# Institute of Actuaries of India 

## Subject CT6 - Statistical Methods

## March 2018 Examination

## INDICATIVE SOLUTION

Introduction

The indicative solution has been written by the Examiners with the aim of helping candidates. The solutions given are only indicative. It is realized that there could be other points as valid answers and examiner have given credit for any alternative approach or interpretation which they consider to be reasonable.

## Solution 1:

i) The payoff matrix is given by,

|  |  | Player A |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\mathbf{1}$ | $\mathbf{2}$ | $\mathbf{3}$ | $\mathbf{4}$ | $\mathbf{5}$ | $\mathbf{6}$ |  |
| Player B | $\mathbf{1}$ | 10 | 10 | - | - | - | - |  |
|  | $\mathbf{2}$ | 10 | 10 | 10 | - | - | - |  |
|  | $\mathbf{3}$ | - | 10 | 10 | 10 | - | - |  |
|  | $\mathbf{4}$ | - | - | 10 | 10 | 10 | - |  |
|  | $\mathbf{5}$ | - | - | - | 10 | 10 | 10 |  |
|  | $\mathbf{6}$ | - | - | - | - | 10 | 10 |  |

ii) Here we can see that for Player A, number 1 and 6 are dominated by number 2 and 5 respectively. Hence the revised payoff matrix can be written as:

|  |  | Player A |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\mathbf{3}$ | $\mathbf{4}$ | $\mathbf{5}$ |  |
| Player B | $\mathbf{1}$ | 10 | - | - | - |
|  | $\mathbf{2}$ | 10 | 10 | - | - |
|  | $\mathbf{3}$ | 10 | 10 | 10 | - |
|  | $\mathbf{4}$ | - | 10 | 10 | 10 |
|  | $\mathbf{5}$ | - | - | 10 | 10 |
|  | $\mathbf{6}$ | - | - | - | 10 |

Now for Player B, number 1 dominates number 2 and 3 and similarly number 6 dominates number 4 and 5 . Hence revised payoff matrix can be written as

|  |  | Player A |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\mathbf{3}$ | $\mathbf{4}$ | $\mathbf{5}$ |  |
| Player B | $\mathbf{1}$ | 10 | - | - | - |
|  | $\mathbf{6}$ | - | - | - | 10 |

Now number 3 and 4 are dominated for Player A, hence the revised payoff matrix is

|  |  | Player A |  |
| :---: | :---: | :---: | :---: |
|  |  | $\mathbf{5}$ |  |
| Player B | $\mathbf{1}$ | 10 | - |
|  | $\mathbf{6}$ | - | 10 |

iii) Player B should select either number 1 or 6 with probability 0.5 .

Hence value for Player $B=-10 * 0.5=-5$.

## Solution 2:

i) Under a proportional reinsurance arrangement, the direct writer (ie the original insurance company) and the reinsurer share the cost of all claims for each risk.

Under a non-proportional reinsurance arrangement, the direct writer pays a fixed premium to the reinsurer. The reinsurer will only be required to make payments where part of the claim amount falls in a particular reinsurance layer.
ii) Proportional reinsurance operates in two forms:

With quota share reinsurance, the proportions are the same for all risks With surplus reinsurance, the proportions can vary from one risk to the next
Two forms of non-proportional reinsurance are:-
Individual excess of loss (XOL) reinsurance, the reinsurer will be required to make a payment when the claim amount for any individual claim exceeds a specified excess point or retention.

