

DEPARTMENT OF MATHEMATICAL STATISTICS

Qualifying Examination

Statistics

Friday November 16, 1979

1 o'clock

Do Any 4 Questions

Answer each question in a separate book



1. Recall Helmert's transformation for  $X_1, \dots, X_n$  iid  $N(\mu, \sigma^2)$  r. v. 's:

$$U_1 = \bar{X}_n$$

$$U_i = X_i - \bar{X}_i, \quad i = 2, \dots, n$$

where  $\bar{X}_i = (X_1 + \dots + X_i)/i$ . The  $U_i$  are independent normal variates.

Let now  $W_1 = [n/(n-1)] \cdot (X_n - \bar{X}_n)^2$ ,  $Q = \sum_1^n (X_i - \bar{X}_n)^2$  and  $Z = W_1/Q$ .

- Express the statistic  $Z$  in terms of the standardized  $U_i$ .
- From that representation and elementary probability methods show that the conditional distribution of  $Z$  given  $Q$  is Beta  $(1/2, (n-2)/2)$ .

Hint: Consider first the independent variates  $W_1$  and  $W_2 = Q - W_1$ . From that the joint density of  $(Z, Q)$  is easily derived.

2. Let  $X$  be an observation from  $P_\theta$  population, where  $P_\theta$  is defined by  $P_\theta[X=i] = \frac{1}{\theta}$  for  $i=1, \dots, \theta$ .  
 $= 0$  otherwise.

$x=1, 2, \dots, \theta$   
 $Eg(x) = 0$  for all  $\theta$   
 $\theta=1 \Rightarrow g(1)=0$   
 $\theta=2 \Rightarrow g(1)+g(2)=0 \Rightarrow g(1)=-g(2)$

(i) Show that  $\mathcal{P} = \{P_\theta, \theta=1, 2, \dots\}$  is complete and find the U.M.V.U. (uniformly minimum variance unbiased) estimate of  $\theta$ .

$g(x) = 0$   
 and sure  $x$  is  
 $EX = \frac{\theta+1}{2}$   
 $T = 2X - 1$   
 U.M.V.

(ii) For any integer  $K$ , let  $\mathcal{P}_K = \{P_\theta, \theta=1, 2, \dots, \theta \neq K\}$ . Show that  $\mathcal{P}_K$  is not complete and your estimate in (i) is not U.M.V.U. in this case.

$Eg(x) = 0 \Rightarrow g(k) + g(k+1) = 0$  is ok.  $\rightarrow$  not complete

$g(i) = 0$  for  $i=1, 2, \dots, k-1$ .  $g(k+2) = 0$ .  $g(k+3) = 0$ .  $g(k+4) = 0$ .  $l \geq 2$

3. An individual belongs to one of two populations  $\Pi_0$  or  $\Pi_1$  with prior probability  $p_\theta$  that a randomly selected individual

the UMVU  
 $2X-1 + I_{(X=K)} - I_{(X=K+1)}$



belongs to  $\Pi_\theta$  ( $\theta = 0, 1$ ). The individuals in  $\Pi_\theta$  have a distribution of scores  $X$  which is  $N(\mu_\theta, 1)$  where  $\mu_0$  and  $\mu_1$  may be assumed known. The loss function  $L(\theta, a)$  for deciding that the individual belongs to population  $\Pi_a$  when in fact he belongs to  $\Pi_\theta$  is given by

$$L(\theta, a) = c_1(1-a)\theta + c_0a(1-\theta) \quad (a=0, 1; \theta=0, 1)$$

where  $c_0$  and  $c_1$  are known positive constants.

What is the Bayes decision rule  $\delta(X)$  against the prior  $(p_0, p_1)$  for deciding population membership based on  $X$ ?

4. Let  $y_t = t\beta + \epsilon_t$ ,  $t=1, 2, \dots, n$ , with  $\epsilon_t$  i.i.d  $N(0, \sigma^2)$ .

Let  $\delta_t = y_{t+1} - y_t$ ,  $t=1, \dots, n-1$

a) Find the mean and variance of  $\delta_t$  and of  $\bar{\delta} = \frac{1}{n-1} \sum_{t=1}^{n-1} \delta_t$ .

$$\begin{aligned} E\delta_t &= E y_{t+1} - E y_t = \beta \\ \text{Var} \delta_t &= \text{Var}(\epsilon_{t+1}) + \text{Var}(\epsilon_t) = 2\sigma^2 \\ E\bar{\delta} &= \beta, \quad \text{Var} \bar{\delta} = \text{Var}\left(\frac{1}{n-1} \sum_{t=1}^{n-1} \delta_t\right) \\ &= \frac{2\sigma^2}{(n-1)^2} \end{aligned}$$

b) Is  $\underline{\delta} = (\delta_1, \dots, \delta_{n-1})'$  sufficient for  $\beta$ ?  $\neq$

c) Give the best linear estimate you can for  $\beta$  based on  $\underline{\delta}$ , and compare it with the least squares estimate based on  $y_1, \dots, y_t$ .

