

### Main Examination period 2019

## **MTH5126: Statistics for Insurance**

**Duration: 2 hours** 

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You should attempt ALL questions. Marks available are shown next to the questions.

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Examiners: D. Boland, S. Liverani

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**Question 1.** [6 marks] An urn  $B_1$  contains 2 white and 3 black balls. A second urn  $B_2$  contains 3 white and 4 black balls. One urn is selected at random and a ball is drawn from it. If the ball drawn is black, find the probability that the urn chosen was  $B_1$ . It should be assumed that each ball in the selected urn is drawn with equal probability.

[6]

Let  $E_1, E_2$  denote the events of selecting  $B_1, B_2$  respectively. Then:

$$P(E_1) = P(E_2) = \frac{1}{2}$$

[1]

Let B denote the event that the ball chosen from the selected urn was black.

Then we have to find 
$$P(E_1|B)$$
 [1]

We have: 
$$P(B|E_1) = \frac{3}{5}$$
 and  $P(B|E_2) = \frac{4}{7}$ 

So, by Bayes' Theorem we get:

$$P(E_1|B) = \frac{P(B|E_1)P(E_1)}{P(B|E_1)P(E_1) + P(B|E_2)P(E_2)}$$

$$\frac{1}{2} \times \frac{3}{5}$$
[1]

$$= \frac{\frac{1}{2} \times \frac{3}{5}}{\frac{1}{2} \times \frac{3}{5} + \frac{1}{2} \times \frac{4}{7}}$$
$$= \frac{21}{41}$$

[2]

## Question 2. [4 marks]

Consider a general game:

$$\begin{array}{c|cccc} & & & \text{Player A} \\ & & \textbf{I} & \textbf{II} \\ \text{Player B} & \textbf{1} & \textbf{a} & \textbf{b} \\ \textbf{2} & \textbf{d} & \textbf{c} \\ \end{array}$$

- (a) Define a saddle point. [2]
- (b) State the conditions for which a saddle point exists in this game. [2]
- (a) A pair of strategies are in equilibrium if and only if the element L(a,b) corresponding to the equilibrium is both the largest in its column and the smallest in its row. Such equilibrium is called a *saddle point*.
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(b) A saddle point is the largest element in its column and the smallest in its row. We have 4 possibilities for a saddle: a,b,c,d. Assuming these are equally likely, then

We have 4 possibilities for a saddle: a,b,c,d. Assuming these are equally likely, then the conditions for a (for example) would be:

$$a > d$$
  
 $a < b$ 

**Question 3.** [15 marks] The table below shows claims paid on a portfolio of general insurance policies. You may assume that claims are fully run off after three years.

Underwriting year	Development Year			ear
	0	1	2	3
2012	650	412	217	91
2013	703	489	262	
2014	711	456		
2015	678			

Past claims inflation has been 2% p.a. However, it is expected that future claims inflation will be 4% p.a. Use the inflation adjusted chain ladder method to calculate the outstanding claims on the portfolio. [15]

We begin by finding the claim amounts uplifted to 2015 prices.

Underwriting year	Development Year					
	0	1	2	3		
2012	689.79	428.64	221.34	91		
2013	731.40	498.78	262			
2014	725.22	456				
2015	678					

[3]

Accumulating gives

Underwriting year	r Development Year			
	0	1	2	3
2012	689.79	1118.43	1339.77	1430.77
2013	731.40	1230.18	1492.18	
2014	725.22	1181.22		
2015	678			