Stop loss reinsurance, the reinsurer will be required to make payments if the total claim amount for a specified group of policies exceeds a specified amount (usually expressed as a percentage of the gross premium).
iii)
a) If $X$ represents the size of an individual claim, then we have:
$E(X)=\frac{\alpha}{\lambda}$
and, Variance $(X)=\frac{\alpha}{\lambda^{2}}$
Hence, $\mathrm{E}(\mathrm{X})=3.33$
Variance $(X)=1.11$
Amount paid by Insurer is $\mathrm{Y}=0.2 \mathrm{X}$

$$
E(Y)=0.2 \times 3.33=0.67
$$

$$
\begin{equation*}
\text { Variance }(\mathrm{Y})=0.2^{2} \times 1.11=0.04 \tag{3}
\end{equation*}
$$

b) Amount paid by Reinsurer $Z=0.8 \mathrm{X}$
$E(Z)=0.8 \times 3.33=2.67$
Variance $(Z)=0.8^{2} \times 1.11=0.71$
iv) $\int_{2500}^{M}(X-2500) \lambda e^{-\lambda x} d x+\int_{M}^{\infty}(M-2500) \lambda e^{-\lambda x} d x=5000$

Where, $\lambda=1 / 10,000$

Solving the above equation using integration by parts, we get:-
$10000 \mathrm{e}^{-2500 \lambda}-10000 \mathrm{e}^{-\mathrm{M} \lambda}=5000$

Solving for $\mathrm{M}=12773$

## Solution 3

i) The Bornhuetter-Ferguson method improves on the crude use of a loss ratio by taking account of the information provided by the latest development pattern of the claims, whilst the addition of the loss ratio to a projection method serves to add some stability against distortions in the development pattern.

The concepts behind the method are:

- That whatever claims have already developed in relation to a given origin year, the future development pattern will follow that experienced for other origin years.
- The past development for a given origin year does not necessarily provide a better clue to future claims than the more general loss ratio.
[2]
ii) The development of claims for various accident year and the expected ultimate claim amount is summarized in the table below:

| Accident Year | Development Year |  |  |  | Earned Premium | Expected loss ratio | Expected loss amount |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 0 | 1 | 2 | 3 |  |  |  |
| 2014 | 40\% | 55\% | 70\% | 75\% | 120 | 80\% | 96 |
| 2015 | 35\% | 50\% | 75\% |  | 150 | 90\% | 135 |
| 2016 | 60\% | 70\% |  |  | 140 | 105\% | 147 |
| 2017 | 55\% |  |  |  | 180 | 95\% | 171 |

The cumulative claim amount based on the earned premium and the development factors are given by:

| Accident Year | Development Year |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  | $\mathbf{0}$ | $\mathbf{1}$ | $\mathbf{2}$ | $\mathbf{3}$ |
|  | 48 | 66 | 84 | 90 |
| 2015 | 52.5 | 75 | 112.5 |  |


| 2016 | 84 | 98 |  |  |
| :--- | :--- | :--- | :--- | :--- |
| 2017 | 99 |  |  |  |


| Total | 283.5 | 239 | 196.5 | 90 |
| :--- | ---: | ---: | ---: | ---: |
| Total - last no | 184.5 | 141 | 84 |  |
| Ratio (r) | 1.295393 | 1.393617 | 1.071429 | 1 |
| DF (f) | 1.93423 | 1.493161 | 1.071429 | 1 |

Revised estimate of total ultimate losses, by accident year is given by

| Accident Year | $\mathbf{2 0 1 7}$ | $\mathbf{2 0 1 6}$ | $\mathbf{2 0 1 5}$ | $\mathbf{2 0 1 4}$ |
| :--- | :---: | :---: | :---: | :---: |
|  |  |  |  |  |
| f | 1.93423 | 1.493161 | 1.071429 | 1 |
| $1-1 / \mathrm{f}$ | 0.482998 | 0.33028 | 0.066667 | 0 |
| Initial UL | 171 | 147 | 135 | 96 |
|  |  |  |  |  |
| Emerging Liability | 82.59 | 48.55 | 9.00 | - |
| Reported Liability | 99.00 | 98.00 | 112.50 | 90.00 |
| Ultimate Liability | 181.59 | 146.55 | 121.50 | 90.00 |

Now, there is 1 year waiting period to make payment from date of intimation of claim.

