## Partial Solutions for h4/2014S: Sampling Distributions

(1) Let  $X_1$  and  $X_2$  be two independent random variables, each with the same probability distribution given as follows.

$$f(x) = \frac{1}{2}e^{-x/2}, \ x \ge 0$$

- (a) Compute the probability distribution function of the new random variable  $Y = X_1 + X_2$ .
- (b) What type of probability distribution is your answer in (a)?

Thus,  $f_Y(y) = \frac{1}{4}ye^{-y/2}, \ 0 < y < \infty$ 

- **S1(b)**  $Y = X_1 + X_2$  has a gamma distribution with parameters  $\alpha = 2$  and  $\theta = 2$  since  $M_Y(t) = M_{X_1}(t) \times M_{X_2}(t) = \frac{1}{1-2t} \times \frac{1}{1-2t} = \frac{1}{(1-2t)^2}.$
- (2) Let the moment-generating function of X be

$$\phi(t) = e^{3t + 2t^2}, \quad -\infty < t < \infty$$

- (a) Find the mean E(X) and variance Var(X).
- (b) Name the distribution of X.
- (c) Write down the p.d.f f(x) for X.
- **S2(a)** E(X) = 3 and Var(X) = 4.

**S2(b)**  $X \sim N(3,4)$ , that is, X has a normal distribution with mean 3 and variance 4.

- **S2(c)** The p.d.f of X is  $f(x) = \frac{1}{\sqrt{8\pi}} e^{-(x-3)^2/8}, -\infty < x < \infty.$
- (3) Let  $X_1 \sim b(n_1, p)$  and  $X_2 \sim b(n_2, p)$  be independent r.v.s. Define  $Y = X_1 + X_2$ .

- (a) What is  $M_Y(t)$ ?
- (b) How is Y distributed?

**S3(a)**  $M_Y(t) = M_{X_1}(t)M_{X_2}(t) = (1 - p + pe^t)^{n_1}(1 - p + pe^t)^{n_2} = (1 - p + pe^t)^{n_1 + n_2}$ **S3(b)**  $Y \sim b(n_1 + n_2, p).$ 

(4) If  $X \sim N(\mu, \sigma^2)$ , show that  $Y = (aX + b) \sim N(a\mu + b, a^2\sigma^2), a \neq 0$ .

**S4** Without loss of generality (W.L.O.G.), let a > 0, then

$$P(Y \le y) = P(aX + b \le y) = P(X \le \frac{y-b}{a})$$
$$= \int_{-\infty}^{(y-b)/a} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(x-\mu)^2/(2\sigma^2)} dx$$

then

$$f(y) = F'(y) = \frac{1}{\sqrt{2\pi}(a\sigma)} e^{-[y - (a\mu + b)]^2 / [2(a\sigma)^2]}, \quad -\infty < y < \infty$$

(5) The joint probability density function for two continuous random variables X and Y is given as follows.

$$f(x,y) = \begin{cases} \beta xy & 0 < x < 2 \text{ and } 0 < y < 4 \\ 0 & elsewhere \end{cases}$$

- (a) Determine the constant  $\beta$  to make it a joint p.d.f.
- (b) Determine the conditional probability distribution function f(x|y).
- (c) Compute the covariance between X and Y.

**S5(a)**  $\beta = \frac{1}{16}, f_X(y) = \int_0^4 \frac{1}{16} xy dx = \frac{x}{2}, \text{ for } 0 < x < 2, f_Y(y) = \int_0^2 \frac{1}{16} xy dx = \frac{y}{8}, \text{ for } 0 < y < 4.$ 

**S5(b)**  $f(x|y) = \frac{f(x,y)}{f(y)} = \frac{x}{2}, \ f(x,y) = \frac{xy}{16} = \frac{x}{2} \times \frac{y}{8} = f_X(x)f_Y(y).$ 

**S5(c)**  $Cov(X, Y) = E[(X - \mu_X)(Y - \mu_Y)]$ , where

$$\mu_x = \int_0^2 x \cdot \frac{x}{2} dx = \frac{4}{3}$$

and

$$\mu_y = \int_0^4 y \cdot \frac{y}{8} dy = \frac{8}{3},$$

Thus

$$Cov(X,Y) = E[XY] - \mu_X \mu_Y = \int_0^2 \int_0^4 \frac{1}{16} x^2 y^2 dy dx - \frac{32}{9} = \frac{8 \times 64}{144} - \frac{32}{9} = 0.$$

Note that X and Y are *uncorrelated*.