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**Turn Over** 

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Hence the development factors are

$$DF_{0,1} = \frac{1118.43 + 1230.18 + 1181.22}{689.79 + 731.40 + 725.22} = \frac{3529.83}{2146.41} = 1.644527$$

$$DF_{1,2} = \frac{1339.77 + 1492.18}{1118.43 + 1230.18} = \frac{2831.95}{2348.61} = 1.205798$$

$$DF_{2,3} = \frac{1430.77}{1339.77} = 1.067922$$

The completed table at 2015 prices is

 Underwriting year
 Development Year

 0
 1
 2
 3

 2012
 3

 2013
 1593.53

 2014
 1424.31
 1521.05

 2015
 1114.99
 1344.45
 1435.77

Differencing gives

Underwriting year	Development Year			ear
	0	1	2	3
2012				
2013				101.35
2014			243.09	96.74
2015		436.99	229.46	91.32

Applying inflation for future years gives

Underwriting year	Development Year			ear
	0	1	2	3
2012				
2013				105.40
2014			252.81	104.63
2015		454.47	248.18	102.72

So the outstanding claims are

$$105.40 + 252.81 + \cdots + 102.72 = 1268.21$$

[1]

[3]

**[2**]

[3]

**[2**]

[1]

### Question 4. [12 marks]

A decision-maker with decreasing risk-aversion has assets of £30 and has a decision problem with the following structure:

	Outcome		
Decision	$\theta_1$	$\theta_2$	
$\phantom{aaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaa$	-10	+5	
$d_2$	+15	-5	
Probability	0.3	0.7	

The entries in the table represent gains or losses in pounds. For example, with decision  $d_1$  and outcome  $\theta_1$  the decision-maker will end up with assets of £20 and with decision  $d_2$  and outcome  $\theta_2$  they will end up with assets of £25.

You are given the following excerpt from the decision-maker's utility table:

£x	u(x)	£x	u(x)
0	0.000	180	0.797
10	0.221	185	0.802
15	0.300	190	0.807
20	0.364	195	0.811
25	0.415	200	0.816
30	0.458	210	0.825
35	0.493	220	0.834
40	0.523	230	0.842
45	0.548	240	0.849
50	0.570	250	0.847

- (a) Advise the decision-maker on whether decision  $d_1$  or  $d_2$  is the better course of action. [7]
- (b) Would your advice remain the same if the decision-maker's assets were £200? [5]

You should justify any advice given and show clear workings.

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(a) The investor currently has £30.

The first thing we should check is the monetary amounts he would have with each decision.

 $d_1$  and  $\theta_1$  means he would lose £10, leaving him with £20.

 $d_1$  and  $\theta_2$  means he would win £5, leaving him with £35.

 $d_2$  and  $\theta_1$  means he would win £15, leaving him with £45.

$$d_2$$
 and  $\theta_2$  means he would lose £5, leaving him with £25. [2]

We can now use the tables provided to check the relevant associated utilities.

$$u(30) = 0.458$$
$$u(20) = 0.364$$
$$u(35) = 0.493$$
$$u(45) = 0.548$$

u(25) = 0.415

Finally we can calculate the expected utilities and check which decision is better on this criteria.

Decision 
$$d_1$$
:  $\bar{u}(x) = 0.3 \times 0.364 + 0.7 \times 0.493 = 0.4543$  [2]

Decision 
$$d_2$$
:  $\bar{u}(x) = 0.3 \times 0.548 + 0.7 \times 0.415 = 0.4549$  [2]

Therefore we prefer decision  $d_2$  as the expected utility is higher at 0.4549.

It is worth noting that a further decision, say  $d_0$  corresponding to not investing at all would have a higher utility than  $d_1$  and  $d_2$  at 0.458.

Should a student point this out then discretionary marks will be awarded.

(b) When the decision-maker has assets of £200 then the expected utilities are:

$$u(200) = 0.816$$
  
 $u(190) = 0.807$   
 $u(205) = (0.816 + 0.825)/2 = 0.8205$   
 $u(215) = (0.825 + 0.834)/2 = 0.8295$   
 $u(195) = 0.811$ 

**[2**]

[2]

[2]

Decision  $d_1$ :  $\bar{u}(x) = 0.3 \times 0.807 + 0.7 \times 0.8205 = 0.81645$ Decision  $d_2$ :  $\bar{u}(x) = 0.3 \times 0.8295 + 0.7 \times 0.811 = 0.81655$ 

So we prefer decision  $d_2$  again and in fact the utility is very close to the initial amount invested.

