

To get 38), sub in $X_{\text{new}} = (1+r)X$ in ⁸⁷ 37) and solve for X for getting the values of X for which the formulas apply, then multiply the formulas by α .

39) *Under 38) for the per loss RV

$$E(Y^L) = \alpha(1+r) \left[E\left(X \wedge \frac{U}{1+r}\right) - E\left(X \wedge \frac{d}{1+r}\right) \right]$$

and for the per payment RV,

$$E(Y^P) = \frac{E(Y^L)}{1 - F_X\left(\frac{d}{1+r}\right)} = \frac{E(Y^L)}{S_X\left(\frac{d}{1+r}\right)}$$

Note: that $\alpha = 1$ for no coinsurance and $r = 0$ for no inflation. Take $U = \infty$ so $E(X \wedge \infty) = E(X)$ if there is no policy limit.

40) *So if there is a deductible and a policy limit $U-d$ (with $U = u-d+d$), then

$$E(Y^L) = E(X \wedge U) - E[X \wedge d] \quad \text{with}$$

$$E(Y^P) = \frac{E(Y^L)}{S_X(d)}$$

ex) with deductible d and
policy limit $U-d$
 $U-d = \text{maximum payment}$

M404 875

$$U = \text{maximum covered loss} \\ = U-d+d$$

math 403

E[dis] 3} $X \sim \text{EXP}$ with mean 1000.

An insurance policy on X has deductible
500 and maximum policy payment per
loss of 1500. Find Y^L and Y^P .

Soln $\theta = 1000$, $U-d = 1500$, $U = 1500+d = 2000$

$$E(Y^L) = E(X \wedge U) - E[X \wedge d]$$

$$= E(X \wedge 2000) - E[X \wedge 500] \quad \text{and}$$

$$E(X \wedge w) = \theta (1 - e^{-w/\theta}).$$

$$\text{So } E(Y^L) = 1000 (1 - e^{-2}) - 1000 (1 - e^{-1/2})$$

$$= 1000 (e^{-1/2} - e^{-2}) = \boxed{471.1954}$$

$$S_X(d) = 1 - F_X(d) = e^{-d/\theta}$$

$$\text{So } E(Y^P) = \frac{E(Y^L)}{S_X(d)} = \frac{471.1954}{e^{-1/2}} = \boxed{776.9698}$$

ex} ^{mat 403} HW 3 # 3 Suppose the loss RV ~~is~~

$X \sim U(0, 100)$. An insurance policy for X has deductible $d = 20$ and a policy limit (max payment) of 60 per loss.

Find $E(Y^L)$.

Soln $Y^L = \begin{cases} 0 & X < 20 \\ X-20 & 20 < X < U = 80 = 60+20 \\ 60 & 80 < X < 100 \end{cases}$

support of $X = [0, 100]$

$$\text{So } E(Y^L) = \int_{20}^{80} (x-20) f(x) dx + \int_{80}^{100} 60 f(x) dx$$

$$= \int_{20}^{80} (x-20) \frac{1}{100} dx + \int_{80}^{100} 60 \frac{1}{100} dx$$

as in the HW problem.

41) LER and inflation

$$X_{\text{new}} = (1+r)X$$

If $X \sim SF(\theta, \tau)$ then

$X_{\text{new}} \sim SF((1+r)\theta, \tau)$ so use formulas for

$$X_{\text{new}}, \text{ Also } (1+r)X \wedge d = (1+r)X \wedge \frac{(1+r)d}{1+r}$$

$$= (1+r) \left(X \wedge \frac{d}{1+r} \right) = (1+r) \min \left(X, \frac{d}{1+r} \right). \text{ So } LER_{\text{new}}(d)$$

$$LER(d) = \frac{E(X_{\text{new}} \wedge d)}{E(X_{\text{new}})}$$

$$= \frac{(1+r)E(X \wedge \frac{d}{1+r})}{(1+r)EX} = \frac{E(X \wedge \frac{d}{1+r})}{EX}$$

42) If $X \sim \text{Exp}(\theta)$, M40#885

$$\text{LER}(d) = 1 - e^{-d/\theta}$$

If $X \sim$ (2 parameter) pareto ($\alpha > 1, \theta$)

$$\text{LER}(d) = 1 - \left(\frac{\theta}{d+\theta}\right)^{\alpha-1}$$

If $X \sim$ single parameter ($\alpha > 1, \theta \leq d$),

$$\text{LER}(d) = 1 - \frac{(\theta/d)^{\alpha-1}}{\alpha}$$

Example

Ch9 Aggregate Loss Models

Once in a while Exam C has a question on a bonus or dividend. The loss ratio r or R is the proportion of aggregate losses to premiums. Express the bonus in terms of the earned premium, usually

$$\max(0, c(rP - X))$$

where P is the earned premium, X the aggregate losses, and $c > 0$ is

For Q11 know how to compute

\bar{X} and $\hat{\sigma}_E^2$ for 1 sample.