(6) Show that the sum of *n* independent Poisson random variables with respective means  $\lambda_1, \lambda_2, \ldots, \lambda_n$  is Poisson with mean  $\lambda = \sum_{i=1}^n \lambda_i$ .

**S6** Let 
$$Y = \sum_{i=1}^{n} X_i$$
, then  

$$E[e^{tY}] = E[e^{t\sum_{i=1}^{n} X_i}] = \sum_{x_1, x_2, \dots, x_n} \left[e^{\sum_{i=1}^{n} (tx_i)}\right] f(x_1) f(x_2) \cdots f(x_n)$$

$$= \sum_{x_1, x_2, \dots, x_n} \prod_{i=1}^{n} [e^{tx_i} f(x_i)] = \prod_{i=1}^{n} \left(\sum_{x_i} e^{tx_i} f(x_i)\right)$$

$$= \prod_{i=1}^{n} e^{\lambda_i (e^t - 1)} = e^{[\sum_{i=1}^{n} \lambda_i](e^t - 1)}$$

Thus, Y has a Poisson distribution with mean  $\lambda = \sum_{i=1}^{n} \lambda_i$ .

- (7) Let  $Z_i \sim N(0,1)$ , for  $1 \le i \le 7$  and define  $W = \sum_{i=1}^7 Z_i^2$ . Find P(1.69< W < 14.07).
- **S7**  $Y \sim \chi^2(7)$ , then  $P(1.69 < W < 14.07) = P(W < 14.07) P(W \le 1.69) = 0.95 0.025 = 0.925.$
- (8) Assume there are 100 obervations, denoted by  $x_i$ ,  $1 \le i \le 100$ , and each is drawn from a population with a continuous distribution in [0,2].
  - (a) Give the formula for the sample mean  $\overline{X}$  and sample variance  $S^2$  in this problem.
  - (b) Compute the sample mean and sample variance for  $\overline{X}$ .
  - (c) Try to give an approximate distribution for  $\overline{X}$  as best as you can and explain your reason.
  - (d) Compute the probability that the sample mean value is larger than 1.02. Note that you can express your solution with the cumulative standard normal distribution function

$$\Phi(r) = \int_{-\infty}^{r} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$$

**S8(a)** 
$$\overline{X} = \frac{1}{100} \sum_{i=1}^{100} X_i, S^2 = \frac{1}{99} \sum_{i=1}^{100} (X_i - \overline{X})^2$$

**S8(b)** 
$$E(X_i) = \int_0^2 (x \cdot \frac{1}{2}) dx = 1, \quad Var(X_i) = E(X_i^2) - [E(X_i)]^2 = \int_0^2 (x^2 \cdot \frac{1}{2}) dx - 1 = \frac{1}{3},$$
  
then  
 $E(\overline{X}) = \frac{1}{100} \sum_{i=1}^{100} E(X_i) = 1$ 

$$Var(\overline{X}) = \frac{1}{10000} \sum_{i=1}^{100} Var(X_i) = \frac{1}{300}$$

**S8(c)** By C.L.T.

$$\frac{\overline{X} - E(\overline{X})}{\sqrt{\frac{1}{100} Var(X_1)}} \to N(0, 1) \quad in \quad probability.$$

Then

$$P(\overline{X} > 1.02) = P(10\sqrt{3}(\overline{X} - 1) > 10\sqrt{3}(1.02 - 1)) = 1 - \Phi(0.02 * 10\sqrt{3})$$
$$= \Phi(-0.3464) \approx 0.3645$$

(9) Let random() be a pseudo random number generator which can randomly generate a real in [0, 1). Let  $X \sim N(0, 1)$ , and  $Y \sim N(3, 16)$ . Give algorithms to sample (simulate) the distributions of X and Y, respectively.

