

Mid 1 Exam

Monday, March 9, 2020	ACTU 466	Academic year 1441-42H
10:00 – 11:30 AM	LOSS	Second Semester

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Instructions

- Switch off your mobile and place it under your seat.
- Time allowed is 90 minutes.
- You are not asked to provide **detailed explanations** for all answers.
- For exercise 5, four responses are given, only one is correct.

Exercise 1 Consider the random variables $X_1 \sim \text{Pareto}(\alpha = 9, \theta = 7)$ with a density f_{X_1} and $X_2 \sim \text{Exponential}(\text{mean} = 3)$ with a density f_{X_2} and $X_3 \sim \text{Exponential}(\text{mean} = 5)$ with a density f_{X_3} . Construct a spliced distribution X from f_{X_1} on $[0, 2)$, $f_{X_2}(x)$ on $[2, 7)$, and $f_{X_3}(x)$ on $[7, \infty)$, such that under the new distribution the probabilities of these intervals are $1/4$ and $2/3$ and $1/12$ respectively.

Hint: write $f(x) = \begin{cases} a \times f_{X_1}(x) & \text{for } x \in [0, 2) \\ b \times f_{X_2}(x) & \text{for } x \in [2, 7) \\ c \times f_{X_3}(x) & \text{for } x \in [7, \infty) \end{cases}$

Exercise 2 The YANG insurance company provided earthquake insurance houses in San Francisco. For 2015, the claims X were distributed as a Pareto distribution with $\alpha = 3$ and $\theta = 100,000$. In 2015, the earthquake insurance did not have any deductibles, but had an upper limit u so that the expected cost $E(X \wedge u)$ for the coverage with policy limit in 2015 was 80% of the expected value EX of the loss. Determine the upper limit u .

Exercise 3 The number of losses N per policy for health insurance is distributed as a Negative binomial with a mean of 6. The amount of each loss Y follows a Weibull distribution with $\tau = 2$ and $\theta = 500$. An insurance policy covering health insurance has an ordinary deductible d such that the expected number of claims per policy (or the expected number of payments) is 5. Calculate the amount of the deductible d .

Exercise 4 The random variable X represents loss under comprehensive coverage for automobile insurance. Losses are distributed as a two point mixture distribution as follows:

Distribution	Weight
Pareto distribution with $\alpha = 5$ and $\theta = 1,000$	0.6
Pareto distribution with $\alpha = 3$ and $\theta = 6,000$	0.4

Calculate the $\text{var}(X)$.

Exercise 5 You are given that X is the random variable representing the amount of a claim from an automobile accident. X is distributed as a two parameter Pareto Distribution with mean of 2,000 and a variance of 6,000,000. The Henry Auto Insurance Company issues automobile policies with a deductible d of 500. This policy will pay for the amount of any claim in excess of 500. Calculate the expected value of payment per claim incurred $E(Y^L)$. This is the same as $E[(X - 500)^+]$.

Hint: calculate first the values of the parameters of the Pareto distribution.

a) 76,435.98 b) 5,476.76 c) 54.54 d) 1,567.05

A.2.3 Two-parameter distributions

A.2.3.1 Pareto (Pareto Type II, Lomax)— α, θ

$$\begin{aligned}
 f(x) &= \frac{\alpha\theta^\alpha}{(x+\theta)^{\alpha+1}} & F(x) &= 1 - \left(\frac{\theta}{x+\theta}\right)^\alpha \\
 E[X^k] &= \frac{\theta^k\Gamma(k+1)\Gamma(\alpha-k)}{\Gamma(\alpha)}, & -1 < k < \alpha \\
 E[X^k] &= \frac{\theta^k k!}{(\alpha-1)\cdots(\alpha-k)}, & \text{if } k \text{ is an integer} \\
 \text{VaR}_p(X) &= \theta[(1-p)^{-1/\alpha} - 1] \\
 \text{TVaR}_p(X) &= \text{VaR}_p(X) + \frac{\theta(1-p)^{-1/\alpha}}{\alpha-1}, & \alpha > 1 \\
 E[X \wedge x] &= \frac{\theta}{\alpha-1} \left[1 - \left(\frac{\theta}{x+\theta}\right)^{\alpha-1} \right], & \alpha \neq 1 \\
 E[X \wedge x] &= -\theta \ln\left(\frac{\theta}{x+\theta}\right), & \alpha = 1 \\
 E[(X \wedge x)^k] &= \frac{\theta^k\Gamma(k+1)\Gamma(\alpha-k)}{\Gamma(\alpha)}\beta[k+1, \alpha-k; x/(x+\theta)] + x^k \left(\frac{\theta}{x+\theta}\right)^\alpha, & \text{all } k \\
 \text{mode} &= 0
 \end{aligned}$$

