A NOTE ON THE EDGEWORTH EXPANSION FOR THE KENDALL RANK CORRELATION COEFFICIENT

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In this note it is shown how to some extent the Edgeworth expansion for the distribution function of Kendall’s \( \tau \) can be established by using a well-known general result on such expansions.

Let \( X_i, Y_1, \ldots, X_N, Y_N \) be independent random variables (rv’s), the \( X_i \) with a continuous distribution function (df) \( F \), the \( Y_i \) with a continuous df \( G \). Let \( R_i \) and \( S_i \) be the ranks of \( X_i \) and \( Y_i \), respectively, then Kendall’s rank correlation coefficient is defined as

\[
\tau = \frac{1}{N(N-1)} \sum_{i=1}^{N} \sum_{j=1}^{N} \text{sign} (R_i - R_j) \text{sign} (S_i - S_j).
\]

Using a direct approach, Prašková–Vizková (1976) establishes the Edgeworth expansion (for a definition see, e.g., Feller (1971), page 542) for the df of \( \tau \). In the present note we shall point out how to some extent this result can be obtained more easily by applying the following standard theorem due to Feller (1971) (see page 548).

**Theorem.** Let \( Z_1, \ldots, Z_N \) be independent rv’s with zero mean and let \( T_N = \sum_{j=1}^{N} Z_j \). Suppose that for some integer \( r \geq 3 \) there exist positive constants \( c \) and \( C \) such that for \( v = 1, \ldots, r + 1 \) and \( j = 1, \ldots, N \),

\[
c < E|Z_j|^r < C.
\]

Moreover, assume that the characteristic functions (ch.f.’s) \( \rho_j \) of \( Z_j \), \( j = 1, \ldots, N \), satisfy

\[
|\prod_{j=1}^{N} \rho_j(t)| = o(N^{-r-1}),
\]

uniformly in \( |t| > \delta \) for all \( \delta > 0 \). Then \( \sup_x |F_N(x) - G_{x,N}(x)| = o(N^{-r/2+1}) \), where \( F_N \) is the df of \( T_N/\sigma(T_N) \) and \( G_{x,N} \) is its Edgeworth expansion to \( O(N^{-r/2+1}) \).

**Remark.** As in Feller (1971), the theorem is formulated here for a single sequence of rv’s \( Z_j \), \( j = 1, \ldots, N \). However, from Feller’s proof it is clear that the theorem also holds for a triangular array \( Z_{j,N} \), \( j = 1, \ldots, N \), \( N = 1, 2, \ldots, \), provided that conditions (2) and (3) hold uniformly for such \( Z_{j,N} \).

To apply the theorem to \( \tau \) in (1), we note that \( \tau \) has the same df as \( 4 T_N/(N-1) \), where \( T_N = \sum_{j=1}^{N} Z_{j,N} \), in which the \( Z_{j,N} \) are independent rv’s with

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\( P(Z_{jn} = k/N) = 1/(j + 1) \), \( k = -j/2, -j/2 + 1, \ldots, j/2 - 1, j/2 \) and \( j = 1, \ldots, N - 1 \) (this follows immediately from Hájek & Šidák (1967), page 115 and Hájek (1955)). The problem is, however, that neither (2) nor (3) is satisfied for these \( Z_{jn} \). We shall demonstrate how these obstacles can be removed. As concerns (2), this is quite simple: just write \( T_N = \sum_{j=1}^{[N/2]} V_{jn} \), where \( V_{jn} = Z_{jn} + Z_{j(N-1-j)N} \), \( j = 1, \ldots, [(N - 1)/2] \), and for \( N \) even, \( V_{(N/2)N} = Z_{(N/2)N} \). For these \( V_{jn} \) condition (2) holds for all \( r \).

The real problem lies in condition (3). Let \( \rho_{jn} \) and \( \rho_{jn} \) be the ch.f. of \( T_N \) and \( Z_{jn} \), respectively, then \( \rho_{jn}(t) = \prod_{j=1}^{N-1} \rho_{jn}(t) = \prod_{j=1}^{N} (\sin jt/(2N))/((\sin t/(2N))) \) (see, e.g., Prašková–Vízková (1976), page 599). Clearly, \(|\rho_{jn}(2k\pi N)| = 1\) for \( k = \pm 1, \pm 2, \ldots \) and hence (3) does not hold. For this reason we introduce \( \tilde{T}_N = T_N + U_N \), with \( U_N = \sum_{i=1}^{[\log N]} U_{in} \). Here the \( U_{in} \) are independent r.v.'s, also independent of the \( Z_{jn} \) and all uniformly distributed on \((-1/(2N), 1/(2N))\). The difference \( U_N = \tilde{T}_N - T_N \) is small with respect to \( T_N \) for two reasons: in the first place \( U_N \) has \([\log N] \) rather than \( N \) terms and furthermore the support of the \( U_N \) is of a smaller order of magnitude than the supports of (most of) the \( Z_{jn} \). Nevertheless, adding \( U_N \) to \( T_N \) suffices to overcome (3): \( \tilde{T}_N \) has ch.f.

