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On a Solution of the Optimal Stopping Problem for Processes with Independent Increments

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Abstract

We discuss a solution of the optimal stopping problem for the case when a reward function is a power function of a process with independent stationary increments (random walks or Levy processes) on an infinite time interval. It is shown that an optimal stopping time is the first crossing time through a level defined as the largest root of the Appell function associated with the maximum of the underlying process.

1. Introduction. Let $X=(X_t)$ be a process with independent stationary increments with a discrete time parameter $t\in \mathbf{Z}^+=\{0,1,2,\ldots\}$ or continuous time parameter $t\in \mathbf{R}^+=[0,\infty),\, X_0=x\in \mathbf{R}=(-\infty,\infty).$ We suppose that X is defined on a probability space (Ω,\mathcal{F},P) with a natural filtration $\mathcal{F}_t=\sigma\{X_s,s\leq t\},\,\mathcal{F}_0=\{\varnothing,\Omega\}$.

The optimal stopping problem we study here consists in finding the "value" function

$$V(x) = \sup_{\tau \in \mathcal{M}} E(e^{-q\tau} g(X_{\tau}) I\{\tau < \infty\}), \qquad x \in \mathbf{R}, \quad q \ge 0,$$

where g(x) is a measurable function, \mathcal{M} is the class of all Markov times τ (with respect to (\mathcal{F}_t)) with values in $[0,\infty]$, $I\{A\}$ is the indicator function. We call τ^* as the *optimal* stopping time if

(1)
$$V(x) = E(e^{-q\tau^*}g(X_{\tau^*})I\{\tau^* < \infty\}), \qquad x \in \mathbf{R}$$

We discuss here only the case of power reward functions that is the case

$$g(x) = (x^+)^{\nu}, \qquad \nu > 0, \quad x^+ = \max(x, 0)$$

though the method developed below is quite general and can be used for finding explicit solutions for monotone functions g(x).

The explicit solution of the problem under consideration for discrete time setting and the case $\nu=1$ was found in [7] and [6]. We generalised their results for the case of integer $\nu=1,2,\ldots$ in [11] (we discussed the case q=0 in [11]) using properties of the so-called Appell polynomials associated with the maximum of the process X_t . Kyprianou and Surya [8] have got an extension of our result to the continuous time setting with q>0.

To solve the problem for arbitrary power $\nu > 0$ we had to study a generalisation of Appell polynomials which we call Appell functions. As in [11] and [8] we show the optimal stopping time has the threshold form that is

$$\tau^* = \tau_a = \inf\{t \ge 0 : X_t \ge a\}$$

where the optimal value of the parameter a is defined as a positive root of the Appell function associated with the maximum of the process X_t , see Theorem 1 in Section 4. Note that for the case $0 < \nu < 1$ this result can be derived also by methods of the

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paper Beibel [1] (we thank Prof. R. Lerche for this reference). Some necessary facts about Appell functions and the maximum of X_t are presented in Sections 2 and 3. In Section 5 we put two simple examples.

2. Appell functions. Appell polynomials (or, Sheffer polynomials, see e.g. [12]) generated by a random variable (r.v.) η such that $E|\eta|^n < \infty$ can be defined as follows:

(2)
$$Q_k(y;\eta) = (-1)^k \frac{d^k}{du^k} \left(\frac{e^{-uy}}{Ee^{-u\eta}} \right) \Big|_{u=0} \qquad k = 1, 2, \dots, n.$$

Based on this definition it is easy to derive the following properties of Appell polynomials which are valid under the assumption $E|\eta|^n < \infty$: for k = 1, ..., n

(3)
$$\frac{d}{dy}Q_k(y;\eta) = kQ_{k-1}(y;\eta),$$

(4)
$$E(Q_k(\eta + y; \eta)) = y^k.$$

Now we define continuous functions (we call them Appell functions) which have these two properties but with real parameter ν instead integer k. At first we find a function $Q_{\nu}(y;\eta)$ which satisfies both (3) and (4) with negative ν instead of integer k.

We assume further that η is a nonnegative random variable and

(5)
$$P(0 \le \eta < \varepsilon) > 0 \text{ for any } \varepsilon > 0.$$

Actually, Appell functions can be defined under more general assumptions but for purposes of this paper we shall need only this case. Condition (5) implies

$$Ee^{-u\eta} \ge P(0 \le \eta < \varepsilon)e^{-u\varepsilon} > 0, \qquad \varepsilon > 0,$$

and, hence, for any $\nu < 0$ and y > 0

(6)
$$\int_0^\infty u^{-\nu-1} \frac{e^{-uy}}{Ee^{-u\eta}} du < \infty.$$

For the proof of main results (see Section 4) we shall use the random variable $\eta = M_{\theta} = \sup_{0 \le t \le \theta} (X_t - X_0)$ which does satisfy (5) (see e.g. Lemma 2 in Section 3).