| Accident Year | $\mathbf{2 0 1 7}$ | $\mathbf{2 0 1 6}$ | $\mathbf{2 0 1 5}$ | $\mathbf{2 0 1 4}$ |
| :--- | :---: | :---: | :---: | :---: |
| Claim paid | 0 | 84 | 75 | 84 |
| Outstanding claim reserve | 181.59 | 62.55 | 46.50 | 6.00 |

Hence the outstanding total claim reserve $=296.64$

## Solution 4:

i) For region Metro, the log-likelihood function using poisson parameter $\lambda 1$ is given by,
$\log L_{1}=\log \lambda_{1} \sum_{i=1}^{9} x_{1 i}-9 \lambda_{1}-\sum_{i=1}^{9} \log x_{1 i}!$..(0.5)
Or, $\log L_{1}=37 \log \lambda_{1}-9 \lambda_{1}-\sum_{i=1}^{9} \log x_{1 i}!$..(0.5)
Taking partial derivative we get,
$\frac{\partial}{\partial \lambda_{1}} \log _{1}=\frac{37}{\lambda_{1}}-9=0, \lambda_{1}=4.111$..(0.5)
Similarly, $\log L_{2}=\log \lambda_{2} \sum_{i=1}^{8} x_{2 i}-8 \lambda_{2}-\sum_{i=1}^{8} \log x_{2 i}!$..(0.5)
Hence, $\lambda_{2}=3.625, \lambda_{3}=5.857$
ii) We need to test, $\lambda_{1}=\lambda_{2}=\lambda_{3}=\lambda$, where $\lambda=107 / 24=4.458$..(0.5)

We can use scaled deviance to compare these models:
The difference in the scaled deviance follows chi-square with (3-1 = 2 ) degrees of freedom.
$2\left(\log L_{1}+\log L_{2}+\log L_{3}-\log L\right)=2\left(37 \log \lambda_{1}-9 \lambda_{1}+29 \log \lambda_{2}-8 \lambda_{2}+41 \log \lambda_{3}-7 \lambda_{3}-107 \log \lambda+24 \lambda\right)$
(factorial terms cancel each other)
$=4.3753$
The critical value at $5 \%$ level $=5.991$

The calculated value is less than the critical level and hence the mean claim rate are not different.

## Solution 5:

Here, $P(X>200)=1 / 250=0.004$

Now, under exponential distribution we get
$\operatorname{Exp}(-\lambda * 200)=0.004$
Hence, $\lambda=0.027607$

Now, $\operatorname{Exp}(-\lambda * Z)=0.0001$
Hence $Z=333.62$

Now, under Weibull distribution we get
$\operatorname{Exp}\left(-\lambda^{*} 200 \wedge 2\right)=0.004$
Hence, $\lambda=0.000138$

Now, $\operatorname{Exp}\left(-\lambda^{*} Z^{\wedge} 2\right)=0.0001$
Hence $Z=258.31$

Hence $Z$ is higher under Exponential
Exponential has longer tail than Weibull and hence $Z$ will be higher under Exponential distribution.
[6 Marks]

## Solution 6:

i) The probability density function of $h(x)$ is given by,
$h(x)=\exp (-x / \lambda) / \lambda, x>0$
Now, $F_{h}(x)=1-\exp (-x / \lambda)=U$,
We get, $X=-\lambda \ln (1-U)=-\lambda \ln U^{\prime}$,
Step 1: Simulate a u from $U(0,1)$
Step 2: Return $X=-\lambda \ln (u)$
ii) Now $f(x)$ follows chi-square with $\alpha$ degrees of freedom ie it follows Gamma( $\alpha / 2,0.5$ )