[3]

### Question 5. [20 marks]

- (a) A new insurance policy is sold to a limited number of existing policy holders. The number of claims in the first six months is 18. Suppose the number of claims each month is assumed to have a Poisson distribution with mean  $\theta$ , independent of all other months. Write down the likelihood function.
- **[4**]
- (b) The prior distribution of  $\theta$  is given by a Gamma distribution. When designing the policy it was thought that based on the claims history of these policy holders the mean number of claims per month would be 4 claims with variance 1. Show that a Gamma distribution Gamma(16,4) is a suitable prior.
- **[2**]

(c) Find the posterior distribution of  $\theta$ .

**[4**]

(d) Find the form of the Bayes estimate of  $\theta$  under squared error loss.

[6]

**[4]** 

(e) Calculate the value of the Bayes estimate. Show it can be written as a weighted average of the prior mean and the data mean and interpret the weights.

(a) The likelihood is given by

$$p(\underline{x}|\theta) = \prod_{i=1}^{n} p(x_i|\theta) \propto e^{-n\theta} \theta^{S}$$

- [2]
- where  $S = \sum x_i$ .
- Here n = 6 and S = 18 so the likelihood is proportional to  $e^{-6\theta}\theta^{18}$  [1]
- (b) The mean and variance of a Ga(a,b) distribution are a/b and  $a/b^2$  (from table of distributions). So Ga(16,4) has mean 4 and variance 1. [2]
- (c) The prior  $p(\theta) \propto \theta^{15} e^{-4\theta}$ 
  - So the posterior  $p(\theta|(\underline{x})) \propto \theta^{15} e^{-4\theta} \theta^{18} e^{-6\theta} \propto \theta^{33} e^{-10\theta}$  [1]
  - So the posterior distribution is Ga(34, 10). [2]

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(d) If we have squared error loss  $l(t,\theta)=(t-\theta)^2$  then the posterior mean is the Bayes estimate. We can show this as follows.

We wish to minimise the expected loss

$$\lim_{t} \int (t-\theta)^2 p(\theta|\underline{x}) d\theta.$$

Differentiating with respect to t we have

$$\int 2(t-\theta)p(\theta|(\underline{x}))d\theta$$

setting this equal to zero

$$\hat{t} \int p(\theta|(\underline{x}))d\theta = \int \theta p(\theta|(\underline{x}))d\theta$$

and since the integral of the posterior density is one we have

$$\hat{t} = \int \theta p(\theta|(\underline{x})) d\theta = E[\theta|(\underline{x})].$$

(e) We have established in part (c) that the posterior distribution is Ga(34, 10), so it follows that the posterior mean is 3.4 and the posterior variance is 0.34. The posterior mean 3.4 is the Bayes estimate.

The prior mean was 4 and the sample mean 18/6 = 3. [1]

The weights are the relative information provided by the data (6 observations) and the prior (equivalent to 4 observations).

We can write the posterior mean as a weighted average of the sample mean and prior mean as follows:

$$3.4 = 3 \times \frac{6}{10} + 4 \times \frac{4}{10}$$

[1]

[1]

[1]

[1]

[2]

[1]

[1]

**Question 6.** [10 marks] The table below shows aggregate annual claim statistics for three risks over a period of eight years. Annual aggregate claims for risk i in year j are denoted by  $X_{ij}$ .

