# THE ROYAL STATISTICAL SOCIETY

# **2004 EXAMINATIONS – SOLUTIONS**

# HIGHER CERTIFICATE

# **PAPER I – STATISTICAL THEORY**

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(i) (a) 
$$P(\text{all four favour the complex}) = (0.6)^4$$
.  
 $P(\text{all four oppose the complex}) = (0.3)^4$ .  
 $P(\text{all four are indifferent}) = (0.1)^4$ .

So *P*(all four think alike) =  $(0.6)^4 + (0.3)^4 + (0.1)^4 = 0.1378$ .

(b) 
$$P(\text{an individual is not opposed}) = 0.6 + 0.1 = 0.7.$$

So  $P(\text{none of the four is opposed}) = (0.7)^4 = 0.2401$ .

(c) Possible favourable results are *FFOI*, *FOOI*, *FOII*, in any order.

 $P(FFOI) = (0.6)^{2}(0.3)(0.1) = 0.0108$   $P(FOOI) = (0.6)(0.3)^{2}(0.1) = 0.0054$  $P(FOII) = (0.6)(0.3)(0.1)^{2} = 0.0018$ 

Each result can be arranged in  $\frac{4!}{2!1!1!} = 12$  ways.

So overall probability is 12(0.0108 + 0.0054 + 0.0018) = 0.216.

(d) From (a),  $P(\text{all four in favour}) = (0.6)^4 = 0.1296$ . From (b), P(none opposed) = 0.2401. So the required conditional probability is

0.1296/0.2401 = 0.5398.

(ii) The number in favour is binomially distributed with n = 4 and p = 0.6. So the expectation (mean) is  $4 \times 0.6 = 2.4$  and the variance is  $4 \times 0.6 \times 0.4 = 0.96$ .

(iii) P(opposed) = P(opposed | young)P(young) + P(opposed | older)P(older)=  $(0.12 \times 0.25) + (p \times 0.75)$ 

where p = P(opposed | older). But we are given that P(opposed) = 0.3. Hence p = 0.36.

(iv) In samples of one "young" and three "olders",  

$$P(\text{exactly one opposes}) = P("young" opposes, "olders" do not)$$
  
 $+ P("young" does not oppose, one "older" opposes)$   
 $= \{(0.12)(0.64)^3\} + \{3(0.88)(0.36)(0.64)^2\} = 0.03146 + 0.38928 = 0.4207.$ 

(i) (a) The moment generating function is

$$M_{X}(t) = E\left(e^{tX}\right) = \sum_{x=0}^{\infty} e^{tx} \frac{e^{-\lambda} \lambda^{x}}{x!} = \sum_{x=0}^{\infty} e^{-\lambda} \frac{\left(\lambda e^{t}\right)^{x}}{x!} = \exp\left(\lambda e^{t} - \lambda\right) = \exp\left(\lambda\left(e^{t} - 1\right)\right)$$

(b) 
$$E(X) = M_X'(0) = \left[\lambda e^t e^{\lambda (e^t - 1)}\right]_{t=0} = \lambda$$

$$E(X^{2}) = M_{X}"(0) = \left[\frac{d}{dt}\left(\lambda e^{t}e^{\lambda(e^{t}-1)}\right)\right]_{t=0}$$
$$= \left[\lambda e^{t} \cdot e^{\lambda(e^{t}-1)}\lambda e^{t} + e^{\lambda(e^{t}-1)} \cdot \lambda e^{t}\right]_{t=0} = \lambda^{2} + \lambda$$

Hence  $\operatorname{Var}(X) = E(X^2) - [E(X)]^2 = \lambda$ .

(Alternatively, the moments could be obtained from the power series expansion of  $M_X(t)$ .)

(Alternatively, though with comparatively lengthy algebra, the moments could be obtained directly by  $E(X) = \Sigma x P(X = x)$  and  $E(X^2) = \Sigma x^2 P(X = x)$ ; or, somewhat easier, use  $E[X(X - 1)] = \Sigma x(x - 1)P(X = x)$  (this is  $\lambda^2$ ) and then  $Var(X) = E[X(X - 1)] + E(X) - \{E(X)\}^2$ .)

