{aligned} \tag{7}$$

and $C^{(a)} = \sum_{i=1}^{N(t)} (Z_i \wedge a)$ satisfying

$$e^{\int_0^t e^{-rs} dC^{(a)}(s)} < \infty, \quad \forall t < \infty. \tag{8}$$

In particular, if the retention level $a = D$, then

$$c^{(a)} = (1 + \eta) \lambda \mu_{\infty} = c^{(D)} = c. \tag{9}$$

Without loss of generality, we always assume the reinsurance is not cheap; that is, $\theta > \eta$ and θ denotes the safety loading of the reinsurer.

Moreover, the insurer also can invest in a financial market consisting of one risk-free asset and one risky asset. The price process $S_0(t)$ of the risk-free asset is given by

$$dS_0(t) = rS_0(t) dt, \tag{10}$$

where $r > 0$ is the risk-free interest rate. The price process $S(t)$ of the risky asset is described by the constant elasticity of variance (CEV) model:

$$dS(t) = S(t) [\mu dt + kS^{\beta}(t) dW_1(t)], \tag{11}$$

where μ and k are positive constants, $\mu > r$ is the expected instantaneous return rate of the risky asset, $kS^{\beta}(t)$ is the instantaneous volatility, β is the elasticity parameter, and $W_1(t)$ is a standard Brownian motion. When $\beta = 0$, a CEV model degenerates into a GBM. In this paper, we assume the standard Brownian motion $W_1(t)$ is independent of $W(t)$.

The control $\alpha = \{(a(t), \pi(t)) : t \in [0, T]\}$ is a two-dimensional $\{\mathcal{F}(t)\}$ adapted stochastic process, where $a(t) \in [0, D]$ is the retention level of the excess-of-loss at time t , in which $a(t) \equiv D$ means “no reinsurance” and $a(t) \equiv 0$ means “full reinsurance,” and $\pi(t)$ represents the proportion invested in the risky asset. Here, short-selling is not allowed; that is, $\pi(t) \geq 0$. The amount invested in the risk-free asset is $X(t)(1 - \pi(t))$, where $X(t)$ is the wealth process for the insurer under the strategy α and the dynamics of $X(t)$ is given by

$$\begin{aligned} dX(t) = & \left[(\eta - \theta) \lambda \mu_{\infty} + (1 + \theta) \lambda \int_0^a \overline{F(x)} dx \right. \\ & \left. + \mu \pi(t) X(t) + r(1 - \pi(t)) X(t) \right] dt \\ & + \sigma dW(t) + \pi(t) X(t) k S^{\beta}(t) dW_1(t) - dC^{(a)}(t) \end{aligned}$$

$$X(0) = x_0. \tag{12}$$

A strategy α is said to be admissible, if $\forall t \in [0, T]$, $(a(t), \pi(t))$ is $\mathcal{F}(t)$ progressively measurable, and $E[\int_0^{\infty} \pi^2(t) X^2(t) S^{2\beta}(t) dt] < \infty$, $a(t) \in [0, D]$, and $\pi(t) \in [0, +\infty)$. The set of all admissible strategies denoted by Λ and (12) has a unique (strong) solution. Suppose the insurer has a utility function $U(x)$ which is strictly concave and continuously differentiable on $(-\infty, +\infty)$ and aims to maximize the expected utility of his/her terminal wealth; that is,

$$\max_{\alpha \in \Lambda} E[U(X(T))]. \tag{13}$$

3. Main Results

In this section, we solve the excess-of-loss reinsurance and investment problem with independence of two Brownian motions $W(t)$ and $W_1(t)$. By maximizing the expected utility of terminal wealth, the optimal strategy and value function are given. At the beginning, let us give the following exponential utility function of a risk aversion insurer:

$$U(x) = -\frac{1}{q} e^{-qx}, \quad q > 0. \tag{14}$$

This utility function has a constant absolute risk aversion parameter q and is the only utility function under the principle of “zero utility” giving a fair premium that is independent of the level of reserves of insurers. For an admissible strategy $\alpha = (a(t), \pi(t))$, we define the value function as

$$H^{\alpha}(t, x, s) = E[U(X(T)) | X(t) = x, S(t) = s] \tag{15}$$

and the optimal value function is

$$H(t, x, s) = \sup_{\alpha \in \Lambda} H^{\alpha}(t, x, s) \tag{16}$$

with the boundary condition

$$H(T, x, s) = -\frac{1}{q} e^{-qx}. \tag{17}$$

The objective of the insurer is to find an optimal strategy $\alpha^* = (a^*(t), \pi^*(t))$ such that $H(t, x, s) = H^{\alpha^*}(t, x, s)$, where $a^*(t)$ is called the optimal reinsurance strategy and $\pi^*(t)$ is called the optimal investment strategy.

For any $H^{\alpha} \in C^{1,2,2}([0, T] \times R_+ \times R_+)$, define the generator

$$\begin{aligned} \mathcal{A}H^{\alpha}(t, x, s) = & H_t^{\alpha} + [(\eta - \theta) \lambda \mu_{\infty} + rx] H_x^{\alpha} \\ & + \mu s H_s^{\alpha} + \frac{1}{2} \sigma^2 H_{xx}^{\alpha} + \frac{1}{2} k^2 s^{2\beta+2} H_{ss}^{\alpha} \\ & + \left\{ (\mu - r) \pi x H_x^{\alpha} + \frac{1}{2} \pi^2 x^2 k^2 s^{2\beta} H_{xx}^{\alpha} + k^2 \pi x s^{2\beta+1} H_{xs}^{\alpha} \right\} \\ & + \left\{ H_x^{\alpha} (1 + \theta) \lambda \int_0^a \overline{F(x)} dx \right. \\ & \left. + \lambda E[H^{\alpha}(t, x - (Z_i \wedge a), s) - H^{\alpha}(t, x, s)] \right\}, \end{aligned} \tag{18}$$

where H_t^{α} , H_x^{α} , H_s^{α} , H_{xx}^{α} , and H_{ss}^{α} denote the corresponding first- and second-order partial derivatives of $H^{\alpha}(t, x, s)$ with respect to (w.r.t.) the corresponding variables, respectively.

Applying the classical tools of stochastic optimal control, we can derive the following Hamilton-Jacobi-Bellman (HJB) equation for problem (16):

$$\sup_{\alpha \in \Lambda} \{\mathcal{A}H^{\alpha}(t, x, s)\} = 0 \tag{19}$$

with the boundary condition (17).

Standard results (e.g., Fleming and Soner [18]) tell us (19) admits the unique strong solution, which, together with the fact that the value function is twice-continuously differentiable, gives the following theorem.

Theorem 1. For the optimal excess-of-loss reinsurance and investment problem with jump-diffusion risk process under the CEV model, a solution to HJB equation (19) with boundary condition (17) is given by $V(t, x, s)$ and the corresponding maximizer is given by $\alpha^* = (a^*, \pi^*)$ in feedback form, where one has the following.

(1) If $D \geq \ln(1 + \theta)/q$, the optimal retention level on the whole interval $[0, T]$ always is

$$a_*^0(t) = \frac{\ln(1 + \theta)}{q} e^{-r(T-t)} \tag{20}$$

and the optimal investment strategy is given by

$$\pi^*(t) = \left[\frac{\mu - r}{xqk^2} - \frac{(\mu - r)^2 (e^{2r\beta(t-T)} - 1)}{2qxrk^2} \right] e^{-r(T-t)} s^{-2\beta}. \tag{21}$$

The optimal value function is

$$\begin{aligned} V(t, x, s) &= -\frac{1}{q} \exp \left\{ -q \right. \\ &\quad \times \left[xe^{r(T-t)} + \frac{(\eta - \theta) \lambda \mu_\infty e^{r(T-t)}}{r} - \frac{\sigma^2 q e^{2r(T-t)}}{4r} \right. \\ &\quad \left. \left. + \int_t^T \left(e^{r(T-y)} (1 + \theta) \lambda \right. \right. \right. \\ &\quad \quad \left. \left. \times \int_0^{(\ln(1+\theta)/q)e^{-r(T-y)}} \overline{F(x)} dx \right) dy \right. \\ &\quad \left. \left. - \int_t^T \left(\lambda e^{r(T-y)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-y)}} e^{qx} e^{r(T-y)} \right. \right. \right. \\ &\quad \quad \left. \left. \times \overline{F(x)} dx \right) dy \right. \\ &\quad \left. - \frac{(\eta - \theta) \lambda \mu_\infty}{r} + \frac{\sigma^2 q}{4r} \right. \\ &\quad \left. + \frac{e^{-2r\beta T} (-e^{2r\beta t} + e^{2r\beta T}) (\mu - r)^2}{4qr\beta k^2} s^{-2\beta} \right. \\ &\quad \left. + (e^{-2r\beta T} (1 + 2\beta) \right. \\ &\quad \quad \times (e^{2r\beta t} + e^{2r\beta T} (-1 + 2r(T-t)\beta)) \\ &\quad \quad \left. \left. \times (\mu - r)^2 \times (8q\beta r^2)^{-1} \right] \right\}. \tag{22} \end{aligned}$$

(2) If $D < \ln(1 + \theta)/q$, the optimal retention level is

$$a^*(t) = \begin{cases} \frac{\ln(1 + \theta)}{q} e^{-r(T-t)}, & t \in [0, t^0), \\ D, & t \in [t^0, T]. \end{cases} \tag{23}$$

And the optimal investment strategy is given by

$$\pi^*(t) = \left[\frac{\mu - r}{xqk^2} - \frac{(\mu - r)^2 (e^{2r\beta(t-T)} - 1)}{2qxrk^2} \right] e^{-r(T-t)} s^{-2\beta}. \tag{24}$$

The optimal value function is given by

$$\begin{aligned} V(t, x, s) &= -\frac{1}{q} \exp \left\{ -q [xe^{r(T-t)} + f^*(t) + g^*(t, s)] \right\} \\ &= \begin{cases} -\frac{1}{q} \exp \left\{ -q [xe^{r(T-t)} + \widehat{g}(t, s) + \widehat{f}(t)] \right\}, & t \in [0, t^0), \\ -\frac{1}{q} \exp \left\{ -q [xe^{r(T-t)} + \widehat{g}(t, s) + \widehat{f}(t)] \right\}, & t \in [t^0, T], \end{cases} \end{aligned}$$

$\widehat{g}(t, s)$

$$\begin{aligned} &= \left[\left(-\frac{(\mu - r)^2 e^{-2rT\beta}}{4qrk^2\beta} \right) e^{2\beta rt} + \frac{(\mu - r)^2}{4qk^2\beta r} \right] s^{-2\beta} \\ &\quad + \left(\frac{(\mu - r)^2 e^{-2rT\beta}}{4qrk^2\beta} \right) \frac{k^2 (2\beta + 1)}{2r} e^{2\beta rt} \\ &\quad - \frac{(2\beta + 1)}{qr} (\mu - r)^2 + (2\beta + 1) (\mu - r)^2 \\ &\quad \times \left(\frac{4 + T - t^0}{4qr} - \frac{1}{8qr^2\beta} \right), \quad t \in [0, t^0), \end{aligned}$$

$\widehat{g}(t, s)$

$$\begin{aligned} &= \frac{(1 - e^{2r\beta(t-T)}) (\mu - r)^2}{4qk^2 r\beta} s^{-2\beta} \\ &\quad + \frac{(1 + 2\beta) [e^{2r\beta(t-T)} - 1 + 2r(T-t)\beta] (\mu - r)^2}{8qr^2\beta}, \quad t \in [t^0, T], \end{aligned}$$