Risk i	$\bar{X}_{i.} = \frac{1}{8} \sum_{j=1}^{8} X_{ij}$	$S_i^2 = \frac{1}{7} \sum_{j=1}^8 (X_{ij} - \bar{X}_{i.})^2$
1	213.11	411.19
2	91.15	94.23
3	134.23	38.6

- (a) What are the assumptions of the Empirical Bayes Credibility Theory (EBCT) Model 1? [2]
- (b) Calculate the credibility premium of each risk under the assumptions of EBCT Model 1. [8]

- (a) The assumptions for EBCT Model 1
  - (i) The distribution of each  $X_j$  depends on a parameter, denoted  $\theta$ , whose value is fixed (and the same for all the  $X_j$ 's) but is unknown.
  - (ii) Given  $\theta$ , the  $X_i$ 's are independent and identically distributed.

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(b) The overall mean is

$$\frac{213.11 + 91.15 + 134.23}{3} = 146.16$$

and

$$E(S^{2}(\theta)) = \frac{1}{3} \sum_{i} S_{i}^{2} = \frac{411.19 + 94.23 + 38.6}{3} = 181.34$$

[1]

[1]

$$Var(m(\theta)) = \frac{1}{2} \sum (X_i - \bar{X})^2 - \frac{1}{8} E(S^2(\theta))$$

$$= \frac{1}{2} \left[ (213.11 - 146.16)^2 + (91.15 - 146.16)^2 + (134.23 - 146.16)^2 \right] - \frac{181.34}{8}$$

$$= \frac{1}{2} (4481.86 + 3026.47 + 142.40) - 22.6675$$

$$= 3825.36 - 22.6675 = 3802.70$$

[2]

So the credibility factor is

$$Z = \frac{8}{8 + \frac{181.34}{3802.70}} = 0.994074$$

[1]

Therefore the credibility premia for the risks are

$$0.994074 \times 213.11 + (1 - 0.994074) \times 146.16 = 212.71$$

[1]

$$0.994074 \times 91.15 + (1 - 0.994074) \times 146.16 = 91.48$$

[1]

$$0.994074 \times 134.23 + (1 - 0.994074) \times 146.16 = 134.30$$

[1]

[8]

**Question 7.** [10 marks] An insurance company has a portfolio of policies under which individual loss amounts follow an exponential distribution with mean  $\lambda^{-1}$ . In one year, the insurer observes 75 claims with mean claim amount 107.8.

- (a) Show that the maximum likelihood estimate of  $\lambda$  is 0.0093.
- (b) Find the probability that an individual loss amount will be greater than 107.8. [2]

(a) The likelihood function for the data is proportional to

$$\left(\lambda e^{-107.8\lambda}\right)^{75}$$

[3]

so the log-likelihood is

$$C + 75 \log \lambda - 75 \times 107.8 \lambda$$
.

[1]

Hence

$$\frac{d\log L}{d\lambda} = \frac{75}{\lambda} - 75 \times 107.8.$$

[1]

Setting this equal to zero and solving we see that the maximum likelihood estimate of  $\boldsymbol{\lambda}$  is

$$\hat{\lambda} = \frac{75}{75 \times 107.8} = \frac{1}{107.8} = 0.0093.$$

**[2**]

This is a maximum as

$$\frac{d^2 \log L}{d\lambda^2} = -\frac{75}{\lambda^2} < 0.$$

[1]

(b) The probability of an individual loss amount exceeding 107.8 is

$$\int_{107.8}^{\infty} \lambda e^{-\lambda x} dx = \left[ -e^{\lambda x} \right]_{107.8}^{\infty} = e^{-107.8\lambda}$$

Using  $\hat{\lambda} = 1/107.8$ , the probability is

$$e^{-107.8 \times \frac{1}{107.8}} = e^{-1}$$
.

**[2**]

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#### Question 8. [15 marks]

A manufacturing company is analysing the number of accidents that occur each year on the factory floor. They believe that the number of accidents per year, *N*, has a geometric distribution with parameter 0.7 so that:

$$P(N = n) = 0.7 \times 0.3^n$$
,  $n = 0, 1, 2, ...$ 

For each accident i, the number of employees injured is  $Y_i$ , where  $Y_i = X_i + 1$  and  $X_i$  is believed to have a Poisson(1.8) distribution.