A.2.3.2 Inverse Pareto— τ, θ

$$\begin{aligned}
 f(x) &= \frac{\tau\theta x^{\tau-1}}{(x+\theta)^{\tau+1}} & F(x) &= \left(\frac{x}{x+\theta}\right)^\tau \\
 E[X^k] &= \frac{\theta^k\Gamma(\tau+k)\Gamma(1-k)}{\Gamma(\tau)}, & -\tau < k < 1 \\
 E[X^k] &= \frac{\theta^k(-k)!}{(\tau-1)\cdots(\tau+k)}, & \text{if } k \text{ is a negative integer} \\
 \text{VaR}_p(X) &= \theta[p^{-1/\tau} - 1]^{-1} \\
 E[(X \wedge x)^k] &= \theta^k \tau \int_0^{x/(x+\theta)} y^{\tau+k-1}(1-y)^{-k} dy + x^k \left[1 - \left(\frac{x}{x+\theta}\right)^\tau \right], & k > -\tau \\
 \text{mode} &= \theta \frac{\tau-1}{2}, & \tau > 1, \text{ else } 0
 \end{aligned}$$

A.2.3.3 Loglogistic (Fisk)— γ, θ

$$\begin{aligned}
 f(x) &= \frac{\gamma(x/\theta)^\gamma}{x[1+(x/\theta)^\gamma]^2} & F(x) &= u, \quad u = \frac{(x/\theta)^\gamma}{1+(x/\theta)^\gamma} \\
 E[X^k] &= \theta^k\Gamma(1+k/\gamma)\Gamma(1-k/\gamma), & -\gamma < k < \gamma \\
 \text{VaR}_p(X) &= \theta(p^{-1} - 1)^{-1/\gamma} \\
 E[(X \wedge x)^k] &= \theta^k\Gamma(1+k/\gamma)\Gamma(1-k/\gamma)\beta(1+k/\gamma, 1-k/\gamma; u) + x^k(1-u), & k > -\gamma \\
 \text{mode} &= \theta \left(\frac{\gamma-1}{\gamma+1}\right)^{1/\gamma}, & \gamma > 1, \text{ else } 0
 \end{aligned}$$

APPENDIX A. AN INVENTORY OF CONTINUOUS DISTRIBUTIONS

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A.3.2.2 Inverse gamma (Vinci)— α, θ

$$\begin{aligned}
 f(x) &= \frac{(\theta/x)^\alpha e^{-\theta/x}}{x\Gamma(\alpha)} & F(x) &= 1 - \Gamma(\alpha; \theta/x) \\
 E[X^k] &= \frac{\theta^k \Gamma(\alpha - k)}{\Gamma(\alpha)}, \quad k < \alpha & E[X^k] &= \frac{\theta^k}{(\alpha - 1) \cdots (\alpha - k)}, \quad \text{if } k \text{ is an integer} \\
 E[(X \wedge x)^k] &= \frac{\theta^k \Gamma(\alpha - k)}{\Gamma(\alpha)} [1 - \Gamma(\alpha - k; \theta/x)] + x^k \Gamma(\alpha; \theta/x) \\
 &= \frac{\theta^k \Gamma(\alpha - k)}{\Gamma(\alpha)} G(\alpha - k; \theta/x) + x^k \Gamma(\alpha; \theta/x), \quad \text{all } k \\
 \text{mode} &= \theta/(\alpha + 1)
 \end{aligned}$$

