

Applications of Mogulskii, and Kurtz-Feng Large Deviation Results to Risk Reserve Processes with Aggregate Claims^{*}

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ABSTRACT

In this paper we examine the large deviations principle (LDP) for sequences of classic Cramér-Lundberg risk processes under suitable time and scale modifications, and also for a wide class of claim distributions including (the non-superexponential) exponential claims. We prove two large deviations principles: first, we obtain the LDP for risk processes on $D\mathbb{R}[0,1]$ with the Skorohod topology. In this case, we provide an explicit form for the rate function, in which the safety loading condition appears naturally. The second theorem allows us to obtain the LDP for Aggregate Claims processes on $D\mathbb{R}[0,\infty)$ with a different time-scale modification. As an application of the first result we estimate the ruin probability, and for the second result we work explicit calculations for the case of exponential claims.

Keywords: Large Deviations; Cramer-Lundberg Reserve Risk Processes; Probability Theory and Mathematical Statistics in Insurance; Stochastic Models for Claim Frequency; Claim Size and Aggregate Claims; Reserves

1. Introduction

There is a wide literature on Large Deviation Techniques and Applications. Relevant to this paper are results by Mogulskii (1993), [1] who proved a Large Deviations result for independent, identically distributed (i.i.d.) random variables with generating functions finite on a neighborhood of the origin. In [2], Lynch and Sethuraman gave large deviations results for stochastic processes with independent and stationary increments. The analysis was done on the space of functions of bounded variation on [0,1] endowed with the weak^{*}-topology. More general results were proved later, as Mogulskii and De Acosta did in [1,3,4] proving large deviations results for Lévy processes in very general settings.

For compound Poisson processes, Li and Pechersky [5], following results by Dobrushin and Pechersky [6], proved the LDP for multi-dimensional compound Poisson processes defined on $[0,\infty)$ with respect to the vague topology, and then strengthened it to the weak-uniform topology introduced in [6].

The LDP for (reserve dependent premium with de-

layed claims) risk process was studied by Ganesh, Massi and Torrisi (2007) [7,8]. They proved the LDP with respect to the uniform topology in the case of superexponential claims *i.e.*, claims for which the moment generating function is finite for every $\theta > 0$. Later, in [7], they illustrated the connection between risk processes and queues. They applied their large deviations result (valid only in the case of super-exponential claims) to obtain an approximation for the probability of ruin and to propose an importance sampling parameter for simulation. The super-exponential claims are an interesting but very particular case, since distributions such as gamma (including exponential), Negative Binomial (including geometric) claims, are not of this type.

One way to deal with large deviations for risk processes is by proving the LDP for a sufficiently similar zero-mean Lévy process, and then using the Contraction Principle. That is the general approach we follow for Theorem 1: we examine the LDP for a sequence of risk processes with respect to the Skorohod topology under suitable time and scale modifications. We follow Mogulskii's approach [1], whose results are based on Lynch and Sethuraman [2] to obtain an LDP for risk processes on $D\mathbb{R}[0,1]$, and give quite explicit forms for the rate

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function.

Another way to prove Large Deviations results is by first analyzing the compound Poisson component $\sum_{i=1}^{N_t} Z_i$, via working with random walks $\sum_{i=1}^{[t]} Z_i$ or their linearized counterpart, proving the LDP, and then dealing with the random time via exponential equivalence of the times. In this direction there is work by Feng and Kurtz, [9] and our Theorem 2 for aggregate claims processes. The difficulty with this approach is that a Poisson process could hardly be exponentially equivalent to a continuous one, and it becomes necessary to use a cumbersome change in time-space scale. We prove Large Deviation results for a wide class of claim distributions in-

claims. We get the LDP for aggregate claims processes on $D\mathbb{R}[0,\infty)$, with a suitable time-scale modification. Both results are LDP's with respect to the Skorohod topology induced by the Skorohod distance, but the first one is in $D\mathbb{R}[0,1]$ and the second is in $D\mathbb{R}[0,\infty)$.

cluding the non-super-exponential case of exponential

Although the vague topology is coarser than Skorohod's, Li and Pechersky's large deviations results do not imply ours (Theorem 1 below) because the space $D\mathbb{R}[0,\infty)$ is not reflexive, and non-trivial exponential tightness should be proved first. On the other hand, we do not work with super exponential claims: we only need the moment generating function to exist on an open neighborhood of the origin, and for this reason our result is more general.

Large Deviations techniques have been used to study ruin probabilities for risk process. A standard reference here is the book by Asmussen [10], references therein, and subsequent work by the author.

On a generalization of the model, Asmussen, Klüppelberg, and Mikosch, in [11,12], studied asymptotic results for the compound Poisson process when the size of the jumps has a heavy tail (the moment generating function of the claims is ∞ on the positive real numbers). In this case, the large deviations theory does not apply, the results are quite different, and that is not the subject of this paper.

The organization of the paper is as follows: First, we have one Section to state the basic notation, to describe previous results, and at the end we have a small discussion about the precise shape of the rate function: that is Section 2. In Section 3, we state the basic Hypotheses that are needed all along the work.

The main results, LDP Theorems 1 and 2, are stated in Sections 4 and 5, respectively. Both are proved in the same Section they are stated. Section 6 is devoted to the explicit calculations for the case of exponential claims. These calculations are combined in Corollary 1, and later used in Section 7 to estimate the probability of ruin for exponential claims, and also for more general claims.

