- 1. Exponential Claims, $X_i \sim Exp(\theta)$:
 - Expected Claim Size

$$E_{\theta}(Y) = E_{\theta}(X) - \int_{0}^{\infty} y f_{X}(y+M;\theta) dy$$
$$= \theta - \int_{0}^{\infty} \frac{y}{\theta} e^{-(y+M)/\theta} dy$$
$$= \theta - e^{-M/\theta} \int_{0}^{\infty} \frac{y}{\theta} e^{-y/\theta} dy$$
$$= \theta (1 - e^{-M/\theta})$$

• Moment Generating Function

$$m_Y(t) = (1 - \theta t)^{-1} \{ 1 - t\theta e^{M(t - \theta^{-1})} \}, \qquad t < \theta^{-1}$$

- 1. Exponential Claims, $X_i \sim Exp(\theta)$ (Continued):
 - Likelihood Estimation:

$$L_{1}(\theta) = \prod_{i=1}^{n} \theta^{-1} \exp(-y_{i}\theta^{-1}) \prod_{j=1}^{m} \left[1 - \{1 - \exp(-M\theta^{-1})\} \right]$$

$$= \theta^{-n} \exp\left(-\theta^{-1} \sum_{i=1}^{n} y_{i}\right) \exp(-mM\theta^{-1})$$

$$\implies l_{1}(\theta) = \ln\{L_{1}(\theta)\}$$

$$= -n \ln \theta - \theta^{-1} \sum_{i=1}^{n} y_{i} - mM\theta^{-1}$$

$$= -n \ln \theta - \theta^{-1} \sum_{i=1}^{n+m} y_{i},$$

since $y_{n+1} = \ldots = y_{n+m} = M$ (if y_i 's sorted).

So, MLE solves:

$$\frac{dl_1(\theta)}{d\theta} = -n\theta^{-1} + \theta^{-2} \sum_{i=1}^{n+m} y_i = 0$$

$$\implies \hat{\theta} = \frac{1}{n} \sum_{i=1}^{n+m} y_i = \frac{n+m}{n} \overline{y} = \left(1 + \frac{m}{n}\right) \overline{y}$$

- 1. Exponential Claims, $X_i \sim Exp(\theta)$ (Continued):
 - Likelihood Estimation (Continued):

Variance of MLE is:

$$\left[-E_{\theta} \left\{ \frac{d^{2}l_{1}(\theta)}{d\theta^{2}} \right\} \right]^{-1} = \left[-E_{\theta} \left\{ n\theta^{-2} - 2\theta^{-3} \sum_{i=1}^{n+m} Y_{i} \right\} \right]^{-1}
= \left[\theta^{-2}E_{\theta}(n) - 2\theta^{-3}(n+m)E_{\theta}(Y_{i}) \right]^{-1}
= \left[-(n+m)\theta^{-2}(1-e^{-M\theta^{-1}})
+ 2(n+m)\theta^{-3} \left\{ \theta(1-e^{-M\theta^{-1}}) \right\} \right]^{-1}
= \theta^{2} \left\{ (n+m)(1-e^{-M\theta^{-1}}) \right\}^{-1},$$

since $n \sim \text{Binomial}[n+m, Pr_{\theta}(X_i \leq M) = 1 - e^{-M/\theta}]$

- Numerical Example:
 - · Use our data, assume M = 15000.
 - · So, m=4 and the MLE of θ is:

$$\frac{1}{92} \left\{ \sum_{i=1}^{92} y_i + 4(15000) \right\} = 2491.38$$

· With approximate standard error:

$$\frac{2491.38}{\sqrt{96(1 - e^{-15000/2491.38})}} = 254.58$$

· Compare to $\hat{\theta}_{MLE}=2898.83,\,\hat{\sigma}(\hat{\theta}_{MLE})=305.15$ for full data.