A.3.2.3 Weibull— θ, τ

$$\begin{aligned}
 f(x) &= \frac{\tau(x/\theta)^\tau e^{-(x/\theta)^\tau}}{x} & F(x) &= 1 - e^{-(x/\theta)^\tau} \\
 E[X^k] &= \theta^k \Gamma(1 + k/\tau), \quad k > -\tau \\
 \text{VaR}_p(X) &= \theta[-\ln(1 - p)]^{1/\tau} \\
 E[(X \wedge x)^k] &= \theta^k \Gamma(1 + k/\tau) \Gamma[1 + k/\tau; (x/\theta)^\tau] + x^k e^{-(x/\theta)^\tau}, \quad k > -\tau \\
 \text{mode} &= \theta \left(\frac{\tau - 1}{\tau} \right)^{1/\tau}, \quad \tau > 1, \text{ else } 0
 \end{aligned}$$

A.3.2.4 Inverse Weibull (log Gompertz)— θ, τ

$$\begin{aligned}
 f(x) &= \frac{\tau(\theta/x)^\tau e^{-(\theta/x)^\tau}}{x} & F(x) &= e^{-(\theta/x)^\tau} \\
 E[X^k] &= \theta^k \Gamma(1 - k/\tau), \quad k < \tau \\
 \text{VaR}_p(X) &= \theta(-\ln p)^{-1/\tau} \\
 E[(X \wedge x)^k] &= \theta^k \Gamma(1 - k/\tau) \{1 - \Gamma[1 - k/\tau; (\theta/x)^\tau]\} + x^k [1 - e^{-(\theta/x)^\tau}], \quad \text{all } k \\
 &= \theta^k \Gamma(1 - k/\tau) G[1 - k/\tau; (\theta/x)^\tau] + x^k [1 - e^{-(\theta/x)^\tau}] \\
 \text{mode} &= \theta \left(\frac{\tau}{\tau + 1} \right)^{1/\tau}
 \end{aligned}$$

A.3.3 One-parameter distributions

A.3.3.1 Exponential— θ

$$\begin{aligned}
 f(x) &= \frac{e^{-x/\theta}}{\theta} & F(x) &= 1 - e^{-x/\theta} \\
 M(t) &= (1 - \theta t)^{-1} & E[X^k] &= \theta^k \Gamma(k + 1), \quad k > -1 \\
 E[X^k] &= \theta^k k!, \quad \text{if } k \text{ is an integer} \\
 \text{VaR}_p(X) &= -\theta \ln(1 - p) \\
 \text{TVaR}_p(X) &= -\theta \ln(1 - p) + \theta \\
 E[X \wedge x] &= \theta(1 - e^{-x/\theta}) \\
 E[(X \wedge x)^k] &= \theta^k \Gamma(k + 1) \Gamma(k + 1; x/\theta) + x^k e^{-x/\theta}, \quad k > -1 \\
 &= \theta^k k! \Gamma(k + 1; x/\theta) + x^k e^{-x/\theta}, \quad k \text{ an integer} \\
 \text{mode} &= 0
 \end{aligned}$$

A.3.3.2 Inverse exponential— θ

$$\begin{aligned}
 f(x) &= \frac{\theta e^{-\theta/x}}{x^2} & F(x) &= e^{-\theta/x} \\
 E[X^k] &= \theta^k \Gamma(1 - k), \quad k < 1 \\
 \text{VaR}_p(X) &= \theta(-\ln p)^{-1} \\
 E[(X \wedge x)^k] &= \theta^k G(1 - k; \theta/x) + x^k (1 - e^{-\theta/x}), \quad \text{all } k \\
 \text{mode} &= \theta/2
 \end{aligned}$$

A.5 Other distributions

A.5.1.1 Lognormal— μ, σ (μ can be negative)

$$\begin{aligned}
 f(x) &= \frac{1}{x\sigma\sqrt{2\pi}} \exp(-z^2/2) = \phi(z)/(\sigma x), \quad z = \frac{\ln x - \mu}{\sigma} & F(x) &= \Phi(z) \\
 E[X^k] &= \exp(k\mu + k^2\sigma^2/2) \\
 E[(X \wedge x)^k] &= \exp(k\mu + k^2\sigma^2/2) \Phi\left(\frac{\ln x - \mu - k\sigma^2}{\sigma}\right) + x^k [1 - F(x)] \\
 \text{mode} &= \exp(\mu - \sigma^2)
 \end{aligned}$$