2. Notation, Previous Results, and the Rate Function

For a random variable *Y* and $\theta \in \mathbb{R}$, we denote by $L_{Y}(\theta) = \mathbb{E}\left[e^{\theta Y}\right]$ the moment generating function for *Y*, whenever it exists; its logarithmic generating function is $\Lambda_{Y}(\theta) = \log \mathbb{E}\left[e^{\theta Y}\right]$, and

 $\Lambda_Y^*(t) = \sup_{\theta \in \mathbb{R}} \left\{ t \theta - \Lambda_Y(\theta) \right\}$ shall denote the Fenchel-Legendre transform of Λ_Y .

Let $(T_-, T_+) = Int \{ \theta | \Lambda_{\xi(1)}(\theta) < \infty \}$ denote the interior of the essential domain for the Laplace transform (and its log-generating function) of $Y = \xi(1)$, where ξ is a process to be specified.

Clearly, $-\infty \leq T_{-} \leq 0 \leq T_{+} \leq \infty$.

The classical risk process is given by

$$R(t) = u + ct - \sum_{k=1}^{N_t} Z_k,$$
 (1)

and the following will be our assumptions regarding this process,

1) $N_t = N(t)$ is a Poisson (λt) process, $\lambda > 0$, that models the number of claims received at time t.

2) Z, Z_1, Z_2, \cdots are non negative i.i.d. random variables with mean μ , independent of the process N_t . We shall always assume the moment generating function of Z, $L_Z(\theta)$ is finite for some $\theta > 0$ (not necessarily for all $\theta \in \mathbb{R}^+$). These variables represent the size of the claims. The compound Poisson term $\sum_{k=1}^{N_t} Z_t$ accounts for the aggregate claims.

3) u = R(0) > 0 is the initial capital or reserve.

4) *c* is positive. The term *ct* represents the (non random and linear) premiums the company charges.

It is usually required to have a safety loading condition $c - \lambda \mu > 0$ to assure ruin does not occur almost surely. We do not need that condition for the moment; however, it shall appear when we give the explicit form of the large deviations rate function.

For each bounded variation $f \in BV[0,1]$ with $f \in D\mathbb{R}[0,1]$ and f(0) = 0, let $f = f_0 + f_1 - f_2$ be its standard decomposition such that $f_0(0) = f_1(0) = f_2(0) = 0$, f_0 is absolutely continuous with respect to the Lebesgue measure (here denoted as $f_0 \ll m$), $f_1 - f_2$ is the Hahn-Jordan decomposition for the singular part of f with respect to the Lebesgue measure. Recall that f_1 and f_2 are hence non-decreasing, and each one is singular with respect to m (which shall be denoted $f_i \perp m, i = 1, 2$).

A standard representation for the characteristic function of a stationary process with independent increments $\xi(t)$ is

$$E\left[e^{\mathrm{i}\theta\xi(t)}\right] = e^{tK(\theta)},$$

with

$$K(\theta) = ia\theta - \frac{1}{2}b^2\theta^2 + \int \left(e^{i\theta x} - 1 - i\theta x \chi_{\{|x| \le 1\}}\right) \pi(dx).$$

A regularity condition on the measure π defined by the expression above, which will be needed in the main result (Theorem 1), is:

$$\lim_{n \to \infty} \frac{1}{n} \log \pi \left(\left(-\infty, -n \right) \right) = T_{-}, \tag{2}$$

and

$$\lim_{n\to\infty}\frac{1}{n}\log\pi([n,\infty])=-T_+$$

We shall apply Theorems 2.5 and 2.7 by Mogulskii [1] (see also [2]). We shall write here together all the results we apply, specialized to our settings, and use the notation defined above:

Proposition 1. (Mogulskii) Let $\xi(t)$ be a stochastic process defined on $[0,\infty)$ with values in \mathbb{R} Assume it has stationary independent increments, and also suppose $\mathbb{E}[\xi(1)] = 0, \mathbb{E}[\xi(1)^2] > 0.$

If
$$-\infty \le T_{-} < 0 < T_{+} \le \infty$$
, then ULDP: $\frac{1}{n} \xi(nt)$ satis-

fies the upper large deviations principle with respect to the completed Skorohod topology, with good rate function I. LLDP: If, additionally, the regularity condition

(2) is satisfied, then the same sequence $\frac{1}{n}\xi(nt)$ satis-

fies the lower large deviations principle with respect to the completed Skorohod topology, with the same rate function I. The rate function I is

$$I(f) = \begin{cases} \int_{0}^{1} \Lambda_{\xi(1)}^{*} (\dot{f}_{0}(t)) dt + T_{+} f_{1}(1) + |T_{-}| f_{2}(1) \\ \text{whenever } f \in BV; f(0) = 0, \\ +\infty \text{ otherwise,} \end{cases}$$
(3)

where

$$f = f_0 + (f_1 - f_2), f_0 \ll m, (f_1 - f_2) \perp m;$$

$$f_0(0) = f_1(0) = f_2(0) = 0,$$

 f_1, f_2 are non-decreasing, and with the understanding that $\infty \times 0 = 0$. If $T_+ = \infty$, then the ULDP holds also with the uniform topology.

