# Institute of Actuaries of India 

## Subject CS1-Actuarial Statistics (Paper A)

## September 2021 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) a) Descriptive analysis: It involves summarising the data or presenting it in a format which highlights any patterns or trends i.e. producing summary statistics like measures of central tendency and dispersion. It describes a data set rather than giving any specific conclusions.

Example (any one point should also fetch marks):

- Calculating mean and standard deviation of number of motor claims in a day
- Plotting graphs on the average rainfall every month to illustrate the months with heaviest rainfall
b) Inferential analysis: This involves estimating the summary parameters of a population based on the sample data set under consideration and testing hypotheses.

Example (any one point should also fetch marks):

- Any example of hypothesis testing (people visit malls more on weekend than on a weekday
- Rate of health claims in India is same as health claims made in Tier I cities
c) Predictive analysis: This extends the principle behind inferential analysis in order for the user to analyse the past data and make predictions about the future event.

Example (any one point should also fetch marks):

- Predicting the number of lapses that will happen in the future years' basis the number of lapses in the last 1 year
- Forecasting the number of customers who would move to using electric vehicles
ii) C


## Solution 2:

Let $X_{i}(i=1,2,3)$ denote the number of hospitalisations in the month of October, November and December respectively.

From the information provided, $X_{1} \sim \operatorname{Poi}(2), X_{2} \sim \operatorname{Poi}(3)$ and $X_{3} \sim \operatorname{Poi}(1)$
Let the total hospitalisation over this period be denoted by $X$ where $X=X_{1}+X_{2}+X_{3}$
Since all $X_{i}$ 's are independent
$X \sim \operatorname{Poi}(2+3+1)$

$$
\text { Thus, } \begin{align*}
P[x<5] & =\sum\left(6^{x} e^{-6}\right) / x!(\text { summation over } x=0 \text { to } 4)  \tag{1}\\
& =0.0248+0.0149+0.0446+0.0892+0.1339 \\
& =0.2851 \tag{1}
\end{align*}
$$

## Solution 3:

i) a) Let $\mathrm{X}_{\mathrm{i}}$ represent each motor claim amount for $\mathrm{i}=1$ to 10

Moment generating function for exponential distribution, $M_{x}(t)=(1-t / \lambda)^{-1}$
Hence, for $Y=\sum X_{i}(i=1$ to 10$)$
$M_{Y}(t)=\left(M_{X}(t)\right)^{\wedge} 10$

$$
\begin{equation*}
=(1-t / \lambda)^{-10} \tag{0.5}
\end{equation*}
$$

which is the moment generating function of gamma distribution with $\alpha=10$ and $\lambda=1.25$
b) MGF of 2.5 Y is $\mathrm{E}\left[\mathrm{e}^{(2.5 t) \mathrm{Y}}\right]$
$=M_{Y}[2.5 t]$
$=(1-2 t)^{-10}$
$=(1-t / 0.5)^{-10}$
which is the moment generating function of gamma $(10,0.5)$
i.e. $\chi^{2}{ }_{20}$ distribution
ii) $B$
iii) From i.a. above, $Y$ ~ gamma ( $10,1.25$ )

Therefore $Y$ has mean 10/1.25 $=8$ and variance $=10 /(1.25)^{2}=6.4$
Applying central limit theorem $Y \sim N(8,6.4)$
Thus, $\mathrm{P}[\mathrm{Y}>10]=\mathrm{P}[\mathrm{Z}>(10-8) /(\mathrm{V} 6.4)=0.791]$

$$
=1-0.786
$$

$$
\begin{equation*}
=0.214 \tag{1}
\end{equation*}
$$

iv) $n$ is not large enough for the central limit theorem to be used, but the approximation is still close to the true probability

## Solution 4:

E

## Solution 5

i) $\quad f(x, y)=(1 / 27)^{*}(2 x+y)$ where $x=0,1,2$ and $y=0,1,2$

Joint probability distribution of $X, Y$ i.e $f(x, y)$ is given by the table
$f(x, y=0,0)=0$
$f(x, y=0,1)=1 / 27$
$f(x, y=0,2)=2 / 27$
$f(x, y=1,0)=2 / 27$
$f(x, y=1,1)=3 / 27$

$$
\begin{aligned}
& f(x, y=1,2)=4 / 27 \\
& f(x, y=2,0)=4 / 27 \\
& f(x, y=2,1)=5 / 27 \\
& f(x, y=2,2)=6 / 27
\end{aligned}
$$

