## LECTURE NOTES

# NON LIFE INSURANCE First Draft

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#### 1. Individual Risk and Distributions

A non negative random variable is called a **loss** and it its distribution a **loss distribution**.  $X \sim Exponential(\alpha)$  means that X has density  $f_X(x) = \alpha e^{-\alpha x}$  and distribution function (d.f)  $F_X(x) = 1 - e^{-\alpha x} \ \forall x > 0$  and  $\alpha > 0$ .

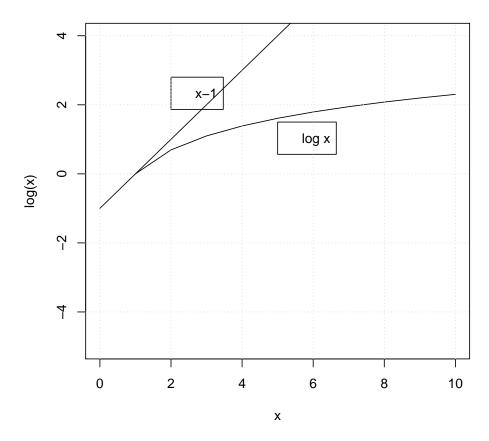
Let  $Y = e^x$ ,

$$F_Y(Y) = F_X(\log Y)$$

$$= 1 - e^{\alpha \log(y)}$$

$$= 1 - y^{-\alpha}$$

Is called the **Pareto Distribution**. If Y follows a Pareto distribution, denoted  $Y \sim Pareto(\alpha)$ 



 $X \sim Exponential(\lambda)$  and  $Y \sim X^{\frac{1}{\tau}} \ \forall \tau > 0$ 

$$F_Y(Y) = F_X(Y^{\tau})$$
$$= 1 - e^{-\lambda y^{\tau}} \quad \forall y > 0$$

Y follows the **Weibull distribution**,  $\tau$  is called the Weibull index. It is denoted by  $Y \sim Weibull(\tau, \lambda)$ 

#### 2. Thursday 09/03/17

#### 2.1. Distribution of the largest claim amount

The distribution of the largest loss is very important in **risk management**.

We will derive asymptotic approximation of standardized maxima.

Let  $X_1, ..., X_n$  be independent losses with distribution function (d.f) F and define

$$M_n = \max\{X_1, ..., X_n\}$$

$$P[M_n \le n] = P[X_1, .., X_n \le x]$$
$$= F^n(x) \quad \forall x > 0$$

Let  $\bar{x} = \sup\{x > 0 | F(x) < 1\}.$ 

Assume 
$$E[M_n] < \infty$$
, then  $E[M_n] = \int_0^{\bar{x}} \{1 - F^n(x)\} dx \xrightarrow{n \to \infty} \bar{x}$ .  
Assume  $E[M_n^2] < \infty$ , then  $E[M_n^2] = \int_0^{\bar{x}} x \{1 - F^n(x)\} dx \xrightarrow{n \to \infty} \bar{x}^2$ 

 $Var(M_n) = E[M_n^2] - E^2[M_n] \xrightarrow{n \to \infty} \bar{x}^2 - \bar{x}^2 = 0$ , assuming  $\bar{x} = 0$ .

Thus the asymptotic distribution of  $M_n$  is degenerate (the total mass is over  $\bar{x}$ ). SO if we want to compute this asymptotic distribution, we must consider the standardization  $\frac{M_n-b_n}{a_n}$ . Before studying these asymptotic approximation we give some examples with finite sample.

#### 2.2. Examples

The distribution of the monthly largest loss is Gumbel  $F(x) = G(\frac{x-\mu}{\sigma})$  where  $G(x) = \exp\{-e^{-x}\}\ x \in \mathbb{R}$ , what is the distribution of the annual maximum?

$$F^{1}2 = exp\{-12e^{-\frac{x-\mu}{\sigma}}\}$$

$$= exp\{-e^{-\frac{x-\mu}{\sigma} + log12}\}$$

$$= exp\{-e^{-\frac{x-(\mu + \sigma log12)}{\sigma}}\}$$

It is thus agian GUmbel, with another location parameter with Frechet monthly largest loss, viz. with  $G(x) = exp\{-x^{-\alpha}\}$  x > 0, we have  $F^{12}(x) = exp\{-12\frac{x-\mu}{\sigma}^{-\alpha}\} = exp\{-(\frac{x-\mu}{12\frac{1}{\alpha}\sigma})^{-\alpha}\}$  It is again Fréchet with another scale parameter. Because of this algebraic closure property, the Gumbel and the Frechet distributions are called max-stable. We consider the slight generalization where the sample size is the random variable N.