A Remark on the Rate Function

Remark 1. *The rate function* I *that appears in Mogulskii* [1] *has two misprints. This can be verified in* [2]. *It says*

$$I(f) = \int_{0}^{1} \Lambda_{\xi(1)}^{*} (f_{0}(t)) dt + T_{+} f_{2}(1) + |T_{-}|| f_{2}(1)|$$

whereas it should say

$$I(f) = \int_{0}^{1} \Lambda_{\xi(1)}^{*} (\dot{f}_{0}(t)) dt + T_{+} f_{1}(1) + |T_{-}|| f_{2}(1)|.$$

Indeed, for this rate function, Mogulskii refers to the paper by Lynch and Sethuraman ([2]), and according to the latter, the value for I(f) is

$$I(f) = \int_{0}^{1} \Lambda_{\xi(1)}^{*} (\dot{f}_{0}(t)) dt + C_{1} f_{1}(1) + C_{2} f_{2}(1)$$

where $C_1 = \lim_{a \to \infty} \frac{\Lambda^*_{\xi(1)}(a)}{a}, C_2 = \lim_{a \to -\infty} \frac{\Lambda^*_{\xi(1)}(a)}{|a|}$ and

 $\Lambda_{\xi(1)}^*(a) = \sup_{t \in \mathbb{R}} \left\{ at - \Lambda_{\xi(1)}(t) \right\}.$ Without loss of generality, suppose that $\Lambda_{\xi(1)}(T_+) = \infty$. Note that for $t > T_+, at - \Lambda_{\xi(1)}(t) = -\infty$, hence

$$\Lambda_{\xi(1)}^{*}\left(a\right) = \sup_{t < T_{+}} \left\{at - \Lambda_{\xi(1)}\left(t\right)\right\}.$$

If $t_{0} < T_{+}$, then $\Lambda(t_{0}) < \infty$, and since
$$\frac{\Lambda_{\xi(1)}^{*}\left(a\right)}{a} = \sup_{t < T_{+}} \left\{t - \Lambda_{\xi(1)}\left(t\right)/a\right\} \ge t_{0} - \Lambda_{\xi(1)}\left(t_{0}\right)/a$$

making a tend to ∞ , we obtain

$$\liminf_{a\to\infty} \frac{\Lambda^*_{\xi(1)}(a)}{a} \ge t_0.$$

Now, if we let $t_0 \rightarrow T_+$, we conclude

$$\lim_{a \to \infty} \inf \frac{\Lambda_{\xi(1)}^*(a)}{a} \ge T_+.$$
(4)

On the other hand, for every $\epsilon > 0$, and $a \in \mathbb{R}$, there is $t_a < T_+$ such that

$$\frac{\Lambda_{\xi(1)}^*\left(a\right)}{a} - \epsilon < t_a - \Lambda_{\xi(1)}\left(t_a\right) / a \leq \frac{\Lambda_{\xi(1)}^*\left(a\right)}{a}.$$

taking limits as $a \to \infty$, we conclude

$$\lim_{a \to \infty} \sup \frac{\Lambda_{\xi(1)}^*(a)}{a} \le T_+.$$
 (5)

In view of formulas (4) and (5) we conclude

$$\lim_{a\to\infty}\frac{\Lambda^*_{\xi(1)}(a)}{a}=T_+$$

With a similar argument we obtain

$$\lim_{a \to -\infty} \frac{\Lambda^*_{\xi(1)}(a)}{|a|} = -T_{-}$$

Finally, since f_2 is non-decreasing and $f_2(0) = 0$,

$$f_{1}(1)\lim_{a\to\infty}\frac{\Lambda_{\xi(1)}^{*}(a)}{a} + f_{2}(1)\lim_{a\to\infty}\frac{\Lambda_{\xi(1)}^{*}(a)}{|a|}$$
$$= T_{+}f_{1}(1) - T_{-}f_{2}(1) = T_{+}f_{1}(1) + |T_{-}|f_{2}(1).$$

In the next Section we shall state the hypotheses we need.

3. Statement of Hypotheses

Let Z, Z_1, Z_2, \cdots be nonnegative independent and identically distributed random variables with mean μ and second moment $\mathbb{E}[Z^2] = \alpha > 0$ (*i.e.* Z is non-degenerated).

Let N(t) be a Poisson process with parameter $\lambda > 0$, defined on the same probability space and independent of the Z, Z_1, Z_2, \cdots All our random variables Z, and Poisson processes N(t) will be as just stated, unless otherwise noted.

If Λ_z is finite in a neighborhood of the origin, we say Condition 1 is satisfied:

Condition 1. There is $0 < \theta_1 \le \infty$ such that $\Lambda_Z(\theta) = \log \mathbb{E}\left[e^{\theta Z}\right] < \infty$ for $\theta < \theta_1$.

Remark 2. Finiteness of the function $\Lambda_z(\theta)$ is equivalent to that of the moment generating function $L_z(\theta)$. Condition 1 is satisfied if $\Lambda_z(\theta)$ is finite for some $\theta > 0$ (the negative part works due to Z being non-negative).

Notice also that this condition implies μ, α , and every moment of the variables $Z, (Z_k)_{k \in \mathbb{N}}$ are finite.