So, $\mathrm{f}(\mathrm{x})=\frac{0.5^{\frac{\alpha}{2}}}{\Gamma(0.5)} x^{\frac{\alpha}{2}-1} e^{-0.5 x}$
Now, $\mathrm{f}(\mathrm{x}) / \mathrm{h}(\mathrm{x})$
$=\frac{\lambda 0.5^{\frac{\alpha}{2}}}{\Gamma(0.5)} x^{\frac{\alpha}{2}-1} e^{-0.5 x+\frac{x}{\lambda}}, x>0$
Taking log both side we get,
$\operatorname{Ln}(\mathrm{f}(\mathrm{x}) / \mathrm{h}(\mathrm{x}))=\ln \left(\frac{\lambda 0.5^{\frac{\alpha}{2}}}{\Gamma(0.5)}\right)+\left(\frac{\alpha}{2}-1\right) \ln x-0.5 x+\frac{x}{\lambda}$
Now taking derivative and equating to zero we get,
$x=\left(\frac{\alpha}{2}-1\right) /\left(0.5-\frac{1}{\lambda}\right) \quad . .(1.0)$

Now the second derivate is $-\left(\frac{\alpha}{2}-1\right) \frac{1}{x^{2}}<0$..(0.5)
Hence, $\mathrm{f} / \mathrm{h}$ attains maximum value at $x_{0}=\left(\frac{\alpha}{2}-1\right) /\left(0.5-\frac{1}{\lambda}\right)$

Hence $\mathrm{C}=\max (\mathrm{f} / \mathrm{h})=\frac{\lambda 0.5^{\frac{\alpha}{2}}}{\Gamma(0.5)} x_{0}^{\frac{\alpha}{2}-1} e^{-\left(0.5-\frac{1}{\lambda}\right) x_{0}}$

Now, the function $\mathrm{g}(\mathrm{x})=\mathrm{f}(\mathrm{x}) /\left[\mathrm{C}^{*} \mathrm{~h}(\mathrm{x})\right]=\frac{x^{\frac{\alpha}{2}-1} e^{-\left(0.5-\frac{1}{\lambda}\right) x}}{x_{0}{ }^{\frac{\alpha}{2}-1} e^{-\left(0.5-\frac{1}{\lambda}\right) x_{0}}}$

Step 1: Simulte U 1 from $\mathrm{U}(0,1)$ and set $Y=-\ln (u 1) / \lambda$
Step 2: Simulate $U 2$ from $U(0,1)$ and if $U 2<g(Y)$ set accept the value by setting $X=Y$ otherwise move to Step 1
iii) The algorithm will be efficient by choosing value of $\lambda$ such that $g(x)$ is maximum or $C$ is minimum.

Now $\mathrm{C}=\frac{\lambda 0.5^{\frac{\alpha}{2}}}{\Gamma(0.5)} x_{0}^{\frac{\alpha}{2}-1} e^{-\left(0.5-\frac{1}{\lambda}\right) x_{0}}=\frac{0.5^{\frac{\alpha}{2}}}{\Gamma(0.5)} \lambda\left[\left(\frac{\alpha}{2}-1\right) /\left(0.5-\frac{1}{\lambda}\right)\right]^{\frac{\alpha}{2}-1} e^{1-\frac{\alpha}{2}}$
Which is minimized if, $\ln \left[\lambda\left[\left(\frac{\alpha}{2}-1\right) /\left(0.5-\frac{1}{\lambda}\right)\right]^{\frac{\alpha}{2}-1}\right]$ is minimized
i.e,
$\ln \lambda-\left(\frac{\alpha}{2}-1\right) \ln \left(0.5-\frac{1}{\lambda}\right)$
Taking derivative and equating to zero we get,
$\frac{1}{\lambda}-\left(\frac{\alpha}{2}-1\right) \frac{1}{\left(0.5-\frac{1}{\lambda}\right)} \frac{1}{\lambda^{2}}=0$
Or, $\left(0.5-\frac{1}{\lambda}\right) \lambda=\left(\frac{\alpha}{2}-1\right)$
Or, $\lambda=\alpha$
[12 Marks]