**S9** Let  $R_i \sim U(0, 1)$  and  $R = \sum_{i=0}^n R_i$ , by C.L.T., we have

$$\frac{R - E(R)}{\sqrt{nVar(R_1)}} \rightarrow N(0, 1) \text{ in probability}$$

or

$$\frac{R/n - E(R_1)}{\sqrt{Var(R_1)/n}} \rightarrow N(0,1) \text{ in probability}$$

In application, if we choose n = 12, then

$$(\sum_{i=0}^{12} R_i) - 6 \approx X \sim N(0,1)$$

- (10) Let  $X_1, X_2, \ldots, X_{30}$  be a random sample of size 30 from a Poisson distribution with a mean 2/3. Approximate
  - (a)  $P(15 < \sum_{i=1}^{30} X_i \le 22).$ (b)  $P(21 \le \sum_{i=1}^{30} X_i < 27).$

S10(a)

$$P(15 < \sum_{i=1}^{30} X_i \le 22) = P(15.5 \le \sum_{i=1}^{30} X_i \le 22.5)$$

$$= P(\frac{(15.5-20)}{\sqrt{30 \times (2/3)}}) \le \frac{\sum_{i=1}^{30} X_i - 20}{\sqrt{30 \times (2/3)}} \le \frac{(22.5-20)}{\sqrt{30 \times (2/3)}})$$

$$= P(\frac{-4.5}{\sqrt{20}} \le \frac{\sum_{i=1}^{30} X_i - 20}{\sqrt{20}}) \le \frac{2.5}{\sqrt{20}})$$

$$= \Phi(\frac{2.5}{\sqrt{20}}) - \Phi(\frac{-4.5}{\sqrt{20}}) = \Phi(0.5590) - \Phi(-1.0062)$$

$$\approx 0.7088 - 0.1630 = 0.5458$$

S10(b)

$$P(21 \le \sum_{i=1}^{30} X_i < 27) = P(\frac{(20.5-20)}{\sqrt{30 \times (2/3)}}) \le \frac{\sum_{i=1}^{30} X_i - 20}{\sqrt{30 \times (2/3)}}) \le \frac{(26.5-20)}{\sqrt{30 \times (2/3)}})$$
$$= P(\frac{0.5}{\sqrt{20}} \le \frac{\sum_{i=1}^{30} X_i - 20}{\sqrt{20}} \le \frac{6.5}{\sqrt{20}})$$
$$= \Phi(\frac{6.5}{\sqrt{20}}) - \Phi(\frac{0.5}{\sqrt{20}}) = \Phi(1.4534) - \Phi(0.1118)$$
$$\approx 0.9259 - 0.5435 = 0.3824$$

- (11) Let  $X_1, X_2, \dots, X_n$  be a random sample of an exponential distribution with mean  $\theta$ , that is,  $f(x) = \frac{1}{\theta} e^{-x/\theta}, x > 0.$ 
  - (a) Show that  $P(min(X_1, X_2, \dots, X_n) > 2) = e^{-2n/\theta}$ .
  - (b) Show that  $P(max(X_1, X_2, \dots, X_n) > 2) = 1 (1 e^{-2/\theta})^n$ .
  - (c) Find a minimum n such that  $P(min(X_1, X_2, \dots, X_n) > 2) \le 3\%$  when  $\theta = 2$ .
  - (d) Find a minimum n such that  $P(max(X_1, X_2, \dots, X_n) > 2) \ge 90\%$  when  $\theta = 2$ .

$$P[min(X_1, X_2, \cdots, X_n) > 2] = \Pi_{i=1}^n P(X_i > 2) = \Pi_{i=1}^n \left[ \int_2^\infty \frac{1}{\theta} e^{-x/\theta} dx \right]$$
$$= \Pi_{i=1}^n \left[ e^{-2/\theta} \right] = e^{\frac{-2n}{\theta}}$$

$$P[max(X_1, X_2, \dots, X_n) > 2] = P(at \ least \ one \ X_i > 2) = 1 - \prod_{i=1}^n P(X_i \le 2)$$
$$= 1 - \prod_{i=1}^n \left( \int_0^2 \frac{1}{\theta} e^{-x/\theta} dx \right) = 1 - [1 - e^{-2/\theta}]^n$$