Let  $M_N = max\{X_1, ..., X_N\}$ . Assume N independent of  $X_1, X_2, ...$ 

$$P[M_N \le x] = \sum_{n=0}^{\infty} P[M_N \le x | N = n] P[N = n]$$
$$= \sum_{n=0}^{\infty} F^n(x) P[N = n]$$
$$= G_N(F(x)) \quad \forall x \ge 0$$

Where  $M_0 = 0$  and  $G_N(v) = \sum_{n=0}^{\infty} v^n P[N=n]$  is the generating function of N. Thus  $P[M_N \le 0]$  if F(0) = 0 **Example 2.2.1.**  $N_k \sim Poisson(k, \lambda)$ , the number of claim amounts during k years.

$$\begin{split} G_{N_k}(v) &= E[v^{N_k}] \\ &= \sum_{n=0}^{\infty} v^n e^{-k\lambda} \frac{(k\lambda)^n}{n!} \\ &= e^{-k\lambda} \sum_{n=0}^{\infty} \frac{(\lambda k v)^n}{n!} \\ &= exp\{-k\lambda + \lambda k v\} \\ &= exp\{\{k\lambda(v-1)\} \quad \forall v \in \mathbb{R} \end{split}$$

 $Let F(x) = 1 - e^{-\frac{x}{\sigma}}$ 

$$\begin{split} P[M_{N_k} \leq x]) &= G_{N_k}(F(x)) \\ &= exp\{-k\lambda e^{-\frac{x}{\sigma}}\} \\ &= exp\{-exp\{-\frac{x}{\sigma + logk\lambda}\}\} \\ &= exp\{-exp\{-\frac{x - \sigma logk\lambda}{\sigma}\}\} \end{split}$$

 $\forall x \geq 0$  which is the Gumbel distribution.

Let 
$$F(x) = 1 - (\frac{x}{\sigma} + 1)^{-\alpha} \quad \forall x \ge 0$$

$$P[M_{N_k} \le x] = exp\{k\lambda(\frac{x}{\sigma} + 1)^{-\alpha}\}$$
$$= exp\{-(\frac{x}{\sigma(k\lambda)^{\frac{1}{\alpha}}} + 1)^{-\alpha}\} \quad \forall x \ge 0$$

Which is the Fréchet distribution.

### 3. Pareto Type Distributions

Extreme value theory is the analysis of the asymptotic distributions of standardized maxima. We search for  $a_1, a_2, ... > 0$ ,  $b_1, b_2, ... \in \mathbb{R}$  and for d.f G s. t

$$P\left[\frac{M_n - b_n}{a_n} \le x\right] \xrightarrow{n \to \infty} G(x)$$

at all continuity points  $x \in \mathbb{R}$  of G

We consider distributions of Pareto-type.

**Definition 3.1.** The d.f F is of Pareto type if

$$\lim_{x \to \infty} \frac{1 - F(tx)}{1 - F(x)} = t^{-\alpha} \quad \forall t > 0$$

for some  $\alpha > 0$ 

Example 3.1.1. 
$$F(x)=1-x^{-\alpha}$$
  $\frac{1-F(tx)}{1-F(x)} = \frac{(tx)^{-\alpha}}{x^{-\alpha}} = t^{\alpha} \quad \forall x > 1$ 

**Definition 3.2.** THe function  $f: \mathbb{R}_+ \to \mathbb{R}_+$  has regular variation (to infinity) with index  $\delta \in \mathbb{R}$ ,

$$\frac{f(tx)}{f(x)} \xrightarrow{x \to \infty} t^{\delta}$$

This means that  $f(tx) \sim t^{\delta} f(x)$ , as  $x \to \infty$  (Remember that a homogeneous function f of degree  $\delta$  satisfies  $f(tx) = t^{\delta} f(x) \ \forall x$ ). Notation  $f \in_{\delta}$  Thus F is of Pareto-type if and only if  $1 - F \in \mathbb{R}_{\alpha}$ 