$f_{Y}(0)=6 / 27$
$f_{Y}(1)=9 / 27$
$f_{Y}(2)=12 / 27$
ii) C

## Solution 6:

i) a) An estimator is said to be consistent when

- mean square error tends to zero
- as ' $n$ ' tends to infinity
- where ' $n$ ' is sample size
b) A good estimator is one that
- has small mean square error
- is unbiased and
- is consistent
ii) a) probability of finding 2 senior grade employees having ESOPs is 0.3637 (i.e., W3) W1 and W2 are more extreme scenarios than W3.
Hence, $p$-value of finding 2 senior grade employees is $\mathrm{P}(\mathrm{W} 1)+\mathrm{P}(\mathrm{W} 2)+\mathrm{P}(\mathrm{W} 3)$
$=0.0606+0.1212+0.3637=0.5455$
(Or p-value of W3 can be found as 1-P(W4) $=1-0.4545=0.5455$ )
b) Required probability is 3 C 3 * 8C2 / 11C5
$=1 *\left(8 * 7 / 1^{*} 2\right) /\left(\left(11^{*} 10 * 9 * 8 * 7 / 1^{*} 2^{*} 3^{*} 4^{*} 5\right)\right.$
$=4 * 7 /\left(11^{*} 7^{*} 3^{*} 2\right)$
=2/33
$=0.0606$
iii) Option A
$(12+18) /(100+3 * 80)=30 / 340=0.088$
iv) If ' $m$ ' is the mean of the lognormal distribution then by invariance property, ' m cap' $=\mathrm{e}^{\wedge(' m u ~ c a p ' ~}+1 / 2$ * ‘sigma square cap') $=e(1.25)=3.49$

If 'var' is the variance of the lognormal distribution then by invariance property, 'var cap' $=\mathrm{e}^{\wedge}\left(2^{* \prime} \mathrm{mu}\right.$ cap' + ‘sigma square cap') * ( $\left.\mathrm{e}^{\wedge(‘ s i g m a ~ s q u a r e ~ c a p ’) ~}-1\right)$ =3.49^2 *(e(0.5)-1)

$$
\begin{aligned}
& =12.1825^{*} 0.6487 \\
& =7.903
\end{aligned}
$$

v)
a) $\mathrm{CRLB}=-1 / \mathrm{E}[$ second derivative of log likelihood with respect to lambda]
$=1 / E[n / l a m b d a \wedge 2]$
$=l a m b d a^{\wedge} 2 / n$
=0.01/20
$=0.0005$
b) 'lambda cap' $\sim N(l a m b d a, ~ C R L B)$ approximately. Hence confidence interval is given by
('lambda cap' -1.96 * sqrt(CRLB), lambda cap' + 1.96 * sqrt (CRLB)
=(0.1-1.96*sqrt(0.0005), 0.1+1.96*sqrt(0.0005))
$=(0.056173,0.143827)$
c) Using, 2*lambda*n*X bar ~ Chi square distribution with 2*n degrees of freedom 40*lambda*X bar ~ chi square distribution with 40 degrees of freedom
$\mathrm{P}(24.43<40 *$ lambda*Xbar<59.34) $=0.95$
Hence $95 \%$ confidence interval for lambda is
(24.43/(40*10), 59.34/(40*10))
$=(0.061075,0.14835)$

Confidence interval using chi square result / exact result is narrower (i.e. better)compared to result in part $b$.
Result in part $b$ is impacted due to smaller sample size.
Larger sample could have resulted in better / narrower interval in part b

## Solution 7:

i) Option A
ii) Chi square with 10 df can be written as $\mathrm{Ga}(5,0.5)$ distribution.

Hence, posterior distribution would be $\mathrm{Ga}(5+\mathrm{x}, 1.5)$ using results from part 1
iii) Option B
iv) Option D
(working is not required)
Bayesian estimate under all-or-nothing loss is mode of the distribution.
Differentiating the log of the posterior distribution and equating with zero, we get,
$4 / p-14 /(1-p)=0$
Hence, 4(1-p)-14p=0
$4-18 p=0$. Hence, $p=4 / 18$
v) a) Posterior distribution of theta can be written as, Normal((A+B)/(C+D), $1 /(C+D))$

Where,
$\mathrm{A}=\left(\mathrm{n}^{*} \mathrm{x}\right.$ bar)$/ 150^{\wedge} 2$
$B=500 / 100^{\wedge} 2$
$\mathrm{C}=\mathrm{n} / 150^{\wedge} 2$
D=1/100^2
b) Mean of the posterior distribution is $(A+B) /(C+D)$