## Solution 7

i) Lundberg's inequality states that: $\psi(\mathrm{U}) \leq \exp (-\mathrm{RU})$
where $U$ is the insurer's initial surplus and $\psi(U)$ is the probability of ultimate ruin. $R$ is a parameter associated with a surplus process known as the adjustment coefficient. Its value depends upon the distribution of aggregate claims and on the rate of premium income. [2]
ii) $\exp (R x) \leq \frac{\mathrm{x}}{\mathrm{M}} \exp (R M)+1-\frac{\mathrm{x}}{\mathrm{M}}$ for $0 \leq \mathrm{x} \leq \mathrm{M}$

Now,

$$
\lambda \mathrm{M}_{\mathrm{x}}(\mathrm{R})=\lambda+\mathrm{cR}
$$

i.e.
$\lambda+c R=\lambda \int_{0}^{M} \exp (R x) f(x) d x$
$\leq \lambda \int_{0}^{\mathrm{M}}\left\{\frac{\mathrm{x}}{\mathrm{M}} \exp (\mathrm{RM})+1-\frac{\mathrm{x}}{\mathrm{M}}\right\} \mathrm{f}(\mathrm{x}) \mathrm{dx}$
This gives, $R>\frac{1}{M} \log \left(\frac{c}{\lambda m_{1}}\right)$
iii) Ruin will occur if the time of the first claim $t$ is such that
$\mathrm{U}+1.5 \lambda \mathrm{dt}<\mathrm{d}$
$\mathrm{t}<\frac{\mathrm{d}-\mathrm{U}}{1.5 \lambda \mathrm{~d}}$
The time until the first claim follows an exponential distribution with parameter $\lambda$, so the probability of ruin in given by:
$\int_{0}^{\frac{d-U}{1.5 \lambda d}} \lambda e^{-\lambda x} d x$
$=1-e^{\frac{-1\left(1-\frac{\mathrm{U}}{\mathrm{d}}\right)}{1.5}}$
iv) $\mathrm{M}_{\mathrm{x}}(\mathrm{t})=\mathrm{E}\left(\mathrm{e}^{\mathrm{tx})}=\frac{0.05^{2}}{(\mathrm{t}-0.05)^{2}}\right.$

The adjustment coefficient is the unique positive solution of
$\mathrm{M}_{\mathrm{x}}(\mathrm{R})=1+1.3 \mathrm{E}(\mathrm{X}) \mathrm{R}$
$E(X)=M_{x}^{\prime}(0)$
$=40$
Thus we need to solve
$\frac{0.05^{2}}{(\mathrm{R}-0.05)^{2}}=1+1.3 \times 40 \mathrm{R}$
$52 R^{2}-4.2 R+0.03=0$
$\mathrm{R}=0.07284$ OR 0.00791
So taking the smaller root we have $\mathrm{R}=0.00791$ since that is less than 0.01

## Solution 8

i)

| Strategy | Minimum Profit |
| :---: | :---: |
| Freekart.com | 14 |
| Appdeal.com | 9 |
| E-mazon.com | 5 |

Hence, the minimax strategy is to sell on Freekart.com
[2]
ii)

| Strategy | Maximum Profit |
| :---: | :---: |
| Freekart.com | 28 |
| Appdeal.com | 30 |
| E-mazon.com | 29 |

Hence the maximmax strategy is to sell on Appedeal.com
iii)

| Strategy | Expected profit |
| :--- | :--- |
| Freekart.com | $\frac{1}{3}(28+19+14)=20.33$ |
| Appdeal.com | $\frac{1}{3}(9+30+15)=18$ |
| E-mazon.com | $\frac{1}{3}(5+16+29)=16.67$ |

Hence the strategy selected by the Bayes criterion is Freekart.com
iv) If probability of medium demand is $p$ then the probabilities of High and Low demands are 0.50.5 P each.