$\widehat{f}(t)$

$$\begin{aligned} &= \frac{(1 + \eta) \lambda \mu_\infty e^{r(T-t)}}{r} - \frac{\sigma^2 q e^{2r(T-t)}}{4r} \\ &\quad - \int_t^T \left[\frac{\lambda}{q} \int_0^D e^{qx} e^{r(T-\tau)} dF(x) \right] d\tau \\ &\quad - \frac{\lambda(t-T)}{q} - \frac{(1 + \eta) \lambda \mu_\infty}{r} + \frac{\sigma^2 q}{4r}, \quad t \in [t^0, T], \end{aligned}$$

$$\begin{aligned}
 \widetilde{f}(t) &= \frac{(\eta - \theta) \lambda \mu_\infty}{r} e^{r(T-t)} - \frac{\sigma^2 q}{4r} e^{2r(T-t)} \\
 &\quad - \int_0^t \left[(1 + \theta) \lambda e^{r(T-\tau)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-\tau)}} \overline{F(x)} dx \right] d\tau \\
 &\quad + \lambda \int_0^t \left[e^{r(T-\tau)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-\tau)}} e^{qx} e^{r(T-\tau)} \overline{F(x)} dx \right] d\tau \\
 &\quad - \frac{(\eta - \theta) \lambda \mu_\infty}{r} e^{r(T-t^0)} \\
 &\quad + (1 + \theta) \lambda \int_0^{t^0} \left[e^{r(T-\tau)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-\tau)}} \overline{F(x)} dx \right] d\tau \\
 &\quad - \lambda \int_0^{t^0} \left[e^{r(T-\tau)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-\tau)}} e^{qx} e^{r(T-\tau)} \overline{F(x)} dx \right] d\tau \\
 &\quad + \frac{(1 + \eta) \lambda \mu_\infty}{r} e^{r(T-t^0)} \\
 &\quad - \frac{\lambda}{q} \int_{t^0}^T \left[\int_0^D e^{qx} e^{r(T-\tau)} dF(x) \right] d\tau \\
 &\quad - \frac{\lambda(t^0 - T)}{q} - \frac{(1 + \eta) \lambda \mu_\infty}{r} + \frac{\sigma^2 q}{4r}, \quad t \in [0, t^0].
 \end{aligned} \tag{25}$$

Proof. See Appendix A. \square

Remark 2. From Theorem 1, we find that the optimal investment strategy is a function of $(\mu - r)$, q , x , ks^β , r , and t , which fully reflects the influence of various factors on the investment strategy. x is the initial wealth, ks^β represents the volatility of risk asset price, and $\mu - r$ is the profit of risk asset appreciation higher than risk-free interest rate. By simple deformation,

$$\begin{aligned}
 \pi^*(t) &= \left[\frac{\mu - r}{xqk^2} - \frac{(\mu - r)^2 (e^{2r\beta(t-T)} - 1)}{2qxrk^2} \right] e^{-r(T-t)} s^{-2\beta} \\
 &= (\mu - r) \left[\frac{1}{q} - \frac{(\mu - r) (1 - e^{-2r\beta(T-t)})}{2qr} \right] \\
 &\quad \cdot e^{-r(T-t)} \cdot \frac{1}{x} \cdot \frac{1}{(ks^\beta)^2}.
 \end{aligned} \tag{26}$$

Obviously, the optimal investment strategy decreases with respect to the initial wealth x and the volatility of risk asset price ks^β . In particular, if $\beta = 0$, then $\pi^*(t) = ((\mu - r)/xqk^2)e^{-r(T-t)}$ which is the same as that in Theorem 1 in Gu et al. [5].

A surprising finding is that the optimal reinsurance strategy has nothing to do with the initial wealth and risk asset. However, insurer's safe load θ and the parameter q in the utility function play an important role in determining reinsurance strategy.

Motivated by the results of Taksar and Zeng [16], Gu et al. [5], and Zhao et al. [6], we obtain the following verification theorem.

Theorem 3 (verification theorem). *For the optimal excess-of-loss reinsurance and investment problem with jump-diffusion risk process under the CEV model, the wealth process $X(t)$ is associated with an admissible strategy $(a(t), \pi(t))$. If the solution to HJB equation (19) with boundary condition (17) is given by $V(t, x, s)$ and the parameters $\mu > r > 0$ satisfy either of the following conditions:*

- (I) $r > (1 - 1/\sqrt{6})\mu$;
- (II) $r < \frac{(1 - 1/\sqrt{6})\mu}{(1/\beta\sqrt{6(\mu - r)^2 - \mu^2}) \arctan(-\sqrt{6(\mu - r)^2 - \mu^2}/\mu)}$; $T < \dots$

then the optimal value function is $H(t, x, s) = V(t, x, s)$, and the optimal strategy is $\alpha^ = (a^*(t), \pi^*(t))$, given in Theorem 1.*

Proof. See Appendix B. \square

Remark 4. In contrast with Gu et al. [5], they also considered the excess-of-loss with a compound Poisson jump under the CEV model; however, they simplified the compound Poisson jump by diffusion approximation process which leads the problem back to the case without jump.

4. Numerical Results

In this section, we analyze the impacts of some parameters on the optimal strategies and the value function. Theoretical analysis and some corresponding numerical examples are given to illustrate the influences of model parameters on the optimal strategy and the optimal value function when $D \geq \ln(1 + \theta)/q$. The analysis for the case of $D < \ln(1 + \theta)/q$ is similar.

Throughout this section, we always assume $\beta \geq 0$; the claim sizes follow exponential distribution $F(x) = 1 - e^{-x/m}$, $x > 0$, and if $\beta < 0$, some of the conclusions will be different. The parameters are given by $m = 2$, $q = 0.5$, $T = 10$, $\sigma = 1$, $\mu_\infty = 0.5$, $\theta = 0.2$, $r = 0.05$, $\mu = 0.12$, $k = 0.2$, $\lambda = 2$, $x = 1$, $\beta = 0.5$, and $s = 1$.

(1) First, let us pay attention to the expression of the optimal strategy when $\beta \geq 0$:

$$\begin{aligned}
 \pi^*(t) &= \left[\frac{\mu - r}{xqk^2} - \frac{(\mu - r)^2 (e^{2r\beta(t-T)} - 1)}{2qxrk^2} \right] e^{-r(T-t)} s^{-2\beta} \\
 &= \left[\frac{\mu - r}{q} - \frac{(\mu - r)^2 (e^{2r\beta(t-T)} - 1)}{2qr} \right] \\
 &\quad \cdot e^{-r(T-t)} \cdot \frac{1}{x} \cdot \frac{1}{(ks^\beta)^2},
 \end{aligned} \tag{27}$$

where x is the initial wealth and ks^β represents the volatility of risk asset price.

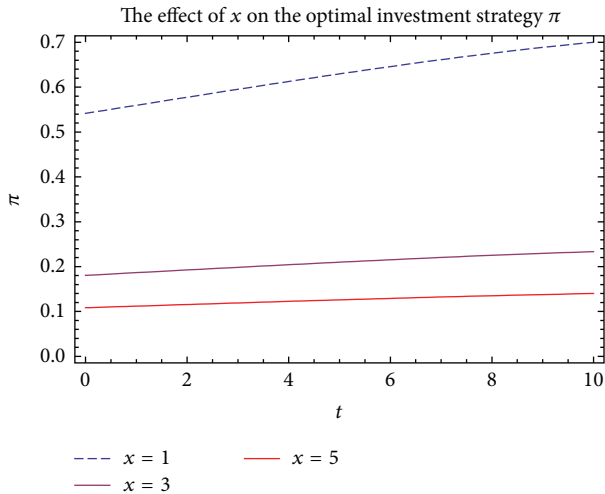


FIGURE 1: The optimal investment strategy π decreases with respect to x .

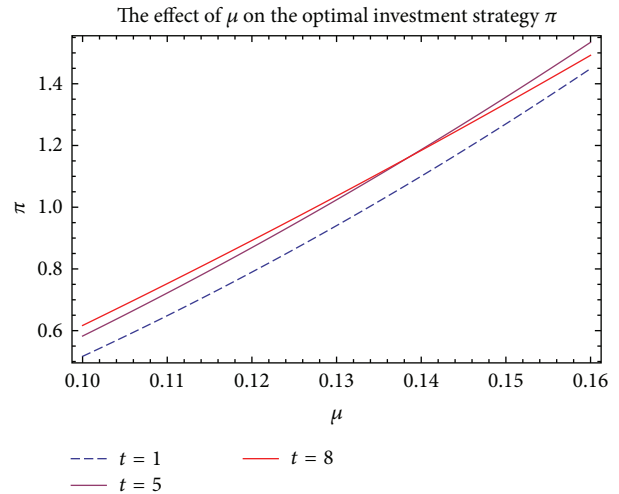


FIGURE 2: The optimal investment strategy increases with the appreciation rate μ of risky asset.

Deriving π^* with respect to x , we have

$$\begin{aligned} \frac{\partial \pi^*}{\partial x} &= \left(e^{r(t-T(1+2\beta))} s^{-2\beta} (r - \mu) \right. \\ &\quad \times \left[\left(e^{2rt\beta} (r - \mu) + e^{2rT\beta} (r + \mu) \right) \right] \times (2k^2 qrx^2)^{-1} \\ &= - \left((\mu - r) s^{-2\beta} e^{r(t-T(1+2\beta))} \right. \\ &\quad \times \left[r \left(e^{2rT\beta} + e^{2rt\beta} \right) + \mu \left(e^{2rT\beta} - e^{2rt\beta} \right) \right] \\ &\quad \times (2k^2 qrx^2)^{-1} < 0, \end{aligned} \tag{28}$$

because $\mu > r > 0$, $T > t$, $s^{-2\beta} > 0$, $e^{r(t-T(1+2\beta))} > 0$, and $\mu(e^{2rT\beta} - e^{2rt\beta}) > 0$.

Remark 5. Obviously, the optimal investment strategy π^* decreases with respect to the initial wealth x , which tells us that, for a risk aversion insurer with the initial fund bigger and bigger, the investment proportion on risk asset must be reduced to make the actual amount invested on risk asset stay at an appropriate level to avoid the potential risk of being unbearable. See Figure 1.

However, the optimal strategy increases with respect to appreciation rate μ of the risky asset.

Since $T > t$, $\mu > r > 0$ so that $e^{2rT\beta} > e^{2rt\beta}$ and $\mu > (\mu - r)$, we have

$$\frac{\partial \pi^*}{\partial \mu} = \frac{e^{r(t-T(1+2\beta))} s^{-2\beta} \left[e^{2rT\beta} \mu - e^{2rt\beta} (\mu - r) \right]}{k^2 qrx} > 0. \tag{29}$$

Figure 2 shows that the optimal strategy increases with respect to the return rate of the risky asset μ , which is obviously established. The higher appreciation rate μ of risky asset will attract more money invested on the risk asset and force the investment proportion π increase. See Figure 2.

Proposition 6. Suppose $\beta \geq 0$ is the volatility parameter in CEV model; π^* is the optimal investment strategy. If $s^2 > e^{(\mu-r)(T-t)}$, $s > 1$, $\mu > r > 0$, $T > t$, then the optimal strategy π^* decreases with respect to k , β , and s . In fact, the optimal investment strategy decreases with the volatility rate of risk asset price denoted by $ks^\beta(t)$.