This can be written in the forms of
[(A/(x bar)) / (C+D) ]* (x bar) + ((B/500)/ (C+D)) * 500
i.e. $(C /(C+D))^{*}(x$ bar $)+(D /(C+D) * 500$
i.e., $Z^{*} \times$ bar $+(1-Z)^{*} 500$ which is a credibility estimate
where $Z=\left(n / 150^{\wedge} 2\right) /\left(\left(n / 150^{\wedge} 2\right)+\left(1 / 100^{\wedge} 2\right)\right)$
MLE of 'theta' $=x$ bar
Prior mean $=500$
c) Impact on Z

If prior variance was $150^{\wedge} 2$ (instead of $100^{\wedge} 2$ ) -

- this will lead to reduction in denominator and $Z$ will increase.
- Increase in prior variance means that prior is less reliable and hence we need to rely more on data and hence $Z$ will increase.
if likelihood variance was $100^{2}$ (instead of $150^{\wedge} 2$ ) -
- $Z$ will increase with more increase in numerator compared to denominator.
- Reduction in variance of the observed data means data is more reliable and hence more weight can be given to it and hence $Z$ increases.


## Solution 8:

i) The sums of squares as given in the question
$S_{x x}=\Sigma\left(x_{i}-\bar{x}\right)^{2}=2,800 \quad S_{x y}=\sum\left(\left(x_{i}-\bar{x}\right)\left(y_{i}-\bar{y}\right)=25,300\right.$
$\bar{x}=\frac{\sum x_{i}}{n}=35, \quad \bar{y}=\frac{\sum y_{i}}{n}=281$
$\hat{\beta}=\frac{s_{x y}}{s_{x}}=\frac{25,300}{2,800}=9.04$
$\hat{\alpha}=\bar{y}-\hat{\beta} \bar{x}=281-9.04 \times 35=-34.82$
Hence the fitted regression line of y on x is $\mathrm{y}=-34.82+9.04 \mathrm{x}$
ii) $\quad S_{y y}=\sum\left(y_{i}-\bar{y}\right)^{2}=2,70,832$
$\hat{\sigma}^{2}=\frac{1}{n-2}\left(S_{y y}-\frac{S_{x y}^{2}}{S_{x x}}\right)=\frac{1}{5}\left(2,70,832-\frac{25,300^{2}}{2,800}\right)=8,445.69$
Now $\frac{5 \hat{\sigma}^{2}}{\sigma^{2}} \sim$ chi square $x_{5}^{2}$ which gives a confidence interval for $\sigma^{2}$ of:
$\left(\frac{5 \times 8,445.69}{11.07}, \frac{5 \times 8,445.69}{1.145}\right)=(3,814.67,36,880.74)$
iii)

The proportion of the variability explained by the model is given by:

$$
\begin{equation*}
\mathrm{R}^{2}=\frac{S_{x y}^{2}}{S_{x x s_{y y}}}=\frac{25,300^{2}}{2800 \times 2,70,832}=84 \% \tag{2}
\end{equation*}
$$

$84 \%$ of the variance is explained by the model, which indicate that the fit is fairly good. It is still might be worthwhile to examine the residuals to double check that a linear model is appropriate.
iv) Testing:
$\mathrm{H}_{0}: \beta=0$ vs $\mathrm{H}_{1}: \beta>0$
Now $\frac{\widehat{\beta}-\beta}{\sqrt{\widehat{\sigma}^{2} / s_{x x}}} \sim t_{5}$
The observed value of test statistics is
$\frac{9.04-0}{\sqrt{8445.63 / 2800}}=5.20$
This exceeds the $0.5 \%$ critical value of the $t_{5}$ distribution of 4.032 . So we have sufficient evidence at the $0.5 \%$ level to reject $\mathrm{H}_{0}$ and the conclusion is that $\beta>0$ hence the data are positively correlated
v) The variance of the distribution of the mean number of COVID claims corresponding to an entry age of 60 is:

$$
\begin{equation*}
\left[\frac{1}{n}+\frac{\left(x_{0}-\bar{x}\right)^{2}}{s_{X X}}\right] \hat{\sigma}^{2}=\left[\frac{1}{7}+\frac{(60-35)^{2}}{2800}\right] \times 8445.69=3,091.72 \tag{1}
\end{equation*}
$$