At first, we define the Appell function of order ν for $\nu < 0$ using the following integral representation:

(7)
$$Q_{\nu}(y;\eta) = \int_{0}^{\infty} u^{-\nu - 1} \frac{e^{-uy}}{Ee^{-u\eta}} \frac{du}{\Gamma(-\nu)} \qquad (y > 0, \ \nu < 0)$$

where $\Gamma(z)$ is the Gamma function. Accordingly to this representation the function $Q_{\nu}(y;\eta)$ is continuous with respect to both parameters ν and y. Note also that for all y>0

(8)
$$\lim_{\nu \uparrow 0} Q_{\nu}(y; \eta) = 1.$$

To see this we write the integral in (7) as a sum of two integrals $\int_0^{\varepsilon} + \int_{\varepsilon}^{\infty}$ and then show that $\int_{\varepsilon}^{\infty}$ vanishes as $\nu \uparrow 0$ for any $\varepsilon > 0$. With help of the fundamental

property of the Gamma function (that is $\Gamma(z+1)=z\Gamma(z)$, $\Gamma(1)=1$) one can show that

$$\int_0^\varepsilon u^{-\nu-1} \frac{e^{-uy}}{Ee^{-u\eta}} \frac{du}{\Gamma(-\nu)} = 1 + o(1), \quad \nu \uparrow 0,$$

and that implies (8). So, to define a continuous function $Q_{\nu}(y;\eta)$ (as a function of parameter ν) we need to set

$$Q_0(y;\eta) = 1$$

for all y > 0.

Definition (7) implies

(10)
$$\frac{d}{dy}Q_{\nu}(y;\eta) = -\int_{0}^{\infty} u^{-\nu} \frac{e^{-uy}}{Ee^{-u\eta}} \frac{du}{\Gamma(-\nu)} = \nu Q_{\nu-1}(y;\eta) \quad (y > 0, \ \nu < 0)$$

where we used the fundamental property of the Gamma function $\Gamma(z+1) = z\Gamma(z)$ and the last equation holds due to (7). Also,

(11)
$$E(Q_{\nu}(y+\eta;\eta)) = \int_0^\infty u^{-\nu-1} \frac{Ee^{-u(y+\eta)}}{Ee^{-u\eta}} \frac{du}{\Gamma(-\nu)} = y^{\nu} \quad (y>0, \ \nu<0).$$

So, we have got properties (3) and (4) with parameter $\nu < 0$ instead of k.

To define $Q_{\nu}(y;\eta)$ for real $\nu>0$ we set as a definition the following relation:

(12)
$$Q_{\nu}(y;\eta) = Q_{\nu}(0;\eta) + \nu \int_{0}^{y} Q_{\nu-1}(z;\eta) dz, \qquad y > 0, \quad \nu > 0,$$

assuming that $Q_{\nu}(0;\eta)$ is a finite constant. In other words, we require the validity of (10) also for $\nu > 0$.

To implement this definition we define at first the function $Q_{\nu}(y;\eta)$ for $\nu \in (0,1)$ using the representation (7), then for $\nu \in (1,2)$ based on (12) and with $Q_{\nu}(y;\eta)$ just defined for $\nu \in (0,1)$ and so on. Doing so, we get the analog of property (3) for any real ν instead of integer k.

Note the defined function $Q_{\nu}(y;\eta)$ is continuous with respect to both parameters ν and y>0 because the right-hand side of (12) is an integral of a continuous function.

To have an analog of property (4) we set

(13)
$$Q_{\nu}(0;\eta) = -\nu E\left(\int_{0}^{\eta} Q_{\nu-1}(z;\eta)dz\right) \quad \nu > 0,$$

assuming that the last expectation is finite. The finiteness of the last integral does hold under the condition

$$E(n^{\nu}) < \infty$$
.

To see this, we may chose the constant A > 0 such that $P(0 \le \eta < A) > 0$ and apply the estimate

$$Ee^{-u\eta} > P(0 < \eta < A)e^{-uA}$$
.

Since

$$Q_{\nu-1}(z;\eta) = \int_0^\infty u^{-\nu} \frac{e^{-uz}}{Ee^{-u\eta}} \frac{du}{\Gamma(1-\nu)} \quad (y > 0, \ \nu < 1),$$

we get the following upper bound with A from above and any y > 0, $\nu \in (0,1)$:

$$\int_{0}^{y} Q_{\nu-1}(z;\eta)dz \leq I(y \leq A) \int_{0}^{A} Q_{\nu-1}(z;\eta) dz$$

$$+ I(y > A) \int_{A}^{y} \left(\int_{0}^{\infty} \frac{u^{-\nu} e^{-u(z-A)}}{\Gamma(1-\nu)P(0 \leq \eta < A)} du \right) dz$$

$$\leq I(y \leq A) A \max_{0 \leq z \leq A} Q_{\nu-1}(z;\eta) + I(y > A) \int_{A}^{y} \frac{(z-A)^{\nu-1}}{P(0 \leq \eta < A)} dz$$

$$\leq C(A,\nu)(I(y \leq A) + I(y > A)(y-A)^{\nu}).$$

where $C(A, \nu)$ is a finite positive constant. By (13) the last estimate implies that for $\nu \in (0, 1)$

(14)
$$|Q_{\nu}(0;\eta)| \le C(A,\nu) (1 + E(I(\eta > A)(\eta - A)^{\nu})) < \infty, \quad \nu \in (0,1),$$
 and also by (12)

$$|Q_{\nu}(y;\eta)| \le |Q_{\nu}(0;\eta)| + C(A,\nu)(I(y \le A) + I(y > A)(y - A)^{\nu}), \quad \nu \in (0,1),$$

where $C(A, \nu)$ is some finite positive constant. Now we can apply this estimate in (12) for the case $\nu \in (1, 2)$ and similarly get

$$|Q_{\nu}(0;\eta)| \le C(A,\nu)(|Q_{\nu}(0;\eta)|E(\eta)+1+E(I(\eta>A)(\eta-A)^{\nu}))<\infty,\nu\in(1,2)$$

with another finite positive constant $C(A, \nu)$. Continuing this procedure we get the estimate with the main term $E(I(\eta > A)(\eta - A)^{\nu})$ like in (14) but for any $\nu > 0$.