B.2.1.2 Geometric— β

$$\begin{aligned} p_0 &= \frac{1}{1+\beta}, & a &= \frac{\beta}{1+\beta}, & b &= 0 & p_k &= \frac{\beta^k}{(1+\beta)^{k+1}} \\ E[N] &= \beta, & \text{Var}[N] &= \beta(1+\beta) & P(z) &= [1-\beta(z-1)]^{-1}. \end{aligned}$$

This is a special case of the negative binomial with $r = 1$.

B.2.1.3 Binomial— $q, m, (0 < q < 1, m \text{ an integer})$

$$\begin{aligned} p_0 &= (1-q)^m, & a &= -\frac{q}{1-q}, & b &= \frac{(m+1)q}{1-q} \\ p_k &= \binom{m}{k} q^k (1-q)^{m-k}, & k &= 0, 1, \dots, m \\ E[N] &= mq, & \text{Var}[N] &= mq(1-q) & P(z) &= [1+q(z-1)]^m. \end{aligned}$$

B.2.1.4 Negative binomial— β, r

$$\begin{aligned} p_0 &= (1+\beta)^{-r}, & a &= \frac{\beta}{1+\beta}, & b &= \frac{(r-1)\beta}{1+\beta} \\ p_k &= \frac{r(r+1)\cdots(r+k-1)\beta^k}{k!(1+\beta)^{r+k}} \\ E[N] &= r\beta, & \text{Var}[N] &= r\beta(1+\beta) & P(z) &= [1-\beta(z-1)]^{-r}. \end{aligned}$$

B.3 The $(a, b, 1)$ class

To distinguish this class from the $(a, b, 0)$ class, the probabilities are denoted $\Pr(N = k) = p_k^M$ or $\Pr(N = k) = p_k^T$ depending on which subclass is being represented. For this class, p_0^M is arbitrary (that is, it is a parameter) and then p_1^M or p_1^T is a specified function of the parameters a and b . Subsequent probabilities are obtained recursively as in the $(a, b, 0)$ class: $p_k^M = (a+b/k)p_{k-1}^M$, $k = 2, 3, \dots$, with the same recursion for p_k^T . There are two sub-classes of this class. When discussing their members, we often refer to the “corresponding” member of the $(a, b, 0)$ class. This refers to the member of that class with the same values for a and b . The notation p_k will continue to be used for probabilities for the corresponding $(a, b, 0)$ distribution.

B.3.1 The zero-truncated subclass

The members of this class have $p_0^T = 0$ and therefore it need not be estimated. These distributions should only be used when a value of zero is impossible. The first factorial moment is $\mu_{(1)} = (a+b)/[(1-a)(1-p_0)]$, where p_0 is the value for the corresponding member of the $(a, b, 0)$ class. For the logarithmic distribution (which has no corresponding member), $\mu_{(1)} = \beta/\ln(1+\beta)$. Higher factorial moments are obtained recursively with the same formula as with the $(a, b, 0)$ class. The variance is $(a+b)[1-(a+b+1)p_0]/[(1-a)(1-p_0)]^2$. For those members of the subclass which have corresponding $(a, b, 0)$ distributions, $p_k^T = p_k/(1-p_0)$.

loss HW 1

Solution

Ex 1

$$f(x) = \begin{cases} \frac{1}{4} \times A f_1(x) & 0 < x < 2 \\ \frac{2}{3} \times B f_2(x) & 2 < x < 7 \\ \frac{1}{12} \times C f_3(x) & x > 7 \end{cases}$$

where $f_1(x)$ density of Pareto ($\alpha=9, \theta=7$)
 $f_2(x)$ " " Exp (mean=3)
 $f_3(x)$ " " Exp (mean=5)

$$\int_0^2 A f_1(x) dx = 1 \Rightarrow A F_1(2) = 1$$

$$\Leftrightarrow A = \frac{1}{1 - (7/9)^9} = 1.1627$$

(you should not calculate the integral. We have new did it)

$$\int_2^7 B f_2(x) dx = 1 \Rightarrow B [S_2(2) - S_2(7)] = 1$$

$$\Rightarrow B = \frac{1}{e^{-2/3} - e^{-7/3}} = 2.401277$$

$$\int_7^{\infty} C f_3(x) dx = 1 \Rightarrow C \times S(7) = 1$$

$$\Rightarrow C \times e^{-7/5} = 1 \Rightarrow C = e^{7/5} = 4.0532$$

$$f(x) = \begin{cases} 0.279 \times f_1(x) \\ 1.600 \times f_2(x) \\ 0.3379 \times f_3(x) \end{cases}$$