The predicted value of number of COVID claims corresponding to age 60 is
$-34.82+9.04 \times 60=507.32$

We have $t_{5}$ distribution. Hence the $95 \%$ confidence interval is
$507.32 \pm 2.571 \times \sqrt{3091.72}=(364.37,650.28)$
vi) a) The completed table of residuals are as follows

| Age | 5 | 15 | 25 | 35 | 45 | 55 | 65 |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| Residual | 91 | 19 | -56 | -95 | -104 | 78 | 67 |

b) Clearly the trend of residual with progression of age is not pattern less. The residuals are not independent of the age. This means that the linear model is missing something and is not appropriate to these data
[21 Marks]

## Solution 9:

i) a) $A$ distribution of the response variable $Y$
b) A "linear predictor" $\eta$
c) A "link function" g
ii) The PDF of exponential distribution can be written as
$\mathrm{f}(\mathrm{y})=\frac{1}{\mu} e^{\frac{-y}{\mu}}=\exp \left\{-\frac{y}{\mu}-\log \mu\right\}$
Comparing the above with the standard PDF of exponential family of distribution $\theta=-1 / \mu, \mathrm{b}(\theta)=\log \mu=-\log (-\theta), \varnothing=1, \mathrm{a}(\varnothing)=\varnothing$ and $\mathrm{c}(\mathrm{y}, \varnothing)=0$
iii)
a) The canonical link function from part (ii) that $\theta=-\frac{1}{\mu}$
b) The variance function is $b^{\prime \prime}(\theta)$. Differentiating $\mathbf{b}(\theta)$ twice , $b^{\prime \prime}(\theta)=1 / \theta^{2}=\mu^{2}$

So the variance function is $\mu^{2}$
c) The dispersion parameter or scale parameter is $\emptyset=1$
iv) The log of the likelihood function is
$\log \mathrm{L}\left(\mu_{i}\right)=-\sum \frac{y_{i}}{\mu_{i}}-\sum \log \mu_{i}$
The canonical link function for the exponential distribution is $\mathrm{g}\left(\mu_{i}\right)=1 / \mu_{i}$.
The canonical link function connects the mean response to the linear predictor, $\mathrm{g}\left(\mu_{i}\right)=$ $\eta_{i}$
Hence we have
$\frac{1}{\mu_{i}}=\alpha+\beta \mathrm{x}_{\mathrm{i}}$
The log likelihood function in terms of $\alpha$ and $\beta$ :
$\log \mathrm{L}(\alpha, \beta)=\sum y_{i}\left(\alpha+\beta x_{i}\right)+\sum \log \left(\alpha+\beta \mathrm{x}_{\mathrm{i}}\right)$
Differentiating the above equation with respect to $\alpha$ and $\beta$ :
$\frac{\partial}{\partial \alpha} \log \mathrm{L}(\alpha, \beta)=-\sum y_{i}+\sum \frac{1}{\alpha+\beta x_{i}}$
$\frac{\partial}{\partial \beta} \log \mathrm{L}(\alpha, \beta)=-\sum x_{i} y_{i}+\sum \frac{x_{i}}{\alpha+\beta x_{i}}$
The equations satisfied by the MLEs of $\alpha$ and $\beta$ are
$-\sum y_{i}+\sum \frac{1}{\hat{\alpha}+\widehat{\beta} x_{i}}=0$
$-\sum x_{i} y_{i}+\sum \frac{x_{i}}{\hat{\alpha}+\widehat{\beta} x_{i}}=0$

Substituting in the given data values gives the following equations
$\frac{1}{\hat{\alpha}+30 \widehat{\beta}}+\frac{1}{\hat{\alpha}+35 \widehat{\beta}}+\frac{1}{\hat{\alpha}+40 \hat{\beta}}+\frac{1}{\hat{\alpha}+45 \widehat{\beta}}+\frac{1}{\hat{\alpha}+50 \widehat{\beta}}-890=0$
$\frac{30}{\hat{\alpha}+30 \widehat{\beta}}+\frac{35}{\hat{\alpha}+35 \widehat{\beta}}+\frac{40}{\hat{\alpha}+40 \hat{\beta}}+\frac{45}{\hat{\alpha}+45 \widehat{\beta}}+\frac{50}{\hat{\alpha}+50 \tilde{\beta}}-39550=0$
(The above is for information purpose only. Students are not expected to provide the above derivation in the answer script)
Correct answer is Option C