Proof. The optimal strategy with respect to β may raise various possible cases under different conditions:

$$\begin{aligned} \frac{\partial \pi^*}{\partial \beta} &= - \left(e^{r(t-T(1+2\beta))} s^{-2\beta} (r - \mu) \right. \\ &\quad \times \left[e^{2rt\beta} (r - \mu) (r(t - T) - \ln [s]) \right. \\ &\quad \quad \left. \left. - e^{2rT\beta} (r + \mu) \ln [s] \right] \right) \times (k^2 qrx)^{-1} \\ &= \left(e^{r(t-T(1+2\beta))} s^{-2\beta} (\mu - r) \right. \\ &\quad \times \left[e^{2rt\beta} (\mu - r) (r(T - t) + \ln [s]) \right. \\ &\quad \quad \left. \left. - e^{2rT\beta} (r + \mu) \ln [s] \right] \right) \times (k^2 qrx)^{-1}. \end{aligned} \tag{30}$$

Since $s > 1$, $\mu > r$, and $T > t$, then $\mu - r > 0$, $T - t > 0$, $e^{2rT\beta} > e^{2rt\beta} > 0$, and $\ln [s] > 0$.

If $s^2 > e^{(\mu-r)(T-t)}$, take logarithm on both sides

$$\ln s^2 > (\mu - r)(T - t), \quad \text{i.e., } 2 \ln s > (\mu - r)(T - t). \tag{31}$$

We have

$$\begin{aligned} &e^{2rT\beta} 2r \ln [s] - e^{2rt\beta} r (\mu - r) (T - t) \\ &> e^{2rt\beta} r (2 \ln [s] - (\mu - r) (T - t)) > 0, \end{aligned} \tag{32}$$

so that

$$\begin{aligned} \frac{\partial \pi^*}{\partial \beta} &= (k^2 q r x)^{-1} e^{r(t-T(1+2\beta))} s^{-2\beta} (\mu - r) \\ &\quad \times [e^{2rt\beta} (\mu - r) \ln [s] \\ &\quad + e^{2rt\beta} (\mu - r) r (T - t) - e^{2rT\beta} \\ &\quad \times (\mu - r) \ln [s] - 2r \ln [s] e^{2rT\beta}] \\ &= (k^2 q r x)^{-1} e^{r(t-T(1+2\beta))} s^{-2\beta} (\mu - r) \\ &\quad \times [-(e^{2rT\beta} - e^{2rt\beta}) (\mu - r) \ln [s] \\ &\quad - (e^{2rT\beta} 2r \ln [s] - e^{2rt\beta} r (\mu - r) (T - t))] < 0, \end{aligned} \tag{33}$$

which means that the optimal strategy decreases with respect to β (see Figure 3(c)):

$$\begin{aligned} \frac{\partial \pi^*}{\partial s} &= (e^{r(t-T(1+2\beta))} s^{-2\beta} (r - \mu) \\ &\quad \times [(e^{2rt\beta} (r - \mu) + e^{2rT\beta} (r + \mu))]) \times (2k^2 q r x^2)^{-1} \\ &= -((\mu - r) s^{-2\beta} e^{r(t-T(1+2\beta))} \\ &\quad \times [r(e^{2rT\beta} + e^{2rt\beta}) + \mu(e^{2rT\beta} - e^{2rt\beta})]) \\ &\quad \times (2k^2 q r x^2)^{-1} < 0, \end{aligned} \tag{34}$$

because $\mu - r > 0$, $\beta > 0$, $s^{-2\beta} > 0$, $e^{r(t-T(1+2\beta))} > 0$, and $\mu(e^{2rT\beta} - e^{2rt\beta}) > 0$.

Consider

$$\frac{\partial \pi^*}{\partial k} = -\frac{e^{-r(T-t)} s^{-2\beta} (\mu - r) (\mu + r - e^{-2r(T-t)\beta} (\mu - r))}{k^3 q r x} < 0. \tag{35}$$

Thus the optimal strategy decreases with respect to s and k , respectively. See Figures 3(a) and 3(b). \square

All the calculations above show that the optimal strategy decreases with respect to k , s , and β simultaneously. So it comes to a conclusion that the optimal investment strategy decreases with the volatility rate $ks^\beta(t)$ of risk asset price.

Remark 7. The conclusion of this proposition is very natural. Risk averse investors always adopt a relatively conservative investment strategy to avoid risk. With the increase of risk asset price volatility rate $ks^\beta(t)$ which means risk increase, the risk averse insurer must reduce the investment proportion π on risk asset to keep the company's risk at an appropriate level. So the optimal investment strategy π^* decreases with the risk asset price volatility rate $ks^\beta(t)$. See Figure 3.

(2) Next, we are concerned with the expression of the value function for the special case when $\eta = \theta = 0$, which

makes the value function simplified completely. Without loss of generality, we also suppose $\beta \geq 0$; the case of $\beta < 0$ is similar. Hence,

$$\begin{aligned} V(t, x, s) &= -\frac{1}{q} \exp \\ &\quad \times \left\{ -q \left[x e^{r(T-t)} - \frac{\sigma^2 q e^{2r(T-t)}}{4r} \right. \right. \\ &\quad + \frac{e^{-2r\beta T} (-e^{2r\beta t} + e^{2r\beta T}) (\mu - r)^2}{4qr\beta k^2} s^{-2\beta} \\ &\quad + (e^{-2r\beta T} (1 + 2\beta) \\ &\quad \times (e^{2r\beta t} + e^{2r\beta T} (-1 + 2r(T-t)\beta)) (\mu - r)^2) \\ &\quad \left. \left. \times (8q\beta r^2)^{-1} + \frac{\sigma^2 q}{4r} \right] \right\}. \end{aligned} \tag{36}$$

Proposition 8. $V(t, x, s)$ is the optimal value function under the optimal strategy (a^*, π^*) and k is the volatility coefficient of risk asset price in CEV model. If $k > 0$, $s > 0$, $\mu > r > 0$, and $T > t$, then the optimal value function $V(t, x, s)$ decreases with respect to k and s , respectively.

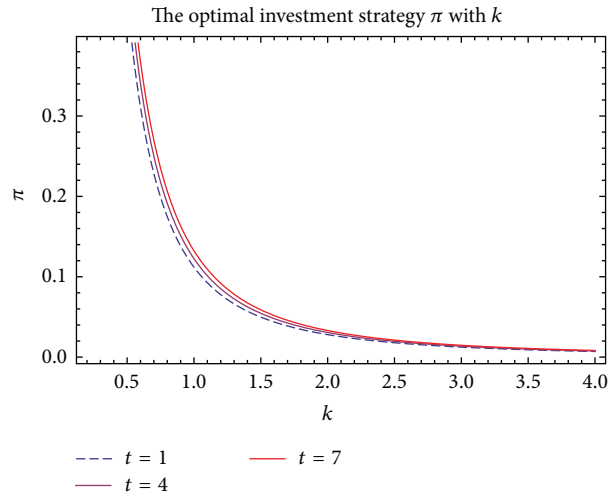
Proof. If $k > 0$, $s > 0$, $\mu > r > 0$, and $T > t$, then

$$(2k^3 q r \beta)^{-1} (e^{2rT\beta} - e^{2rt\beta}) s^{-2\beta} (r - \mu)^2 > 0, \tag{37}$$

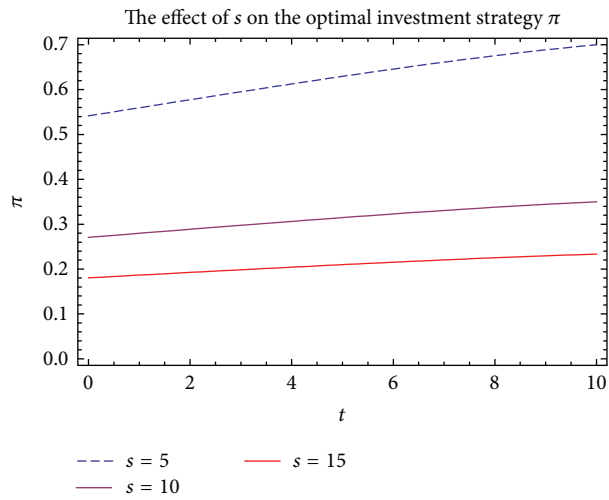
so that

$$\begin{aligned} \frac{\partial V}{\partial k} &= -(2k^3 q r \beta)^{-1} (e^{2rT\beta} - e^{2rt\beta}) s^{-2\beta} (r - \mu)^2 \\ &\quad \cdot \exp \left\{ -2rT\beta - \frac{1}{8} q \right. \\ &\quad \times \left[8e^{r(-t+T)} x - \frac{2(-1 + e^{2r(-t+T)}) q \sigma^2}{r} \right. \\ &\quad - (k^2 q r^2 \beta)^{-1} e^{-2rT\beta} s^{-2\beta} (r - \mu)^2 \\ &\quad \times [2(e^{2rt\beta} - e^{2rT\beta}) r + k^2 s^{2\beta} (1 + 2\beta) \\ &\quad \left. \left. \times (-e^{2rt\beta} + e^{2rT\beta} (1 + 2r(t-T)\beta)) \right] \right\} \\ &< 0, \end{aligned} \tag{38}$$

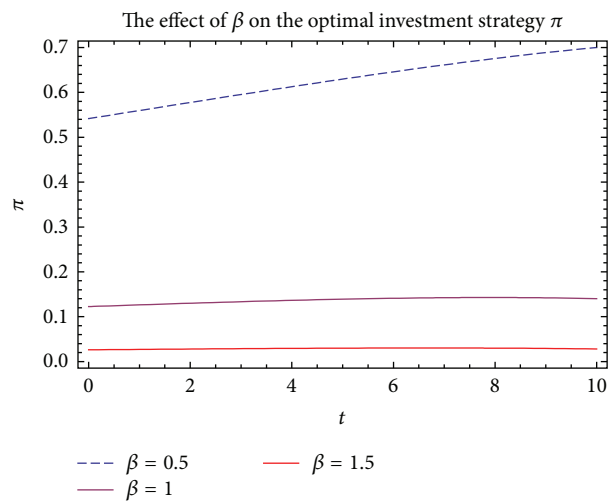
which means the value function decreases with respect to k . See Figure 4(b).