Now we claim that under the condition $E(\eta^{\nu}) < \infty$ we have the analog of the property (4) for any ν :

(15)
$$E(Q_{\nu}(\eta + y; \eta)) = y^{\nu}, \quad y > 0.$$

Indeed, we have shown above that it is true for $\nu < 0$ (see (11)). For $\nu = 0$ it is true by definition (9). For $\nu \in (0,1)$ and y > 0 we have by definitions (12) and (13) that

$$E(Q_{\nu}(\eta + y; \eta)) = \nu E \int_{\eta}^{\eta + y} Q_{\nu - 1}(z; \eta) dz = \nu \int_{0}^{y} EQ_{\nu - 1}(z + \eta; \eta) dz,$$

where $EQ_{\nu-1}(z+\eta;\eta) = z^{\nu-1}$ due to (11). So we have shown the validity of (15) for $\nu \in (0,1)$. Applying this consideration recursively (of course, always assuming the existence of integrals) we get the validity of (15) for all real $\nu > 0$.

The case of integer $\nu = 0, 1, 2, \ldots$ now can viewed as a limiting case of functions $Q_{\nu}(y; \eta)$ but, of course, the original definition (2) is easier to use.

Below we shall use the following property of the Appell functions.

Lemma 1. Let (5) hold and let $E(\eta^{\nu}) < \infty$. Then for any $\nu > 0$ there exists a_{ν} such that

(16)
$$Q_{\nu}(y;\eta) < 0 \text{ for } 0 < y < a_{\nu}, \qquad Q_{\nu}(a_{\nu};\eta) = 0$$

and $Q_{\nu}(y;\eta)$ is an increasing function for $y \geq a_{\nu}$.

Proof. For integer $\nu = 1, 2, \dots$ the statement of this lemma was proved in [11, Lemma 5].

For the case $\nu \in (0,1)$ we note at first that due to the assumption $\eta \geq 0$ we have the estimate

$$Q_{\nu-1}(y;\eta) \ge \int_0^\infty u^{-\nu-1} e^{-uy} \frac{du}{\Gamma(-\nu)} = y^{\nu-1} \qquad (\nu < 1)$$

and so by (13) $Q_{\nu}(0;\eta) < 0$. Also, by (12) we have that $Q_{\nu}(y;\eta)$ is a nondecreasing function (of the variable y) such that

$$Q_{\nu}(y;\eta) \ge Q_{\nu}(0;\eta) + y^{\nu}.$$

So, it grows to infinity and, hence, Lemma 1 does hold for $\nu \in (0,1)$.

Next, consider the case $\nu \in (1,2)$ and y>0. Then due the fact just proved

$$Q_{\nu-1}(y;\eta) < 0 \text{ for } y \in (0, a_{\nu-1}).$$

So, on the interval $(0, a_{\nu-1})$ the function $Q_{\nu}(y; \eta)$ is negative and decreasing. Obviously, it reaches its minimum at point $y = a_{\nu-1}$. For $y \ge a_{\nu-1}$ the function $Q_{\nu}(y; \eta)$ is ultimately increasing to infinity due to the estimate

$$Q_{\nu}(y;\eta) \ge Q_{\nu}(0;\eta) + Q_{\nu-1}(0;\eta)y + y^{\nu} \text{ for } y > 0.$$

Hence, there exists a root $a_{\nu} > a_{\nu-1} > 0$.

Using (12) recursively and the consideration presented above we see that the statement of Lemma 1 holds for all $\nu > 0$.

3. Some facts about the distribution of maximum

Writing $t \in \mathbf{Z}^+$ or $t \in \mathbf{R}^+$ we will indicate that the discrete time or continuous time cases are under consideration correspondingly. We formulate here all results in a form which is valid for the both cases $t \in \mathbf{Z}^+$ and $t \in \mathbf{R}^+$ but proofs of corresponding results we will have to discuss separately.

We assume always below that (X_t) is a process with independent homogeneous increments, $X_0 = x$. Let a random variable θ be independent of X_t such that

$$P(\theta > t) = e^{-tq}, \qquad q > 0,$$

where $t \in \mathbf{Z}^+$ or $t \in \mathbf{R}^+$. Set

$$M_{\theta,q} = \sup_{0 < t < \theta} (X_t - x),$$

and by definition set for the case q = 0

$$M_{\theta,0} = M_{\infty} \stackrel{\text{def}}{=} \sup_{0 \le t < \infty} (X_t - x).$$

Further we always assume that for the case q = 0

(17)
$$E(X_1^+) < \infty \quad \text{nand} \quad E(X_1 - x) < 0.$$

Lemma 2. If $q \ge 0$ then

(18)
$$P\{M_{\theta,q} < \varepsilon\} > 0 \text{ for any } \varepsilon > 0.$$

Proof.