- **S11(c)** Find a minimum *n* such that  $P(min(X_1, X_2, \dots, X_n) > 2) = e^{-2n/\theta} = e^{-2n/2} \le 0.03$ , that is,  $e^{-n} \le 0.03$ , or  $n \ge \lceil -ln(0.03) = 3.5066 \rceil = 4$
- **S11(d)** Find a minimum *n* such that  $P(max(X_1, X_2, \dots, X_n) > 2) = 1 (1 e^{-2/\theta})^n = 1 [1 e^{-2/2}]^n \ge 0.9$ , that is,  $1 [1 e^{-1}]^n \ge 0.9$  or  $[1 e^{-1}]^n \le 0.1$ . In other words, we want to find *n* such that  $(e^{-1} + (0.1)^{1/n}) \ge 1$ , or  $(0.1)^{1/n} \ge (1 e^{-1})$ , or  $n \ln(0.1) \ge \ln(\frac{e-1}{e})$ . Thus,  $n \ge 6$ .
- (12) Let  $W_1 < W_2 < \cdots < W_n$  be the order statistics of a random sample of zise *n* from the uniform distribution U(0,1).
  - (a) Find the probability density function of  $W_1$ .
  - (b) Find the probability density function of  $W_n$ .
  - (c) Use the result of (a) to find  $E(W_1)$ .
  - (d) Use the result of (b) to find  $E(W_n)$ .

## Solution:

$$G_{r}(y) = P(W_{r} \leq y) = \sum_{k=r}^{n} \binom{n}{k} [F(y)]^{k} [1 - F(y)]^{n-k}$$
$$= \sum_{k=r}^{n-1} \binom{n}{k} [F(y)]^{k} [1 - F(y)]^{n-k} + [F(y)]^{n}$$

Thus the p.d.f. of  $W_r$  could be derived as

$$g_r(y) = G_r'(y) = \frac{n!}{(r-1)!(n-r)!} [F(y)]^{r-1} [1 - F(y)]^{n-r} f(y), \quad a < y < b$$
(1)

In particular,

$$g_1(y) = n[1 - F(y)]^{n-1}f(y), \quad a < y < b$$
  
$$g_n(y) = n[F(y)]^{n-1}f(y), \quad a < y < b$$

- (1) If  $X_i$  has a U(0,1) distribution, then  $E(W_r) = \frac{r}{n+1}$ . In particular,  $g_1(y) = n(1-y)^{n-1}$ , 0 < y < 1,  $E(W_1) = \frac{1}{n+1}$ ;  $g_n(y) = ny^{n-1}$ , 0 < y < 1,  $E(W_n) = \frac{n}{n+1}$ .
- (2) If  $X_j$  has an exponential distribution with mean 2, then  $g_1(y) = ne^{-ny}$ , y > 0 and  $E(W_1) = \frac{1}{n}$ .
- \*(13) One of the most popular distributions used to model the lifetimes of electric components is the Weibull distribution, whose probability density function is given by

$$f(t) = \alpha t^{\alpha - 1} e^{-t^{\alpha}}, \text{ for } t > 0, \ \alpha > 0.$$

Determine for which values of  $\alpha$  the hazard function of a Weibull random variable is increasing, for which values it is decreasing, and for which values it is constant.

**\*S13**  $f'(t) = \alpha(\alpha - 1)t^{\alpha - 2}e^{-t^{\alpha}} - \alpha^{2}t^{2\alpha - 2}e^{-t^{\alpha}}$ , thus,

$$f'(t) > 0 \quad if \quad t < \left(\frac{\alpha - 1}{\alpha}\right)^{1/\alpha}$$
$$< 0 \quad if \quad t > \left(\frac{\alpha - 1}{\alpha}\right)^{1/\alpha}$$