We shall fix $0 < \theta_0 \le \infty$ to be the maximum possible by letting $\theta_0 = \sup \left\{ \theta | \mathbb{E} \left[e^{(\theta Z)} \right] < \infty \right\}.$

To have the LLDP in the topology we want, we require to ask an additional regularity condition:

Condition 2.

$$\lim_{n\to\infty}\frac{1}{n}\log(1-F_Z(n))\geq-\theta_0.$$

Remark 3. All the conditions on Z are satisfied if Z is, for example, exponentially distributed. This important case, for which many calculations can be made explicit, is discussed in Sections 6 and 7.

4. Large Deviations for the Risk Process on $D\mathbb{R}[0,1]$

Now we are ready to state the LDP in the space $D\mathbb{R}[0,1]$ with the Skorohod topology:

Theorem 1. Let
$$R(t) = u + ct - \sum_{k=1}^{N(t)} Z_k$$
 be a risk pro-

cess as defined in (1). Under conditions 1 and 2, the sequence of processes

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$$R_n = u + ct - \frac{1}{n} \sum_{k=1}^{N(nt)} Z_k, \qquad (6)$$

satisfies the large deviations principle in the space $D\mathbb{R}[0,1]$ with the completed topology induced by the Skorohod distance, and rate function \hat{I} defined as

$$\hat{I}(g) = \begin{cases} \int_{0}^{1} \Lambda_{\xi(1)}^{*} \left(c - \lambda \mu - \dot{g}_{0}(t) \right) dt + T_{+} g_{2}(1) \\ \text{whenever } g \in BV; g(0) = u, \\ +\infty \text{ otherwise,} \end{cases}$$
(7)

where g is decomposed as

$$g - u = g_0 + (g_1 - g_2), g_0(0) = g_1(0) = g_2(0) = 0,$$

$$g_0 \ll m, (g_1 - g_2) \perp m,$$

and g_1, g_2 are non-decreasing.

Remark 4. The form of the rate function is given by

$$\hat{I}(g) = I\left(u + (c - \lambda \mu) \cdot -g(\cdot)\right), \quad (8)$$

with *I* as in expression (3). We also notice that, in this case, $T_{-} = -\infty$.

Proof of Theorem 1.

Let $\xi(t) = \sum_{k=1}^{N(t)} Z_k - \lambda \mu t$ be the compound Poisson process. Using the assumptions about Z_1, Z_2, \cdots and N we conclude that $\xi(t)$ has stationary and independent increments, $\mathbb{E}[\xi(t)] = 0$ and that

$$\mathbb{E}\left[\xi(1)^{2}\right] = \lambda\alpha + 2\mathbb{E}\left[\sum_{1 \leq i < j \leq N} Z_{i}Z_{j}\right] - \lambda^{2}\mu^{2} = \lambda\alpha > 0.$$

The Laplace transform $L_{\xi(1)}(\theta)$ of $\xi(1)$, is

$$\mathbb{E}\left[e^{\theta\xi(1)}\right]$$

= exp{ $-\theta\lambda\mu$ } $\mathbb{E}\left(\mathbb{E}\left(\exp\left\{\theta\sum_{k=1}^{n}Z_{k}\right\}|N=n\right)\right)$
= exp{ $-\lambda\left[\theta\mu-\left(L_{Z}\left(\theta\right)-1\right)\right]$ },

We notice that the log-Laplace

 $\Lambda_{\xi(1)}(\theta) = -\lambda \Big[\theta \mu - (L_z(\theta) - 1) \Big] \text{ is finite as long as}$

 Λ_{z} is finite. By Condition 1, we can guarantee that $\Lambda_{\xi(1)}(\theta)$ is finite for $\theta < \theta_{0}$, in fact both, Λ_{z} and $\Lambda_{\xi(1)}$, have the same support. By similar calculations,

$$\mathbb{E}\left[e^{i\lambda_{z}(t)}\right]$$

= exp{(-i\theta t\lambda \mu - t\lambda)} $\sum_{n=0}^{\infty} \exp\left(n\Lambda_{z}(i\theta)\right) \frac{(t\lambda)^{n}}{n!}$
= exp{-t\lambda [i\theta\mu - (exp{\lambda_{z}(i\theta)}-1)]} = e^{iK(\theta)}

where

$$K(\theta) = -i\lambda\theta\mu + \int (e^{i\theta x} - 1)d[\lambda F_{Z}(x)],$$

which means that the Lévy measure π of this process is λ times the distribution of the claims Z. We notice that if $(A_n)_{n \in \mathbb{N}}$ is a sequence of sets, then

$$\lim_{n \to \infty} \frac{1}{n} \log \pi(A_n) = \lim_{n \to \infty} \frac{1}{n} \log \left(\lambda \mathbb{P} [Z \in A_n] \right)$$
$$= \lim_{n \to \infty} \frac{1}{n} \log \left(\mathbb{P} [Z \in A_n] \right).$$

Regarding the Lévy measure π , we see that for each $n \in \mathbb{N}, \pi[(-\infty, -n)] = \lambda \mathbb{P}[Z \le -n] \equiv 0$, therefore

 $\lim_{n \to \infty} \frac{1}{n} \log \pi \left[\left(-\infty, n \right) \right] = -\infty \equiv T_{-}, \text{ and if } \theta_1 < \theta_0, \text{ by}$ the Markov inequality,

 $\frac{1}{n}\log\mathbb{P}[Z\geq n]\leq -\theta_1+\frac{1}{n}\Lambda_Z(\theta_1).$

