

1 Part I Q1

The MGF of Y_n is given by $M(t) = (1 - \frac{\pi}{n} + \frac{\pi}{n}e^t)^n = (1 + (e^t - 1)\frac{\pi}{n})^n$. As $n \rightarrow \infty$, the limit of the MGF is

$$\exp(\pi(e^t - 1)),$$

which is the MGF of Poisson with mean π . Note that the MGF is unique. ■

2 Part I Q2

(a)

It is clear that $\bar{Z} = \frac{\bar{X} - \mu}{\sigma}$. So we have $X_i - \bar{X} = \sigma(Z_i - \bar{Z})$. Plugging this into Y_n gives

$$\begin{aligned} Y_n &= \frac{\sigma^4 \frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^4}{\sigma^4 \left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2\right)^2} - 3 \\ &= \frac{\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^4 - 3 \left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2\right)^2}{\left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2\right)^2}. \blacksquare \end{aligned}$$

(b)

By the hint, $\left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2\right)^2 = \left(\frac{1}{n} \sum_{i=1}^n Z_i^2 - \left(\frac{1}{n} \sum_{i=1}^n Z_i\right)^2\right)^2$. Since $Z_i \stackrel{iid}{\sim} N(0, 1)$, $Z_i^2 \stackrel{iid}{\sim} \chi(1)$. Note that $E(Z_i) = 0$, $E(Z_i^2) = 1$ and $Var(Z_i) = 1 < \infty$, $Var(Z_i^2) = 2 < \infty$. By the WLLN, $\frac{1}{n} \sum_{i=1}^n Z_i^2 = 1 + o_p(1)$ and $\frac{1}{n} \sum_{i=1}^n Z_i = 0 + o_p(1)$. By Slutsky Theorem,

$$\begin{aligned} \left(\frac{1}{n} \sum_{i=1}^n Z_i^2 - \left(\frac{1}{n} \sum_{i=1}^n Z_i\right)^2\right)^2 &= (1 + o_p(1) - (0^2 + o_p(1)))^2 \\ &= (1 + o_p(1))^2 \\ &= 1^2 + o_p(1). \blacksquare \end{aligned}$$

(c)

Define $W_i = \begin{bmatrix} Z_i^4 \\ Z_i^3 \\ Z_i^2 \\ Z_i \end{bmatrix}$. W_i is i.i.d. By the hint, $E[W_i] = \begin{bmatrix} 3 \\ 0 \\ 1 \\ 0 \end{bmatrix}$. Also,

$$\begin{aligned} \text{Var}(W_i) &= E[W_i W_i'] - E[W_i]E[W_i'] \\ &= E \begin{bmatrix} Z_i^8 & Z_i^7 & Z_i^6 & Z_i^5 \\ Z_i^7 & Z_i^6 & Z_i^5 & Z_i^4 \\ Z_i^6 & Z_i^5 & Z_i^4 & Z_i^3 \\ Z_i^5 & Z_i^4 & Z_i^3 & Z_i^2 \end{bmatrix} - \begin{bmatrix} 9 & 0 & 3 & 0 \\ 0 & 0 & 0 & 0 \\ 3 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \\ &= \begin{bmatrix} 105 & 0 & 15 & 0 \\ 0 & 15 & 0 & 3 \\ 15 & 0 & 3 & 0 \\ 0 & 3 & 0 & 1 \end{bmatrix} - \begin{bmatrix} 9 & 0 & 3 & 0 \\ 0 & 0 & 0 & 0 \\ 3 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \\ &= \begin{bmatrix} 96 & 0 & 12 & 0 \\ 0 & 15 & 0 & 3 \\ 12 & 0 & 2 & 0 \\ 0 & 3 & 0 & 1 \end{bmatrix}. \end{aligned}$$

We want to know the limiting distribution of $\sqrt{n}(\frac{1}{n} \sum W_i - E[W_i])$. By CLT,

$$\sqrt{n}(\frac{1}{n} \sum W_i - E[W_i]) \xrightarrow{d} N(0, \text{Var}(W_i)) = N \left(0, \begin{bmatrix} 96 & 0 & 12 & 0 \\ 0 & 15 & 0 & 3 \\ 12 & 0 & 2 & 0 \\ 0 & 3 & 0 & 1 \end{bmatrix} \right). \blacksquare$$

(d)

CAUTION: we want to know the limiting distribution of $\sqrt{n} \left[\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^4 - 3 \left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2 \right)^2 \right]$, not the one of $\frac{1}{\sqrt{n}} \left[\sum_{i=1}^n (Z_i - \bar{Z})^4 - 3 \left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2 \right)^2 \right]$ stated in the question.