5. Consider the regression model,  $y_i = x_{i1}\beta_1 + \dots + x_{ip}\beta_p + \epsilon_i$ ,  $i=1, 2, \dots$

where  $\epsilon_i$  are i.i.d. random variables with mean zero and variance  $\sigma^2 > 0$ . Let  $X_n$  denote the design matrix  $(x_{ij})$   $1 \leq i \leq n$ ;  $1 \leq j \leq p$ . Assume that  $(X_n' X_n)^{-1}$  exists for  $n \geq p$ . Let  $b_n$  be the least squares estimate of  $\beta_1$ .  $\therefore \text{rk}(X_n) = p \rightarrow \beta_1$  is estimable

(i) Show that  $\text{var}(b_n)$  is a decreasing function of  $n$  ( $n \geq p$ ).

除非  $x_{n+1, j} \equiv 0 \forall j$

$$\begin{aligned} b_n &= c_1 y_1 + \dots + c_n y_n \\ b_{n+1} &= c_1 y_1 + \dots + c_n y_n + c_{n+1} y_{n+1} \end{aligned}$$

无偏线性集中最小者  
选择  $c_{n+1} = 0$  则  $b_{n+1} = b_n$  可也



(ii) If  $p = 2$ ,  $E|\epsilon_i|^3 < \infty$ ,  $\text{var}(b_n) \rightarrow 0$  and all design points  $(x_{i1}, x_{i2})$  are taken from a compact subset of  $\{(x_1, x_2) | x_2 = 1\}$  then  $(b_n - \beta_1) / \{\text{var}(b_n)\}^{\frac{1}{2}} \xrightarrow{d} N(0, 1)$ .

$$\frac{\sum (x_i - \bar{x}) \epsilon_i}{\sqrt{\sigma^2 \sum (x_i - \bar{x})^2}} = \frac{\sum (x_i - \bar{x})(y_i - E y_i)}{\sqrt{\sigma^2 \sum (x_i - \bar{x})^2}} \xrightarrow{d} N(0, 1)$$

6. Let  $p_0, p_1, p_2$  be unknown probabilities, with  $\sum p_i = 1$ , and let  $A_i$  be corresponding events, with  $UA_i = S$ , the certain event. We may interpret these as indifference, preference for treatment I, and preference for treatment II, resp., on a single trial of an experiment. Consider now  $n$  independent such trials with constant  $(p_i, i = 0, 1, 2)$ , and let  $X_i$  be the number of times  $A_i$  occurs.

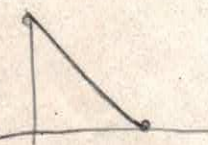
a) Consider first the conditional statistical model that describes the assumed behaviour of  $(X_1, X_2)$  given that  $X_0$  equals a fixed given constant. Give the UMP level  $\alpha$  test for  $p_1 = p_2$  vs.  $p_1 > p_2$  under such conditional model. Specifically: give the test statistic, the distribution of the test statistic, and the form of the critical function.

b) Consider now such conditional test as a test depending on  $X_0, X_1$  and  $X_2$ , for  $p_1 = p_2$  vs.  $p_1 > p_2$ , under the original unconditional model. Show that this test is UMP among all tests  $\phi$  such that  $E(\phi | X_0) \equiv \alpha$  under the null hypothesis (i.e., whenever  $p_1 = p_2$ ).

a) given  $x_0 = k \Rightarrow x_1 + x_2 = n - k$

$$P(X_1 = l | X_0 = k) = \binom{n-k}{l} \left(\frac{p_1}{p_1 + p_2}\right)^l \left(\frac{p_2}{p_1 + p_2}\right)^{n-k-l}$$

$l + m = n - k$



在定条件下分布

for such  $l$ ,  $\frac{P_{p_1 > p_2}(X_1 = l | X_0 = k)}{P_{p_1 = p_2}(X_1 = l | X_0 = k)} > 1$  reject  $H_0$   
 $\Rightarrow l > k$  reject

b) size  $P_{p_1 = p_2}(X_1 = l | X_0 = k) = \binom{n-k}{l} \left(\frac{1}{2}\right)^{n-k}$   
 $E(\phi | X_0) \equiv \alpha$  for  $X_0 = 1, 2, \dots$