Brand name
 Discrete RVs: $X = N$ counts something.
 usually number of claims.

contin RV! X is usually loss amount

A) MLE I) Let X_1, \dots, X_n are iid $\text{Ber}(g)$

$= \text{bin}(g, m=1)$ and $\sum_{i=1}^n X_i = \# \text{ 1s.}$

If $X_1 = x_1, \dots, X_n = x_n$ $x_i \in \{0, 1\}$ are

known, the likelihood is $L(g|x) = g^{\sum x_i} (1-g)^{n-\sum x_i}$.

If only $\sum x_i = k$ is known, use the

$\text{bin}(g, m=n)$ likelihood $L(g|k) = \binom{n}{k} g^k (1-g)^{n-k}$

II) (Positive) payments = Y_i^P

losses = X_i usually left

truncated and often right censored
 deductible \uparrow maximum payment $U-d$
 maximum covered loss $U = U-d+d$

47h) - 54) are important

$X = \frac{Y^P}{\alpha} + d$ where usually $d=1$

X is left truncated at d and right censored at U

$I + n-k$ (losses) cases are uncensored and k are censored at U

$$L(\theta) = \frac{\prod_{i=1}^{n-k} f(x_i)}{[1-F(d)]^n}$$

if all n cases are truncated at d

MLE problem a) set up likelihood
 b) find MLE . . . b) often has easy formula in case a) was done wrong.

ex) C79 Losses come from a mixture of an EXP(100) dist with prob p and an EXP(10000) dist with prob $1-p$. Losses of 100 and 2000 are observed. Find $L(p)$.

Soln) $f(x) = p \cdot f_1(x) + (1-p) \cdot f_2(x)$

$$= p \frac{1}{100} e^{-x/100} + (1-p) \frac{1}{10000} e^{-x/10000}$$

$f(x_1, x_2) = f(x_1) \cdot f(x_2) =$

$$\left(p \frac{1}{100} e^{-x_1/100} + (1-p) \frac{1}{10000} e^{-x_1/10000} \right) \left(p \frac{1}{100} e^{-x_2/100} + (1-p) \frac{1}{10000} e^{-x_2/10000} \right)$$

ex) C146) X_1, \dots, X_n are iid single parameter Pareto ($\alpha = \alpha_1, \theta = 1$)
 and Y_1, \dots, Y_m are iid \dots ($\alpha = \alpha_2, \theta = 1$)

$$EY_2 = \frac{\alpha_2}{1-\alpha_2} = 1.5, EY_1 = 1.5 \frac{\alpha_1}{1-\alpha_1}$$

want MLE for α_1 using both samples.

Find $\frac{d}{d\alpha_1} \log L(\alpha_1)$.

Soln $\alpha_2 = 1.5 \frac{\alpha_1}{1-\alpha_1} \implies \alpha_2 (1-\alpha_1) = 1.5 \alpha_1$

$$\alpha_2 \left(1 + 1.5 \frac{\alpha_1}{1-\alpha_1} \right) = \alpha_2 \frac{1+1.5\alpha_1}{1-\alpha_1} = 1.5 \frac{\alpha_1}{1-\alpha_1} \implies \alpha_2 = \frac{1.5\alpha_1}{1+1.5\alpha_1} = \frac{3\alpha_1}{2+\alpha_1}$$

$$L = \prod_{i=1}^n \frac{\alpha_1}{x_i^{\alpha_1+1}} \prod_{i=1}^m \frac{\alpha_2}{y_i^{\alpha_2+1}} = \frac{\alpha_1^n}{\prod x_i^{\alpha_1+1}} \frac{\left(\frac{3\alpha_1}{2+\alpha_1}\right)^m}{\prod y_i^{\frac{3\alpha_1}{2+\alpha_1}+1}}$$

$$\ln L(\alpha_1) = n \ln \alpha_1 - (\alpha_1+1) \sum \ln x_i + m \ln \left(\frac{3\alpha_1}{2+\alpha_1} \right)$$

$$- \left(\frac{2+4\alpha_1}{2+\alpha_1} \right) \sum \ln y_i$$

$$\frac{d}{d\alpha_1} \ln L(\alpha_1) = \frac{n}{\alpha_1} - \sum \ln x_i + m \frac{2+\alpha_1-\alpha_1}{\alpha_1(2+\alpha_1)} \left(\frac{2+\alpha_1}{3} - 3\alpha_1 \right) \frac{1}{(2+\alpha_1)^2}$$

$$\frac{d\alpha_1 - n d\alpha_1}{d^2}$$

$$- \frac{(2+\alpha_1)^4 - (2+4\alpha_1)}{(2+\alpha_1)^2} \sum \ln y_i =$$

$$\frac{n}{\alpha_1} - \sum \ln x_i + \frac{2m}{\alpha_1(2+\alpha_1)} - \frac{6}{(2+\alpha_1)^2} \sum \ln y_i \stackrel{\text{set}}{=} 0$$

ex) C152 Observe losses 600, 700, 900.