- (c) The binomial distribution with parameters n and p may be approximated by the Poisson distribution with parameter np if n is large and p is small. As a "rule of thumb",  $\frac{1}{2} \le np \le 10$  gives an indication of how large n should be and how small p should be. (If np > 10, a Normal approximation to the binomial may be better.)
- (ii) Let X = number of wrong calculations. We have  $X \sim B(200, 0.0075)$ .

$$P(X=1) = {\binom{200}{1}} (0.0075) (0.9925)^{199} = 200 \times 0.0075 \times 0.9925^{199} = 0.3353(2) .$$
$$P(X=4) = {\binom{200}{4}} (0.0075)^4 (0.9925)^{196}$$
$$= \frac{200 \times 199 \times 198 \times 197}{4 \times 3 \times 2 \times 1} \times 0.0075^4 \times 0.9925^{196} = 0.0468(0) .$$

(iii) We approximate using  $X \sim \text{Poisson}(200 \times 0.0075 = 1.5)$ . With this,

$$P(X=1) = 1.5e^{-1.5} = 0.3347(0)$$

giving a percentage error of  $\frac{100(0.33532 - 0.33470)}{0.33532} = 0.18\%$ , and

$$P(X=4) = \frac{e^{-1.5} (1.5)^4}{4!} = 0.0470(7)$$

giving a percentage error of 
$$\frac{100(0.04707 - 0.04680)}{0.04680} = 0.58\%$$
.

[Note. These percentage errors might come out *slightly* differently if more accuracy is kept in the binomial and Poisson probabilities.]

Both approximations are remarkably accurate, with percentage errors well below 1%. The approximation for X = 1 (one wrong calculation) is the more accurate of the two. That approximation is an underestimate; the other is an overestimate.

Actual volume  $X \sim N(1010, 8^2)$ . Let  $Z \sim N(0,1)$ .

(i) 
$$P(X < 1000) = P(Z < \frac{1000 - 1010}{8}) = P(Z < -1.25) = 0.1056.$$

(ii) Let *Y* be the total volume in a 6-pack.

We have  $Y \sim N(6 \times 1010, 64 + 64 + 64 + 64 + 64 + 64)$ , i.e.  $Y \sim N(6060, 384)$ .

$$P(Y < 6000) = P\left(Z < \frac{6000 - 6060}{\sqrt{384}}\right) = P(Z < -3.06) = 0.0011.$$

(Alternatively, could use  $\overline{X} \sim N(1010, 64/6)$  and calculate  $P(\overline{X} < 1000)$ .)

This probability is considerably smaller than that in part (i). In practical terms, this is because there will be a tendency for heavier and lighter cartons in a 6-pack to balance each other out. Alternatively, in terms of probability distributions, consider X and  $\overline{X}$ :  $\overline{X}$  has the same mean as X but only one-sixth of the variance, so less of the lower tail of the distribution of  $\overline{X}$  is below the nominal volume of 1000.

(iii) The new volume  $W \sim N(\mu, 4^2)$ , where  $\mu$  is the new mean. So we have that  $P(W < 1000) = P\left(Z < \frac{1000 - \mu}{4}\right)$ . We require that this probability must be no greater than 0.1056. Thus the cut-off point for Z is to be z = -1.25 (as before). Hence  $\frac{1000 - \mu}{4} = -1.25$ , giving  $\mu = 1005$ .

This means that 5 ml per carton could be saved, i.e. a cost saving per carton of  $\frac{5}{1000} \times \pounds 1$ . To recover the £200, the number of cartons required is  $\frac{200}{5/1000} = 40000$ .

(i) The binomial distribution with parameters n, p can be approximated by N(np, np(1-p)) when n is large and p is not too near to 0 or 1. As a "rule of thumb", the approximation is likely to be good if both np and np(1-p) are > 10.

Let  $X \sim B(n, p)$  and let  $\Phi$  denote the c.d.f. of N(0, 1). Using a continuity correction,  $P(X \le x) \approx \Phi\left(\frac{x + \frac{1}{2} - np}{\sqrt{np(1-p)}}\right)$  and  $P(X < x) \approx \Phi\left(\frac{x - \frac{1}{2} - np}{\sqrt{np(1-p)}}\right)$ .

The 95% confidence interval for *p* uses the estimated variance  $\hat{p}(1-\hat{p})/n$ , giving the approximate interval

$$\hat{p} - 1.96\sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$$

The estimate of *p* is  $\hat{p} = \frac{30}{50} = 0.6$ , so  $\sqrt{\frac{\hat{p}(1-\hat{p})}{n}} = \sqrt{\frac{(0.6)(0.4)}{50}} = 0.0693$ . Thus the approximate interval is

$$0.6 - (1.96 \times 0.0693)$$
,  $0.6 + (1.96 \times 0.0693)$ 

i.e. (0.464, 0.736).

