

December 6, 2005

Exercise 1

$$\begin{aligned}
P' &= (X(X'X)^{-1}X')' = (X')'((X'X)^{-1})'X' = X(X'(X')^{-1})^{-1}X' = P \\
PP &= X(X'X)^{-1}X'X(X'X)^{-1}X' = X(X'X)^{-1}X' = P \\
M' &= (I-P)' = I-P' = I-P = M \quad MM = (I-P)(I-P) = I-2P+PP = I-P = M \\
PX &= PX = X(X'X)^{-1}X'X = X, \quad MX = (I-P)X = (X-X(X'X)^{-1}X'X) = X-X = 0
\end{aligned}$$

Exercise 2

$$\begin{aligned}
\varepsilon &\sim N(0, \sigma^2 I), \quad \hat{\beta} = (X'X)^{-1}X'y = (X'X)^{-1}X'(X\beta + \varepsilon) = \beta + (X'X)^{-1}X'\varepsilon \\
\hat{\beta} &\sim N(\beta, \sigma^2(X'X)^{-1}X'X(X'X)^{-1}) = N(\beta, \sigma^2(X'X)^{-1}) \\
e &= M\varepsilon \sim N(0, \sigma^2 MM') = N(0, \sigma^2 MM) = N(0, \sigma^2 M)
\end{aligned}$$

Hence, e and $\hat{\beta}$ have joint normal distribution.

$$E[(X'X)^{-1}X'\varepsilon \cdot (M\varepsilon)'] = \sigma^2(X'X)^{-1}X'M' = \sigma^2(X'X)^{-1}(MX)' = 0$$

Therefore, they are independent.

$$\text{trace}(M) = \text{trace}(I-P) = \text{trace}(I) - \text{trace}(P) = n - k$$

$$\frac{e'e}{\sigma^2} = \frac{\varepsilon'M\varepsilon}{\sigma^2} \sim \chi^2(n-k), \quad E\left[\frac{e'e}{\sigma^2}\right] = n-k \quad \Rightarrow \quad E\left[\frac{e'e}{n-k}\right] = \sigma^2.$$

Exercise 3

Because Φ is monotonically increasing, we have $Y \leq y$ if and only if $\Phi(Y) \leq \Phi(y)$. Hence, $\Pr[Y \leq y] = \Pr[\Phi(Y) \leq \Phi(y)] = \Pr[\Phi(\Phi^{-1}(U)) \leq \Phi(y)] = \Pr[U \leq \Phi(y)] = \Phi(y)$. Therefore, if $U \sim U(0, 1)$, then $\Phi^{-1}(U)$ has c.d.f. Φ , i.e. $\Phi^{-1}(U) \sim N(0, 1)$.

Exercise 4

$X_n \sim \chi^2(n)$. If we had $Y_i \stackrel{iid}{\sim} N(0, 1)$, then $\sum_{i=1}^n Y_i^2 \sim \chi^2(n)$.

We can treat X_n as a sum of some i.i.d. normal Y_i^2 because their distributions are equivalent.

By construction $Y_i \stackrel{iid}{\sim} N(0, 1)$ and $Z_i = Y_i^2 \stackrel{iid}{\sim} \chi^2(1)$. $E[Z_i] = 1$ $Var[Z_i] = 2$

Replacing $\frac{X_n - n}{\sqrt{n}} = \frac{\sum_{i=1}^n Y_i^2 - n}{\sqrt{n}} = \sqrt{n}(\bar{Y}^2 - 1) = \sqrt{n}(\bar{Z} - 1) = \sqrt{n}(\bar{Z} - E[Z])$

By LLN $\sqrt{n}(\bar{Z} - E[Z]) \xrightarrow{d} N(0, Var[Z]) = N(0, 2)$.

Exercise 5

$$Z = \begin{bmatrix} Y \\ X \end{bmatrix} \sim N\left(\begin{bmatrix} \mu_Y \\ \mu_X \end{bmatrix}, \begin{bmatrix} \sigma_Y^2 & \rho\sigma_Y\sigma_X \\ \rho\sigma_Y\sigma_X & \sigma_X^2 \end{bmatrix}\right) = N(\mu, \Sigma)$$

$$\tilde{Z} = Z - \mu = \begin{bmatrix} Y - \mu_Y \\ X - \mu_X \end{bmatrix} \sim N(0, \Sigma)$$

$$\begin{bmatrix} U \\ X - \mu_X \end{bmatrix} = \begin{bmatrix} (Y - \mu_Y) - \rho\frac{\sigma_Y}{\sigma_X}(X - \mu_X) \\ X - \mu_X \end{bmatrix} = \begin{bmatrix} 1 & -\rho\frac{\sigma_Y}{\sigma_X} \\ 0 & 1 \end{bmatrix} \begin{bmatrix} Y - \mu_Y \\ X - \mu_X \end{bmatrix} = A\tilde{Z}$$

$$A\tilde{Z} \sim N(A0, A\Sigma A') = N\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} \sigma_Y^2 - \rho^2\sigma_Y^2 & 0 \\ 0 & \sigma_X^2 \end{bmatrix}\right) \quad \Rightarrow \quad VarU = \sigma_Y^2 - \rho^2\sigma_Y^2$$

U and $X - \mu_X$ have joint normal distribution with zero covariance and hence are independent.

$$Y = \mu_Y + \rho \frac{\sigma_Y}{\sigma_X} (X - \mu_X) + U$$

$$Y | (X = x) = \mu_Y + \rho \frac{\sigma_Y}{\sigma_X} (x - \mu_X) + U \sim N \left(\mu_Y + \rho \frac{\sigma_Y}{\sigma_X} (x - \mu_X), \text{Var} U \right)$$

$$E[Y^2 | X = x] = \text{Var}[Y | X = x] + E[Y | X = x]^2 = \sigma_Y^2 - \rho^2 \sigma_Y^2 + \left(\mu_Y + \rho \frac{\sigma_Y}{\sigma_X} (x - \mu_X) \right)^2$$