No information is available about losses of 500 or less. Losses $\sim \text{EXP}(\theta)$. Find the MLE of θ .

soln Losses are left truncated at 500 = d.

$$\text{So } L(\theta) = \frac{\pi f(x_i)}{(1-F_{500})^3} = \frac{\frac{1}{\theta} e^{-600/\theta} \frac{1}{\theta} e^{-700/\theta} \frac{1}{\theta} e^{-900/\theta}}{\left(e^{-500/\theta} \right)^3}$$

$$= \theta^{-3} e^{-\frac{(600+700+900)-(1500)}{\theta}} = \theta^{-3} e^{-\frac{700}{\theta}}$$

$$\ln L(\theta) = -3 \ln \theta - \frac{700}{\theta}$$

$$\frac{d \ln L(\theta)}{d\theta} = -\frac{3}{\theta} + \frac{700}{\theta^2} \stackrel{\text{set}}{=} 0 \text{ or } 3\theta = 700 \text{ or } \hat{\theta} = \frac{700}{3}$$

$$c4) f(x) = \frac{\alpha}{x^{\alpha+1}} \quad x > 1$$

Fred 3

$\alpha > 0$,

observe 3 losses 3, 6, 14 and 2
losses above 25
censored

Find MLE of α .

$$\left(\frac{1}{25}\right)^\alpha$$

$$\text{Soln } L(\alpha) = \prod_{i=1}^m f(x_i) \left(1 - F(25)\right)^2$$

$$= \frac{\alpha^3}{(3(6)14)^{\alpha+1}} \frac{1}{(25)^{2\alpha}}$$

$$\ln L(\alpha) = 3 \ln \alpha - (\alpha+1) \ln(3(6)14) - 2\alpha \ln 25$$

$$\frac{d \ln L(\alpha)}{d\alpha} = \frac{3}{\alpha} - \ln[3(6)14] - 2 \ln 25$$

$$= \frac{3}{\alpha} - 11.9672 \stackrel{\text{set}}{=} 0$$

$$\hat{\alpha} = \frac{3}{11.9672} = 0.2507$$

~~$n=2 \quad \bar{x} = 6 \quad x_1, x_2 \text{ iid Exp}(\theta)$~~

~~Find asymptotic variance of the~~

~~MLE of $g(\theta) = 1 - e^{-10/\theta} \quad (1+10\theta^{-1})$~~

~~Soln $g'(\theta) = e^{-10/\theta} \frac{10}{\theta^2} (1+10\theta^{-1}) - e^{-10/\theta} (-10\theta^{-2})$~~

A good formal model be Q11

$E1: 2, E2: 3, 4, 6, E3: 2, 4, 6$

U1E21

ex} simulate $\hat{\theta}$ 5 times: get

10, 28, 15, 22, 18. If these values are equally likely and the true value of θ is 19, estimate bias.

Soln bias = $\frac{\sum_{i=1}^B \hat{\theta}_i - \theta}{B} = \frac{93}{5} - 19 = -0.40$

MSE = $\frac{\sum (\hat{\theta}_i - \theta)^2}{B}$

no longer on formal

ex) $n = 200$ There are 200 claims

Fred 4

with aggregate loss 20000.

$X_i \sim \text{Exp}(\theta)$ $H_0: \theta = 80$ $H_1: \theta = 100$

Find LRT test statistic T .

soln $n = 200$ $\sum X_i = 20000$ $\bar{X} = \hat{\theta}_{MLE} = 100$

$$L(\theta) = \prod_{i=1}^{200} \theta^{-1} e^{-x_i/\theta} = \theta^{-200} e^{-\sum x_i/\theta}$$

$$\ln L(\theta) = -200 \ln(\theta) - \frac{\sum x_i}{\theta}$$

$$\ln L_0 = -200 \ln(80) - \frac{20000}{80} = -1126.4053$$

$$\ln L_1 = -200 \ln(100) - \frac{20000}{100} = -1121.0340$$

$$T = 2(\ln L_1 - \ln L_0) = 10.7427$$

ex) UIE2 is # claims # obs's

0	6
1	0
2	4
3+	0

Above table is for a negative binomial data.
Determine method of moments estimators
of r and β then estimate the prob
of one or more claims in a time period