The company has taken out an insurance policy, which provides a benefit of £5,000 to each injured employee, up to a maximum of three employees per accident, irrespective of the level of injury. There is no limit on the number of accidents that may be claimed for in a year.

- (a) Show that E(S) = 1.0166 and var(S) = 3.6859, where S is the total number of employees claiming benefit in a year under this policy. [9]
- (b) Hence find the mean and variance of the aggregate amount paid out under this policy in a year. [6]

(a) The aggregate amount paid out by the company is 5,000S, where:

$$S = Z_1 + \ldots + Z_n$$

is the total number of employees claiming benefit in a year. [1] N has the distribution given in the question and:

$$Z_i = \begin{cases} Y_i & Y_i < 3\\ 3 & Y_i \ge 3 \end{cases}$$

where  $Y_i = X_i + 1$  and  $X_i$  has a Poisson(1.8) distribution. We note that:

$$P(Y_i = 1) = e^{-1.8}$$
 (The Poisson probability that  $X_i = 0$ )

$$P(Y_i = 2) = 1.8e^{-1.8}$$
 (The Poisson probability that  $X_i = 1$ )  
 $P(Y_i \ge 3) = 1 - e^{-1.8} - 1.8 \cdot e^{-1.8} = 1 - 2.8e^{-1.8}$ 

$$P(Y_i \ge 3) = 1 - e^{-1.8} - 1.8 \cdot e^{-1.8} = 1 - 2.8e^{-1.8}$$
[3]

We must first find the mean and variance of  $Z_i$ :

$$E(Z_i) = \sum_{z=1}^{2} zP(Y_i = z) + 3P(Y_i \ge 3) = e^{-1.8} + 3.6e^{-1.8} + 3(1 - 2.8e^{-1.8}) = 3 - 3.8e^{-1.8} = 2.37186$$

[3]

$$E(Z_i^2) = \sum_{z=1}^{2} z^2 P(Y_i = z) + 3^2 P(Y_i \ge 3) = e^{-1.8} + 7.2e^{-1.8} + 9(1 - 2.8e^{-1.8}) = 9 - 17e^{-1.8} = 6.1899$$

$$var(Z_i) = E(Z_i^2) - (E(Z_i))^2 = 6.1899 - 2.37186^2 = 0.5642$$

**[2**]

The expectation and the variance of the number of claims per year are as follows.

$$E(N) = \frac{1-p}{p} = \frac{0.3}{0.7} = 0.4286$$

[1]

$$var(N) = \frac{1-p}{p^2} = \frac{0.3}{0.7^2} = 0.6122$$

[1]

So the mean and variance of *S* are:

$$E(S) = E(Z_i)E(N) = 2.37186 \times 0.4286 = 1.0166$$

[1]

$$var(S) = (E(Z_i))^2 var(N) + var(Z)E(N) = 2.37186^2 \times 0.6122 + 0.5642 \times 0.4286 = 3.6859$$

**[2**]

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(b) The mean and variance of 5,000S are:

$$E(5,000S) = 5,000E(S) = 5,082.5$$

[3]

$$var(5,000S) = 5,000^2 var(S) = 92,152,500$$

[3]

[Follow through marks were awarded where a calculation error was made]

## Question 9. [8 marks]

- (a) Explain the difference between **IBNR** and an **Outstanding Claims Reserve.** [4]
- (b) Explain the difference between **Proportional** and **Non-Proportional** reinsurance and give two examples of each. [4]

(a) An IBNR reserve is required in respect of claims that have been incurred but not [1] reported. This means that the claim event has occurred but the claim has not yet been reported to the insurer. [1] An Outstanding Claims reserve is required in respect of claims that may have been reported but have not yet been closed. [1] OCR's are usually calculated by run-off triangle method whereas IBNR uses Loss **Ratios** [1] IBNR happens earlier in the typical four part claims process Claim event occurs - Claim reported - Claim payment made - Claim file closed. [1] Any other sensible comment award one mark to a max of four.

