THE ROYAL STATISTICAL SOCIETY

2005 EXAMINATIONS – SOLUTIONS

HIGHER CERTIFICATE

PAPER I – STATISTICAL THEORY

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(i)
$$\binom{49}{6} = \frac{49!}{6!43!} = \frac{49.48.47.46.45.44}{6.5.4.3.2.1} = 13983816.$$

(ii) The exact distribution of X is binomial with n = 10,000,000 and p = 1/13983816.

Its Poisson approximation has parameter (mean) $\lambda = np = 0.7151124$.

(a)
$$P(X=0) = e^{-\lambda} = e^{-0.7151124} = 0.48914$$
.

(b)
$$P(X=1) = \lambda e^{-\lambda} = 0.34979$$
.

(iii)
$$\binom{31}{6} = \frac{31!}{6!25!} = \frac{31.30.29.28.27.26}{6.5.4.3.2.1} = 736281.$$

Hence (using also the result of part (i)) we have

$$P(\text{winning set contains no number} > 31) = \frac{736281}{13983816} = 0.05265.$$

(iv) Let U be the number out of the 3,000,000 players choosing from (1, 2, ..., 31) who match all 6 winning numbers, and let V be the number out of the 7,000,000 players choosing from all the numbers (1, 2, ..., 49) who match all 6 winning numbers.

Then Y = U + V, where (given the winning numbers) U and V are independent.

(a) If all six winning numbers are in the list (1, 2, ..., 31), then U is binomial with n = 3,000,000 and p = 1/736281, which is approximated by Poisson(*np*) i.e. Poisson (4.07453).

Similarly, V is binomial with n = 7,000,000 and p = 1/13983816, which is approximated by Poisson(0.50058).

Thus, using the result that $Poisson(\lambda_1) + Poisson(\lambda_2) = Poisson(\lambda_1 + \lambda_2)$ [where the two Poisson distributions are independent], we have that the approximate distribution of Y is Poisson(4.57511).

(b) In this case, U must be zero so we simply have Y = V, i.e. Poisson(0.50058) approximately.

Cycle ~ N(27, 6.25) Bus ~ N(13, 20) Walk₁ ~ N(7, 4) Walk₂ ~ N(5, 1) Car ~ N(23, 36) The sum of independent N(μ_i , σ_i^2) distributions is N($\Sigma\mu_i$, $\Sigma\sigma_i^2$).

(i) The distribution of total journey time by bus is N(7+13+5, 4+20+1), i.e. N(25,25).

(ii) Cycle:
$$P(N(27, 6.25) < 30) = \Phi\left(\frac{30-27}{\sqrt{6.25}}\right) = \Phi(1.2) = 0.8849$$
.
Bus: $P(N(25, 25) < 30) = \Phi\left(\frac{30-25}{\sqrt{25}}\right) = \Phi(1) = 0.8413$.
Car: $P(N(23, 36) < 30) = \Phi\left(\frac{30-23}{\sqrt{36}}\right) = \Phi(1.1667) = 0.8783$.

Cycling is best, with a probability of 0.8849.

(iii) Cycle:
$$P(N(27, 6.25) > 35) = 1 - \Phi\left(\frac{35 - 27}{\sqrt{6.25}}\right) = 1 - \Phi(3.2) = 0.0007$$

Bus: $P(N(25, 25) > 35) = 1 - \Phi\left(\frac{35 - 25}{\sqrt{25}}\right) = 1 - \Phi(2) = 0.0228$.
Car: $P(N(23, 36) > 35) = 1 - \Phi\left(\frac{35 - 23}{\sqrt{36}}\right) = 1 - \Phi(2) = 0.0228$.

Again cycling is best, with a probability of 0.0007.