(a)

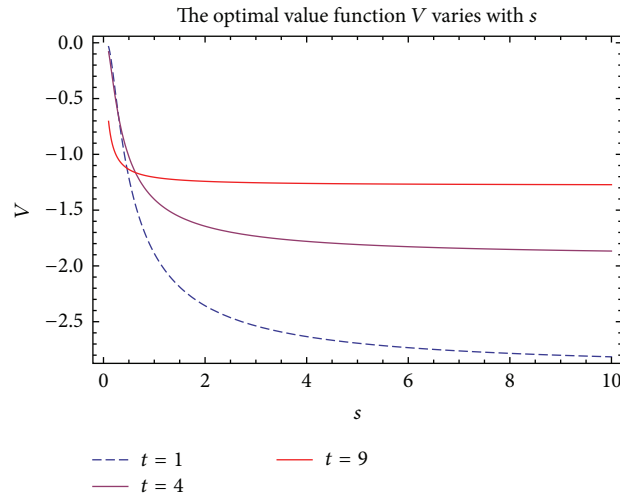


(b)

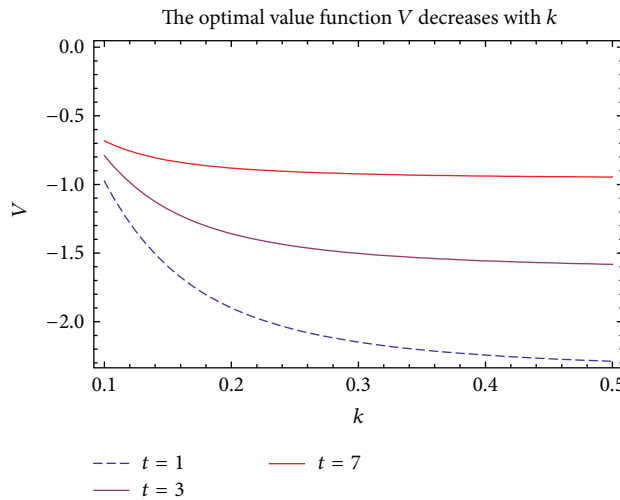


(c)

FIGURE 3: (a) The optimal investment strategy decreases with respect to k . (b) The optimal investment strategy π decreases with respect to s . (c) The optimal investment strategy π decreases with respect to β .



(a)



(b)

FIGURE 4: (a) The optimal value function V decreases with s . (b) The value function V decreases with k .

Completely similar to the analysis in (1), we also have

$$\begin{aligned} \frac{\partial V}{\partial s} &= -(2k^2qr)^{-1} (e^{2rT\beta} - e^{2rt\beta}) s^{-1-2\beta} (r - \mu)^2 \\ &\cdot \exp \left\{ -2rT\beta - \frac{1}{8}q \right. \\ &\quad \times \left[8e^{r(-t+T)} x - \frac{2(-1 + e^{2r(-t+T)}) q\sigma^2}{r} \right. \\ &\quad \left. - (k^2qr^2\beta)^{-1} e^{-2rT\beta} s^{-2\beta} (r - \mu)^2 \right. \\ &\quad \left. \cdot [2(e^{2rt\beta} - e^{2rT\beta})r + k^2s^{2\beta}(1 + 2\beta) \right. \\ &\quad \left. \left. \times (-e^{2rt\beta} + e^{2rT\beta}(1 + 2r(t - T)\beta)) \right] \right\} \\ &< 0, \end{aligned} \tag{39}$$

which shows that the value function decreases with respect to s . See Figure 4(a). \square

Let

$$\begin{aligned} D^\beta(t) &\triangleq k^2s^{2\beta} \\ &\quad \times (e^{2rT\beta}(1 + 4r(T - t)\beta^2) \\ &\quad \quad - e^{2rt\beta}(1 + 2r(T - t)\beta(1 + 2\beta))) \\ &\quad + 2e^{2rt\beta}r(1 + 2r(T - t)\beta + 2\beta \ln[s]) \\ &\quad - 2e^{2rT\beta}r(1 + 2\beta \ln[s]) \\ \frac{\partial V}{\partial \beta} &= (8k^2qr^2\beta^2)^{-1} s^{-2\beta} (r - \mu)^2 \\ &\quad \cdot \{k^2s^{2\beta} (e^{2rT\beta}(1 + 4r(T - t)\beta^2) - e^{2rt\beta} \\ &\quad \quad \times (1 + 2r(T - t)\beta(1 + 2\beta))) \\ &\quad + 2e^{2rt\beta}r(1 + 2r(T - t)\beta + 2\beta \ln[s]) \\ &\quad - 2e^{2rT\beta}r(1 + 2\beta \ln[s]) \} \end{aligned}$$

$$\begin{aligned} & \cdot \exp \left\{ -2rT\beta - \frac{1}{8}q \right. \\ & \times \left[8e^{r(-t+T)}x - \frac{2(-1 + e^{2r(-t+T)})q\sigma^2}{r} \right. \\ & \quad - (k^2qr^2\beta)^{-1}e^{-2rT\beta}s^{-2\beta}(r-\mu)^2 \\ & \quad \times [2(e^{2rt\beta} - e^{2rT\beta})r + k^2s^{2\beta}(1+2\beta) \\ & \quad \times (-e^{2rt\beta} + e^{2rT\beta} \\ & \quad \times (1+2r(t-T)\beta))] \left. \right] \left. \right\}. \end{aligned} \tag{40}$$

Obviously, $D^\beta(T) = 0$. Deriving D^β defined above with respect to t , we have

$$\begin{aligned} \frac{\partial D^\beta}{\partial t} = & 4r\beta^2 \left\{ -e^{2rT\beta}k^2s^{2\beta} + 2e^{2rt\beta}r \ln [s] \right. \\ & + e^{2rt\beta} (2r^2(-t+T) + k^2s^{2\beta} \\ & \left. \times (1+r(t-T)(1+2\beta))) \right\}. \end{aligned} \tag{41}$$

Remark 9. If t is big enough relative to T , then $\partial D^\beta/\partial t > 0$, which tells us that $D^\beta(t) \rightarrow 0$ increases as time t closes to T . So $D^\beta(t) < 0$ as $t \rightarrow T$. The sign of $\partial V/\partial\beta$ is the same as $D^\beta(t)$; that is, $\partial V/\partial\beta < 0$ when t is large enough relative to T , which means the value function decreases with respect to β when t is large enough. Else if t is small enough relative to T , the value function may increase with respect to β .

In conclusion, the volatility ks^β of risk asset has a significant negative influence on the expected utility of insurance company; through varying parameters k , s , and β some approximate rule can be found.

Proposition 10. $V(t, x, s)$ is the optimal value function under the optimal strategy (a^*, π^*) and μ is the expected return of risk asset price in CEV model. If $\beta \geq 0, s > 0, \mu > r > 0$, and $T > t$, then the optimal value function $V(t, x, s)$ increases with respect to the appreciation rate of risk asset μ .

Proof. Letting

$$\begin{aligned} D^\mu(t) \triangleq & [k^2s^{2\beta}(1+2\beta) - 2r] (e^{2rT\beta} - e^{2rt\beta}) \\ & - 2r(T-t)\beta k^2s^{2\beta}(1+2\beta)e^{2rT\beta}, \end{aligned} \tag{42}$$

obviously, $D^\mu(T) = 0$ and

$$\begin{aligned} \frac{\partial D^\mu}{\partial t} = & [k^2s^{2\beta}(1+2\beta) - 2r] (-2r\beta e^{2rt\beta}) \\ & + 2r\beta k^2s^{2\beta}(1+2\beta)e^{2rT\beta} \end{aligned}$$

$$\begin{aligned} = & -2r\beta e^{2rt\beta}k^2s^{2\beta}(1+2\beta) \\ & + 4r^2\beta e^{2rt\beta} + 2r\beta k^2s^{2\beta}(1+2\beta)e^{2rT\beta} \\ = & 2r\beta k^2s^{2\beta}(1+2\beta)(e^{2rT\beta} - e^{2rt\beta}) \\ & + 4r^2\beta e^{2rt\beta} > 0, \end{aligned} \tag{43}$$

which shows that $D^\mu(t)$ increases to $D^\mu(T) = 0$ with respect to time t . Thus

$$\begin{aligned} & [k^2s^{2\beta}(1+2\beta) - 2r] (e^{2rT\beta} - e^{2rt\beta}) \\ & - 2r(T-t)\beta k^2s^{2\beta}(1+2\beta)e^{2rT\beta} \leq 0. \end{aligned} \tag{44}$$

So

$$\begin{aligned} \frac{\partial V}{\partial \mu} = & -(4k^2qr^2\beta)^{-1}s^{-2\beta}(\mu-r) \\ & \cdot \{ [k^2s^{2\beta}(1+2\beta) - 2r] (e^{2rT\beta} - e^{2rt\beta}) \\ & \quad - 2r(T-t)\beta k^2s^{2\beta}(1+2\beta)e^{2rT\beta} \} \\ & \cdot \exp \left\{ -2rT\beta - \frac{1}{8}q \right. \\ & \times \left[8e^{r(-t+T)}x - \frac{2(-1 + e^{2r(-t+T)})q\sigma^2}{r} \right. \\ & \quad - (k^2qr^2\beta)^{-1}e^{-2rT\beta}s^{-2\beta}(r-\mu)^2 \\ & \quad \cdot [2(e^{2rt\beta} - e^{2rT\beta})r + k^2s^{2\beta}(1+2\beta) \\ & \quad \times (-e^{2rt\beta} + e^{2rT\beta}(1+2r(t-T)\beta))] \left. \right] \left. \right\} \\ \geq & 0. \end{aligned} \tag{45}$$

□

The value function V increases with μ referring to Figure 5, which shows that the higher appreciation rate of risk asset will lead to relatively more greater expected wealth utility. As a matter of fact, appreciation rate always is the main driving factor to make the wealth increase. Of course, the insurer's utility maximization can be better realized for a larger appreciation rate.

5. Conclusion

In this paper, we describe the dynamics of the risky assets' prices with the CEV model, under which the optimal excess-of-loss reinsurance and investment problem with jump-diffusion risk process is investigated by maximizing the insurer's exponential utility of terminal wealth. Applying

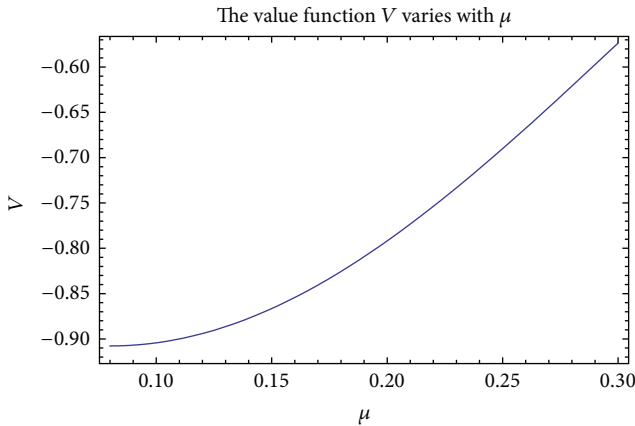


FIGURE 5: The value function V varies with μ .

stochastic control of jump diffusion, a Hamilton-Jacobi-Bellman (HJB) equation associated with a compound Poisson jump risk process is established. The explicit solution for the exponential utility function is given. In the last part of the text, some numerical examples are given to illustrate the effects of model parameters on the optimal strategy and the optimal value function. Meanwhile, we give some propositions and remarks which enriched the innovation of the paper.

Appendices

A. Proof of Theorem 1

This appendix collects the proofs of the main results stated in Section 3. Other necessary lemmas are also presented and proved in this appendix.