Taking the limit in the previous inequality as *n* approaches to ∞ , and using $\Lambda_Z(\theta_1) < \infty$ gives us $\limsup_{n \to \infty} \frac{1}{n} \log \mathbb{P}[Z \ge n] \le -\theta_1$ and since this assertion is valid for each $\theta_1 < \theta_{0,2}$, we obtain

$$\lim_{n\to\infty}\sup\frac{1}{n}\log\mathbb{P}[Z\geq n]\leq-\theta_0.$$

We observe that by Condition 2,

$$\lim_{n \to \infty} \inf \frac{1}{n} \log \pi [(n, \infty)]$$
$$= \lim_{n \to \infty} \inf \frac{1}{n} \log \mathbb{P} [Z \ge n] \ge -\theta_0$$

Since $T_+ = \theta_0$, we conclude that

$$\liminf_{n\to\infty} \frac{1}{n} \log \pi \left[\left(n, \infty \right) \right] = -T_+$$

which is the regularity Condition 2. In summary, the process $\xi(t)$ satisfies the hypotheses of Proposition 1, therefore the process

$$X_n = \frac{1}{n} \xi(nt) = \frac{1}{n} \sum_{k=1}^{N(nt)} Z_k - \lambda \mu t,$$

satisfies the LDP with respect to the Skorohod topology with rate function provided by 3.

We now consider $(D\mathbb{R}[0,1],s)$ the metric space $D\mathbb{R}[0,1]$ endowed with the completed topology induced by the Skorohod distance and we also consider the functional $F: D\mathbb{R}[0,1] \rightarrow D\mathbb{R}[0,1]$ defined by

$$F(f)(t) = u + (c - \lambda \mu)t - f(t).$$

To verify continuity of *F* on this specific topology, take $\epsilon > 0$, and x_n with $\lim_{n \to \infty} x_n = x$ in the Sko-

rohod topology. Recall that the Skorohod distance in $D\mathbb{R}[0,1]$ is given by

$$s(f,g) = \inf \left\{ \max \left\{ \|f - g \circ r\|, \|I - r\| \right\} \mid r : [0,1] \rightarrow [0,1] \right\}$$

is non-decreacing and continuous

Then, there is $N \in \mathbb{N}$ so that for each $n \ge N$, we can find $r^* = r^*(\epsilon, n)$, where $r^* : [0,1] \rightarrow [0,1]$ is an non-decreasing continuous function for which

$$s(x_n, x) \leq \max\left\{ \left\| x_n - x \circ r^* \right\|, \left\| I - r^* \right\| \right\} < \epsilon.$$

Now,

$$F(x_{n})(t) - F(x) \circ r^{*}(t)$$

= $u + (c - \lambda \mu)t - x_{n}(t)$
- $[u + (c - \lambda \mu)r^{*}(t) - x \circ r^{*}(t)]$
= $(c - \lambda \mu)(t - r^{*}(t)) - [x_{n}(t) - x \circ r^{*}(t)]$

therefore

$$s(F(x_n), F(x)) \le \max\left\{ |c - \lambda \mu| \|I - r^*\|, \|I - r^*\|\right\}$$
$$+ \max\left\{ \|x_n - x \circ r^*\|, \|I - r^*\|\right\}$$
$$\le \max\left\{ |c - \lambda \mu| \epsilon, \epsilon \right\} + \epsilon,$$

hence $F(x_n)$ tends to F(x). Therefore F is continuous with respect to the Skorohod topology. By the Contraction Principle,

$$F(X_n)(t) = u + (c - \lambda \mu)t - \frac{1}{n}\xi(nt)$$
$$= u + ct - \frac{1}{n}\sum_{k=1}^{N(nt)}Z_k$$

satisfies the LDP with rate function given by

$$\hat{I}(g) = \inf \{I(h) : F(h) = g\}$$
$$= I(F^{-1}(g)) = I(u + (c - \lambda\mu) - g(\cdot)).$$

If we consider the Hahn-Jordan decomposition for $g - u = g_0 + g_1 - g_2$, and we observe that $(c - \lambda \mu)t$ is absolutely continuous with respect to the Lebesgue measure *m*, we conclude that the Hahn-Jordan decomposition $f_0 + f_1 - f_2$ for the function $u + (c - \lambda \mu)t - g(t)$ is $f_0(t) = (c - \lambda \mu)t - g_0(t), f_1(t) = g_2(t)$, and

 $f_2(t) = g_1(t)$. Evaluating \hat{I} and using (3), we obtain formula (7), this concludes the proof of Theorem 1. \Box

5. Large Deviations for the Aggregate Claims Process on $D\mathbb{R}[0,\infty)$

In this Section we prove the LDP for the process of aggregate claims, on a different space, with another

time-scale parametrization, and using very different results.

Theorem 2. Assume Condition 1, but now let us denote the Poisson process as \hat{N} , with parameter λ , defined on the same probability space and independent of the $(Z_k)_{k\in\mathbb{N}}$

Let
$$N_n(t) = \left[\frac{\hat{N}(\sqrt{nt})}{\sqrt{n}}\right]$$
. Then, the sequence of pro-

cesses

$$X_n = \frac{1}{n} \sum_{k=1}^{N_n(nt)} Z_k$$

satisfies the large deviations principle in the space $D\mathbb{R}[0,\infty)$ with the topology induced by the Skorohod distance.