(iv) P(cycle) = 0.3 P(bus) = 0.3 P(car) = 0.4

 $P(\text{cycle}|<30) = \frac{P(<30|\text{cycle})P(\text{cycle})}{P(<30)}$, and similarly for the other modes of travel.

$$P(<30) = P(<30|\text{cycle})P(\text{cycle}) + P(<30|\text{bus})P(\text{bus}) + P(<30|\text{car})P(\text{car})$$
$$= (0.8849 \times 0.3) + (0.8413 \times 0.3) + (0.8783 \times 0.4)$$
$$= 0.26547 + 0.25239 + 0.35132 = 0.86918.$$

Hence

P(cycle|<30) = 0.26547 / 0.86918 = 0.3054P(bus|<30) = 0.25239 / 0.86918 = 0.2904P(car|<30) = 0.35132 / 0.86918 = 0.4042.

(i)
$$P(T > t) = P(\text{no failure in time } t) = P(X = 0) = e^{-\lambda t}$$
.

 $\therefore 1 - F(t) = e^{-\lambda t} \text{ and so the pdf of } T \text{ is } \frac{d}{dt} F(t) = -\frac{d}{dt} (e^{-\lambda t}) = \lambda e^{-\lambda t}.$

(ii)
$$P(\text{all } n \text{ systems still functioning at time } t) = \prod_{i=1}^{n} P(T_i > t) = \prod_{i=1}^{n} e^{-\lambda t} = e^{-n\lambda t}$$

: the pdf of T_{\min} is $-\frac{d}{dt}(e^{-n\lambda t}) = n\lambda e^{-n\lambda t}$ (for t > 0).

Thus T_{\min} is exponential with parameter $n\lambda$.

(iii)
$$P(\text{all } n \text{ systems have failed by time } t) = \prod_{i=1}^{n} P(T_i \le t) = \prod_{i=1}^{n} (1 - e^{-\lambda t})$$

$$=(1-e^{-\lambda t})^n$$
, and this is $P(T_{\max}) \le t$.

: the pdf of
$$T_{\max}$$
 is $\frac{d}{dt} \left\{ \left(1 - e^{-\lambda t} \right)^n \right\} = n\lambda e^{-\lambda t} \left(1 - e^{-\lambda t} \right)^{n-1}$ (for $t > 0$).

(Note that this is not exponential.)

We now have n = 10 and $\lambda = 0.002$. We require t_1 and t_2 such that $1 - e^{-n\lambda t_1} = 0.05$ and $(1 - e^{-\lambda t_2})^n = 0.95$.

: $1 - e^{-0.02t_1} = 0.05$, giving $0.02t_1 = -\log(0.95) = 0.051293$ so that $t_1 = 2.565$.

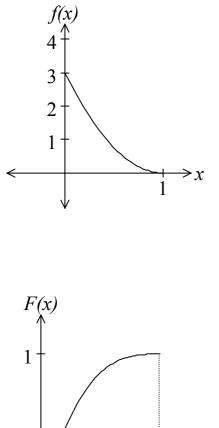
Also, $1 - e^{-0.002t_2} = (0.95)^{1/10} = 0.994884$, giving $0.002t_2 = -\log(0.005116) = -5.2754$ so that $t_2 = 2638$.

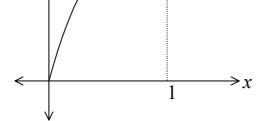
$$f(x) = \alpha (1-x)^{\alpha-1}, \qquad 0 < x < 1, \qquad \alpha > 0.$$

(i)
$$F(x) = \int_0^x \alpha (1-u)^{\alpha-1} du = \left[-(1-u)^{\alpha} \right]_0^x = 1 - (1-x)^{\alpha}$$
 (for $0 < x < 1$ and $\alpha > 0$).

The median *m* is given by $F(m) = \frac{1}{2}$, so we have $1 - (1 - m)^{\alpha} = \frac{1}{2}$ or $(1 - m)^{\alpha} = \frac{1}{2}$, so that $1 - m = 2^{-1/\alpha}$, i.e. $m = 1 - 2^{-1/\alpha}$.