Proof. Differentiating (19) with respect to π gives the optimal investment policy

$$\pi^* = -\frac{(\mu - r)H_x^\alpha + k^2s^{2\beta+1}H_{xs}^\alpha}{xk^2s^{2\beta}H_{xx}^\alpha}. \tag{A.1}$$

Putting (A.1) into (19), after simplification, we have

$$\begin{aligned} & H_t^\alpha + [(\eta - \theta)\lambda\mu_\infty + rx]H_x^\alpha + \mu sH_s^\alpha \\ & + \frac{1}{2}\sigma^2H_{xx}^\alpha + \frac{1}{2}k^2s^{2\beta+2}H_{ss}^\alpha \\ & - \frac{[(\mu - r)H_x^\alpha + k^2s^{2\beta+1}H_{xs}^\alpha]^2}{2k^2s^{2\beta}H_{xx}^\alpha} \\ & + \max_a \left\{ H_x^\alpha (1 + \theta)\lambda \int_0^a \overline{F}(x)dx \right. \\ & \left. + \lambda E [H^\alpha(t, x - (Z_i \wedge a), s) - H^\alpha(t, x, s)] \right\} \\ & = 0. \end{aligned} \tag{A.2}$$

That is,

$$\begin{aligned} & H_t^\alpha + [(\eta - \theta)\lambda\mu_\infty + rx]H_x^\alpha + \mu sH_s^\alpha + \frac{1}{2}\sigma^2H_{xx}^\alpha \\ & + \frac{1}{2}k^2s^{2\beta+2}H_{ss}^\alpha - \frac{(\mu - r)^2(H_x^\alpha)^2}{2k^2s^{2\beta}H_{xx}^\alpha} \\ & - \frac{k^2s^{2\beta+2}(H_{xs}^\alpha)^2}{2H_{xx}^\alpha} - \frac{(\mu - r)sH_x^\alpha H_{xs}^\alpha}{H_{xx}^\alpha} \\ & + \max_a \left\{ H_x^\alpha (1 + \theta)\lambda \int_0^a \overline{F}(x)dx \right. \\ & \left. + \lambda E [H^\alpha(t, x - (Z_i \wedge a), s) - H^\alpha(t, x, s)] \right\} \\ & = 0. \end{aligned} \tag{A.3}$$

We try to conjecture (A.2) having a solution $V(t, x, s)$ of this form

$$\begin{aligned} & V(t, x, s) = -\frac{1}{q}e^{\{-q[xe^{r(T-t)} + f(t) + g(t,s)]\}}, \quad t \in [0, T] \\ & f(T) = 0, \\ & g(T, s) = 0, \\ & V(T, x, s) = -\frac{1}{q}e^{-qx}. \end{aligned} \tag{A.4}$$

Differentiating the solution (A.4) with respect to the corresponding variables,

$$\begin{aligned} & V_t = V(-q) [-rx e^{r(T-t)} + f'(t) + g_t], \\ & V_x = V(-q) e^{r(T-t)}, \quad V_{xx} = V(-q)^2 e^{2r(T-t)}, \\ & V_s = V(-q) g_s, \quad V_{ss} = V [(-q)^2 g_s^2 + (-q) g_{ss}], \\ & V_{xs} = V(-q)^2 e^{r(T-t)} g_s, \\ & E [H^\alpha(t, x - (Z_i \wedge a), s) - H^\alpha(t, x, s)] \\ & = Vq e^{r(T-t)} \int_0^a e^{qx e^{r(T-t)}} \overline{F}(x) dx. \end{aligned} \tag{A.5}$$

Plugging the above derivatives (A.5) into (A.1) gives

$$\pi^* = \frac{\mu - r}{xqk^2s^{2\beta}e^{r(T-t)}} - \frac{sg_s}{xe^{r(T-t)}} \tag{A.6}$$

and substituting the above derivatives into (A.2) yields

$$\begin{aligned}
 f'(t) + g_t + (\eta - \theta) \lambda \mu_{\infty} e^{r(T-t)} + r s g_s \\
 - \frac{1}{2} \sigma^2 q e^{2r(T-t)} + \frac{1}{2} k^2 s^{2\beta+2} g_{ss} + \frac{(\mu - r)^2}{2qk^2} s^{-2\beta} \\
 + \max_a \left\{ (1 + \theta) \lambda e^{r(T-t)} \int_0^a \overline{F(x)} dx \right. \\
 \left. - \lambda e^{r(T-t)} \int_0^a e^{qx} e^{r(T-t)} \overline{F(x)} dx \right\} = 0.
 \end{aligned} \tag{A.7}$$

Differentiating with respect to a in (A.7), we find that the minimizer $a_*^0(t)$ satisfies

$$\left[(1 + \theta) - e^{qa^0(t)e^{r(T-t)}} \right] \overline{F(a^0(t))} = 0. \tag{A.8}$$

If $a^0(t) \in [0, D)$, then $0 < \overline{F(a^0(t))} \leq 1$ and $(1 + \theta) - e^{qa^0(t)e^{r(T-t)}} = 0$, which gives

$$a_*^0(t) = \frac{\ln(1 + \theta)}{q} e^{-r(T-t)}. \tag{A.9}$$

To ensure that $0 \leq a_*^0(t) < D$, that is, $0 \leq (\ln(1 + \theta)/q) e^{-r(T-t)} < D$, it must be ensured that

$$t < T + \frac{1}{r} \ln \left[\frac{qD}{\ln(1 + \theta)} \right] \triangleq t^0, \tag{A.10}$$

which leads to two different cases.

Case 1. If $D \geq (\ln(1 + \theta)/q)$, then $t^0 \geq T$ and

$$a_*^0(t) = \frac{\ln(1 + \theta)}{q} e^{-r(T-t)} \tag{A.11}$$

always is optimal retention level on the whole interval $[0, T]$.

Inserting (A.5) into (A.1), the optimal investment policy is given by

$$\pi^*(t) = \frac{\mu - r}{xqk^2 s^{2\beta} e^{r(T-t)}} - \frac{sg_s}{x e^{r(T-t)}} \tag{A.12}$$

and $a^* = a^0(t) = (\ln(1 + \theta)/q) e^{-r(T-t)}$. The corresponding HJB equation (A.7) becomes

$$\begin{aligned}
 f'(t) + g_t + r s g_s + (\eta - \theta) \lambda \mu_{\infty} e^{r(T-t)} \\
 - \frac{1}{2} \sigma^2 q e^{2r(T-t)} + \frac{1}{2} k^2 s^{2\beta+2} g_{ss} \\
 + \frac{(\mu - r)^2}{2qk^2 s^{2\beta}} + e^{r(T-t)} (1 + \theta) \lambda \\
 \times \int_0^{(\ln(1+\theta)/q)e^{-r(T-t)}} \overline{F(x)} dx - \lambda e^{r(T-t)} \\
 \times \int_0^{(\ln(1+\theta)/q)e^{-r(T-t)}} e^{qx} e^{r(T-t)} \overline{F(x)} dx = 0.
 \end{aligned} \tag{A.13}$$

To solve (A.13), we decompose (A.13) into two equations:

$$\begin{aligned}
 f'(t) + (\eta - \theta) \lambda \mu_{\infty} e^{r(T-t)} - \frac{1}{2} \sigma^2 q e^{2r(T-t)} \\
 + e^{r(T-t)} (1 + \theta) \lambda \int_0^{(\ln(1+\theta)/q)e^{-r(T-t)}} \overline{F(x)} dx \\
 - \lambda e^{r(T-t)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-t)}} e^{qx} e^{r(T-t)} \overline{F(x)} dx \\
 = 0,
 \end{aligned} \tag{A.14}$$

$$g_t + r s g_s + \frac{1}{2} k^2 s^{2\beta+2} g_{ss} + \frac{(\mu - r)^2}{2qk^2 s^{2\beta}} = 0. \tag{A.15}$$

By simple transposition and integration for (A.14) with $f(T) = 0$, we have

$$\begin{aligned}
 f(t) = \frac{(\eta - \theta) \lambda \mu_{\infty} e^{r(T-t)}}{r} - \frac{\sigma^2 q e^{2r(T-t)}}{4r} \\
 + \int_t^T \left(e^{r(T-y)} (1 + \theta) \lambda \int_0^{(\ln(1+\theta)/q)e^{-r(T-y)}} \overline{F(x)} dx \right) dy \\
 - \int_t^T \left(\lambda e^{r(T-y)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-y)}} e^{qx} e^{r(T-y)} \overline{F(x)} dx \right) dy \\
 - \frac{(\eta - \theta) \lambda \mu_{\infty}}{r} + \frac{\sigma^2 q}{4r}.
 \end{aligned} \tag{A.16}$$

To solve (A.20), we conjecture a solution with the following form with $A(T) = 0$ and $B(T) = 0$:

$$g(t, s) = A(t) s^{-2\beta} + B(t). \tag{A.17}$$

Differentiating (A.17) with respect to the corresponding variables, respectively,

$$\begin{aligned}
 g_t = A'(t) s^{-2\beta} + B'(t), \quad g_s = -2\beta s^{-2\beta-1} A(t), \\
 g_{ss} = 2\beta(2\beta + 1) s^{-2\beta-2} A(t).
 \end{aligned} \tag{A.18}$$

Substituting the above derivatives into (A.20) yields

$$\begin{aligned}
 A'(t) s^{-2\beta} + B'(t) - 2r\beta s^{-2\beta} A(t) \\
 + k^2 \beta(2\beta + 1) A(t) + \frac{(\mu - r)^2}{2qk^2} s^{-2\beta} = 0.
 \end{aligned} \tag{A.19}$$

To solve (A.20), we decompose (A.20) into two equations:

$$\left[A'(t) - 2r\beta A(t) + \frac{(\mu - r)^2}{2qk^2} \right] s^{-2\beta} = 0, \tag{A.20}$$

$$B'(t) + k^2 \beta(2\beta + 1) A(t) = 0. \tag{A.21}$$

Obviously, $s^{-2\beta} > 0$, (A.20) is equivalent to

$$A'(t) - 2r\beta A(t) + \frac{(\mu - r)^2}{2qk^2} = 0. \tag{A.22}$$

Equation (A.22) with $A(T) = 0$ is easy to be solved and the solution is

$$A(t) = \frac{e^{-2r\beta T} (-e^{-2r\beta t} + e^{2r\beta T}) (\mu - r)^2}{4qr\beta k^2}. \tag{A.23}$$

Inserting (A.23) into (A.21) with $B(T) = 0$, we have

$$B(t) = \left(e^{-2r\beta T} (1 + 2\beta) \right. \\ \left. \times (e^{2r\beta t} + e^{2r\beta T} (-1 + 2r(T-t)\beta)) \right. \\ \left. \times (\mu - r)^2 \right) \times (8q\beta r^2)^{-1}. \tag{A.24}$$

Substituting (A.23) and (A.24) back into (A.17), we obtain

$$g(t, s) = \frac{e^{-2r\beta T} (-e^{-2r\beta t} + e^{2r\beta T}) (\mu - r)^2}{4qr\beta k^2} s^{-2\beta} \\ + \left(e^{-2r\beta T} (1 + 2\beta) \right. \\ \left. \times (e^{2r\beta t} + e^{2r\beta T} (-1 + 2r(T-t)\beta)) \right. \\ \left. \times (\mu - r)^2 \right) \times (8q\beta r^2)^{-1}. \tag{A.25}$$