1) For the case $t \in \mathbf{Z}^+$ and q = 0 this result is a consequence of the observation that

$$P\{M_{\infty} < \varepsilon\} \ge P\{M_{\infty} = 0\}$$

and the fact that under imposed conditions

(19)
$$P\{M_{\infty} = 0\} > 0$$

(see e.g. [3, pp. 91-92]).

2) For the case $t \in \mathbf{Z}^+$ and q > 0 note that if $\theta = 1$ then $M_{\theta,q} = 0$ and, hence,

$$P\{M_{\theta,q} < \varepsilon\} \ge P\{\theta = 1\} = e^{-q} > 0.$$

So (18) holds as well.

3) Consider now the case $t \in \mathbf{R}^+$ and q = 0. Let (R_t) be a compound Poisson process generated by jumps of (X_t) which are greater than 1. Set

$$Q_t = X_t - x - R_t.$$

Due to this definition the process (Q_t) does not contain jumps exceeding 1. Note that (R_t) and (Q_t) are independent processes with stationary increments, $E(R_1) \ge 0$

To prove Lemma 2 we note that for any m

$$M_{\infty} = \sup_{t \ge 0} (Q_t + mt + R_t - mt) \le \sup_{t \ge 0} (Q_t + mt) + \sup_{t \ge 0} (R_t - mt)$$

and so due to independency of (R_t) and (Q_t) we have for any $\varepsilon > 0$

(20)
$$P\{M_{\infty} < \varepsilon\} \ge P\left\{\sup_{t>0} (Q_t + mt) < \varepsilon/2\right\} P\left\{\sup_{t>0} (R_t - mt) < \varepsilon/2\right\}.$$

We shall estimate the both last probabilities separately under a proper choice of $\,m\,.$

To estimate $P\{\sup_{t\geq 0}(Q_t+mt)<\varepsilon/2\}$ note that we may choose the constant $m>E(R_1)\geq 0$ such that

$$E(Q_1) + m = E(X_1 - x) + m - E(R_1) < 0$$

(see assumption (17)). Now we show that with the such choice of m for any $\varepsilon > 0$

(21)
$$P\left(\sup_{t>0}(Q_t + mt) < \varepsilon\right) > 0.$$

To see this, consider the exponential martingale

$$Z_t(u) = \exp\{u(Q_t + mt) - t\varphi(u)\}\$$

with

$$\varphi(u) = \log E e^{u(Q_1 + m)}.$$

Then

$$\varphi(0) = 0, \quad \varphi'(0) = E(Q_1 + m) < 0.$$

and, as well known, $\varphi(u)$ is a continuous convex function.

Suppose the function $\varphi(u)$ has a root $u^* > 0$. (It is certainly true when $\varphi(u) \to \infty$, e.g. when Q_t contains a diffusion component or a component with positive jumps.) Then the process $Z_t(u^*) = \exp\{u^*Q_t\}$ is an exponential martingale. Applying the optional stopping theorem for the stopping time

$$\tau_{\varepsilon} = \inf\{t : Q_t + mt \ge \varepsilon\}$$

and the fact that $E(\exp\{u^*(Q_t+mt)\})=1$ we get the inequality

(22)
$$EI\{\tau_{\varepsilon} < \infty\} \exp\{u^*(Q_{\tau_{\varepsilon}} + m\tau_{\varepsilon})\} \le 1.$$

Since $Q_{\tau_{\varepsilon}} + m\tau_{\varepsilon} \ge \varepsilon$ on the set $\{\tau_{\varepsilon} < \infty\} = \{\sup_{t > 0} (Q_t + mt) \ge \varepsilon\}$ it implies that $P\{\tau_{\varepsilon} < \infty\} \le e^{-u^* \varepsilon}$ and so

(23)
$$P\left\{\sup_{t>0}(Q_t+mt)<\varepsilon\right\} \ge 1 - e^{-u^*\varepsilon} > 0.$$

Hence, (21) does hold under the assumption that the function $\varphi(u)$ has a root $u^* > 0$. Consider now the alternative case when $\varphi(u) \leq 0$ for all u > 0. Then we may choose u = 1 and similar to (22) we get

(24)
$$EI\{\tau_{\varepsilon} < \infty\} \exp\{Q_{\tau_{\varepsilon}} + m\tau_{\varepsilon} - \tau_{\varepsilon}\varphi(1)\} \le 1.$$

Since $\varphi(1) \leq 0$ and $Q_{\tau_{\varepsilon}} + m\tau_{\varepsilon} \geq \varepsilon$ on the set $\{\tau_{\varepsilon} < \infty\}$, this inequality implies (23) with $u^* = 1$ and so (21) does hold for all possible cases.