When $\alpha = 3$, $f(x) = 3(1-x)^2$ and $F(x) = 1-(1-x)^3$ (in [0, 1]).





The solution to part (ii) is on the next page

(ii)
$$L = \prod_{i=1}^{n} \left[\alpha (1-x_i)^{\alpha-1} \right] = \alpha^n \prod_{i=1}^{n} (1-x_i)^{\alpha-1}$$

Hence $\log L = n \log \alpha + (\alpha - 1) \sum_{i=1}^{n} \log (1 - x_i).$

 $\therefore \frac{d \log L}{d\alpha} = \frac{n}{\alpha} + \sum_{i=1}^{n} \log(1 - x_i) \text{ which on setting equal to zero gives that the maximum likelihood estimate is <math>\hat{\alpha} = \frac{-n}{\sum_{i=1}^{n} \log(1 - x_i)}$, as required. [Consideration of $\frac{d^2 \log L}{d\alpha^2}$

(see below) confirms that this is a maximum.]

 $\frac{d^2 \log L}{d\alpha^2} = -\frac{n}{\alpha^2}$. Hence, using the result quoted in the question, $\hat{\alpha}$ is approximately Normally distributed with mean α and variance $\frac{\alpha^2}{n}$. We estimate the variance by $\frac{\hat{\alpha}^2}{n}$, so that we have $\hat{\alpha} \sim N\left(\alpha, \frac{\hat{\alpha}^2}{n}\right)$, approximately.

Hence an approximate 90% confidence interval is given by

$$0.90 \approx P\left(-1.645 < \frac{\hat{\alpha} - \alpha}{\hat{\alpha}/\sqrt{n}} < 1.645\right),$$

leading to the interval $\left(\hat{\alpha} - \frac{1.645\hat{\alpha}}{\sqrt{n}}, \hat{\alpha} + \frac{1.645\hat{\alpha}}{\sqrt{n}}\right)$.

For the given sample, we have n = 5 and the values of $1 - x_i$ are 0.88, 0.57, 0.93, 0.13 and 0.71. Therefore

$$\Sigma \log(1 - x_i) = -0.1278 - 0.5621 - 0.0726 - 2.0402 - 0.3425 = -3.1452$$

giving $\hat{\alpha} = \frac{5}{3.1452} = 1.5897$.

Also, $\frac{1.645\hat{\alpha}}{\sqrt{n}} = \frac{1.645 \times 1.5897}{\sqrt{5}} = 1.1695$, so the confidence interval is (0.420, 2.759).

(i) We have $Y \sim B(n, p)$, so $P(Y = y) = \frac{n!}{y!(n-y)!} p^y (1-p)^{n-y}$ (for y = 0, 1, 2, ..., nand 0). The likelihood*L*is simply <math>P(Y = y).

Hence $\log L = \text{constant} + y \log p + (n - y) \log(1 - p)$.

 $\therefore \frac{d \log L}{dp} = \frac{y}{p} - \frac{n-y}{1-p}$ which on setting equal to zero gives that the maximum likelihood estimate is $\hat{p} = \frac{y}{n}$. [Consideration of $\frac{d^2 \log L}{dp^2}$ confirms that this is a maximum.]

$$\operatorname{Var}(\hat{p}) = \frac{1}{n^2} \operatorname{Var}(Y) = \frac{1}{n^2} np(1-p) = \frac{p(1-p)}{n}.$$
 We may estimate p by \hat{p} in this and

thus obtain an estimate of the standard error of \hat{p} as $SE(\hat{p}) = \sqrt{\frac{p(1-p)}{n}}$.

For n = 100 and y = 20, we have $\hat{p} = 0.2$ and $SE(\hat{p}) = \sqrt{\frac{0.2 \times 0.8}{100}} = 0.04$.

(ii) P("yes") is given by P(takes drugs and coin shows "takes drugs") + <math>P(does not take drugs and coin shows "does not take drugs").

Hence $\theta = P("yes") = 0.75p + 0.25(1-p) = 0.25 + 0.5p$.

