

1. A. Using empirical distribution function notation, $NIH_N = mIF_m + nIG_n$, so

$$\begin{aligned}
 mnU_{m,n} &= \int mIF_m d nIG_n = \int N IH_N d nIG_n - \int nIG_n d nIG_n \\
 &= \sum_{j=1}^n NIH_N(Y_j) - \sum_{j=1}^n nIG_n(Y_j) \\
 &= \sum_{j=1}^n R_{m+j} - \sum_{j=1}^n j \\
 &= \sum_{j=1}^n R_{m+j} - n(n+1)/2.
 \end{aligned}$$

- B. The expectation is easy:

$$E(U_{m,n}) = \frac{1}{mn} \sum_{j=1}^m \sum_{i=1}^n P(X_i \leq Y_j) = P(X_1 \leq Y_1) = \int F dG.$$

For the variance, we first calculate

$$\begin{aligned}
 E[mnU_{m,n}]^2 &= \sum_{i=1}^m \sum_{j=1}^n \sum_{k=1}^m \sum_{l=1}^n E1_{[X_i \leq Y_j, X_k \leq Y_l]} \\
 &= \sum_{i=1}^m \sum_{j=1}^n E1_{[X_i \leq Y_j]} + \sum_{i \neq k} \sum_{j=1}^n P(X_i \leq Y_j, X_k \leq Y_j) \\
 &\quad + \sum_{i=1}^m \sum_{j \neq l} P(X_i \leq Y_j, X_i \leq Y_l) + \sum_{i \neq k} \sum_{j \neq l} P(X_i \leq Y_j, X_k \leq Y_l) \\
 &= mn P(X_1 \leq Y_1) + m(m-1)n P(X_1 \leq Y_1, X_2 \leq Y_1) \\
 &\quad + mn(n-1) P(X_1 \leq Y_1, X_1 \leq Y_2) \\
 &\quad + m(m-1)n(n-1) P(X_1 \leq Y_1, X_2 \leq Y_2)
 \end{aligned}$$

where

$$\begin{aligned}
 P(X_1 \leq Y_1) &= \int F dG, \\
 P(X_1 \leq Y_1, X_2 \leq Y_1) &= EP(X_1 \leq Y_1, X_2 \leq Y_1 | Y_1) = \int F^2(x) dG(x), \\
 P(X_1 \leq Y_1, X_1 \leq Y_2) &= EP(X_1 \leq Y_1, X_1 \leq Y_2 | X_1) = \int (1 - G(x-))^2 dF(x),
 \end{aligned}$$

and

$$P(X_1 \leq Y_1, X_2 \leq Y_2) = P(X_1 \leq Y_1)^2 = \left(\int F dG \right)^2 .$$

It follows by algebra that

$$\begin{aligned} \text{Var}(mnU_{m,n}) &= E(mnU_{m,n})^2 - \{E(mnU_{m,n})\}^2 \\ &= mn \int F dG + m(m-1)n \int F^2 dG \\ &\quad + mn(n-1) \int (1-G(x-))^2 dF(x) \\ &\quad + m(m-1)n(n-1) \left\{ \int F dG \right\}^2 - (mn \int F dG)^2 \\ &= m(m-1)n \left\{ \int F^2 dG - \left(\int F dG \right)^2 \right\} \\ &\quad + mn(n-1) \left\{ \int (1-G(x-))^2 dF(x) - \left(\int F dG \right)^2 \right\} \\ &\quad - mn \int F dG \left(1 - \int F dG \right) . \end{aligned}$$

By noting that

$$\begin{aligned} \int F dG &= P(X \leq Y) = 1 - P(X > Y) \\ &= 1 - \int G(x-) dF(x) = \int (1-G(x-)) dF(x) , \end{aligned}$$

this yields the claimed variance formula (to within a left limit):

$$\begin{aligned} \text{Var}(\sqrt{mn}U_{m,n}) &= (m-1) \text{Var}(F(Y)) + (n-1) \text{Var}(1-G(X-)) \\ &\quad + \int F dG \left(1 - \int F dG \right) . \end{aligned}$$

C. When $F = G$ continuous we find that

$$E(U_{mn}) = \int F dF = 1/2 ,$$

and, since now $\text{Var}[F(Y)] = \text{Var}[G(X)] = 1/12$,

$$\begin{aligned} \text{Var}(\sqrt{mn}U_{m,n}) &= (m-1) \frac{1}{12} + (n-1) \frac{1}{12} + \frac{1}{4} \\ &= (N-2) \frac{1}{12} + \frac{1}{4} = (N+1) \frac{1}{12} . \end{aligned}$$

Hence from part A it follows that

$$E\left(\sum_{j=1}^n Q_j\right) = n(n+1)/2 + mnE(U_{m,n}) = n(N+1)/2$$

and

$$\begin{aligned} \text{Var}\left(\sum_{j=1}^n Q_j\right) &= mn \text{Var}(\sqrt{mn}U_{m,n}) \\ &= mn(N+1) \frac{1}{12}, \end{aligned}$$

both of which agree with the finite sampling calculations of problem 4 of problem set 4.