We will use results from Feng and Kurtz [9], Lemma 4.9 and Theorem 10.1, that we compile in next two Propositions:

Proposition 2. For each n, let \hat{X}_n be a process in $D_E[0,\infty)$. Let Λ_n . be a nonnegative, nondecreasing process $D\mathbb{R}[0,\infty)$ valued, independent of X_n . Define

 $X_n = \hat{X}_n \left(\Lambda_n \left(t \right) \right).$

Suppose that for each t > 0 and $\eta > 0$,

$$\lim_{n \to \infty} \frac{1}{n} \log P \left[\sup_{s \le t} \left| s - \Lambda_n(s) \right| > \eta \right] = -\infty.$$
(9)

If \hat{X}_n is exponentially tight, then the LDP holds for X_n if and only if it holds for \hat{X}_n .

Proposition 3. Let A_n be given by

$$A_{n}g(x) = n \int_{\mathbb{R}^{d}} \left(g\left(x + \frac{1}{n}z\right) - g(x) - \frac{1}{n}z \cdot \nabla g(x) \right) \eta(dz) + \frac{1}{2n} \sum_{i,j} a_{i,j} \partial_{i} \partial_{j}g(x) + b \cdot \nabla g(x),$$

for $g \in D_d$. Suppose that X_n is a solution of the martingale problem for A_n and that the LDP holds for $X_n(0)$ in \mathbb{R}^d with a good rate function I_0 and rate transform Λ_0 . Then the LDP holds for $\{X_n\}$ in $D\mathbb{R}[0,\infty)$ with good rate function:

$$I(x) = \sup_{\{t_i\} \subset \Delta_x^c} I_{t_1, \dots, t_m} (x(t_1), \dots, x(t_m)), \qquad (10)$$

where

$$I_{t_{1},\cdots,t_{m}}(x_{1},\cdots,x_{m}) = \sup_{f_{1},\cdots,f_{k}\in D_{d}} \left\{ f_{1}(x_{1}) + \cdots + f_{k}(x_{k}) - \Lambda_{0} \left(V(t_{1}) (f_{1} + V(t_{1} - t_{1}) (f_{2} + \cdots + V(t_{k} - t_{k-1}) f_{k}) \cdots) \right) \right\}$$

And if $x \in D\mathbb{R}[0,\infty) - C\mathbb{R}[0,\infty)$, then $I(x) = \infty$. Proof of Theorem 2.

Let
$$\hat{X}_n = \frac{1}{n} \sum_{k=1}^{[nt]} Z_k$$

Due to the fact that $\Lambda_z(\theta) = \log \mathbb{E}\left[e^{\theta Z}\right] < \infty$ for some $\theta > 0$, Proposition 3 can be applied to sequences of processes with independent increments such as \hat{X}_n —as discussed in Section 10.2 of Feng and Kurtz [9]. Therefore \hat{X}_n satisfies the LDP with a good rate function defined as in (10). Since this rate function is good, and our space is Polish, then the sequence $\{\hat{X}_n\}$ is exponentially tight in $D\mathbb{R}[0,\infty)$.

Now, let
$$\Lambda_n(t) = \frac{1}{n^{3/2}} N(n^{3/2}t)$$
. By the Mason,

Shorack, Wellner inequality [13] (p. 545) we have,

$$P\left[\sup_{s \le t} \left| s - \Lambda_n(s) \right| > \eta \right]$$

$$\leq P\left[\sup_{s \le t} \left| n^{3/2} s - N(n^{3/2} s) \right| > n^{3/2} \eta \right]$$

$$\leq 2e^{-n^{3/2} \cdot \psi(\eta/t) \frac{\eta^2}{2t}},$$

where $\psi(t) = 2t^{-2} ((1+t)\ln(1+t) - t)$, hence

$$\lim_{n \to \infty} \frac{1}{n} \log P \left[\sup_{s \le t} \left| s - \Lambda_n(s) \right| > \eta \right]$$
$$\leq \lim_{n \to \infty} -n^{1/2} \psi(\eta/t) \cdot \frac{\eta^2}{2t} = -\infty$$

which is (9). Applying Proposition 2 we conclude that

$$X_{n} = \hat{X}_{n} \left(\Lambda_{n} \left(t \right) \right) = \frac{1}{n} \sum_{k=1}^{N_{n}(nt)} Z_{k}$$

satisfies the LDP with a good rate function. \Box

6. Exponential Claims

In this Section we give the explicit calculations when the law of the claims is exponential, and we also find the particular shape for the rate function. We have the following Corollary to Theorem 1:

Corollary 1. Consider a classical risk process as defined in (1) with exponentially distributed claims:

$$R(t) = u + ct - \sum_{k=1}^{N(t)} Z_k, \text{ with } Z \sim \exp(\beta), \text{ and } \beta > 0.$$

The sequence of processes

$$R(t) = u + ct - \sum_{k=1}^{N(t)} Z_k, \qquad (11)$$

satisfies the LDP in the space $D\mathbb{R}[0,1]$ with the completed topology induced by the Skorohod distance, and rate function given by

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$$I(g) = \begin{cases} \int_{0}^{1} \left(\sqrt{\beta(c - \dot{g}_{0}(t))} - \sqrt{\lambda}\right)^{2} dt + \beta g_{2}(1) \\ \text{whenever } g \in BV; g(0) = u, \\ +\infty, \text{otherwise}, \end{cases}$$
(12)

with

$$g - u = g_0 + g_1 - g_2, g_0(0) = g_1(0) = g_2(0) = 0,$$

$$g_0 \perp m, g_1 - g_2 \perp m, g_1, g_2$$

non-decreasing, and $\dot{g}_0(t) \leq c$.