Hence revised expected profits from each strategy are:-

| Strategy | Expected profit |
| :---: | :--- |
| Freekart.com | $28(0.5-0.5 \mathrm{P})+19 \mathrm{P}+14(0.5-0.5 \mathrm{P})=21-2 \mathrm{P}$ |
| Appdeal.com | $9(0.5-0.5 \mathrm{P})+30 \mathrm{P}+15(0.5-0.5 \mathrm{P})=12+18 \mathrm{P}$ |
| E-mazon.com | $5(0.5-0.5 \mathrm{P})+16 \mathrm{P}+29(0.5-0.5 \mathrm{P})=17-1 \mathrm{P}$ |

As there are more than even chances of it being medium, so $p$ is expected to be in between 0.5 t0 1
Hence Appdeal.com is the optimal decision.

## Solution 9

i) GLMs are widely used both in general and life insurance. They are used to:

- determine which rating factors to use (rating factors are measurable or categorical factors that are used as proxies for risk in setting premiums)
- Estimate an appropriate premium to charge for a particular policy given the level of risk present.

For example, in motor insurance, there are a large number of factors that may be used as proxies for the level of risk (type of car driven, age of driver, number of years past driving experience, etc). We can use a GLM both to determine which of these factors are significant to
the assessment of risk (and hence which should be included) and to suggest an appropriate premium to charge for a risk that represents a particular combination of these factors.
ii) The three components of a GLM are:

- A distribution for the data (Poisson, exponential, gamma, normal or binomial)
- A linear predictor (a function of the covariates that is linear in the parameters)
- A link function (that links the mean of the response variable to the linear predictor).


## iii)

a) In parameterized form, the linear predictors are (with $\mathrm{i}, \mathrm{j}$ and k corresponding to the levels of A, B, C respectively

Model $1=\alpha_{\mathrm{i}}+\beta_{\mathrm{j}}+\gamma_{\mathrm{k}}$; total 4 parameters
There is one parameter to set the base level for the combination $\mathrm{A}, \mathrm{B}$ and C and one additional parameter for each of the higher levels of the three factors

Model $2=\alpha_{\mathrm{ij}}+\gamma_{\mathrm{k}}$; total 5 parameters
There are four parameters for the $2 \times 2$ combinations of $A$ and $B$ and one additional parameter for the higher level of TC.

Model $3=\alpha_{\mathrm{ijk}}$; total 8 parameters
There are eight parameters for the $2 \times 2 \times 2$ combinations of $A, B$ and $C$
b)

| Model | Scaled <br> Deviance | Degrees of <br> Freedom | Change in Scaled <br> Deviance | Chance in Degrees of <br> Freedom |
| :---: | :---: | :---: | :---: | :---: |
| 1 | 40 | $\mathrm{n}^{*}$ (say) |  |  |
| $\mathrm{A}+\mathrm{B}+\mathrm{C}$ | 11 | $\mathrm{n}-4$ | 29 | 3 |
| $\mathrm{~A}+\mathrm{B}+\mathrm{C}+\mathrm{A} . \mathrm{B}$ | 7 | $\mathrm{n}-3$ | 4 | 1 |
| $\mathrm{~A}^{*} \mathrm{~B}^{*} \mathrm{C}$ | 0 | $\mathrm{n}-7$ | 7 | 3 |

*Candidates will also get full marks if they estimate n as (2*2*2-1=7)

|  | If Change in Deviance <br> $>2 \times$ change in df | Conclusion |
| :--- | :--- | :--- |
| Constant Model vs Model 1 | $29>2 \times 3$ | Model 1 is better than constant <br> model |
| Model 1 vs Model 2 | $4>2 \times 1$ | Model 2 i better than model 1 |
| Model 2 vs Model 3 | $7>2 \times 3$ | Model 3 is better than Model 2 |

Hence Model 3 is the most appropriate model