Exercise 6

$$X_i \stackrel{iid}{\sim} N(\mu, \sigma^2) \Rightarrow X_i - \mu \stackrel{iid}{\sim} N(0, \sigma^2) \Rightarrow Z_i = \frac{X_i - \mu}{\sigma} \stackrel{iid}{\sim} N(0, 1)$$

$$X_i = \mu + \sigma Z_i \quad \bar{X} = \frac{1}{n} \sum_{i=1}^n X_i = \mu + \sigma \frac{1}{n} \sum_{i=1}^n Z_i = \mu + \sigma \bar{Z}$$

$$Y_n = \frac{\frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^3}{\left(\frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2 \right)^{3/2}} = \frac{\frac{1}{n} \sum_{i=1}^n (\mu + \sigma Z_i - \mu - \sigma \bar{Z})^3}{\left(\frac{1}{n} \sum_{i=1}^n (\mu + \sigma Z_i - \mu - \sigma \bar{Z})^2 \right)^{3/2}} = \frac{\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^3}{\left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2 \right)^{3/2}}$$

$$\text{plim}_{n \rightarrow \infty} \left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2 \right)^{3/2} = \text{plim}_{n \rightarrow \infty} \left(\frac{1}{n} \sum_{i=1}^n Z_i^2 - \left(\frac{1}{n} \sum_{i=1}^n Z_i \right)^2 \right)^{3/2} =$$

$$\left(\text{plim}_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n Z_i^2 - \left(\text{plim}_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n Z_i \right)^2 \right)^{3/2} = (E[Z_i^2] - (E[Z_i])^2)^{3/2} = 1$$

$$E[Z^{2k}] = \frac{(2k)!}{2^k k!} \quad E[Z^{2k-1}] = 0 \quad \frac{1}{2^k} \frac{(2k)!}{(k)!} \Big|_{k=2} = 3 \quad \frac{1}{2^k} \frac{(2k)!}{(k)!} \Big|_{k=3} = 15$$

$$E \begin{bmatrix} Z_i^3 \\ Z_i^2 \\ Z_i \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \quad \text{Var} \begin{bmatrix} Z_i^3 \\ Z_i^2 \\ Z_i \end{bmatrix} = E \begin{bmatrix} Z_i^6 & Z_i^5 & Z_i^4 \\ Z_i^5 & Z_i^4 & Z_i^3 \\ Z_i^4 & Z_i^3 & Z_i^2 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \end{bmatrix} =$$

$$\begin{bmatrix} 15 & 0 & 3 \\ 0 & 3 & 0 \\ 3 & 0 & 1 \end{bmatrix} - \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 15 & 0 & 3 \\ 0 & 2 & 0 \\ 3 & 0 & 1 \end{bmatrix}$$

The vector is i.i.d. with finite expectation and variance. Hence, by LLN

$$\sqrt{n} \left(\begin{bmatrix} \frac{1}{n} \sum_{i=1}^n Z_i^3 \\ \frac{1}{n} \sum_{i=1}^n Z_i^2 \\ \frac{1}{n} \sum_{i=1}^n Z_i \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right) \xrightarrow{d} N \left(\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 15 & 0 & 3 \\ 0 & 2 & 0 \\ 3 & 0 & 1 \end{bmatrix} \right)$$

$$\left[\frac{1}{n} \sum_{i=1}^n Z_i^3 \quad \frac{1}{n} \sum_{i=1}^n Z_i^2 \quad \frac{1}{n} \sum_{i=1}^n Z_i \right]' \equiv [A_n \quad B_n \quad C_n]'$$

$$\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^3 = A_n - 3C_n B_n + 2(C_n)^3 = u \left([A_n \quad B_n \quad C_n]' \right)$$

$$G = \frac{\partial u(x)}{\partial x} \Big|_{(0,1,0)} = [1 \quad -3C_n \quad -3B_n + 6(C_n)^2] \Big|_{(0,1,0)} = [1 \quad 0 \quad -3]$$

$$u \left([0 \quad 1 \quad 0]' \right) = 0 \quad w = G \Sigma G' = [1 \quad 0 \quad -3] \begin{bmatrix} 15 & 0 & 3 \\ 0 & 2 & 0 \\ 3 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ -3 \end{bmatrix} = 6$$

$$\text{By delta-method:} \quad \frac{1}{\sqrt{n}} \sum_{i=1}^n (Z_i - \bar{Z})^3 = \sqrt{n} \left(u \left(\begin{bmatrix} \frac{1}{n} \sum_{i=1}^n Z_i^3 \\ \frac{1}{n} \sum_{i=1}^n Z_i^2 \\ \frac{1}{n} \sum_{i=1}^n Z_i \end{bmatrix} \right) - 0 \right) \xrightarrow{d} N(0, 6)$$

$$\text{Summing up, } \frac{1}{\sqrt{n}} \sum_{i=1}^n (Z_i - \bar{Z})^3 \xrightarrow{d} N(0, 6), \quad \text{plim}_{n \rightarrow \infty} \left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2 \right)^{3/2} = 1$$

$$\text{By continuity } \text{plim}_{n \rightarrow \infty} \frac{1}{\left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2 \right)^{3/2}} = 1.$$

$$\text{By Slutsky, } \sqrt{n} Y_n = \frac{\frac{1}{\sqrt{n}} \sum_{i=1}^n (Z_i - \bar{Z})^3}{\left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2 \right)^{3/2}} \xrightarrow{d} N(0, 6).$$