Substituting (A.16) and (A.25) back into (A.4), the optimal value function is given by

$$V(t, x, s) \\ = -\frac{1}{q} \exp \\ \times \left\{ -q \right. \\ \left. \times \left[xe^{r(T-t)} + \frac{(\eta - \theta) \lambda \mu_\infty e^{r(T-t)}}{r} - \frac{\sigma^2 q e^{2r(T-t)}}{4r} \right. \right. \\ \left. \left. + \int_t^T \left(e^{r(T-y)} (1 + \theta) \lambda \right. \right. \right. \\ \left. \left. \times \int_0^{(\ln(1+\theta)/q)e^{-r(T-y)}} \overline{F(x)} dx \right) dy \right. \\ \left. - \int_t^T \left(\lambda e^{r(T-y)} \right. \right. \\ \left. \left. \times \int_0^{(\ln(1+\theta)/q)e^{-r(T-y)}} e^{qx} e^{r(T-y)} \overline{F(x)} dx \right) dy \right. \\ \left. - \frac{(\eta - \theta) \lambda \mu_\infty}{r} + \frac{\sigma^2 q}{4r} \right\}$$

$$+ \left(e^{-2r\beta T} (1 + 2\beta) \right. \\ \left. \times (e^{2r\beta t} + e^{2r\beta T} (-1 + 2r(T-t)\beta)) \right. \\ \left. \times (\mu - r)^2 \right) \times (8q\beta r^2)^{-1} \\ \left. + \frac{e^{-2r\beta T} (-e^{-2r\beta t} + e^{2r\beta T}) (\mu - r)^2}{4qr\beta k^2} s^{-2\beta} \right\}, \\ t \in [0, T]. \tag{A.26}$$

Case 2. If $D < \ln(1 + \theta)/q$, then $t^0 < T$; the whole interval $[0, T]$ is divided into two parts, $[0, t^0]$ and $[t^0, T]$. $a^0(t) = (\ln(1 + \theta)/q)e^{-r(T-t)}$ is the optimal retention level only on $[0, t^0]$; however, $a^0(t) \geq D$ on $[t^0, T]$; we can only take D as the retention level naturally. Therefore, the optimal retention level is denoted by

$$a^*(t) = \begin{cases} \frac{\ln(1 + \theta)}{q} e^{-r(T-t)}, & t \in [0, t^0], \\ D, & t \in [t^0, T]. \end{cases} \tag{A.27}$$

The solving process depends on two different time intervals corresponding to the different optimal retention level, respectively.

(1) For

$$a^*(t) = D, \quad t \in [t^0, T], \tag{A.28}$$

the corresponding HJB equation is

$$H_t^\alpha + [(1 + \eta) \lambda \mu_\infty + rx] H_x^\alpha + \mu s H_s^\alpha \\ + \frac{1}{2} \sigma^2 H_{xx}^\alpha + \frac{1}{2} k^2 s^{2\beta+2} H_{ss}^\alpha \\ + \max_\pi \left\{ (\mu - r) \pi x H_x^\alpha + \frac{1}{2} \pi^2 x^2 k^2 s^{2\beta} H_{xx}^\alpha \right. \\ \left. + k^2 \pi x s^{2\beta+1} H_{xs}^\alpha \right\} \\ + \lambda E [H^\alpha(t, x - (Z_i \wedge D), s) - H^\alpha(t, x, s)] = 0. \tag{A.29}$$

The first-order maximizing condition for the optimal investment strategy is

$$\pi^*(t) = -\frac{(\mu - r) H_x^\alpha + k^2 s^{2\beta+1} H_{xs}^\alpha}{x k^2 s^{2\beta} H_{xx}^\alpha}. \tag{A.30}$$

Inserting (A.30) into (A.29) and simplifying, the HJB equation becomes

$$\begin{aligned}
 &H_t^\alpha + [(1 + \eta) \lambda \mu_\infty + rx] H_x^\alpha + \mu s H_s^\alpha \\
 &+ \frac{1}{2} \sigma^2 H_{xx}^\alpha + \frac{1}{2} k^2 s^{2\beta+2} H_{ss}^\alpha \\
 &- \frac{[(\mu - r) H_x^\alpha + k^2 s^{2\beta+1} H_{xs}^\alpha]^2}{2k^2 s^{2\beta} H_{xx}^\alpha} \\
 &+ \lambda E [H^\alpha(t, x - (Z_i \wedge D), s) - H^\alpha(t, x, s)] = 0.
 \end{aligned} \tag{A.31}$$

We conjecture (A.31) having a solution $V(t, x, s)$ as follows:

$$\begin{aligned}
 V(t, x, s) &= -\frac{1}{q} \exp \left\{ -q \left[x e^{r(T-t)} + \widehat{f}(t) + \widehat{g}(t, s) \right] \right\}, \\
 & t \in [t^0, T], \\
 V(T, x, s) &= U(x), \\
 \widehat{f}(T) &= 0, \\
 \widehat{g}(T, s) &= 0.
 \end{aligned} \tag{A.32}$$

Differentiating the conjecture (A.32) with respect to the corresponding variables,

$$\begin{aligned}
 V_t &= V(-q) \left[-rx e^{r(T-t)} + \widehat{f}'(t) + \widehat{g}_t \right], \\
 V_x &= V(-q) e^{r(T-t)}, \quad V_{xx} = V(-q)^2 e^{2r(T-t)}, \\
 V_s &= V(-q) \widehat{g}_s, \quad V_{ss} = V \left[(-q)^2 \widehat{g}_s^2 + (-q) \widehat{g}_{ss} \right], \\
 V_{xs} &= V(-q)^2 e^{r(T-t)} \widehat{g}_s, \\
 E [H^\alpha(t, x - (Z_i \wedge D), s) - H^\alpha(t, x, s)] \\
 &= V \int_0^D e^{qx e^{r(T-t)}} dF(x) - V.
 \end{aligned} \tag{A.33}$$

Substituting the above derivatives (A.33) into (A.31) and (A.30), after simplifying it gives

$$\begin{aligned}
 \widehat{\pi}^*(t) &= \frac{(\mu - r)}{qk^2 xs^{2\beta} e^{r(T-t)}} - \frac{s \widehat{g}_s}{x e^{r(T-t)}}, \\
 \widehat{f}'(t) + \widehat{g}_t + (1 + \eta) \lambda \mu_\infty e^{r(T-t)} - \frac{1}{2} \sigma^2 q e^{2r(T-t)} \\
 + rs \widehat{g}_s + \frac{1}{2} k^2 s^{2\beta+2} \widehat{g}_{ss} + \frac{(\mu - r)^2}{2qk^2} s^{-2\beta} \\
 - \frac{\lambda}{q} \int_0^D e^{qx e^{r(T-t)}} dF(x) + \frac{\lambda}{q} &= 0.
 \end{aligned} \tag{A.34}$$

$$\begin{aligned}
 + rs \widehat{g}_s + \frac{1}{2} k^2 s^{2\beta+2} \widehat{g}_{ss} + \frac{(\mu - r)^2}{2qk^2} s^{-2\beta} \\
 - \frac{\lambda}{q} \int_0^D e^{qx e^{r(T-t)}} dF(x) + \frac{\lambda}{q} &= 0.
 \end{aligned} \tag{A.35}$$

Decompose (A.35) into two equations

$$\begin{aligned}
 \widehat{f}'(t) + (1 + \eta) \lambda \mu_\infty e^{r(T-t)} - \frac{1}{2} \sigma^2 q e^{2r(T-t)} \\
 - \frac{\lambda}{q} \int_0^D e^{qx e^{r(T-t)}} dF(x) + \frac{\lambda}{q} &= 0,
 \end{aligned} \tag{A.36}$$

$$\widehat{g}_t + rs \widehat{g}_s + \frac{1}{2} k^2 s^{2\beta+2} \widehat{g}_{ss} + \frac{(\mu - r)^2}{2qk^2} s^{-2\beta} = 0. \tag{A.37}$$

By transposition and integration for (A.36) with $\widehat{f}(T) = 0$,

$$\begin{aligned}
 \widehat{f}(t) &= \frac{(1 + \eta) \lambda \mu_\infty e^{r(T-t)} - \frac{1}{2} \sigma^2 q e^{2r(T-t)}}{r} \\
 &- \int_t^T \left[\frac{\lambda}{q} \int_0^D e^{qx e^{r(T-\tau)}} dF(x) \right] d\tau \\
 &- \frac{\lambda(t - T)}{q} - \frac{(1 + \eta) \lambda \mu_\infty}{r} + \frac{\sigma^2 q}{4r}.
 \end{aligned} \tag{A.38}$$

For (A.37), we try to find a solution as follows:

$$\widehat{g}(t, s) = \widehat{A}(t) s^{-2\beta} + \widehat{B}(t) \tag{A.39}$$

satisfying $\widehat{A}(T) = 0$ and $\widehat{B}(T) = 0$. Equation (A.37) turns into

$$\begin{aligned}
 \widehat{A}'(t) s^{-2\beta} + \widehat{B}'(t) - 2r\beta \widehat{A}(t) s^{-2\beta} + k^2 \beta (2\beta + 1) \widehat{A}(t) \\
 + \frac{(\mu - r)^2}{2qk^2} s^{-2\beta} &= 0.
 \end{aligned} \tag{A.40}$$

Decompose (A.40) into two equations

$$\begin{aligned}
 \left[\widehat{A}'(t) - 2r\beta \widehat{A}(t) + \frac{(\mu - r)^2}{2qk^2} \right] s^{-2\beta} &= 0, \\
 \widehat{B}'(t) + k^2 \beta (2\beta + 1) \widehat{A}(t) &= 0.
 \end{aligned} \tag{A.41}$$

Solving the above simple ordinary differential equations, we get

$$\begin{aligned}
 \widehat{A}(t) &= \frac{(1 - e^{2r\beta(t-T)}) (\mu - r)^2}{4qk^2 r \beta}, \\
 \widehat{B}(t) &= \frac{(1 + 2\beta) [e^{2r\beta(t-T)} - 1 + 2r(T-t)\beta] (\mu - r)^2}{8qr^2 \beta}.
 \end{aligned} \tag{A.42}$$

So

$$\begin{aligned}
 \widehat{g}(t, s) &= \frac{(1 - e^{2r\beta(t-T)}) (\mu - r)^2}{4qk^2 r \beta} s^{-2\beta} \\
 &+ \frac{(1 + 2\beta) [e^{2r\beta(t-T)} - 1 + 2r(T-t)\beta] (\mu - r)^2}{8qr^2 \beta}, \\
 \widehat{\pi}^*(t) &= \frac{2r(\mu - r) + (\mu - r)^2 (1 - e^{2r\beta(t-T)})}{2qk^2 r x s^{2\beta} e^{r(T-t)}}.
 \end{aligned} \tag{A.43}$$

(2) For

$$a^*(t) = \frac{\ln(1 + \theta)}{q} e^{-r(T-t)}, \quad t \in [0, t^0], \quad (\text{A.44})$$

the corresponding HJB equation is the same as (A.29), which corresponds to $[0, t^0]$ having a solution $V(t, x, s)$ as follows:

$$V(t, x, s) = -\frac{1}{q} \exp \left\{ -q \left[x e^{r(T-t)} + \overline{f(t)} + \overline{g(t, s)} \right] \right\}, \quad t \in [0, t^0],$$

$$\overline{f(t^0)} = \overline{f(t^0)},$$

$$\overline{g(t^0, s)} = \overline{g(t^0, s)}. \quad (\text{A.45})$$

After the same calculation as (1), we obtain the following results:

$$\begin{aligned} \overline{f'(t)} + \overline{g_t} + (\eta - \theta) \lambda \mu_\infty e^{r(T-t)} - \frac{1}{2} \sigma^2 q e^{2r(T-t)} \\ + \frac{1}{2} k^2 s^{2\beta+2} \overline{g_{ss}} + r s \overline{g_s} + \frac{(\mu - r)^2}{2qk^2} s^{-2\beta} \\ + (1 + \theta) \lambda e^{r(T-t)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-t)}} \overline{F(x)} dx \\ - \lambda e^{r(T-t)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-t)}} e^{qx e^{r(T-t)}} \overline{F(x)} dx = 0. \end{aligned} \quad (\text{A.46})$$