(ii)  $P(\text{neither hits}) = (1-p)^2$ . Therefore  $P(\text{at least 1 hit}) = 1 - (1-p)^2 = p(2-p)$ . We estimate this by (0.6)(2-0.6) = 0.84.

(iii) When p = 0.464 (lower limit of interval in part (i)), we have p(2 - p) = 0.713. Similarly, when p = 0.736, we have p(2 - p) = 0.930. Thus (0.713, 0.930) is the required interval.

(iv) When *n* pairs are fired,  $P(\text{all miss}) = [(1-p)^2]^n$ , estimated by  $(0.16)^n$ . Hence  $(0.16)^n < 0.0005$  is required. Solving this by taking logarithms to base 10, we have  $n \log_{10}(0.16) < \log_{10}(0.0005)$ , i.e. -0.79588n < -3.30103 which gives n > 4.148. so *n* must be at least 5.

(i) 
$$f(t) = \lambda e^{-\lambda t}, \quad t > 0; \quad \lambda > 0$$

(a) Sketch of f(t).

[*NOTE.* The curve should of course appear as a smooth decaying exponential; it might not do so, due to the limits of electronic reproduction.]



(c) 
$$P(a < T \le b) = F(b) - F(a) = e^{-\lambda a} - e^{\lambda b}$$
.

### (ii) Assume all settlements of invoices are independent.

 $P(50 \text{ in first week}) = \left\{ F(1) \right\}^{50} = \left(1 - e^{-\lambda}\right)^{50}, \text{ because } T \le 1 \text{ for all these 50.}$ 

Likewise,  $1 < T \le 2$  for the 35 in the second week, so we have  $P(35 \text{ in second week}) = \{F(2) - F(1)\}^{35} = (e^{-\lambda} - e^{-2\lambda})^{35}$ .

The remaining 15 have T > 2, which has probability  $1 - P(T \le 2) = e^{-2\lambda}$ , and thus  $P(15 \text{ after week } 2) = (e^{-2\lambda})^{15}$ .

The likelihood is therefore the product

$$L(\lambda) = k (1 - e^{-\lambda})^{50} (e^{-\lambda} - e^{-2\lambda})^{35} (e^{-2\lambda})^{15}$$

where *k* is a constant of proportionality.

Taking logarithms (base e),

$$\log L(\lambda) = \log k + 50 \log(1 - e^{-\lambda}) + 35 \log \left\{ e^{-\lambda} (1 - e^{-\lambda}) \right\} + 15 \log(e^{-2\lambda})$$
$$= \log k + 85 \log(1 - e^{-\lambda}) - (35 + 30)\lambda = \log k + 85 \log(1 - e^{-\lambda}) - 65\lambda.$$

$$\therefore \frac{d}{d\lambda} \log L = \frac{85e^{-\lambda}}{1 - e^{-\lambda}} - 65 = \frac{85}{e^{\lambda} - 1} - 65$$

Equating to zero,  $85 = 65(e^{\lambda} - 1)$  or  $e^{\lambda} = 150/65$ , so that  $\hat{\lambda} = \log(150/65) = 0.836$ .

[It is easy to check that this is indeed a maximum; e.g.  $\frac{d^2}{d\lambda^2}\log L = -\frac{85}{(e^{\lambda}-1)^2} < 0$ .]

(iii)  $1-e^{-0.836} = 0.5666$ ;  $e^{-0.836} - e^{-1.672} = 0.43344 - 0.18787 = 0.2456$ . Hence, out of 100 invoices, 56.66, 24.56 and 18.78 would be expected to be paid, on this model, in weeks 1, 2 and later. The actual numbers were 50, 35 and 15. The prediction for the second week is a long way from what happened, balanced by smaller discrepancies in the other two periods. This does not seem very satisfactory.

$$f(x) = \frac{k}{x^{k+1}}, \qquad x \ge 1; \quad k > 0$$

(i) Sketch of f(x).

[**NOTE.** The curve should of course appear as a smooth curve; it might not do so, due to the limits of electronic reproduction.]



C.d.f. is 
$$F(x) = P(X \le x) = \int_{1}^{x} \frac{k}{u^{k+1}} du = \left[-\frac{1}{u^{k}}\right]_{1}^{x} = 1 - \frac{1}{x^{k}} \quad (\text{for } x \ge 1).$$

(ii) Median *M* has  $\frac{1}{2} = F(m) = 1 - M^{-k}$ , so  $\frac{1}{2} = M^{-k}$  and hence  $M = 2^{1/k}$ . Lower quartile  $Q_1$  has  $\frac{1}{4} = F(Q_1) = 1 - Q_1^{-k}$ , so  $\frac{3}{4} = Q_1^{-k}$ , i.e.  $Q_1 = (\frac{4}{3})^{1/k}$ . Upper quartile  $Q_3$  has  $\frac{3}{4} = F(Q_3) = 1 - Q_3^{-k}$ , so  $Q_3 = (4)^{1/k}$ .