Next, we show that for any constant $m > E(R_1) \ge 0$ and any $\varepsilon > 0$

(25)
$$P\left\{\sup_{t>0}(R_t - mt) < \varepsilon\right\} > 0.$$

This estimate is, actually, a consequence of the fact (19) and the estimate (23) proved above. Recall that (R_t) is a compound Poisson process generated by jumps of (X_t) which are greater than 1 and so it has the representation

$$R_t = \sum_{k=1}^{N_t} \xi_k,$$

where (N_t) is a Poisson process with the rate $\lambda \geq 0$, ξ_k are independent identically distributed (iid) random variables, $\xi_k > 1$, $\{\xi_k\}$ and N_t are independent. Assume further $\lambda > 0$ (otherwise $P\{\sup_{t>0}(R_t - mt) < \varepsilon\} = 1$), choose b such that

$$E(R_1) = \lambda E(\xi_1) < \lambda b < m$$

and note

$$P\Big\{\sup_{t>0}(R_t - mt) < \varepsilon\Big\} \ge P\Big\{\sup_{t>0}(R_t - bN_t) = 0, \sup_{t>0}(bN_t - mt) < \varepsilon\Big\}.$$

Here $R_t - bN_t = \sum_{k=1}^{N_t} (\xi_k - b)$ and so

(26)
$$\sup_{t \ge 0} (R_t - bN_t) = \sup_{k \ge 1} S_k^+,$$

where (S_k) is a random walk with negative drift as $E(S_1) = E(\xi_k - b) < 0$. This implies

$$P\Big\{\sup_{t>0}(R_t - bN_t) = 0\Big\} = P\Big\{\sup_{k>1}S_k^+ = 0\Big\} > 0$$

(see the step 1).

Since the set $\{\sup_{k\geq 1} S_k^+ = 0\}$ and the process (N_t) are independent we get

$$P\Big\{\sup_{t\geq 0}(R_t - mt) < \varepsilon\Big\} \geq P\Big\{\sup_{k\geq 1}S_k^+ = 0\Big\} P\Big\{\sup_{t\geq 0}(N_t - tm/b) < \varepsilon/b\Big\}.$$

Now we need just to note that the inequality

$$P\Big\{\sup_{t>0}(N_t - tm/b) < \varepsilon\Big\} > 0$$

is a particular case of (23) because (N_t) is a Poisson process with unit jumps and $E(N_1) = \lambda < m/b$.

To complete the proof of Lemma 2 we need just to note that under the choice of m indicated above we have shown that the lower bound in (20) $\varepsilon > 0$.

4) For the case $t \in \mathbf{R}^+$ and q > 0 note that due to independency of (R_t) and (Q_t) we have for any $\varepsilon > 0$

$$P\{M_{\theta,q} < \varepsilon\} \ge P\Big\{ \sup_{0 \le t < \theta} R_t < \varepsilon/2 \Big\} P\Big\{ \sup_{0 \le t < \theta} Q_t < \varepsilon/2 \Big\}.$$

Due to independency of θ and $\sup_{0 \le s \le t} R_s$ we have

$$P\Big\{\sup_{0\leq s<\theta}R_s<\varepsilon/2\Big\}\geq P\Big\{\sup_{0\leq s\leq 1}R_s<\varepsilon/2\Big\}\,P\{\theta<1\},$$

where $P\{\theta < 1\} = 1 - e^{-q} > 0$ and so, obviously,

$$P\left\{\sup_{0 \le s \le 1} R_s < \varepsilon/2\right\} \ge P\left\{\sup_{0 \le s \le 1} R_s = 0\right\} > 0$$

for any $\varepsilon > 0$.

To estimate from below $P\{\sup_{0 \le t < \theta} Q_t < \varepsilon\}$ we can use the consideration from the previous step 3) in the part related to the process Q_t . At first note

$$P\Big\{\sup_{0\leq s<\theta}Q_s<\varepsilon\Big\}\geq P\Big\{\sup_{0\leq s\leq t}Q_s<\varepsilon\Big\}P\{\theta< t\}.$$

Consider the exponential martingale $Z_t(u) = \exp\{uQ_t - t\varphi(u)\}$, u > 0. Applying the optional stopping theorem for the stopping time $\tau_{\varepsilon} = \inf\{t : Q_t > \varepsilon\}$ we get

$$E(I\{\tau_{\varepsilon} < t\}e^{uQ_{\tau_{\varepsilon}} - \tau_{\varepsilon}\varphi(u)}) \le 1.$$

Then due to the estimate $Q_{\tau_{\varepsilon}} \geq \varepsilon$ we get

$$P\{\tau_{\varepsilon} < t\} \le e^{-u\varepsilon + t\varphi(u)}$$

and so

$$P\Big\{\sup_{0 \le s \le t} (Q_s) < \varepsilon\Big\} = 1 - P\{\tau_{\varepsilon} < t\} \ge 1 - e^{-u\varepsilon + t\varphi(u)}.$$

Fixing $\varepsilon > 0$ we can find small t such that $u\varepsilon > t\varphi(u)$ and this implies the required fact that for any $\varepsilon > 0$

$$P\{\sup_{0 \le s < \theta} (Q_s) < \varepsilon\} > 0.$$

The proof of Lemma 2 is completed.

Lemma 3. Let $\nu > 0$,

$$q = 0$$
, $E(X_1) < 0$, $E((X_1^+)^{\nu+1}) < \infty$

or

$$q > 0$$
, $E((X_1^+)^{\nu}) < \infty$.

Then

(27)
$$E(M_{\theta,q}^{\nu}) < \infty.$$

Proof. 1) For the case $t \in \mathbf{Z}^+$ and q = 0 this result is well know, see e.g. [3, pp. 91–92].