\[
\tilde{\rho}(t) = \prod_{j=1}^{N-1} \frac{\sin \{jt/(2N)\}}{j \sin \{t/(2N)\}} \prod_{j=N-2\log N + 1}^{N} \frac{\sin \{jt/(2N)\}}{j \sin \{t/(2N)\}}
\]

and it follows that for \( \tilde{\rho}(t) \) condition (3) holds for all \( r \). Hence the replacement of some of the lattice r.v.'s \( Z_{jn} \) in \( T_N \) by smooth r.v.'s \( Z_{jn} + U_{in} \)—which in fact are uniformly distributed on \((-1/(2N), 1/(2N))\)—enables us to apply the theorem for arbitrary \( r \) to the resulting r.v \( \tilde{T}_N \).

Let \( F_N \) and \( \tilde{F}_N \) be the df of \( T_N/\sigma(T_N) \) and \( \tilde{T}_N/\sigma(T_N) \), respectively, and let \( G_{rN} \) and \( \tilde{G}_{rN} \) be their Edgeworth expansions. According to the above, \( \sup_{x} \left| \tilde{F}_N(x) - \tilde{G}_{rN}(x) \right| = o(N^{-r/2+1}) \) for all \( r \). It remains to find out what this means for \( \sup_{x} \left| F_N(x) - G_{rN}(x) \right| \). We note in the first place that \( |E[U_N|^k = O(N^{-r/2N} \log^N N \log^k N) \) and \( \sigma^2(T_N) = \sigma^2(T_N) + O(N^{-r/2N}) \). Then we observe that \( P(T_N/\sigma(T_N) \leq x) \leq P(\tilde{T}_N/\sigma(T_N) \leq x + \varepsilon) \) for all \( \varepsilon > 0 \). Combining this with a similar inequality in the opposite direction, we obtain that

\[
|F_N(x) - \tilde{F}_N(x\sigma(T_N)/\sigma(T_N))| \\
\leq P(x - \varepsilon \leq \tilde{T}_N/\sigma(T_N) \leq x + \varepsilon) + P(|U_N| \geq \varepsilon\sigma(T_N)).
\]

Using the results above and Chebyshev’s inequality, we find for \( r \geq 3 \) and \( k \) sufficiently large that

\[
\sup_{x} \left| F_N(x) - \tilde{F}_N(x\sigma(T_N)/\sigma(T_N)) \right| \\
\leq \sup_{x} \left| \tilde{G}_{rN}(x + \varepsilon\sigma(T_N)/\sigma(T_N)) - \tilde{G}_{rN}(x) \right| + o(N^{-r/2+1}) + \{\varepsilon\sigma(T_N)\}^{-k}E[U_N]^k
\]

\[
= O(\varepsilon + \varepsilon^{-k}N^{-\eta k/2} \log^k N) + o(N^{-r/2+1}) = O(N^{-\eta k/2} + \varepsilon) + o(N^{-r/2+1})
\]

for all \( \eta > 0 \), where the last step follows by choosing \( \varepsilon = N^{-\eta/2} \).
Next we note that sup_x |G_{r,N}(x) - \tilde{G}_{r,N}(x)| = O(N^{-\eta}) for all r and that
sup_x |G_{r,N}(x\sigma(T_N)/\sigma(\tilde{T}_N)) - \tilde{G}_{r,N}(x)| = O(|\sigma(T_N)/\sigma(\tilde{T}_N) - 1|) = O(N^{-4}) for all r.

Hence, for all r \geq 3, we have
\[
\begin{align*}
\sup_x |\tilde{F}_N(x\sigma(T_N)/\sigma(\tilde{T}_N)) - G_{r,N}(x)| \\
\leq \sup_x |\tilde{F}_N(x\sigma(T_N)/\sigma(\tilde{T}_N)) - \tilde{G}_{r,N}(x\sigma(T_N)/\sigma(\tilde{T}_N))| \\
+ \sup_x |\tilde{G}_{r,N}(x) - \tilde{G}_{r,N}(x\sigma(T_N)/\sigma(\tilde{T}_N))| + \sup_x |\tilde{G}_{r,N}(x) - G_{r,N}(x)| \\
= o(N^{-r/4+1}) + O(N^{-1}).
\end{align*}
\]

Combining the results above we finally arrive at the conclusion that
sup_x |F_N(x) - G_{r,N}(x)| = O(N^{-1+\eta}) + o(N^{-r/4+1}) for all r \geq 3 and \eta > 0. As G_{r,N} is continuous while F_N has jumps of order N^{-1}, this is, apart from \eta, the best result possible when ordinary Edgeworth expansions are used. However, it should be clear that the present result is weaker than the one obtained by Prašková-Vizková (1976), using methods of Esseen (1945). She adds terms to the G_{r,N} to account for the lattice character. With the generalized Edgeworth expansions thus obtained one can approximate F_N to every order desired by using sufficiently many terms of the expansion, whereas with the G_{r,N} the rate of convergence will never be better than O(N^{-2}).

REFERENCES


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