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A process variability control chart

Muhammad Riaz · Ronald J. M. M. Does

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Abstract In this study a Shewhart type control chart namely the V_t chart, is proposed for improved monitoring of the process variability of a quality characteristic of interest Y . The proposed control chart is based on the ratio type estimator of the variance using a single auxiliary variable X . It is assumed that (Y, X) follows a bivariate normal distribution. The design structure of the V_t chart is developed for Phase-I quality control and its comparison is made with those of the S^2 chart (a well-known Shewhart control chart) and the V_r chart (a Shewhart type control chart proposed by Riaz (Comput Stat, 2008a) used for the same purpose. It is observed that the proposed V_t chart outperforms the S^2 and V_r charts, in terms of discriminatory power, for detecting moderate to large shifts in the process variability. It is observed that the performance of the V_t chart keeps improving with an increase in $|\rho_{yx}|$, where ρ_{yx} is the correlation between Y and X .

Keywords Auxiliary information · Normality · Power curves · Simulations · S^2 charts · V_t chart · V_r chart

1 Introduction

The monitoring of any process output demands an early detection of shifts in the process parameters. The shift may be in the process variability or the process mean

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level or both. The variability of any process is controlled first, followed by controlling of the mean level. In 1920s Walter A. Shewhart introduced the idea of control charts to monitor any process for variability or mean level. The commonly used control charts for monitoring the process variability include the R chart, S chart, and S^2 chart and for the process mean level include the \bar{X} chart, median chart, trimmed mean chart, and mid-range chart.

Following the pioneering work of Walter A. Shewhart, researchers have developed a variety of control charts to monitor the parameters of any process using different classifications and different approaches. Farnum (1994) classified two basic types of control: threshold control and deviation control. Threshold control is concerned with detecting large shifts while deviation control is concerned with detecting small shifts in process parameters. The Shewhart type control charts are regarded as threshold control charts while non-Shewhart control charts (e.g., CUSUM and EWMA charts) are regarded as deviation control charts. There are different approaches which have been used to improve the efficiency of the control charts in detecting changes in the process parameters: for example, Battaglia (1993) used a regression based approach, Chun (2000) used a nonparametric control charting approach, Muttlak and Al-Sabah (2001) used a ranked sampling approach, Reynolds and Arnold (2001) used a variable sample size approach, Jones et al. (2004) used a CUSUM approach, Knoth (2005) used an EWMA approach, Chen and Huang (2005) used a synthetic control charting approach, He and Grigoryan (2006) used a double sampling approach, Muhammad and Riaz (2006) used a probability weighted moments based approach, Riaz and Saghir (2007) used a gini mean difference based approach, and Riaz (2008a,b) used an auxiliary information based approach.

The idea of using information on some auxiliary variable(s) along with the study variable(s) has been widely used in different areas of statistical analysis for the sake of gain in efficiency. The information on the relationship between the study and auxiliary variables helps improving the precision with which the parameters are estimated. According to Singh and Mangat (1996) “the prior information on an auxiliary variable can be used to enhance the precision of an estimator”. There is a wide variety of literature available on using such auxiliary information to achieve higher efficiency. To refer but a few of these: Olkin (1958), Rao and Mudholkar (1967), Adhvaryu (1975), Isaki (1983), Naik and Gupta (1991), Magnus (2002), and Singh et al. (2004).

In the quality control literature, the idea of exploiting correlation of the quality characteristic(s) of interest with some other associated quality characteristic(s) of interest had been used by different researches. The cause-selecting and regression adjusted charts are very commonly used methods of capitalizing on the correlation between the study characteristic(s) and the auxiliary characteristic(s) for the sake of improved process monitoring, for example, see Mandel (1969), Zhang (1984, 1985), Hawkins (1991, 1993), Wade and Woodall (1993), and Shu et al. (2005).

Riaz (2008a,b) considered the information of an auxiliary characteristic X for improved monitoring of the variance and the mean respectively of a quality characteristic of interest Y . Riaz (2008a) proposed a process variance chart (namely the V_r chart) and claimed its superiority over the well known S^2 chart whereas Riaz (2008b) proposed a process mean chart (namely the M_r chart) and claimed its superiority over the well known \bar{X} chart, the cause-selecting and regression adjusted control charts.

The V_r and M_r charts exploit the correlation of the auxiliary characteristic(s) with the study characteristic on the regression pattern for improved monitoring of the process variability and location parameters.

In this study, the information about an auxiliary characteristic X is introduced for improved monitoring of process variability of a quality characteristic of interest Y , following Riaz (2008a). Assuming bivariate normality of (Y, X) a new Shewhart type process variability control chart namely, the V_t chart (a threshold control chart) is proposed which is based on a ratio type estimator of variance. The focus of this proposal would be Phase-I quality control. The ratio type estimator for variance of Y using a single auxiliary variable X , is defined for a bivariate random sample $(y_1, x_1), (y_2, x_2), \dots, (y_n, x_n)$ of size n as (see Garcia and Cebrian 1996):

$$V_t = s_y^2 \left(\sigma_x^2 / s_x^2 \right)^{\rho_{yx}^2}. \quad (1)$$

where s_y^2 is the sample variance of Y , s_x^2 is the sample variance of X , σ_x^2 is the population variance of X (assumed to be known) and ρ_{yx} is the correlation (linear relationship) between Y and X ("In many practical situations we have information about the correlations between X and Y " see Garcia and Cebrian 1996).