Proof. The Moment Generating function in this case is

$$L_{\mathrm{Z}}(\theta) = \begin{cases} \frac{\beta}{\beta - \theta} & \text{if } \theta < \beta \\ \infty & \text{if } \theta \ge \beta, \end{cases}$$

hence $\log \mathbb{E}\left[e^{\theta Z}\right] < \infty$ if and only if $\theta < \beta$. Therefore, Condition 1 holds. In this case

$$(T_{-},T_{+})=(-\infty,\beta)=\left\{\theta|\log \mathbb{E}\left[e^{\theta Z}\right]<\infty\right\}.$$

Also,

$$\lim_{n\to\infty}\frac{1}{n}\log(1-F_Z(n)) = \lim_{n\to\infty}\frac{1}{n}\log e^{-\beta n} = -\beta,$$

so Condition 2 also holds.

Applying Theorem 1, we conclude that R_n satisfies the LDP with rate function given by (7). It remains to show that the rate function has the explicit form (12). Let us calculate, for $x < \beta$,

$$L_{\xi(1)}(x) = \exp\left\{-\frac{\lambda}{\beta}x\right\} \mathbb{E}\left[\exp\left\{x\sum_{k=1}^{N(1)}Z_{k}\right\}\right]$$
$$= \exp\left\{-\frac{\lambda}{\beta}x\right\} \mathbb{E}\mathbb{E}\left[\exp\left\{x\sum_{k=1}^{N(1)}Z_{k}\right\} \middle| N(1)\right]$$
$$= \exp\left\{\frac{\lambda x^{2}}{\beta(\beta-x)}\right\}$$

Now

$$\Lambda_{\xi(1)}^{\star}(t) = \sup_{x \in \mathbb{R}} \left\{ tx - \Lambda_{\xi(1)}(x) \right\} = tx_0 - \frac{\lambda x_0^2}{\beta(\beta - x_0)}$$

with $x_0 = \beta \left| 1 - \sqrt{\frac{\lambda}{\beta t + \lambda}} \right|$, if $t \ge -\frac{\lambda}{\beta}$. Therefore,

$$\Lambda_{\xi(1)}^{\star}(t) = \begin{cases} \left(\sqrt{\beta t + \lambda} - \sqrt{\lambda}\right)^{2}, \text{ if } t \ge -\frac{\lambda}{\beta} \\ \infty \text{ if } t < -\frac{\lambda}{\beta} \end{cases}$$
(13)

Here, again the Hahn-Jordan decomposition

$$f_{0} + f_{1} - f_{2} \text{ for the function } u + (c - \lambda\mu)t - g(t) \text{ is}$$

$$f_{0}(t) = (c - \lambda\mu)t - g_{0}(t), f_{1}(t) = g_{2}(t), \text{ and}$$

$$f_{2}(t) = g_{1}(t). \text{ Therefore,}$$

$$\Lambda_{\xi(1)}^{*}(f_{0}(t)) = \Lambda_{\xi(1)}^{*}(c - \lambda\mu - \dot{g}_{0}(t))$$

$$= \left(\sqrt{\lambda + \beta(c - \lambda/\beta) - \beta\dot{g}_{0}(t)} - \sqrt{\lambda}\right)^{2}$$

$$= \left(\sqrt{\beta c - \beta\dot{g}_{0}(t)} - \sqrt{\lambda}\right)^{2}$$

provided $c \ge \dot{g}_0(t)$, (otherwise $\Lambda^*_{\xi(1)}(\dot{f}_0(t)) = \infty$.). Since $T_- = -\infty$, the value

 $\int_{0}^{1} \Lambda_{\xi(1)}^{*} (f_{0}(t)) dt + T_{+} f_{1}(1) + |T_{-}| |f_{2}(1)| \text{ in the rate fun-}$ ction (7) provides us information only when $f_2(1) = 0$, and since f_2 is non-decreasing and $f_2(0) = 0$, this implies that $f_2 \equiv 0$, in our case $f_2 = g_1 \equiv 0$. Finally, $T_{+} = \beta$, hence the rate function *I* takes the form (12). This proves Corollary 1. □

7. Estimating the Ruin Probability of the Process R_n

In this section we give an upper bound for the Probability of Ruin for the process R_n as studied in Section 6 and in Corollary 1; i.e. when the claims are exponentially distributed with parameter β .

Theorem 3. For the process R_n defined in Corollary 1, we have that

$$P\left[\inf_{t\geq0}R_{n}\left(t\right)\leq0\right]$$

$$\leq\exp\left(-n\left[\beta u-2\sqrt{\lambda\beta}\sqrt{c+u}\right]\right)$$
(14)

for sufficiently large n. The estimate is meaningful when

$$u > \frac{2\left(\lambda + \sqrt{\lambda\left(\lambda + c\beta\right)}\right)}{\beta}.$$
 (15)

Proof of Theorem 3.