2) Consider here the case $t \in \mathbb{Z}^+$ and q > 0. At first note that

$$M_{\theta,q} \le \sum_{k=1}^{\theta} (\Delta X_k)^+,$$

where ΔX_k are iid r.v. For the case $0 < \nu \le 1$ by Hölder inequality and Wald's identity

$$E(M_{\theta}^{\nu}) \le E\left(\sum_{k=1}^{\theta} ((\Delta X_k)^+)^{\nu}\right) = E(\theta)E((\Delta X_k)^+)^{\nu}) < \infty.$$

For the case $1 < \nu \le 2$ we note that

(28)
$$Z_{t} = \sum_{k=1}^{t} [(\Delta X_{k})^{+} - E(\Delta X_{k})^{+}]$$

is a martingale and so we can use well-known martingale inequalities (see e.g. [5]) which lead to the estimate

$$E(M_{\theta}^{\nu}) \leq C_{\nu}(E(|Z_{\theta}|^{\nu}) + C_{\nu}E(\theta^{\nu})(E((\Delta X_{k})^{+})^{\nu}))$$

and

$$E(|Z_{\theta}|^{\nu}) \leq C_{\nu} E(\theta^{\nu}) E((\Delta X_k)^+)^{\nu}) < \infty$$

with some finite constants C_{ν} .

For $\nu \geq 2$ the process (Z_t) from (28) is a square integrable martingale and with help of the same martingale inequalities from [5] we get

$$E(M_{\theta}^{\nu}) \le C_{\nu} (E(|Z_{\theta}|^{2})^{\nu/2} + C_{\nu} E(\theta^{\nu}) E((\Delta X_{k})^{+})^{\nu})$$

and

$$E(|Z_{\theta}|^2) \le C_{\nu} E(\theta^{\nu}) E((\Delta X_k)^+)^{\nu}) < \infty$$

with some finite constants C_{ν} .

So, (27) is proved for the case $t \in \mathbf{Z}^+$ and q > 0.

3) Consider now the case $t \in \mathbf{R}^+$ and q = 0. We may use considerations which are similar to the proof of Lemma 2. With the same choice of constants m and b as in the proof of Lemma 2 (step 3) we have for any x > 0

$$P\{M_{\infty} > x\} \le P\left\{\sup_{t \ge 0} (R_t - bN_t) > \frac{x}{3}\right\} + P\left\{\sup_{t \ge 0} (Q_t + mt) > \frac{x}{3}\right\} + P\left\{\sup_{t \ge 0} (N_t - tm/b) > \frac{x}{3b}\right\}.$$

Integrating both sides of this inequality with respect to the measure

$$I\{x > 0\}\nu x^{\nu - 1} dx$$

we get

(29)
$$E(M_{\infty}^{\nu}) \leq 3^{\nu} E\left(\sup_{t \geq 0} (R_t - bN_t)^{\nu}\right) + 3^{\nu} E\left(\sup_{t \geq 0} (Q_t + mt)^{\nu}\right) + (3b)^{\nu} E\left(\sup_{t \geq 0} (N - tm/b)^{\nu}\right).$$

Note that due to the relation (26)

$$E\left(\sup_{t>0}(R_t - bN_t)^{\nu}\right) = E\left(\sup_{k>1}(S_k^+)^{\nu}\right)$$

and so this term in (29) is finite because we have assumed that $E(X_1^+)^{\nu+1} < \infty$ (see step 1) above).

Now we show that for any $\nu > 0$

$$(30) E\left(\sup_{t>0} (Q_t + mt)^{\nu}\right) < \infty$$

and

$$(31) E(\sup_{t>0} (N - tm/b)^{\nu}) < \infty.$$

To prove these facts we can use again the exponential martingale $Z_t(u^*) = \exp\{u^*(Q_t + mt)\}$ and apply the optional stopping theorem with the stopping time

$$\tau_x = \inf\{t : Q_t + mt > x\}.$$

With some standard considerations involving uniform integrability we get the identity

$$E(I\{\tau_x < \infty\} \exp\{u^*(Q_{\tau_x} + m\tau_x)\}) = 1.$$

Since $Q_{\tau_{\varepsilon}} \geq x$ it implies that $P\{\tau_x < \infty\} \leq e^{-u^*x}$ and so

$$P\Big\{\sup_{t\geq 0}(Q_t+mt) > x\Big\} = P\{\tau_x < \infty\} \leq e^{-u^*x}$$

and so (30) does hold. To complete the step 3) we need just note that the inequality (31) is a particular case of (30).

4) The result of this part of Lemma 3 was proved in [8]. We would like to mention here that, actually, it is a simple consequence of general martingale inequalities proved originally in [10].

Lemma 4.