$$\bar{X} = \mu = \frac{0(6) + 2(4)}{10} = 0.8$$

$$\hat{\sigma}_E^2 = s - \mu^2 = \frac{0^2(6) + 2^2(4)}{10} - (0.8)^2 = 0.96$$

$$\left. \begin{array}{l} \hat{r}_{\hat{\beta}} \stackrel{\text{set}}{=} 0.8 \\ \hat{r}_{\hat{\beta}} (1 + \hat{\beta}) \stackrel{\text{set}}{=} 0.96 \end{array} \right\} \hat{r} = \frac{0.8}{\hat{\beta}}$$

$$1 + \hat{\beta} = \frac{0.96}{0.8}$$

$$\hat{\beta} = 0.2$$

$$\hat{r} = \frac{0.8}{0.2} = 4$$

$$P(W \geq 1) = 1 - \hat{P}_0 = 1 - (1 + \hat{\beta})^{-\hat{r}}$$

$$= 1 - (1.2)^{-4} = 0.518$$

$$\left(\hat{P}_0 = \binom{0 + \hat{r}}{0} (1 + \hat{\beta})^{-\hat{r}} \left(\frac{\hat{\beta}}{1 + \hat{\beta}} \right)^0 = (1 + \hat{\beta})^{-\hat{r}} \right)$$

ex C229 X_1, \dots, X_n iid with $\theta > 0, x > 0$ Fred 5

$$f(x) = \frac{\theta}{(\theta+x)^2}$$

Find the asymptotic variance of the MLE $\hat{\theta}$ of θ

Soln } $\ln f(x) = \ln(\theta) - 2 \ln(\theta+x)$

$$\frac{d \ln f(x)}{d\theta} = \frac{1}{\theta} - \frac{2}{\theta+x}$$

$$\frac{d^2 \ln f(x)}{d\theta^2} = -\frac{1}{\theta^2} + \frac{2}{(\theta+x)^2}$$

$$E \frac{d^2 \ln f(x)}{d\theta^2} = -\frac{1}{\theta^2} + \int_0^{\infty} \frac{2}{(\theta+x)^2} \frac{\theta}{(\theta+x)^2} dx$$

$$= -\frac{1}{\theta^2} + \int_0^{\infty} \frac{2\theta}{(\theta+x)^4} dx = -\frac{1}{\theta^2} + 2\theta \left. \frac{(\theta+x)^{-3}}{-3} \right|_0^{\infty}$$

$$= -\frac{1}{\theta^2} + \frac{2\theta}{3} \frac{1}{\theta^3} = -\frac{1}{\theta^2} + \frac{2}{3} \frac{1}{\theta^2} = -\frac{1}{3\theta^2}$$

$$I_1(\theta) = \frac{1}{3\theta^2}$$

$$V(\hat{\theta}) = \frac{1}{I_n(\theta)} = \frac{3\theta^2}{n}$$

076 Bayesian posterior

$$x e^{-x}$$

$N_k \sim \text{Pois}(\theta)$, θ has pdf $\propto e^{-\theta} \theta^k$ dummy variable is θ

$f(\theta) = \theta e^{-\theta}$, $\theta > 0$ $P_k = \frac{e^{-\theta} \theta^k}{k!}$

$$\int_0^{\infty} \theta e^{-k\theta} d\theta = \frac{1}{k^2}$$

$$\left(\pi(\theta | x=k) \propto \pi(\theta) f(k|\theta) \propto \theta e^{-\theta} e^{-\theta} \theta^k \right.$$

$$\left. \begin{matrix} \alpha^* = \alpha + k = k + 2 \\ \theta^* = \frac{\theta'}{1 + \theta'} = \frac{1}{1+1} = \frac{1}{2} \end{matrix} \right\} = \theta e^{-2\theta} \text{ gamma } \left(\alpha = k+2, \theta = \frac{1}{2} \right)$$

A randomly selected policyholder had at least one claim last year. What posterior prob this same policyholder will have at least one claim this year. We will find the posterior pdf of

$$\pi(\theta | N > 0) \propto \pi(\theta) P(N > 0 | \theta)$$

$$\propto \theta e^{-\theta} (1 - e^{-\theta})$$

$$\text{So } \int_0^{\infty} \theta e^{-\theta} - \theta e^{-2\theta} d\theta = C - C \int_0^{\infty} \theta e^{-2\theta} d\theta$$

Gamma PDF

$$= C - C \frac{1}{4} = \frac{3C}{4} \text{ so } C = \frac{4}{3}$$

$$\text{Posterior pdf} = \frac{4}{3} \theta e^{-\theta} (1 - e^{-\theta})$$

The final is a lot like the old

final. 2 pages from Q1/ 4 pages

from ex 0-3, ^{11 problems} 12 parts 25 points each.

MLE problem: given $L(\theta)$. So easier than usual.

MLE I like 23 i) ^{class} iv) $\text{parab}(a, 0)$

MLE 43), 46