By Corollary 1, R_n satisfies the LDP with good rate function given by (12), therefore

$$\lim_{n\to\infty}\sup\frac{1}{n}\log P\left[\inf_{t\geq 0}R_n(t)\leq 0\right]\leq -I(g^*),$$

for some function g^* in the essential domain of I, hence g^* satisfies:

 $g^* = u + g_0 - g_2$ where $g_0(0) = g_2(0) = 0$,

 $g_0 \ll m, g_2 \perp m, g_2$ non-decreasing, there is a first time $t_0 > 0$ when $g^*(t_0) \le 0$.

By definition of t_0 , we have

 $0 \le \lim_{t \to t_0} (u + g_0(t) - g_2(t))$, since g_0 is continuous and g_2 is non-decreasing, we conclude that

$$0 \le g_2(t_0 -) \le u + g_0(t_0).$$

Therefore

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$$\begin{split} &\lim_{n \to \infty} \sup \frac{1}{n} \log P \bigg[\inf_{t \ge 0} R_n(t) \le 0 \bigg] \\ &\le -\int_0^{t_0} \bigg(\sqrt{\beta c - \beta \dot{g}_0(t)} - \sqrt{\lambda} \bigg)^2 \, \mathrm{d}t - \beta g_2(t_0) \\ &\le -\beta c t_0 + \beta g_0(t_0) - \lambda t_0 \\ &+ 2\sqrt{\lambda\beta} \sqrt{c t_0 - g_0(t_0)} - \beta g_2(t_0) \\ &\le 2\sqrt{\lambda\beta} \sqrt{c t - \mu} - \beta \mu, \end{split}$$

the second inequality is due to Jensen's inequality. Note that $2\sqrt{\lambda\beta}\sqrt{c+u} - \beta u < 0$ if and only if u is above the largest root of the polynomial $\beta x^2 - 4\lambda x - 4\lambda c$, such root is precisely the right hand side of inequality (15). This proves Theorem 3. \Box

Remark 5. The estimate of the probability ruin provided here for the process $R_n = u + ct - \frac{1}{n} \sum_{k=1}^{N(nt)} Z_k$ is $\exp\left(-n\left\lceil \beta u-2\sqrt{\lambda\beta}\sqrt{c+u}\right\rceil\right)$, whereas the exact pro-

bability of ruin for the process $R = u + ct - \sum_{k=1}^{N(t)} Z_k$, (see,

for instance, Section 1.2 of [14]) is

 $\frac{\lambda}{c\beta} \exp\left(-\left(1-\frac{\lambda}{c\beta}\right)u\right).$ Note that the exponent in our

estimation also behaves as a polynomial on u of first degree when *u* is large.

If the claims are not exponentially distributed, we can still make the following observation regarding the probability of ruin.

Corollary 2. Consider the rate function for the process in Theorem 1, \hat{I} , and let g^* be the dominating point of $\{g \in D\mathbb{R}[0,1] : g \in BV, \inf_{t \in [0,1]} g(t) \le 0, g(0) = u\}$ with

respect to \hat{I} . Consider also the decomposition

 $g^{*} = u + g_0 - g_2$, where $g_0(0) = g_2(0) = 0, g_0 \ll m, g_2 \perp m$, and with g_2 nondecreasing.

If x_t is a solution of

$$E\left[Ze^{x_{r}Z}\right] = \frac{c - \dot{g}_{0}\left(t\right)}{\lambda},$$
(16)

then, for the process R_n defined in (6), we have

$$P\left[\inf_{t\geq0}R_{n}(t)\leq0\right]$$

$$\leq\exp\left(-n\left[\int_{0}^{1}\left(c-\lambda\mu-\dot{g}_{0}(t)\right)\cdot x_{t}dt+T_{+}g_{2}(1)\right]\right)$$
(17)

for sufficiently large *n*.

Recall that the LDP proved in Theorem 1 holds for claims with moment generating functions much less restrictive than the super-exponential case.

Proof of Corollary 2. By Corollary 1, R_n satisfies the LDP with good rate function \hat{I} given by (7), therefore

$$\lim_{n\to\infty}\sup\frac{1}{n}\log P\left[\inf_{t\geq 0}R_n(t)\leq 0\right]\leq -\hat{I}(g^*),$$

where the function g^* is as in the statement above. Since

$$\Lambda_{\xi(1)}^{*}(t) = \sup_{x \in \mathbb{R}} \left\{ tx + \lambda \mu x + \lambda - \lambda L_{Z}(x) \right\},\$$

and since x_t is as in (16), then

$$\Lambda_{\xi(1)}^{*}\left(c-\lambda\mu-\dot{g}_{0}\left(t\right)\right)=\lambda+\left(c-\dot{g}_{0}\left(t\right)\right)\cdot x_{t}-\lambda L_{Z}\left(x_{t}\right).$$

Finally,

$$\begin{split} &\lim_{n \to \infty} \sup \frac{1}{n} \log P \Big[\inf_{t \ge 0} R_n(t) \le 0 \Big] \\ &\le -\int_0^1 \Lambda_{\xi(1)}^* \Big(c - \lambda \mu - \dot{g}_0(t) \Big) dt - T_+ g_2(1) \\ &= \int_0^1 \Big[\lambda \Big(L_Z(x_t) - 1 \Big) - x_t \cdot \big(c - \dot{g}_0(t) \big) \Big] dt - T_+ g_2(1) \\ &\le \int_0^1 \Big(\lambda \mu - c + \dot{g}_0(t) \Big) \cdot x_t dt - T_+ g_2(1) \end{split}$$

For the second inequality we used the fact that $e^x \leq x+1$. \Box

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