Hence the semi-interquartile range is  $\frac{1}{2} \left\{ 4^{1/k} - \left(\frac{4}{3}\right)^{1/k} \right\}$ .

(iii) 
$$E(X) = \int_{1}^{\infty} xf(x) dx = \int_{1}^{\infty} \frac{k}{x^{k}} dx = \left[\frac{-k}{(k-1)x^{k-1}}\right]_{1}^{\infty} = \frac{k}{k-1}.$$

$$E(X^{2}) = \int_{1}^{\infty} x^{2} f(x) dx = \int_{1}^{\infty} \frac{k}{x^{k-1}} dx = \left[\frac{-k}{(k-2)x^{k-2}}\right]_{1}^{\infty} = \frac{k}{k-2}.$$

:. Var 
$$(X) = E(X^2) - \{E(X)\}^2 = \frac{k}{k-2} - \frac{k^2}{(k-1)^2}$$

$$=\frac{k}{(k-2)(k-1)^{2}}\left\{(k-1)^{2}-k(k-2)\right\}=\frac{k}{(k-1)^{2}(k-2)}.$$

 $P(X > E(X)) = \int_{k/(k-1)}^{\infty} \frac{k}{x^{k+1}} dx = \left[-\frac{1}{x^k}\right]_{k/(k-1)}^{\infty} = \left(\frac{k-1}{k}\right)^k$ , or this can be written down directly from the c.d.f. found in part (i).

(iv) For the case k = 3,

- (a)  $M = 2^{1/3}$  in the units given, or £12599,
- (b) mean = 3/2 in the units given, or £15000,
- (c) inserting X = 10,  $P(X \le 10) = 1 \frac{1}{10^3}$ , so  $P(X > 10) = \frac{1}{10^3}$ , i.e. 0.1%.

(i) 
$$E(X) = \int_0^\theta \frac{x}{\theta} dx = \frac{1}{\theta} \left[ \frac{1}{2} x^2 \right]_0^\theta = \frac{1}{2} \theta.$$

$$E(X^{2}) = \int_{0}^{\theta} \frac{x^{2}}{\theta} dx = \frac{1}{\theta} \left[ \frac{1}{3} x^{3} \right]_{0}^{\theta} = \frac{1}{3} \theta^{2} .$$
  
$$\therefore \operatorname{Var}(X) = E(X^{2}) - \left\{ E(X) \right\}^{2} = \frac{1}{3} \theta^{2} - \left( \frac{1}{2} \theta \right)^{2} = \frac{1}{12} \theta^{2} .$$

(ii)  $P(\text{longest offcut is } \le x) = P(\text{all } n \text{ offcuts are } \le x).$ 

The c.d.f. for each  $X_i$  is  $F(x) = P(X \le x) = \int_0^x \frac{du}{\theta} = \left[\frac{u}{\theta}\right]_0^x = \frac{x}{\theta}$ , and the  $X_i$  are all independent. Therefore  $P(\text{all } n \text{ offcuts are } \le x) = \left\{F(x)\right\}^n = \left(\frac{x}{\theta}\right)^n$ , and this is also  $P(\text{longest offcut is } \le x)$ , i.e. the c.d.f. of the sample maximum  $X_{(n)}$ . Thus the p.d.f. of  $X_{(n)}$  is the derivative of this, i.e.  $nx^{n-1}/\theta^n$ . This is for the interval  $(0, \theta)$ .

$$\therefore E(X_{(n)}) = \int_{0}^{\theta} \frac{nx^{n}}{\theta^{n}} dx = \frac{n}{\theta^{n}} \left[ \frac{x^{n+1}}{n+1} \right]_{0}^{\theta} = \frac{n\theta}{n+1}.$$

$$E(X_{(n)}^{2}) = \int_{0}^{\theta} \frac{nx^{n+1}}{\theta^{n}} dx = \frac{n}{\theta^{n}} \left[ \frac{x^{n+2}}{n+2} \right]_{0}^{\theta} = \frac{n\theta^{2}}{n+2}.$$

$$\therefore \operatorname{Var}(X_{(n)}) = E(X_{(n)}^{2}) - \left\{ E(X_{(n)}) \right\}^{2} = \frac{n\theta^{2}}{n+2} - \frac{n^{2}\theta^{2}}{(n+1)^{2}}$$

$$= n\theta^{2} \left( \frac{(n+1)^{2} - n(n+2)}{(n+2)(n+1)^{2}} \right) = \frac{n\theta^{2}}{(n+1)^{2}(n+2)}.$$