- 1. Exponential Claims, $X_i \sim Exp(\theta)$ (Continued):
 - The Reinsurer's Perspective:
 - · Distribution of amounts for which reinsurer is liable, Z|Z>0:

$$f_{Z|Z>0}(z;\theta) = f_X(z+M;\theta)\{1 - F_X(M;\theta)\}^{-1}$$

$$= \theta^{-1} \exp\{-(z+M)\theta^{-1}\} \left[1 - \{1 - \exp(-M\theta^{-1})\}\right]^{-1}$$

$$= \theta^{-1} \exp\{-(z+M)\theta^{-1}\} \exp(M\theta^{-1})$$

$$= \theta^{-1} \exp(-z\theta^{-1})$$

· Distribution remains $Exp(\theta)$, "memoryless" property.

- 2. Pareto Claims, $X_i \sim Pareto(\alpha, \delta)$:
 - Expected Claim Size

$$\begin{split} E_{\alpha,\delta}(Y) &= E_{\alpha,\delta}(X) - \int_0^\infty y f_X(y+M;\alpha,\delta) dy \\ &= \frac{\delta}{\alpha - 1} - \int_0^\infty y \frac{\alpha \delta^\alpha}{\{\delta + (y+M)\}^{\alpha + 1}} dy \\ &= \frac{\delta}{\alpha - 1} - \frac{\delta^\alpha}{(\delta + M)^\alpha} \int_0^\infty y \frac{\alpha (\delta + M)^\alpha}{\{(\delta + M) + y\}^{\alpha + 1}} dy \\ &= \frac{\delta}{\alpha - 1} - \left\{ \frac{\delta^\alpha}{(\delta + M)^\alpha} \right\} \left(\frac{\delta + M}{\alpha - 1} \right) \\ &= \frac{\delta \{(\delta + M)^{\alpha - 1} - \delta^{\alpha - 1}\}}{(\alpha - 1)(\delta + M)^{\alpha - 1}} \end{split}$$

- Numerical Example
 - · With M = 15000, we see that

$$E_{\alpha,\delta}(X) - E_{\alpha,\delta}(Y) = \frac{\delta^{\alpha}}{(\alpha - 1)(\delta + 15000)^{\alpha - 1}}$$

- · For our data, $\hat{\alpha}_{MLE} = 1.909$ and $\hat{\delta}_{MLE} = 2704.47$
- · Predict a drop in expected claim size using these MLEs as:

$$\frac{\hat{\delta}_{MLE}^{\hat{\alpha}_{MLE}}}{(\hat{\alpha}_{MLE} - 1)(\hat{\delta}_{MLE} + 15000)^{\hat{\alpha}_{MLE} - 1}} = 539.23.$$

• Actual drop is 2989.83 - 2387.57 = 602.26

- 2. Pareto Claims, $X_i \sim Pareto(\alpha, \delta)$ (Continued):
 - Likelihood Estimation:

$$l_1(\alpha, \delta) = n \ln \alpha + (n+m)\alpha \ln \delta - (\alpha+1) \sum_{i=1}^n \ln(\delta + y_i) - m\alpha \ln(\delta + M)$$

· Score equations:

$$\frac{\partial l_1(\alpha, \delta)}{\partial \alpha} = \frac{n}{\alpha} + (n+m) \ln \delta - \sum_{i=1}^{n+m} \ln(\delta + y_i)$$
$$\frac{\partial l_1(\alpha, \delta)}{\partial \delta} = \frac{(n+m)\alpha}{\delta} - (\alpha+1) \sum_{i=1}^{n+m} \frac{1}{\delta + y_i} + \frac{m}{\delta + M}$$

- · Require iterative (computer-based) solution
- · For our data, with M=15000: $\hat{\alpha}_{MLE}=1.853$, $\hat{\delta}_{MLE}=2602.47$ [Compare to no censoring case: $\hat{\alpha}_{MLE}=1.909$, $\hat{\delta}_{MLE}=2704.47$]