(a) Let $\tau_a = \inf\{t \ge 0 : X_t \ge a\}$, $a \ge x$. Then for all $u \le 0$

(32)
$$E(I\{\tau_a < \infty\} e^{uX_{\tau_a}} e^{-q\tau_a}) = \frac{E(I\{M_\theta + x \ge a\} e^{u(M_\theta + x)})}{E(e^{uM_\theta})}$$

(b) Let the conditions of Lemma 3 hold. Then for all $a \ge x$ and ν

(33)
$$E(I\{\tau_a < \infty\}X_{\tau}^{\nu} e^{-q\tau_a}) = E(I\{M_{\theta} + x \ge a\} Q_{\nu}(M_{\theta} + x; M_{\theta})).$$

Proof. (a) This result for the case $t \in \mathbf{Z}^+$ and q = 0 was proved in [11]. For the case $t \in \mathbf{Z}^+$ with q > 0 the proof needs just some minor modifications by taking into account, in particular, the memoryless property of geometric distribution. We need to note that on the set $\{M_{\theta} + x \geq a\} = \{\tau_a < \theta\}$

$$\widehat{M}_{\theta} := M_{\theta} - (X_{\tau_{\theta}} - x) \stackrel{law}{=} M_{\theta}$$

and \widehat{M}_{θ} is independent of X_{τ_a} on the event $\{\tau_a < \theta\}$. Here we show details of the proof for the case $t \in \mathbf{Z}^+$ and q > 0 only.

We have $P(\theta = k) = e^{-kq}(e^q - 1)$ for k = 1, 2, ... and so

$$E(e^{uM_{\theta}}) = \sum_{k=1}^{\infty} E(I\{\theta = k\}e^{uM_{k-1}}) = \sum_{k=1}^{\infty} e^{-kq}(e^q - 1)E(e^{uM_{k-1}}).$$

Note also that $X_{\tau_a} = x + M_{\tau_a}$ and so

$$E(I\{M_{\theta} + x \ge a\} e^{u(M_{\theta} + x)}) = E(I\{\tau_a < \theta\} e^{u(M_{\theta} + x - X_{\tau_a})} e^{uX_{\tau_a}})$$

$$= \sum_{k=1}^{\infty} E(I\{\tau_a = k\} I\{k < \theta\} e^{u(M_{\theta} - M_k)} e^{uX_k})$$

$$= E(\sum_{k=1}^{\infty} \sum_{n=k}^{\infty} I\{\tau_a = k\} I\{k < n\} I\{\theta = n\} e^{u(M_{n-1} - M_k)} e^{uX_k})$$

$$= E(\sum_{k=1}^{\infty} I\{\tau_a = k\} e^{uX_k} I\{k < n\} \sum_{n=k+1}^{\infty} E(I\{\theta = n\} e^{u\hat{M}_{n-1-k}} | \mathcal{F}_k)) =$$

(setting n = k + i and taking into account that both θ and M_{i-1} are independent and also independent of \mathcal{F}_k)

$$\begin{split} &= E[\sum_{k=1}^{\infty} I\{\tau_a = k\} e^{uX_k} \sum_{i=1}^{\infty} P\{\theta = i + k\} E e^{u\hat{M}_{i-1}}] \\ &= \sum_{k=1}^{\infty} E(I\{\tau_a = k\} e^{uX_k} e^{-kq} \sum_{i=1}^{\infty} e^{-iq} (e^q - 1) E e^{u\hat{M}_i}) \\ &= \sum_{k=1}^{\infty} E(I\{\tau_a = k\} e^{uX_k} e^{-kq} E e^{uM_{\theta}}) = E(e^{uX_{\tau_a}} I\{\tau_a < \infty\} e^{uX_{\tau_a}} e^{-q\tau_a}) E(e^{uM_{\theta}}). \end{split}$$

For the case $t \in \mathbb{R}^+$ considerations are similar (see [8] for more details).

(b). Relation (33) for $\nu < 0$ is obtained by integrating of both sides of (32) with respect to the measure

$$I\{u > 0\}u^{-\nu - 1}du/\Gamma(-\nu), \qquad \nu < 0.$$

Then use relation (12) for $\nu \in (0,1)$ and so on.

Lemma 5. Let $t \in \mathbf{Z}^+$, $q \ge 0$ and let f(x) and g(x) be nonnegative functions such that for all x

$$(34) f(x) \ge g(x)$$

and

$$(35) f(x) \ge e^{-q} Ef(X_1).$$

Then for all x

(36)
$$f(x) \ge \sup_{\tau \in M} E I\{\tau < \infty\} e^{-q\tau} g(X_{\tau}).$$

Proof. Condition (35) implies the fact that the process $e^{-qt}(X_t)$ is a nonnegative supermartingale and, hence, by the supermartingale property we have for any stopping time τ

$$f(x) \ge E I\{\tau < \infty\} e^{-q\tau} f(X_{\tau}).$$

Now one can see that inequality (36) is a consequence of condition (34).

Remark. Lemma 5 is just a slight generalisation of Lemma 7 from [11] (see also [2]).

4. Main result

Theorem 1. Let $g(x) = (x^+)^{\nu}$, $\nu > 0$, the conditions of Lemma 3 hold and let a_{ν} be the positive root of the equation

$$Q_{\nu}(y; M_{\theta}) = 0.$$

Then the stopping time τ_{a_n} is optimal and

$$\nu(x) = E(e^{-q\tau_a} X_{\tau_{a\nu}} I\{\tau_{a\nu} < \infty\}) = E(Q_{\nu}(M_{\theta} + x; M_{\theta}) I\{M_{\theta} + x \ge a_{\nu}\}).$$

Proof. For integer $\nu=1,2,\ldots$ the proof was given for the case $t\in\mathbf{Z}^+$ and q=0 in [11] and for the case $t\in\mathbf{R}^+$ and $q\geq0$ in [8]. (Note that the condition of Theorem 2 in [8] for the case q=0 should be changed as we formulated in Lemma 3 above.) The proof for real $\nu>0$, actually, coincides with the lines of the proof in the mentioned papers. By this reason we just outline it here, omitting obvious details.