Splitting (A.46) into two equations,

$$\overline{g_t} + \frac{1}{2} k^2 s^{2\beta+2} \overline{g_{ss}} + r s \overline{g_s} + \frac{(\mu - r)^2}{2qk^2} s^{-2\beta} = 0. \quad (\text{A.47})$$

The equation (A.47) is a linear second-order partial differential equation with variable coefficients, which exhibits a unique solution according to the appendix of Badaoui and Fernandez [19] or is ensured by the theorem of Friedman [20]:

$$\begin{aligned} \overline{f'(t)} + (\eta - \theta) \lambda \mu_\infty e^{r(T-t)} - \frac{1}{2} \sigma^2 q e^{2r(T-t)} \\ + (1 + \theta) \lambda e^{r(T-t)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-t)}} \overline{F(x)} dx \\ - \lambda e^{r(T-t)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-t)}} e^{qx e^{r(T-t)}} \overline{F(x)} dx = 0. \end{aligned} \quad (\text{A.48})$$

By transposition and integration to (A.48), the result is given by

$$\begin{aligned} \overline{f(t)} = \frac{(\eta - \theta) \lambda \mu_\infty}{r} e^{r(T-t)} - \frac{\sigma^2 q}{4r} e^{2r(T-t)} \\ - \int_0^t \left[(1 + \theta) \lambda e^{r(T-\tau)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-\tau)}} \overline{F(x)} dx \right] d\tau \\ + \lambda \int_0^t \left[e^{r(T-\tau)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-\tau)}} e^{qx e^{r(T-\tau)}} \overline{F(x)} dx \right] d\tau \\ + C_3, \end{aligned} \quad (\text{A.49})$$

where w can be determined by $\overline{f(t^0)} = \overline{f(t^0)}$. Consider

$$\begin{aligned} C_3 = -\frac{(\eta - \theta) \lambda \mu_\infty}{r} e^{r(T-t^0)} + (1 + \theta) \lambda \\ \times \int_0^{t^0} \left[e^{r(T-\tau)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-\tau)}} \overline{F(x)} dx \right] d\tau \\ - \lambda \int_0^{t^0} \left[e^{r(T-\tau)} \int_0^{(\ln(1+\theta)/q)e^{-r(T-\tau)}} e^{qx e^{r(T-\tau)}} \overline{F(x)} dx \right] d\tau \\ + \frac{(1 + \eta) \lambda \mu_\infty}{r} e^{r(T-t^0)} \\ - \frac{\lambda}{q} \int_{t^0}^T \left[\int_0^D e^{qx e^{r(T-\tau)}} dF(x) \right] d\tau \\ - \frac{\lambda (t^0 - T)}{q} - \frac{(1 + \eta) \lambda \mu_\infty}{r} + \frac{\sigma^2 q}{4r}. \end{aligned} \quad (\text{A.50})$$

So

$$f^*(t) = \begin{cases} \overline{f(t)}, & t \in [0, t^0], \\ \overline{f(t)}, & t \in [t^0, T]. \end{cases} \quad (\text{A.51})$$

With the same calculations as before, we obtain

$$\begin{aligned} \overline{g(t, s)} = \left[C_1 e^{2\beta r t} + \frac{(\mu - r)^2}{4qk^2 \beta r} \right] s^{-2\beta} - C_1 \frac{k^2 (2\beta + 1)}{2r} e^{2\beta r t} \\ - \frac{(2\beta + 1)}{qr} (\mu - r)^2 + C_2; \end{aligned} \quad (\text{A.52})$$

here, the undetermined constants C_1 and C_2 will be determined by using the continuity of $V(t, x, s)$.

By the continuity of $V(t, x, s)$ at $t = t^0$, we have

$$\overline{g(t^0, s)} = \overline{g(t^0, s)}. \quad (\text{A.53})$$

Comparing the items and the coefficients on both sides of (A.53), we obtain the undetermined constants C_1 and C_2 :

$$C_1 = -\frac{(\mu - r)^2 e^{-2rT\beta}}{4qrk^2\beta}, \tag{A.54}$$

$$C_2 = (2\beta + 1)(\mu - r)^2 \left(\frac{4 + T - t^0}{4qr} - \frac{1}{8qr^2\beta} \right).$$

So

$$g^*(t, s) = \begin{cases} \overline{g(t, s)}, & t \in [0, t^0), \\ \underline{g(t, s)}, & t \in [t^0, T]. \end{cases} \tag{A.55}$$

And the optimal value function is

$$V(t, x, s) = -\frac{1}{q} \exp \left\{ -q \left[x e^{r(T-t)} + f^*(t) + g^*(t, s) \right] \right\}. \tag{A.56}$$

□

B. Proof of Theorem 3

A new lemma is necessary to prove Theorem 3. For convenience, denote $Q := [0, +\infty) \times [0, +\infty)$.

Lemma B.1. *Take a sequence of bounded open sets Q_1, Q_2, Q_3, \dots , with $Q_i \subset Q_{i+1} \subset Q$, $i = 1, 2, \dots$ and $Q = \cup_i Q_i$. Let τ_i denote the exit time of $(X^{\alpha^*}(t), S(t))$ from Q_i . If the conditions in Theorem 3 hold, then one has $E[V^2(\tau_i \wedge T, X^{\alpha^*}(\tau_i \wedge T), S(\tau_i \wedge T)) \mid X(t) = x, S(t) = s] < \infty$ for $i = 1, 2, \dots$*

Proof. According to Øksendal and Sulem [17], we rewrite the compound Poisson process $C(t) = \sum_{i=1}^{N(t)} Z_i$ in terms of a Poisson random measure. Suppose γ is the Poisson random measure; then

$$\sum_{i=1}^{N(t)} Z_i = \int_0^t \int_{R^+} z \gamma(dz, du), \tag{B.1}$$

$$C^a(t) = \sum_{i=1}^{N(t)} Z_i = \int_0^t \int_{R^+} (z \wedge a) \gamma(dz, du).$$

The compensator ν of the random measure γ is

$$\nu(dz, dt) = \lambda dF(z) dt, \tag{B.2}$$

so the compensated Poisson random measure of $C(t)$ is

$$\tilde{\gamma}(dz, dt) = \gamma(dz, dt) - \lambda dF(z) dt. \tag{B.3}$$

Set $\tilde{V}(t) = V(t, X^{\alpha^*}(t), S(t))$. Applying Itô's formula to $\tilde{V}^2(t)$,

$$\begin{aligned} d\tilde{V}^2(t) &= 2\tilde{V}(t) \left\{ V_x \sigma dW(t) + kS^\beta(t) \right. \\ &\quad \times [V_x \pi(t) X(t) + V_s S(t)] dW_1(t) + \mathcal{A}\tilde{V}(t) \left. \right\} \\ &\quad + [V_x^2 \sigma^2 + V_x^2 \pi^2 X(t)^2 k^2 S^{2\beta}(t) + V_s^2 k^2 S^{2\beta+2}(t) \\ &\quad + 2V_x V_s \pi(t) X(t) k^2 S^{2\beta+1}(t)] dt \\ &\quad + \int_{R^+} [V^2(t, X(t-), S(t)) \\ &\quad - V^2(t, X(t-), S(t)) - 2V(t, X(t-), S(t)) \\ &\quad \cdot V_x \cdot (z \wedge a)] \lambda dF(z) dt \\ &\quad + \int_{R^+} [V^2(t, X(t-), S(t)) \\ &\quad - V^2(t, X(t-), S(t))] \tilde{\gamma}(dz, dt). \end{aligned} \tag{B.4}$$

Since α^* is the optimal strategy for (19), then $\mathcal{A}^{\alpha^*} \tilde{V}(t) = 0$. Plugging the expression of V_x, V_s , and $a^*(t)$ into (B.4), we obtain the following equation by simple calculations:

$$\begin{aligned} \frac{d\tilde{V}^2(t)}{\tilde{V}^2(t)} &= \left\{ -2q\sigma e^{r(T-t)} dW(t) - \frac{2(\mu - r) s^{-\beta}}{k} \right. \\ &\quad \times \left[1 + \frac{(1 - e^{-2r\beta(T-t)})(\mu - r)}{2r} \right] dW_1(t) \\ &\quad \left. + \frac{(1 - e^{-2r\beta(T-t)})(\mu - r)^2 s^{-\beta}}{rk} dW_1(t) \right\} \\ &\quad + \left\{ q^2 \sigma^2 e^{2r(T-t)} \right. \\ &\quad + \frac{(\mu - r)^2 s^{-2\beta}}{k^2} \left[1 + \frac{(1 - e^{-2r\beta(T-t)})(\mu - r)}{2r} \right]^2 \\ &\quad + \frac{(1 - e^{-2r\beta(T-t)})^2 (\mu - r)^4 s^{-2\beta}}{4r^2 k^2} \\ &\quad - \frac{(1 - e^{-2r\beta(T-t)})(\mu - r)^3 s^{-2\beta}}{rk^2} \\ &\quad \left. \times \left[1 + \frac{(1 - e^{-2r\beta(T-t)})(\mu - r)}{2r} \right] \right\} dt \end{aligned}$$

$$\begin{aligned}
 & + \int_{R^+} \left[e^{-2q(z \wedge a^*)e^{r(T-t)}} \right. \\
 & \quad \left. + 2q(z \wedge a^*)e^{r(T-t)} - 1 \right] \lambda dF(z) dt \\
 & + \int_{R^+} \left[e^{-2q(z \wedge a^*)e^{r(T-t)}} - 1 \right] \tilde{\gamma}(dz, dt).
 \end{aligned} \tag{B.5}$$