**Definition 3.3.** The function  $f: \mathbb{R}_+ \to \mathbb{R}_+$  is a slow varying function if

$$\frac{f(tx)}{f(x)} \xrightarrow{x \to \infty} 1 \quad \forall t > 0$$

 $f \in \mathbb{R}_{\delta} <=> f(x) = x^{\delta} l(x)$  where  $l \in \mathbb{R}_0$ 

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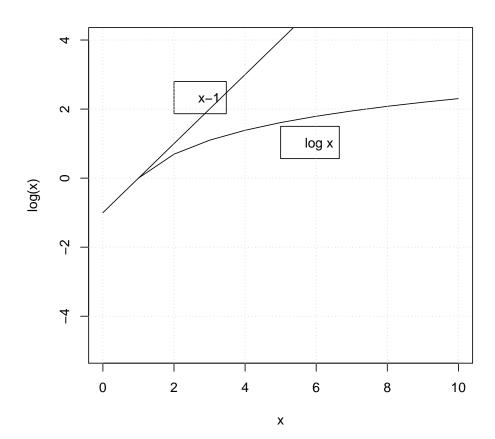
$$\frac{(tx)^{-\delta}f(tx)}{x^{-\delta}f(x)} = t^{-\delta}\frac{f(tx)}{f(x)} \xrightarrow{x\to\infty} t^{-\delta}t^{\delta} = 1$$

<=

$$\frac{f(tx)}{f(x)} = \frac{(tx)^{\delta}l(tx)}{x^{\delta}l(x)} = t^{\delta}\frac{l(tx)}{l(x)} \xrightarrow{x \to \infty} t^{\delta}$$

We want to show that if the distribution of the individual losses is of Pareto type, then the simple maxima is Fréchet distribution.

$$logP[\frac{M_n - b_n}{a_n} \le x] = logF^n(a_n x + b_n)$$
$$= nlogF(a_n x + b_n)$$
$$\sim \{1 - F(a_n x + b_n)\}$$



as  $n \to \infty$ , provided that  $a_n x + b_n \xrightarrow{n \to \infty} \infty$  where  $a_1, a_2, ... > 0$  and  $b_1, b_2, ... \in \mathbb{R}$ . Let us consider  $F(x) = 1 - x^{-\alpha} \ \forall x \ge 1$  and  $b_1 = b_2 = ... = 0$ .

$$n\{1 - F(a_n x)\} = n(a_n x)^{-\alpha} = x^{-\alpha}$$

would give us

$$logP[\frac{M_n}{a_n} \le x] \xrightarrow{n \to \infty} exp\{-x^{-\alpha}\}$$

<=>

$$P\left[\frac{M_n}{a_n} \le x\right] \xrightarrow{n \to \infty} exp\{-x^{-\alpha}\}$$

$$\frac{M_n}{a_n} \xrightarrow{d} Fr\'{e}chet(\alpha)$$

$$na_n^{-\alpha} = 1 <=> a_n^{-\alpha} = n^{-1} <=> a_n = n^{1/\alpha}$$

Thus  $n^{1/\alpha}M_n \xrightarrow{d} Frechet(\alpha)$  as can be expressed in terms of F as follows.

$$1 - x^{-\alpha} = u <=> x = (1 - u)^{-1/\alpha}$$

$$F^{(-1)}(u) = (1 - u)^{-1/\alpha}$$

$$F^{-1}(1 - \frac{1}{n}) = (1 - \{1 - frac + 1n\})^{-\frac{1}{\alpha}} = (\frac{1}{n})^{-\frac{1}{\alpha}}$$

$$= n^{\frac{1}{\alpha}} = a_n$$

Thus  $1 - \frac{1}{n} = F(a_n) <=>$ 

$$\frac{1}{n} <=> 1 - F(a_n) <=> n = \{1 - F(a_n)\}^{-1}$$

Let us keep this relation for a more general distribution function  ${\cal F}.$  Thus

$$n\{1 - F(a_n x)\} = \frac{1 - F(a_n x)}{1 - F(a_n)}$$
$$\xrightarrow{n \to \infty} x^{-\alpha}$$

if F is of Pareto-type.

Therefore, from the previous computations

$$M_n \xrightarrow{d} Fr\'echet(\alpha)$$

where  $a_n = F^{(-1)}(1 - \frac{1}{n})$