At first we show that the function $E(X_{\tau_a}^{\nu} I\{\tau_a < \infty\})$ achieves its maximum at the point $a = a_{\nu}$ where a_{ν} is the positive root of the equation (37) and so by Lemma 4

$$\widehat{\nu}(x) = E(X_{\tau_{a_{\nu}}}^{\nu} I\{\tau_{a_{\nu}} < \infty\}) = E(Q_{\nu}(M+x; M) I\{M+x \ge a_{\nu}\}).$$

This fact is a direct consequence of Lemmas 1, 3, 4(b) and (15).

Next, we note that, obviously,

$$\hat{V}(x) \le V(x).$$

At the final step we show that

$$\hat{V}(x) \ge \nu(x)$$

and conclude that the optimal stopping time is $\, \tau = \tau_{a_{\nu}} \, .$

The proof of (38) for the case $t \in \mathbf{Z}^+$ and q = 0 follows the lines of the proof from the paper [11] given there for integer ν ; for the case $t \in \mathbf{R}^+$ and $q \ge 0$ it follows the lines of the paper [8].

Here we present some details of the proof for (38) only for the case $t \in \mathbf{Z}^+$ and q > 0. The idea of our proof is similar to that one used in [6] and [11] and it is based on Lemma 5 and the following fact known as *Lindley recursion*:

$$\hat{M}_{\theta} = (\gamma \hat{M}_{\theta} + \xi)^{+}$$
 (by law)

where \hat{M}_{θ} , γ , and ξ are independent r.v.'s,

$$\xi = X_1 - x$$
, $P(\gamma = 1) = e^{-q} = 1 - P(\gamma = 0)$.

Using this equation we check that the function

$$f(x) = \nu(x) = E(Q_{\nu}(M_{\theta} + x; M_{\theta}) I\{M_{\theta} + x \ge a_{\nu}\})$$

satisfies conditions (35) and (34) with $g(x) = (x^+)^{\nu}$ and so by Lemma 5 it implies the required inequality (38).

Condition (35) holds because the function $f(x) = \widehat{\nu}(x)$ is nonnegative increasing function as x increases and, therefore,

$$f(x) = E(I\{(\gamma M_{\theta} + \xi)^{+} + x \ge a_{\nu}\} Q_{\nu}((\gamma M_{\theta} + \xi)^{+} + x))$$

$$= e^{-q} E(I\{(M_{\theta} + \xi)^{+} + x \ge a_{\nu}\} Q_{\nu}((M_{\theta} + \xi)^{+} + x))$$

$$\ge e^{-q} E(I\{x \ge a^{*}, M + \xi < 0\} Q_{\nu}(M + \xi + x))$$

$$+ e^{-q} E(I\{M + \xi + x \ge a^{*}, M + \xi \ge 0\} Q_{\nu}(M + \xi + x))$$

$$= e^{-q} E(f(x + \xi)).$$

Condition (34) holds because for any x > 0 $f(x) = E(Q_{\nu}(M_{\theta} + x; M_{\theta})^{+}$ and so by Jensen's inequality and Lemma 4

$$f(x) > (E(Q_{\nu}(M_{\theta} + x; M_{\theta}))^{+} = (x^{+})^{\nu}) = q(x).$$

5. Examples.

(1) Discrete time case. Let ξ_1^+ have the density $pe^{-\lambda x}$, x>0; p>0, $0<\lambda<\infty$, $E(\xi_1)<0$. Then using martingale considerations like in the proof of Lemma 3 we can show that

$$P(M_{\infty} > a) = \frac{\lambda e^{-u_0 a}}{\lambda + u_0}, \quad P(M_{\infty} = 0) = \frac{u_0}{\lambda + u_0},$$

where u_0 is a positive root of the equation

$$E(e^{u\xi_1}) = 1.$$

By direct calculations one can get that for any ν and y>0

$$Q_{\nu}(y;\eta) = (2/3)^{\nu} \exp(3y/2)(-3\nu\Gamma[\nu,y/2] + \Gamma[1+\nu,3y/2]).$$

where $\Gamma[z,y]$ is the incomplete Gamma function.

(2) Brownian Motion case. Let $L_t = W_t - mt$, $g(x) = (x^+)^{\nu}$, $\nu > 0$. If m > 0 then $M_{\infty} = \sup_{t \ge 0} (W_t - mt) \sim \operatorname{Exp}(2m)$ and the Appell function for any ν and y > 0 is

$$Q_{\nu}(y; M) = y^{\nu - 1} \left(y - \frac{\nu}{2m} \right).$$

So, the optimal threshold is

$$a_{\nu} = \frac{\nu}{2m}.$$

If q > 0, m = 0 then $M_{\theta} = \sup_{s \leq \theta} (W_s) = |W_{\theta}|$ (by distribution), where $\theta \sim \operatorname{Exp}(q)$. We get

$$a_1 = E(|W_\theta|) = \sqrt{\frac{1}{2q}},$$

$$a_2 = E(|W_\theta|) + \sqrt{\text{var}(W_\theta)} = \sqrt{\frac{1}{2q}}(1 + \sqrt{2}),$$

and so on.

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