

Ph.D. Qualifying Exam A

Probability Distributions and Regression Analysis

January 13, 2021

1. Let $\{\mathbf{X}_n\}$ be a sequence of independent identically distributed random vectors with common mean vector $\boldsymbol{\mu}$ and variance-covariance matrix $\boldsymbol{\Sigma}$, which is positive definite. Assume that the common moment generating function $M(\mathbf{t})$ of \mathbf{X}_n exists in an open neighborhood of $\mathbf{0}$. State and prove the Multivariate Central Limit Theorem.
2. Let T be Geometric with parameter p_1 in $(0, 1)$ and U be an independent Geometric with parameter p_2 in $(0, 1)$. (a) Compute $P(T < U)$. (b) Derive the conditional probability mass function of T given the event $\{T < U\}$.
3. Let Y_1, \dots, Y_n be a random sample of size $n = 6$ from a normal population with mean 0 and variance 1 and let \bar{Y} denote the sample mean based on this sample of size $n = 6$. Let Y_7 be another independent observation from the same population. (a) Prove the independence of \bar{Y} and $\sum_{i=1}^6 (Y_i - \bar{Y})^2$. (b) What is the distribution of

$$\frac{6\bar{Y}}{\sqrt{\sum_{i=1}^6 (Y_i - \bar{Y})^2 + Y_7^2}},$$

and why?

4. For a normal error simple linear regression model:

$$Y_i = \beta_0 + \beta_1 X_i + \epsilon_i,$$

where ϵ_i are i.i.d. $N(0, \sigma^2)$ random errors for $i = 1, 2, \dots, n$. Prove the following fact that the ANOVA F test and the two-sided t test for $H_0 : \beta_1 = 0$ versus $H_1 : \beta_1 \neq 0$ are equivalent.

5. For a normal error general linear regression model,

$$Y_i = \beta_0 + \beta_1 X_{i1} + \beta_2 X_{i2} + \dots + \beta_{p-1} X_{ip-1} + \epsilon_i,$$

where ϵ_i are i.i.d. $N(0, \sigma^2)$ random errors for $i = 1, 2, \dots, n$, show that the matrices associated with quadratic forms of the regression sum of squares and that of the residual sum of squares are both symmetric and idempotent.

6. Consider the linear model with vector of observations $Y' = (Y_1, \dots, Y_n)$,

$$E(Y_i) = \beta_0 + x_{i1}\beta_1 + \dots + x_{ip-1}\beta_{p-1}, i = 1, \dots, n,$$

satisfying the assumptions of normality, independence and homoscedasticity, with design matrix X of dimension $n \times p$ and rank $p, p < n$. Consider $r_i = e_i / \sqrt{SSE(1 - h_{ii}) / (n - p)}$, where e_i is the i th residual, h_{ii} is the i^{th} diagonal element of the hat matrix $H = X(X^T X)^{-1} X^T$ and SSE is the sum of squared errors. Here, ' T ' denotes the transpose of a matrix. Show that r_i^2 has $(n-p)F_{1, n-p-1} / [(n-p-1) + F_{1, n-p-1}]$ distribution. Use the fact that $SSE - \{e_i^2 / (1 - h_{ii})\}$ is independent of e_i^2 .