Immediately we have  $E\left(\frac{n+1}{n}X_{(n)}\right) = \theta$ , so  $\frac{n+1}{n}X_{(n)}$  is an unbiased estimator of  $\theta$ .

$$\operatorname{Var}\left(\frac{n+1}{n}X(n)\right) = \frac{(n+1)^{2}}{n^{2}}\operatorname{Var}\left(X_{(n)}\right) = \frac{(n+1)^{2}}{n^{2}}\frac{n\theta^{2}}{(n+1)^{2}(n+2)} = \frac{\theta^{2}}{n(n+2)}$$

(iii) We have (see part (i)) that  $E(X) = \theta/2$ . Thus the method of moments estimator of  $\theta/2$  is  $\overline{X}$ , and so the method of moments estimator of  $\theta$  is  $2\overline{X}$  or  $\frac{2}{n}\sum X_i$  as required.

$$\operatorname{Var}\left(\frac{2}{n}\sum X_{i}\right) = \operatorname{Var}\left(2\overline{X}\right) = 4\operatorname{Var}\left(\overline{X}\right) = \frac{4}{n}\operatorname{Var}\left(X\right) = \frac{4}{n}\cdot\frac{\theta^{2}}{12} = \frac{\theta^{2}}{3n}.$$



Simple linear regression analysis seems quite suitable.

(ii) The model is  $y_i = \alpha + \beta x_i + e_i$ , where  $\{e_i\}$  are uncorrelated with zero mean and (constant) variance  $\sigma^2$  (independent identically distributed N(0,  $\sigma^2$ ) for the purpose of undertaking statistical tests, as in part (iii)). Estimating by the method of least squares gives

$$\hat{\beta} = \frac{S_{xy}}{S_{xx}}, \qquad \hat{\alpha} = \overline{y} - \hat{\beta}\overline{x},$$

where (standard notation)

$$S_{xy} = \Sigma (x_i - \overline{x}) (y_i - \overline{y}) = \sum x_i y_i - \frac{\sum x_i \sum y_i}{n} ,$$
$$S_{xx} = \sum (x_i - \overline{x})^2 = \sum x_i^2 - \frac{(\sum x_i)^2}{n} .$$

We have

$$\hat{\beta} = \frac{S_{xy}}{S_{xx}} = \frac{4440 - (150 \times 220/10)}{3200 - (150^2/10)} = \frac{1140}{950} = 1.20 \text{ and } \hat{\alpha} = 22 - (1.20 \times 15) = 4,$$

so the line is

$$y = 4 + 1.2x$$
.

## **Continued on next page**

(i)

The total sum of squares is  $S_{yy} = \sum (y_i - \overline{y})^2 = \sum y_i^2 - \frac{(\sum y_i)^2}{10} = 1440.$ 

The sum of squares for regression is  $\hat{\beta}S_{xy}$  (or  $S_{xy}^2/S_{xx}$ ) = 1368.

Therefore the residual sum of squares is 1440 - 1368 = 72.

This has 8 degrees of freedom, so the residual mean square ( $\hat{\sigma}^2$ ) is 72/8 = 9.

The coefficient of determination  $R^2 = 1368/1440 = 0.95$  (usually given as 95%).

(iii) The estimated variance of  $\hat{\beta}$  is 9/950 = 0.009474. So the test statistic for testing the null hypothesis  $\beta = 1$  is  $\frac{1.2 - 1}{\sqrt{0.009474}} = 2.05$ , which we refer to  $t_8$ .

This is not significant at the 5% level, so the null hypothesis  $\beta = 1$  cannot be rejected.

(iv) The model here is  $y_i = bx_i + e_i$ . Estimating *b* by least squares, we minimise  $\Omega = \sum_{i=1}^{n} (y_i - bx_i)^2$ . Differentiating with respect to *b*, we have  $\frac{d\Omega}{db} = -2\sum (y_i - bx_i)x_i$ . Setting this equal to zero gives  $\Sigma x_i y_i = \hat{b}\Sigma x_i^2$ , i.e.  $\hat{b} = \Sigma x_i y_i / \Sigma x_i^2$ . (Note that  $\frac{d^2\Omega}{db^2} = 2\sum x_i^2 > 0$ , so this is a minimum.)

Thus we have  $\hat{b} = 4440/3200 = 1.3875$ .