Ex 2 $E(x|u) = 0.8 \times E(x)$

$$E(x) = \frac{\theta}{\alpha-1} = 50,000$$

$$E(x|u) = \frac{\theta}{\alpha-1} \left[1 - \left(\frac{\theta}{u+\theta} \right)^{\alpha-1} \right] = 0.8 \frac{\theta}{\alpha-1}$$

$$\Rightarrow 50,000 - 50,000 \left(\frac{\theta}{u+\theta} \right)^2 = 0.8 \times 50,000$$

$$\Rightarrow 0.2 \times 50,000 = 50,000 \left(\frac{\theta}{u+\theta} \right)^2$$

$$\Rightarrow \left(\frac{\theta}{u+\theta} \right) = \sqrt{0.2} \Rightarrow u = \frac{(1 - \sqrt{0.2}) 100,000}{\sqrt{0.2}}$$

(all calculations should be done)
= 123,607

Ex 3

Solution 1

For a given value of N , the number of losses, N^P , is the number of losses for which loss $y > d$.
(= the number of success)

Then $N^P | N \sim \text{Binomial}(N, p = P(Y > d))$
 $\Rightarrow E(N^P | N) = Np \Rightarrow E(N^P) = E(E(N^P | N)) = E(Np)$
 $= p E(N) \Rightarrow p = 5/6$
 indep

Solution 2 ~~(p-value)~~
 $N^P = \sum_{i=1}^N \mathbb{1}_{y_i > d}$

$\Rightarrow N^P$ is a compound distribution

$$E(N^P) = E(N) \times E(\mathbb{1}_{y > d}) = E(N) \times P(Y > d)$$

Ex 4

$$E(x) = 0.6 E(x_1) + 0.4 E(x_2)$$

$$= 0.6 \times \frac{1000}{4} + 0.4 \times 3000 = 1350$$

$$E(x^2) = 0.6 E(x_1^2) + 0.4 E(x_2^2)$$

$$= 0.6 \times 2 \times \frac{1000^2}{12} + 0.4 \times 2 \times \frac{6000^2}{2}$$

$$= 14,500,000$$

$$V(x) = E(x^2) - [E(x)]^2 = 12,677,500$$

Ex 5

$$E(x) = \frac{\theta}{\alpha-1} = 2000 \Rightarrow \theta = (\alpha-1)2000$$

$$E(x^2) = [E(x)]^2 + V(x) = 10^7$$

$$\Rightarrow E(x^2) = \frac{2\theta^2}{(\alpha-1)(\alpha-2)} = 10^7$$

$$\Rightarrow \frac{2 \times 4 \times 10^6 \times (\alpha-1)^2}{(\alpha-1)(\alpha-2)} = 10^7$$

$$\Rightarrow 8 \times 10^6 \times \frac{\alpha-1}{\alpha-2} = 10^7$$

$$\Rightarrow \frac{\alpha-1}{\alpha-2} = \frac{10}{8} = \frac{5}{4}$$

$$\Rightarrow 4(\alpha-1) = 5(\alpha-2) \Rightarrow \alpha = 6; \theta = 2000(\alpha-1) = 10,000$$

(all calculations should be given)

~~sketch~~

$$E(x) = E(x_1 | \alpha) + E(y_1)$$

2000

$$\left(\frac{\theta}{\alpha-1} \right) \left[1 - \left(\frac{\theta}{\alpha+\theta} \right)^{\alpha-1} \right] = 432.95$$

$$\Rightarrow E(y_1) = 2000 - 432.95 = 1567.05$$