Lecture 20

Risk theory and premium principles



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Risk theory is also called ruin theory.

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We will mainly study the most basic model, but will also give some directions of extensions.

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Using these we construct the surplus process:

$$X_t = x + pt - \sum_{i=1}^{N_t} Y_i.$$

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$$\psi(x,T) = P\left(\inf_{0 \le t \le T} X_t < 0\right)$$

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denote the probability of being ruined before time T, and let

$$\psi(x) = P\left(\inf_{t\geq 0} X_t < 0\right)$$

denote the probability of ultimate ruin.

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Goal

Calculate $\psi(x)$ in this model.

First of all we note that

$$X_0 = x$$
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= x + (p - \lambda \mu)t.

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This is called the net profit condition (NPC) and it means that we have a strictly positive drift in the expected value of X_t .

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This is called the net profit condition (NPC) and it means that we have a strictly positive drift in the expected value of X_t .

Note that p is the income per time unit for the insurance company, and $\lambda\mu$ is the cost per time unit for the company; it follows that $p - \lambda\mu$ is the net profit.

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The idea is to study the stochastic process X_t under the interval $[0, \min(T_1, h)]$, where T_1 is the first jump time of the Poisson process and h > 0.

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The key observaton is that on $[0, \min(T_1, h)]$

$$\psi_{c}(x) = \psi_{c}(x+ph) \cdot e^{-\lambda h} + \int_{0}^{h} \left[\int_{0}^{x+ph} \psi_{c}(x+pt-y) \frac{1}{\mu} e^{-y/\mu} dy \right] \lambda e^{-\lambda t} dt$$

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- The first term represents no jumps on [0, h].
- The second term represents the fact that the first jump will not ruin us.

We can rewrite this equation as

$$p\frac{\psi_c(x+ph)-\psi_c(x)}{ph} = \frac{1-e^{-\lambda h}}{h}\psi_c(x+ph)$$
$$-\frac{\lambda}{\mu}\cdot\frac{1}{h}\int_0^h \left[\int_0^{x+pt}\psi_c(x+pt-y)e^{-y/\mu}dye^{-\lambda t}\right]$$

(We have reshuffled and then divided by h.)

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Now let $h \downarrow 0$.

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This yields

$$p\psi_c'(x) = \lambda\psi_c(x) - \frac{\lambda}{\mu}\int_0^x\psi_c(x-y)e^{-y/\mu}dy.$$

Here we used that

$$\lim_{h\downarrow 0}\frac{1}{h}\int_0^h f(t)dt = f(0)$$

if f is nice.

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In order to solve the equation for $\psi_c(x)$ we start by rewriting the integral.

$$\int_0^x \psi_c(x-y) e^{-y/\mu} dy = \{ \text{Replace } x - y \text{ with } u \}$$

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The Cramér-Lundberg model (8)

The equation for $\psi_c(x)$ now becomes

$$p\psi'_{c}(x) = \lambda\psi_{c}(x) - \frac{\lambda}{\mu}e^{-x/\mu}\int_{0}^{x}\psi_{c}(u)e^{u/\mu}du.$$



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Differenting with respect to x yields (after simplification!)

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This is a first-order ODE in $\psi'_c(x)$,

$$[\psi_c'(x)]' + \left(\frac{1}{\mu} - \frac{\lambda}{p}\right)\psi_c'(x) = 0.$$

The solution is

$$\psi_c'(x) = Ae^{-(1/\mu - \lambda/p)x},$$

where A is a constant.



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Integrating once more yields

$$\psi_c(x) = B_1 + B_2 e^{-(1/\mu - \lambda/p)x},$$

where B_1, B_2 are two constants.

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How can we find B_1 and B_2 ?



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Recall the NPC:

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How can we find B_1 and B_2 ?

Recall the NPC:

$$p-\lambda\mu>0 \quad \Leftrightarrow \quad \frac{1}{\mu}-\frac{\lambda}{p}>0.$$

Hence

$$R:=\frac{1}{\mu}-\frac{\lambda}{p}>0.$$

This constant is called the Lundberg exponent.

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Now

$$\psi_c(x) \to 1 \text{ as } x \to \infty,$$

i.e. the probability of not going into ruin goes to one as the initial capital goes to infinity.

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Using $\psi(x) = 1 - \psi_c(x)$ we get

$$\psi(x) = 1 - (1 + B_2 e^{-Rx}) = -B_2 e^{-Rx}.$$

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The Cramér-Lundberg model (12)

Note that

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so we can write

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This is the qualitative behavior of the ultimate ruin probability in the Cramér-Lundberg model.

Using the integro-differential equation for $\psi_c(x)$ one can show that

$$\psi_c(0) = 1 - \frac{\lambda}{\mu p}$$

It follows that

$$\psi(\mathbf{0}) = 1 - \psi_c(\mathbf{0}) = \frac{\lambda}{\mu p}$$

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Theorem

In the Cramér-Lundberg model, if the NPC

 $p > \lambda \mu$

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is satisfied, then the ultimate ruin probability is given by

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where the Lundberg exponent R is given by

$$R=\frac{1}{\mu}-\frac{\lambda}{p}.$$

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$$\mathsf{R} = rac{1}{\mu} - rac{\lambda}{p}.$$

If the NPC is not satsified, then

 $\psi(x) = 1$ for every $x \in \mathbb{R}$.

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- Use another claim size distribution than the exponential distribution. In order to model the possibility of extreme claims, a subexponential distribution can be used. In this case the Lundberg exponent does not exist.
- Use a more general claim arrival process than the Poisson process. If a renewal process is used, then the model is often referred to as the Sparre Andersen model.

The premium rate p is set by the insurance company. We know from the NPC that in the Cramér-Lundberg model we must have

 $p > \lambda \mu$

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But the question is how large should p be, and how should we set it?

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In some way the premium p must reflect the claim's properties.



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We will also normalise and set $\lambda=1.$ This implies that we need to ensure that

 $p > \mu$.

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• The expected value principle

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In this case $\theta > 0$ is known as the safety loading.



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In this case $\theta > 0$ is known as the safety loading.

• The variance principle

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• The standard deviation principle

$$p = \mu + \theta \sigma$$

We can also use utility functions to set the premium.

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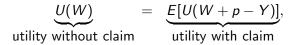
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Assume that the insurance company has utility function U and wealth W. If the insurance company sets the premium p accodring to



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then the company is indifferent between not taking the claim Y and taking the claim.

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I have used

• Lecture notes on Risk theory by Hanspeter Schmidli

for this lecture.

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