- 2. Pareto Claims, $X_i \sim Pareto(\alpha, \delta)$ (Continued):
 - Likelihood Estimation (Continued):
 - · Fisher Information:

$$-E_{\alpha,\delta} \left\{ \frac{\partial^2 l_1(\alpha,\delta)}{\partial \alpha^2} \right\} = E_{\alpha,\delta} \left(\frac{n}{\alpha^2} \right) = \frac{n+m}{\alpha^2} \left\{ 1 - \frac{\delta^{\alpha}}{(\delta+M)^{\alpha}} \right\}$$

$$-E_{\alpha,\delta} \left\{ \frac{\partial^2 l_1(\alpha,\delta)}{\partial \alpha \partial \delta} \right\} = -E_{\alpha,\delta} \left(\frac{n+m}{\delta} - \sum_{i=1}^{n+m} \frac{1}{\delta+Y_i} \right)$$

$$= -\frac{n+m}{(\alpha+1)\delta} \left\{ 1 - \frac{\delta^{\alpha+1}}{(\delta+M)^{\alpha+1}} \right\}$$

$$-E_{\alpha,\delta} \left\{ \frac{\partial^2 l_1(\alpha,\delta)}{\partial \delta^2} \right\} = E_{\alpha,\delta} \left\{ \frac{(n+m)\alpha}{\delta^2} - \sum_{i=1}^{n+m} \frac{(\alpha+1)}{(\delta+Y_i)^2} + \frac{m}{(\delta+M)^2} \right\}$$

$$= \frac{(n+m)\alpha}{(\alpha+2)\delta^2} \left\{ 1 - \frac{\delta^{\alpha+2}}{(\delta+M)^{\alpha+2}} \right\}$$

Since $m \sim Binomial[n+m, \delta^{\alpha}(\delta+M)^{-\alpha}]$ and:

$$E_{\alpha,\delta}\{(\delta+Y_i)^{-k}\} = \int_0^M (\delta+y)^{-k} \alpha \delta^{\alpha} (\delta+y)^{-(\alpha+1)} dy$$
$$+ (\delta+M)^{-k} Pr_{\alpha,\delta}(Y_i = M)$$
$$= \alpha \delta^{\alpha} \int_0^M (\delta+y)^{-\{(\alpha+k)+1\}} dy$$
$$+ (\delta+M)^{-k} \left\{ \frac{\delta^{\alpha}}{(\delta+M)^{\alpha}} \right\}$$
$$= \frac{\alpha}{(\alpha+k)\delta^k} \left\{ 1 + \frac{k\delta^{\alpha+k}}{\alpha(\delta+M)^{\alpha+k}} \right\}$$

· Using MLEs and M = 15000 for our data:

$$\widehat{Var}_{\alpha,\delta}(\hat{\alpha}_{MLE}, \hat{\delta}_{MLE}) = \begin{pmatrix} 0.3546 & 670.16 \\ 670.16 & 1413396.83 \end{pmatrix}$$

 \cdot So, 95% confidence intervals are:

$$\alpha:$$
 1.853 \pm 1.96 $\sqrt{0.3546} = (0.6859, 3.0201)$
 $\delta:$ 2602.48 \pm 1.96 $\sqrt{1413396.83} = (272.30, 4932.64)$

- 2. Pareto Claims, $X_i \sim Pareto(\alpha, \delta)$ (Continued):
 - The Reinsurer's Perspective:
 - · Distribution of amounts for which reinsurer is liable, Z|Z>0:

$$f_{Z|Z>0}(z;\alpha,\delta) = \frac{f_X(z+M;\alpha,\delta)}{1 - F_X(M;\alpha,\delta)}$$

$$= \frac{\alpha\delta^{\alpha}}{\{\delta + (z+M)\}^{\alpha+1}} \left\{ \frac{\delta^{\alpha}}{(\delta+M)^{\alpha}} \right\}^{-1}$$

$$= \frac{\alpha(\delta+M)^{\alpha}}{\{(\delta+M)+z\}^{\alpha+1}}$$

since

$$Pr_{\alpha,\delta}(X_i > x) = 1 - F_X(x;\alpha,\delta) = \frac{\delta^{\alpha}}{(\delta + x)^{\alpha}}.$$

· Thus, Z|Z>0 is $\mathrm{Pareto}(\alpha,\delta+M)$