The solution of (B.5) is

$$\begin{aligned}
 & \frac{\tilde{V}^2(t)}{\tilde{V}^2(0)} \\
 & = \exp \left\{ \int_0^t -2q\sigma e^{r(T-u)} dW(u) \right. \\
 & \quad - \frac{1}{2} \int_0^t 4q^2 \sigma^2 e^{2r(T-u)} du \\
 & \quad + \int_0^t \frac{2(\mu-r)s^{-\beta}}{k} \\
 & \quad \quad \times \left[1 + \frac{(1-e^{-2r\beta(T-u)})(\mu-r)}{2r} \right] dW_1(u) \\
 & \quad - \frac{1}{2} \int_0^t \frac{4(\mu-r)^2 s^{-2\beta}}{k^2} \\
 & \quad \quad \times \left[1 + \frac{(1-e^{-2r\beta(T-u)})(\mu-r)}{2r} \right]^2 du \\
 & \quad + \int_0^t \frac{(1-e^{-2r\beta(T-u)})(\mu-r)^2 s^{-\beta}}{rk} dW_1(u) \\
 & \quad - \frac{1}{2} \int_0^t \frac{(1-e^{-2r\beta(T-u)})^2 (\mu-r)^4 s^{-2\beta}}{r^2 k^2} du \\
 & \quad + \int_0^t 3q^2 \sigma^2 e^{2r(T-u)} du \\
 & \quad + \int_0^t \frac{3(1-e^{-2r\beta(T-u)})^2 (\mu-r)^4 s^{-2\beta}}{4r^2 k^2} du \\
 & \quad + \int_0^t \frac{3(\mu-r)^2 s^{-2\beta}}{k^2} \\
 & \quad \quad \times \left[1 + \frac{(1-e^{-2r\beta(T-u)})(\mu-r)}{2r} \right]^2 du \\
 & \quad + \int_0^t \frac{(1-e^{-2r\beta(T-u)})(\mu-r)^3 s^{-2\beta}}{rk^2} \\
 & \quad \quad \times \left[1 + \frac{(1-e^{-2r\beta(T-u)})(\mu-r)}{2r} \right] du
 \end{aligned}$$

$$\begin{aligned}
 & + \int_0^t \int_{R^+} \left[e^{-2q(z \wedge a^*)e^{r(T-u)}} \right. \\
 & \quad \left. + 2q(z \wedge a^*)e^{r(T-u)} - 1 \right] \lambda dF(z) du \\
 & + \int_0^t \int_{R^+} \left[e^{-2q(z \wedge a^*)e^{r(T-u)}} - 1 \right] \tilde{\gamma}(dz, du) \Big\},
 \end{aligned} \tag{B.6}$$

where (x_0, s_0) is the state at time 0. Applying Itô's formula to (11), we can derive

$$\begin{aligned}
 dS^{-2\beta}(t) & = (\beta(2\beta+1)k^2 - 2\beta r S^{-2\beta}(t)) dt \\
 & \quad - 2\beta k \sqrt{S^{-2\beta}(t)} dW_1(t).
 \end{aligned} \tag{B.7}$$

According to the results of Taksar and Zeng [16] and Gu et al. [5] and other existing literature, $\mu > r > 0$, we know that

$$\begin{aligned}
 & \exp \left\{ \int_0^t -2q\sigma e^{r(T-u)} dW(u) - \frac{1}{2} \int_0^t 4q^2 \sigma^2 e^{2r(T-u)} du \right\} \\
 & \exp \left\{ \int_0^t \frac{2(\mu-r)s^{-\beta}}{k} \right. \\
 & \quad \times \left[1 + \frac{(1-e^{-2r\beta(T-u)})(\mu-r)}{2r} \right] dW_1(u) \\
 & \quad - \frac{1}{2} \int_0^t \frac{4(\mu-r)^2 s^{-2\beta}}{k^2} \\
 & \quad \quad \times \left[1 + \frac{(1-e^{-2r\beta(T-u)})(\mu-r)}{2r} \right]^2 du \Big\}, \\
 & \exp \left\{ \int_0^t \frac{(1-e^{-2r\beta(T-u)})(\mu-r)^2 s^{-\beta}}{rk} dW_1(u) \right. \\
 & \quad - \frac{1}{2} \int_0^t \frac{(1-e^{-2r\beta(T-u)})^2 (\mu-r)^4 s^{-2\beta}}{r^2 k^2} du \Big\}
 \end{aligned} \tag{B.8}$$

are martingales. Since $z \in [0, D]$, $D = \sup\{z : F(z) \leq 1\} < +\infty$, and

$$E \left[\int_0^T \int_{R^+} \left[e^{-2q(z \wedge a^*)e^{r(T-u)}} - 1 \right]^2 \lambda dF(z) du \right] < \infty, \tag{B.9}$$

then the process

$$\int_0^t \int_{R^+} \left[e^{-2q(z \wedge a^*)e^{r(T-u)}} - 1 \right] \tilde{\gamma}(dz, du) \tag{B.10}$$

is also a martingale.

According to Taksar and Zeng [16],

$$E \left[\exp \left\{ \int_0^t \frac{3(\mu-r)^2 s^{-2\beta}(u)}{k^2} du \right\} \right] < \infty \tag{B.11}$$

when either of conditions (I) and (II) in Theorem 3 is satisfied. Taking expectation from both sides of (B.6) yields

$$\begin{aligned}
 E[\tilde{V}^2(t)] &= \tilde{V}^2(0) E \\
 &\times \left[\exp \left\{ \int_0^t 3q^2 \sigma^2 e^{2r(T-u)} du \right. \right. \\
 &\quad + \int_0^t \frac{3(1 - e^{-2r\beta(T-u)})^2 (\mu - r)^4 s^{-2\beta}(u)}{4r^2 k^2} du \\
 &\quad + \int_0^t \frac{3(\mu - r)^2 s^{-2\beta}(u)}{k^2} \\
 &\quad \times \left[1 + \frac{(1 - e^{-2r\beta(T-u)}) (\mu - r)}{2r} \right]^2 du \\
 &\quad + \int_0^t -\frac{(1 - e^{-2r\beta(T-u)}) (\mu - r)^3 s^{-2\beta}(u)}{rk^2} \\
 &\quad \times \left[1 + \frac{(1 - e^{-2r\beta(T-u)}) (\mu - r)}{2r} \right] du \\
 &\quad \left. \left. + \int_0^t \int_{R^+} \left[e^{-2q(z \wedge a^*) e^{r(T-u)}} \right. \right. \right. \\
 &\quad \quad \left. \left. + 2q(z \wedge a^*) e^{r(T-u)} - 1 \right] \right. \\
 &\quad \left. \left. \times \lambda dF(z) du \right\} \right]. \tag{B.12}
 \end{aligned}$$

If $\beta \geq 0$, then $0 \leq (1 - e^{-2r\beta(T-u)}) < 1$. For $\mu > r > 0$ and $T > 0$ are constants, we have

$$\begin{aligned}
 &\exp \left\{ \int_0^t \frac{3(\mu - r)^2 s^{-2\beta}(u)}{k^2} \right. \\
 &\quad \left. \times \left[1 + \frac{(1 - e^{-2r\beta(T-u)}) (\mu - r)}{2r} \right]^2 du \right\} \\
 &< \exp \left\{ \int_0^t \frac{3(\mu - r)^2 s^{-2\beta}(u)}{k^2} \left[1 + \frac{(\mu - r)}{2r} \right]^2 du \right\} \\
 &= \exp \left\{ \left[1 + \frac{(\mu - r)}{2r} \right]^2 \int_0^t \frac{3(\mu - r)^2 s^{-2\beta}(u)}{k^2} du \right\},
 \end{aligned}$$

$$\begin{aligned}
 &\exp \left\{ \int_0^t \frac{3(1 - e^{-2r\beta(T-u)})^2 (\mu - r)^4 s^{-2\beta}(u)}{4r^2 k^2} du \right\} \\
 &< \exp \left\{ \int_0^t \frac{3(\mu - r)^4 s^{-2\beta}(u)}{4r^2 k^2} du \right\}, \\
 &\exp \left\{ \int_0^t \left| -\frac{(1 - e^{-2r\beta(T-u)}) (\mu - r)^3 s^{-2\beta}(u)}{rk^2} \right. \right. \\
 &\quad \left. \left. \times \left[1 + \frac{(1 - e^{-2r\beta(T-u)}) (\mu - r)}{2r} \right] \right| du \right\} \\
 &< \exp \left\{ \int_0^t \frac{(\mu - r)^3 s^{-2\beta}(u)}{rk^2} \left[1 + \frac{(\mu - r)}{2r} \right] du \right\}. \tag{B.13}
 \end{aligned}$$

For the case of $\beta < 0$, $|1 - e^{-2r\beta(T-u)}| < e^{-2r\beta T}$, which is similar. After simple calculation and analysis, we obtain

$$E[V^2(t, X^{\alpha^*}(t), S(t))] = E[\tilde{V}^2(t)] < \infty. \tag{B.14}$$

So $E[V^2(\tau_i \wedge T, X^{\alpha^*}(\tau_i \wedge T), S(\tau_i \wedge T))] < \infty$ for $i = 1, 2, \dots$ \square

We use Lemma B.1 to prove the verification Theorem 3 as follows.

Proof. By Itô's formula for Itô-Lévy process, we have

$$\begin{aligned}
 dV(t, X^\alpha(t), S(t)) &= \mathcal{A}^\alpha V(t, X^\alpha(t), S(t)) + V_x \sigma dW(t) \\
 &\quad + kS^\beta(t) [V_x \pi(t) X(t) + V_s S(t)] dW_1(t) \\
 &\quad + \int_{R^+} [V(t, X(t-) + (z \wedge a), S(t)) \\
 &\quad \quad - V(t, X(t-), S(t))] \tilde{\gamma}(dz, dt). \tag{B.15}
 \end{aligned}$$

Take a sequence of bounded open sets Q_1, Q_2, Q_3, \dots , with $Q_i \subset Q_{i+1} \subset Q$, $i = 1, 2, \dots$. For $(x, s) \in Q_1$, let τ_i denote the exit time of (x, s) from Q_i . Then $\tau_i \wedge T \rightarrow T$ when $i \rightarrow \infty$. Integrating both sides of the above equation,

$$\begin{aligned}
 &V(\tau_i \wedge T, X^\alpha(\tau_i \wedge T), S(\tau_i \wedge T)) \\
 &= V(t, x, s) + \int_t^{\tau_i \wedge T} \mathcal{A}^\alpha V(u, X^\alpha(u), S(u)) du \\
 &\quad + \int_t^{\tau_i \wedge T} V_x \sigma dW(u) \\
 &\quad + \int_t^{\tau_i \wedge T} kS^\beta(u) [V_x \pi x + V_s S(u)] dW_1(u) \\
 &\quad + \int_t^{\tau_i \wedge T} \int_{R^+} [V(u, X(u-) + (z \wedge a), S(u)) \\
 &\quad \quad - V(u, X(u-), S(u))] \tilde{\gamma}(dz, du). \tag{B.16}
 \end{aligned}$$

Since $\sup_{\alpha \in \Lambda} \{\mathcal{A}^\alpha V(t, x, s)\} = 0$, then $\mathcal{A}^\alpha V(t, x, s) < 0$. The last three terms of (B.16) are square-integrable martingales with zero expectation, taking conditional expectation given (t, x, s) on two sides of the above formula:

$$\begin{aligned} & E [V(\tau_i \wedge T, X^\alpha(\tau_i \wedge T), S(\tau_i \wedge T)) \\ & \quad | X^\alpha(t) = x, S(t) = s] \\ & = V(t, x, s) \\ & \quad + E \left[\int_t^{\tau_i \wedge T} \mathcal{A}^\alpha V(u, X^\alpha(u), S(u)) du \right. \\ & \quad \left. | X^\alpha(t) = x, S(t) = s \right] \\ & \leq V(t, x, s). \end{aligned} \tag{B.17}$$

From Lemma B.1, $V(\tau_i \wedge T, X^\alpha(\tau_i \wedge T), S(\tau_i \wedge T))$, $i = 1, 2, \dots$, are uniformly integrable. Thus we have

$$\begin{aligned} & H(t, x, s) \\ & = \sup_{\alpha \in \Lambda} E [U(X(T)) | X^{\alpha^*}(t) = x, S(t) = s] \\ & = \lim_{t \rightarrow \infty} E [V(\tau_i \wedge T, X^\alpha(\tau_i \wedge T), S(\tau_i \wedge T))] \\ & \leq V(t, x, s). \end{aligned} \tag{B.18}$$

When $\alpha = \alpha^*$, the inequality in the above formula becomes an equality; that is, $H(t, x, s) = V(t, x, s)$. The proof is finished. \square

Conflict of Interests

The authors declare that they have no conflict of interests regarding the publication of this paper.

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