2. A. If $\mathbf{F}_{3+\delta, M}$ is the collection of all distribution functions with $E_F|X|^{3+\delta} \leq M$ for some $\delta > 0$ and $M < \infty$, then $|X|^k$ is uniformly integrable over \mathbf{F} for $k = 0, 1, 2, 3$: for example,

$$\begin{aligned} E_F|X|^3 1_{\{|X|>K\}} &\leq E_F\{|X|^3 \frac{|X|^\delta}{K^\delta} 1_{\{|X|>K\}}\} \\ &\leq \frac{E_F\{|X|^{3+\delta}\}}{K^\delta} \leq \frac{M}{K^\delta} \end{aligned}$$

and hence

$$\lim_{K \rightarrow \infty} \sup_{F \in \mathbf{F}} E_F|X|^3 1_{\{|X|>K\}} = 0,$$

and this implies that if we have a sequence of distribution functions $\{F_n\} \subset \mathbf{F}_{3+\delta, M}$ with $F_n \rightarrow_d F$, then

$$E_{F_n} X^k \rightarrow E_F X^k, \quad k = 1, 2, 3.$$

Hence it also holds for such a sequence $\{F_n\}$ that

$$\gamma_1(F_n) \rightarrow \gamma_1(F).$$

For a complete arbitrary sequence $\{F_n\}$ with $F_n \rightarrow_d F$ there are no such guarantees, and in fact it is not hard to construct a sequence $\{F_n\}$ with

$$0 = \liminf_n \gamma_1(F_n) < \limsup_n \gamma_1(F_n) = \infty.$$

[Take $F_n = \Phi$, the standard normal df for n even, $F_n = (1 - 1/n)\Phi + (1/n)\delta_{n^{1/2}}$ for n odd.]

Similarly, if $\mathbf{F}_{4+\delta, M}$ is the collection of all distribution functions with $E_F|X|^{4+\delta} \leq M$ for some $\delta > 0$ and $M < \infty$, then $|X|^k$ is uniformly integrable over $\mathbf{F}_{4+\delta, M}$ for $k = 0, 1, 2, 3, 4$, and this implies that if we have a sequence of distribution functions $\{F_n\} \subset \mathbf{F}_{4+\delta, M}$ with $F_n \rightarrow_d F$, then

$$E_{F_n} X^k \rightarrow E_F X^k, \quad k = 1, 2, 3, 4.$$

Hence it also holds for such a sequence $\{F_n\}$ that

$$\gamma_2(F_n) \rightarrow \gamma_2(F).$$

B. One way to proceed is to write the standard coefficient of skewness, $\gamma_1(F)$, in terms of F^{-1} :

$$\frac{E_F(X - \mu_F)^3}{\{Var_F(X)\}^{3/2}} = \frac{\int_0^1 [F^{-1}(t) - \int_0^1 F^{-1}(s) ds]^3 dt}{\left\{ \int_0^1 [F^{-1}(t) - \int_0^1 F^{-1}(s) ds]^2 dt \right\}^{3/2}} .$$

This suggests defining a "trimmed skewness coefficient" γ_{1t} by

$$\gamma_{1t}(F) = \frac{\frac{1}{1-2\alpha} \int_{\alpha}^{1-\alpha} [F^{-1}(t) - (1-2\alpha)^{-1} \int_{\alpha}^{1-\alpha} F^{-1}(s) ds]^3 dt}{\left\{ \frac{1}{1-2\alpha} \int_{\alpha}^{1-\alpha} [F^{-1}(t) - (1-2\alpha)^{-1} \int_{\alpha}^{1-\alpha} F^{-1}(s) ds]^2 dt \right\}^{3/2}}$$

Then, by the same argument we used in class to prove continuity of the trimmed mean, $\gamma_{1t}(F)$ is weakly continuous: if $F_n \rightarrow_d F$, then $\gamma_{1t}(F_n) \rightarrow \gamma_{1t}(F)$. Does $\gamma_{1t}(F)$ measure "skewness" of F ? If F is symmetric about some point μ_s , then the trimmed mean $(1-2\alpha)^{-1} \int_{\alpha}^{1-\alpha} F^{-1}(t) dt = \mu_s$ and it is easy to see that $\gamma_{1t}(F) = 0$. Moreover, if F is asymmetric with $|F^{-1}(1/2-t) - \mu_s| \neq |F^{-1}(1/2+t) - \mu_s|$ for some t with $\alpha \leq 1/2-t < 1/2+t \leq 1-\alpha$, then $\gamma_{1t}(F) \neq 0$. But if the asymmetry of F occurs outside $[F^{-1}(\alpha), F^{-1}(1-\alpha)]$ then we can still have F asymmetric and $\gamma_{1t}(F) = 0$. Thus while γ_{1t} is not an ideal measure of skewness, it is weakly continuous: we have traded ease of interpretability for continuity.