 $Z \sim B(n, \theta)$, so from part (i) we have $\hat{\theta} = z/n$ and $SE(\hat{\theta}) = \sqrt{\hat{\theta}(1-\hat{\theta})/n}$.

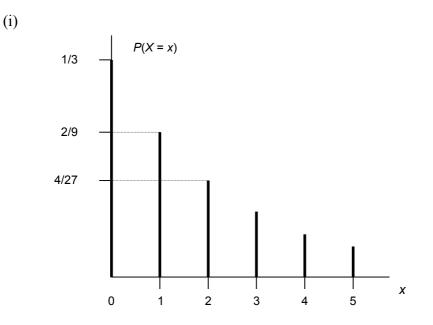
We have $p = 2\theta - \frac{1}{2}$, so the MLE of p is $\tilde{p} = 2\hat{\theta} - \frac{1}{2}$.

Thus $\operatorname{SE}(\tilde{p}) = 2 \operatorname{SE}(\hat{\theta}) = 2 \sqrt{\hat{\theta}(1-\hat{\theta})/n}$.

For n = 100 and z = 45, we have $\hat{\theta} = 0.45$ and so $\tilde{p} = 0.4$ and SE $(\tilde{p}) = 0.0995$.

(iii) Both p and the standard error are estimated to be larger by the second survey. The larger p is plausible, as some people are likely not to admit to taking drugs when asked directly as in the first survey. There is a much better expectation of truthful answers in the second survey. We should not claim that the first is better just because the SE is smaller; the first survey is likely to be biased. (Is it clear what the journalist means by "reliable"?)

Probability mass function: $f(x) = (1-p)^{x} p, \quad x = 0, 1, 2, ..., \quad 0$



(ii) Probability generating function G(s) is

$$G(s) = E\left[s^{X}\right] = \sum_{x=0}^{\infty} s^{x} (1-p)^{x} p = p \sum_{x=0}^{\infty} \left\{(1-p)s\right\}^{x} = \frac{p}{1-(1-p)s}$$

(requires |s| < 1/(1-p) for convergence).

The mean is given by E[X] = G'(1). We have $G'(s) = p \frac{1-p}{\left\{1-(1-p)s\right\}^2}$ and inserting s = 1 gives $G'(1) = \frac{p(1-p)}{p^2}$, i.e. the mean is $\frac{1-p}{p}$.

The variance is given by $\operatorname{Var}(X) = G''(1) + \operatorname{mean} - \operatorname{mean}^2$. $G''(s) = \frac{2p(1-p)^2}{\{1-(1-p)s\}^3}$, so that $G''(1) = \frac{2(1-p)^2}{p^2}$. Thus the variance is given by $\frac{2(1-p)^2}{p^2} + \frac{1-p}{p} - \left(\frac{1-p}{p}\right)^2$ $= \frac{1}{p^2} \{(1-p)^2 + p(1-p)\} = \frac{1-p}{p^2}$.

The solution to parts (iii) and (iv) is on the next page

(iii)
$$P(X \ge x) = \sum_{r=x}^{\infty} p(1-p)^r$$
 [geometric series] $= \frac{p(1-p)^x}{1-(1-p)} = (1-p)^x$ (for $x =$

0, 1, 2, ...). We now use $P(A|B) = \frac{P(A \cap B)}{P(B)}$ and take the event *A* as " $X \ge l + m$ " and the event *B* as " $X \ge l$ ", so that $A \cap B = A$. Thus

$$P(X \ge l + m | X \ge l) = \frac{(1-p)^{l+m}}{(1-p)^{l}} = (1-p)^{m} = P(X \ge m).$$

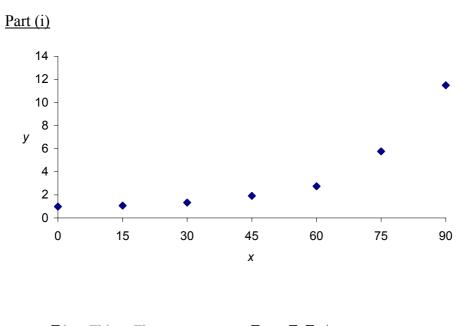
This is the "lack of memory" property of a geometric distribution.