(b)	Under a proportional reinsurance arrangement, the ceding company and the reinsurer	
	share in all aspects of the risk. Most importantly, this means they share in the cost of	
	claims for the risk. The cedant must pay a premium for this. The premium calculations	
	will be based on the reinsurance treaty.	[1]
	Examples are:	
	Quota share reinsurance: Where the proportions are the same for all risks.	
	Surplus reinsurance: Where the proportions can vary from risk to risk.	[1]
	With non-proportional reinsurance the cedant pays a fixed premium to the reinsurer.	
	The reinsurer will then pay the part of the claim that lies in a particular reinsurance	
	layer. The layer itself will be defined by a lower limit called the retention limit and an	
	upper limit. The upper limit can be infinite if the cover is unlimited.	[1]
	Examples are:	
	Individual Excess of Loss: The reinsurer will make payment when the amount of a	
	claim exceeds a specified excess point.	[1]
	For example a reinsurer might pay the excess when the amount of a particular claim	
	from a car insurance policy exceeds £50,000 but to an upper limit of £1m	[1]
	Stop Loss: With Stop Loss insurance the reinsurer will make a payment if the total	
	claim amount for a specified group of policies exceeds a specified amount. The amount	
	is usually expressed as a percentage of the premium.	[1]
	For example a reinsurer might agree to pay 85% of the excess when the total claim	
	amount for all car insurance policies exceeds 100% of the total premium with no upper	
	limit.	[1]
	Maximum of four marks available. Two marks must be for definition of	
	proportional, non-proportional and two for the examples. One mark for either	
	proportional, non-proportional and two for the champion one mark for either	

defining the specific type of reinsurance or giving an example of its use

End of Paper - An appendix of 2 pages follows.

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# **Statistics – Common Distributions**

## **Discrete Distributions**

Distribution	Density	Range of Variates	Mean	Variance	
Uniform	$\frac{1}{N}$	$N = 1, 2, \dots$ $x = 1, 2, \dots, N$	$\frac{N+1}{2}$	$\frac{N^2-1}{12}$	
Bernoulli	$p^x(1-p)^{1-x}$	$0 \le p \le 1, x = 0, 1$	p	p(1-p)	
Binomial	$\binom{n}{x}p^x(1-p)^{n-x}$	$0 \le p \le 1, n = 1, 2, \dots$ $x = 0, 1, \dots n$	np	np(1-p)	
Poisson	$\frac{e^{-\lambda}\lambda^x}{x!}$	$\lambda > 0, x = 0, 1, 2, \dots$	λ	λ	
Geometric	$p(1-p)^x$	$0$	$\frac{(1-p)}{p}$	$\frac{(1-p)}{p^2}$	
Continuous Distributions					