Define $\bar{W} = \frac{1}{n} \sum_i W_i = \begin{bmatrix} \frac{1}{n} \sum_i Z_i^4 \\ \frac{1}{n} \sum_i Z_i^3 \\ \frac{1}{n} \sum_i Z_i^2 \\ \frac{1}{n} \sum_i Z_i \end{bmatrix} \equiv \begin{bmatrix} A_n \\ B_n \\ C_n \\ D_n \end{bmatrix}$. By the hint,

$$\begin{aligned} \frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^4 - 3 \left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2 \right)^2 &= A_n - 4D_n B_n + 12D_n^2 C_n - 3C_n^2 - 6D_n^4 \\ &\equiv u(\bar{W}). \end{aligned}$$

Note that $u(E[W_i]) = 3 - 3 \cdot 1^2 = 0$. We have

$$\sqrt{n} \left[\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^4 - 3 \left(\frac{1}{n} \sum_{i=1}^n (Z_i - \bar{Z})^2 \right)^2 \right] = \sqrt{n} [u(\bar{W}) - u(E[W_i])].$$

The first derivative of u evaluated at $\bar{W} = E[W_i]$ is

$$\begin{aligned} \left. \frac{\partial u}{\partial \bar{W}'} \right|_{\bar{W}=E[W_i]} &= [1 \quad -4D_n \quad 12D_n^2 - 6C_n \quad -4B_n + 24D_n C_n - 24D_n^3] \Big|_{\bar{W}=E[W_i]} \\ &= [1 \quad 0 \quad -6 \quad 0] \equiv G. \end{aligned}$$

Recall $\sqrt{n}(\bar{W} - E[W_i]) \xrightarrow{d} N(0, \text{Var}(W_i))$. By the Delta method, we have

$$\sqrt{n}(u(\bar{W}) - u(E[W_i])) \xrightarrow{d} N(0, G\text{Var}(W_i)G').$$

So we have

$$\begin{aligned} \omega &= G\text{Var}(W_i)G' \\ &= \begin{bmatrix} 1 & 0 & -6 & 0 \end{bmatrix} \begin{bmatrix} 96 & 0 & 12 & 0 \\ 0 & 15 & 0 & 3 \\ 12 & 0 & 2 & 0 \\ 0 & 3 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & -6 & 0 \end{bmatrix}' \\ &= 24. \blacksquare \end{aligned}$$

(e)

$\sqrt{n}Y_n = \frac{\sqrt{n}\left(\frac{1}{n}\sum_{i=1}^n(Z_i - \bar{Z})^4 - 3\left(\frac{1}{n}\sum_{i=1}^n(Z_i - \bar{Z})^2\right)^2\right)}{\left(\frac{1}{n}\sum_{i=1}^n(Z_i - \bar{Z})^2\right)^2}$. Since $\left(\frac{1}{n}\sum_{i=1}^n(Z_i - \bar{Z})^2\right)^2 = 1 + o_p(1)$, Slutsky Theorem gives $\frac{1}{\left(\frac{1}{n}\sum_{i=1}^n(Z_i - \bar{Z})^2\right)^2} = 1^{-1} + o_p(1)$. Also, note that $\sqrt{n}\left(\frac{1}{n}\sum_{i=1}^n(Z_i - \bar{Z})^4 - 3\left(\frac{1}{n}\sum_{i=1}^n(Z_i - \bar{Z})^2\right)^2\right) \xrightarrow{d} N(0, 24)$. By Slutsky, we have

$$\sqrt{n}Y_n \xrightarrow{d} N(0, 24). \blacksquare$$