(iv) By independence,
$$P(Z \ge z) = P(X \ge z) P(Y \ge z) = (1-p)^{z} (1-\theta)^{z}$$
.
 $P(Z = z) = P(Z \ge z) - P(Z \ge z+1) = \{(1-p)(1-\theta)\}^{z} - \{(1-p)(1-\theta)\}^{z+1}$
 $= \{(1-p)(1-\theta)\}^{z} (1-1+p+\theta-p\theta) = \{(1-p)(1-\theta)\}^{z} (p+\theta-p\theta), \text{ for } z = 0, 1,$

This is a geometric distribution as given at the start of the question with p replaced by $p + \theta - p\theta$. Hence, from part (ii),

$$E[Z] = \frac{1-p-\theta+p\theta}{p+\theta-p\theta}, \quad Var(Z) = \frac{1-p-\theta+p\theta}{(p+\theta-p\theta)^2}.$$

<u>Note</u> There are many equivalent forms of the formulae that are required to be stated. The basic expressions are $\Sigma(x-\overline{x})(y-\overline{y})$, $\Sigma(x-\overline{x})^2$ and $\Sigma(y-\overline{y})^2$. Convenient computing expressions are $\Sigma xy - (\Sigma x \Sigma y)/n$ and similarly for the others. [Where appropriate, numerators and denominators of fractions could both be multiplied by *n* (7) to avoid possible slight inaccuracies caused by rounding when dividing by 7.]



$$r = \frac{\Sigma(x - \overline{x})(y - \overline{y})}{\sqrt{\Sigma(x - \overline{x})^2 \Sigma(y - \overline{y})^2}} = \frac{\Sigma xy - \Sigma x \Sigma y/n}{\sqrt{(\Sigma x^2 - (\Sigma x)^2/n)(\Sigma y^2 - (\Sigma y)^2/n)}}$$
$$= \frac{1773.795 - (315 \times 25.305/7)}{\sqrt{(20475 - 315^2/7)(180.474 - 25.305^2/7)}} = \frac{635.07}{748.784} = 0.848$$

This indicates a strong linear association between x and y, but nevertheless the scatter diagram clearly suggests that the relationship is curved.

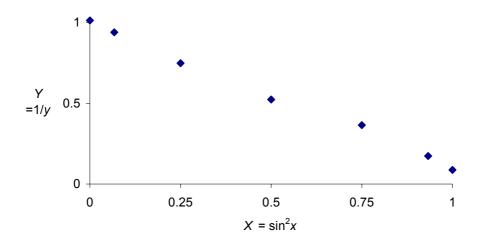
Part (ii)

We have $\frac{1}{y} = \frac{1-\sin^2 x}{a} + \frac{\sin^2 x}{b} = \frac{1}{a} + \sin^2 x \left(\frac{1}{b} - \frac{1}{a}\right)$, i.e. $Y = \frac{1}{a} + \left(\frac{1}{b} - \frac{1}{a}\right)X$ where X and Y are as given. So $A = \frac{1}{a}$ and $B = \frac{1}{b} - \frac{1}{a}$.

The solution to part (iii) is on the next page

<u>Part (iii)</u>							
$X = \sin^2 x$	0	0.067	0.250	0.500	0.750	0.933	1
Y = 1/y	1.013	0.940	0.748	0.523	0.365	0.173	0.087

[Note. Summary statistics for these are given in the question.]



The scatter diagram indicates that the relationship between *X* and *Y* is very close to linear, with little scatter about a straight line. Linear regression should be suitable.

We have $\overline{X} = 3.5/7 = 0.500$ and $\overline{Y} = 3.849/7 = 0.550$.