Uniform	$\frac{1}{b-a}$	$-\infty < a < b < \infty$ $a < x < b$	$\frac{a+b}{2}$	$\frac{(b-a)^2}{12}$
Normal $N(\mu, \sigma^2)$	$\frac{1}{\sqrt{2\pi\sigma^2}}\exp\left[\frac{-(x-\mu)^2}{2\sigma^2}\right]$	$-\infty < \mu < \infty$	μ	$\sigma^2$
Lognormal $(\mu, \sigma^2)$	$\frac{1}{r\sqrt{2\pi\sigma^2}}\exp\left[\frac{-(logx-\mu)^2}{2\sigma^2}\right]$	$\sigma > 0, -\infty < x < \infty$ $-\infty < \mu < \infty$	$e^{(\mu+rac{1}{2}\sigma^2)}$	$e^{(2\mu+\sigma^2)}(e^{\sigma^2}-1)$
Exponential	$\lambda e^{-\lambda x}$	$\sigma > 0, -\infty < x < \infty$ $\lambda > 0, x \ge 0$	$\frac{1}{\lambda}$	$\frac{1}{\lambda^2}$
Gamma $(\alpha, \lambda)$	$\frac{\lambda^{\alpha} x^{\alpha-1} e^{-\lambda x}}{\Gamma(\alpha)}$	$\lambda > 0,  \alpha > 0,  x > 0$	$rac{lpha}{\lambda}$	$rac{lpha}{\lambda^2}$
Weibull $(c, \gamma)$	$c\gamma x^{\gamma-1}e^{-cx^{\gamma}}$	$c > 0, \gamma > 0, x > 0$	$c^{-\frac{1}{\gamma}}\Gamma(1+\gamma^{-1})$	$c^{-\frac{2}{\gamma}} [\Gamma(1+2\frac{1}{\gamma})]$
Pareto $(\alpha, \lambda)$	$\frac{\alpha\lambda^{\alpha}}{(\lambda+x)^{\alpha+1}}$	$\alpha > 0, \lambda > 0, x > 0$	$rac{\lambda}{(lpha-1)}$	$\frac{-\Gamma^2(1+\frac{1}{\gamma})]}{\frac{\lambda^2\alpha}{(\alpha-1)^2(\alpha-2)}}$
$\mathrm{Burr}(\alpha,\lambda,\gamma)$	$\frac{\alpha\gamma\lambda^{\alpha}x^{\gamma-1}}{(\lambda+x^{\gamma})^{\alpha+1}}$	$\alpha > 0, \lambda > 0, \gamma > 0, x > 0$	Not required	Not required

## **Useful Formulae**

### **EBCT Model 1**

$$E[m(\theta)] = \bar{X}$$

$$E[s^{2}(\theta)] = \frac{1}{N} \sum_{i=1}^{N} \frac{1}{(n-1)} \sum_{j=1}^{n} (X_{ij} - \bar{X}_{i})^{2}$$

$$var[m(\theta)] = \frac{1}{(N-1)} \sum_{i=1}^{N} (\bar{X}_{i} - \bar{X})^{2} - \frac{1}{Nn} \sum_{i=1}^{N} \frac{1}{(n-1)} \sum_{i=1}^{n} (X_{ij} - \bar{X}_{i})^{2}$$

### **EBCT Model 2**

$$E[m(\theta)] = \bar{X}$$

$$E[s^{2}(\theta)] = \frac{1}{N} \sum_{i=1}^{N} \frac{1}{(n-1)} \sum_{j=1}^{n} P_{ij} (X_{ij} - \bar{X}_{i})^{2}$$

$$var[m(\theta)] = \frac{1}{P^{*}} \left[ \frac{1}{Nn-1} \sum_{i=1}^{N} \sum_{j=1}^{n} P_{ij} (X_{ij} - \bar{X}_{i})^{2} - \frac{1}{N} \sum_{i=1}^{N} \frac{1}{(n-1)} \sum_{j=1}^{n} P_{ij} (X_{ij} - \bar{X}_{i})^{2} \right]$$

## **Intermediate calculations**

$$\sum_{j=1}^{n} P_{ij} = \bar{P}_{i} \qquad \sum_{i=1}^{N} \bar{P}_{i} = \bar{P} \qquad \frac{1}{(Nn-1)} \sum_{i=1}^{N} \bar{P}_{i} \left(1 - \frac{\bar{P}_{i}}{\bar{P}}\right) = P^{*}$$

$$\sum_{j=1}^{n} \frac{P_{ij}X_{ij}}{\bar{P}_{i}} = \bar{X}_{i} \qquad \qquad \sum_{i=1}^{N} \sum_{j=1}^{n} \frac{P_{ij}X_{ij}}{\bar{P}} = \bar{X}$$

End of Appendix.