So the fitted linear regression is Y = A + BX where $B = \frac{\Sigma (X - \overline{X}) (Y - \overline{Y})}{\Sigma (X - \overline{X})^2}$ and

$$A = \overline{Y} - B\overline{X} \; .$$

Carrying out the calculations as in part (i), we get

$$B = \frac{1.03385 - (3.5 \times 3.849/7)}{2.75 - (3.5^2/7)} = -\frac{0.89065}{1} = -0.89065$$

and hence A = 0.550 + (0.89065)(0.500) = 0.9953.

Thus the corresponding estimates of *a* and *b* are given by $\hat{a} = \frac{1}{0.9953} = 1.0047$ and

$$\hat{b} = \left(B + \frac{1}{\hat{a}}\right)^{-1} = \frac{1}{0.10465} = 9.556$$
.

The correlation coefficient for *X* and *Y* is

$$\frac{\Sigma(X-\bar{X})(Y-\bar{Y})}{\sqrt{\Sigma(X-\bar{X})^2 \Sigma(Y-\bar{Y})^2}} = \frac{-0.89065}{1\times\sqrt{\Sigma(Y-\bar{Y})^2}} = \frac{-0.89065}{\sqrt{2.9136-(3.849^2/7)}} = -0.9975.$$

This is an even stronger indication of a linear relationship than in part (i), and we can see from the scatter diagram that the relationship appears almost purely linear (very little random scatter, certainly no curved component).

		l l	Values of Y	Marginal distribution	
		0	1	2	of X
Values of X	-1	1/6	1/12	1/12	1/3
	0	1/12	1/6	1/12	1/3
	1	1/12	1/12	1/6	1/3
Marginal distribution of Y		1/3	1/3	1/3	

(i) The marginal distributions of X and Y are as shown, appended to the table.

Hence E[X] = 0 and E[Y] = 1 (by symmetry; by noting that X and Y are both discrete uniform; or by explicit calculation).

Var(X) = $\Sigma(x-0)^2 P(X=x) = \{(-1)^2 + 0^2 + 1^2)/3 = 2/3.$

Var(Y) can be calculated similarly; or, since Y = X + 1, we have Var(Y) = Var(X).

(ii) The conditional distributions of *Y* for each value of *X* are as follows.

Y =	0	1	2
X = -1	1/2	1/4	1/4
0	1/4	1/2	1/4
1	1/4	1/4	1/2

Hence $E[Y | X = -1] = (0)(\frac{1}{2}) + (1)(\frac{1}{4}) + (2)(\frac{1}{4}) = 3/4.$

Similarly, E[Y | X = 0] = 1 and E[Y | X = 1] = 5/4.

Thus we have $E[Y | X = x] = 1 + \frac{x}{4}$.

(iii) $E[XY] = (-1)(0)(1/6) + (-1)(1)(1/12) + \dots + (1)(2)(1/6) = 1/6.$ ∴ Cov(X, Y) = $E[XY] - E[X]E[Y] = \frac{1}{6} - (0)(1) = 1/6.$ ∴ Corr(X, Y) = $\frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X) \text{Var}(Y)}} = \frac{1/6}{2/3} = \frac{1}{4}.$

X and Y are <u>not</u> independent: their correlation (or covariance) is non-zero. (In fact we saw in part (i) that they are linearly related: Y = X + 1.)

(iv) $Z = X^3 + (Y-1)^3$. The values of Z and their probabilities are as shown:

Y =	0	1	2
X = -1	Z = -2; p = 1/6	Z = -1; p = 1/12	Z = 0; p = 1/12
0	Z = -1; p = 1/12	Z = 0; p = 1/6	Z =1; p = 1/12
1	Z = 0; p = 1/12	Z = 1; p = 1/12	Z = 2; p = 1/6

Thus we have

Values of Z	-2	-1	0	1	2
Probabilities	1/6	1/6	1/3	1/6	1/6