The possibilities regarding the nature of an auxiliary characteristic X may include: (i) X is also a property that would be monitored (see Alt 1985), for example, Consider a polymer that has specifications on the density Y and the tensile strength X ; (ii) X is a crude but simple to obtain measurement on the process (see Singh and Mangat 1996), for example, Consider a bottling operation where the important property is the net weight Y , but it is easy to record the filling speed X ; (iii) X is an early measurement in the process (see Shu et al. 2005), for example, Consider a polishing process where X could be the time required to remove the large burrs on the product (an early step) and Y is the finished thickness.

Hawkins (1993) concludes in his paper that (i) Multivariate control charts capitalize on the correlation between different correlated characteristics at the cost of losing simplicity, (ii) Individual univariate control charts provide better visual picture and clues for process improvement at the cost of missing out the possibility of capitalizing on the correlation between different correlated characteristics, (iii) Quadratic form based control charts exploit the correlation between different correlated quality characteristics at the cost of losing the benefit of interpretability and performance loss. He proposed an alternative approach which overcomes the aforementioned three problems keeping their respective benefits. He argued about his approach as: "In this approach, information on the dynamics of the process itself is used to uncover the directions of causality giving rise to the correlations in the data. This diagnosis in turn leads to prescriptions for the regression adjustment of different variables". He classified the processes into two main categories (i) Cascade: "where each variable that changes distribution affects those below but not above it", (ii) Without Cascade: "where each variable may undergo a distributional change without affecting any others". The focus of this article would be the processes without cascade property.

In the following sections (i) the design structure of the V_t chart is developed for improved monitoring of a process variability following the pioneering work of

Shewhart (1931), Pearson (1932), Pappanastos and Adams (1996), Ramalhoto and Morais (1999), Gonzalez and Viles (2000, 2001) and Riaz (2008a,b), (ii) the power curves are constructed as a performance measure of the V_t chart following Scheffe (1949), Duncan (1951), Nelson (1985) and Riaz (2008a,b), (iii) the performance of the V_t chart is compared with those of the conventional S^2 chart (a well-known Shewhart control chart) and the V_r chart (a Shewhart type control chart proposed by Riaz 2008a,b) used for the same purpose following Tuprah and Ncube (1987), Acosta-Mejia et al. (1999), Ding et al. (2005) and Riaz (2008a).

2 The proposed chart

Assuming bivariate normality of (Y, X) a relationship between σ_y^2 (the unknown process variability of the quality characteristic of interest Y which is to be monitored) and V_t (the ratio type estimator of σ_y^2 defined in (1)) is required to develop the structure of the proposed V_t chart. Let $(y_1, x_1), (y_2, x_2), \dots, (y_n, x_n)$ be a bivariate random sample of size n from the bivariate normal distribution, and let A be a random variable that defines a relationship between σ_y^2 and V_t as:

$$A = \frac{V_t}{\sigma_y^2}, \quad (2)$$

which helps in determining the parameters (i.e., the centerline, lower control limit and upper control limit) of the proposed V_t chart.

Now if the distributional behavior of A is available then the sample statistic V_t can easily be used for testing hypotheses about shifts in σ_y^2 . When (Y, X) follows a bivariate normal distribution, the distributional behavior of A depends only on ρ_{yx} and n (see Garcia and Cebrian 1996). The distributional behavior of A , in terms of its mean, standard error and quantile points, is required for developing the design structure of the V_t chart, and is explored in the following paragraphs when (Y, X) follows a bivariate normal distribution.

First, for the mean, applying expectations to (2) gives:

$$E(A) = E(V_t/\sigma_y^2) = E(V_t)/\sigma_y^2. \quad (3)$$

Here $E(V_t)$ can safely be replaced by its estimate \bar{V}_t (the mean of the sample V_t 's) using an appropriate number of random samples, as discussed in Hillier (1969) and Yang and Hiller (1970), from the process under study when the process is in the state of statistical control as written in Shewhart (1939, p. 26) just like \bar{R} replaces $E(R)$ for the R chart. Thus from (3) an estimate of σ_y^2 , after rearranging the terms, is given as:

$$\hat{\sigma}_y^2 = \bar{V}_t/E(A). \quad (4)$$

Let $E(A) = r_0$, on the same pattern as v_0 for V_r chart in Riaz (2008a). As V_t is unbiased estimator for σ_y^2 (see Garcia and Cebrian 1996) we have $r_0 = 1$. Thus (4)

results in the following:

$$\hat{\sigma}_y^2 = \bar{V}_t. \tag{5}$$

Also from (3) we have:

$$E(V_t) = \sigma_y^2. \tag{6}$$

Replacement of the estimate of σ_y^2 (given in (5)) in (6) gives:

$$E(V_t) \simeq \bar{V}_t. \tag{7}$$

Thus, the V_t chart works without constants, such as d_2 for the R chart and c_4 for the S chart, for an unbiased estimation of process variability using (5).

Secondly, for the standard error, let the standard deviation of A (i.e., σ_A) be

$$\sigma_A = r_2. \tag{8}$$

It is not easy to get analytical results for r_2 because $E(V_t^2)$ is difficult to obtain analytically. So simulation results are obtained for r_2 in this paper (In practice, simulation methods are often used to evaluate the expectation of a statistic, see Ross 1990). The coefficient r_2 entirely depends on ρ_{yx} and n for the case of bivariate normal distribution. Using 10,000 random samples generated from the standard bivariate normal distribution without loss of generality, the results of r_2 have been obtained, for different combinations of ρ_{yx} and n , 1,000 times for each combination. Based on these results the mean values of r_2 , along with their respective standard errors, are provided in Table 1 for $n = 5, 6, \dots, 15, 20, 25, 30, 50, 100$ at some representative values of ρ_{yx} , in the Appendix. Similar results can easily be obtained for any combination of ρ_{yx} and n .

Also taking the variance of A followed by simplification gives the expression for σ_A as:

$$\sigma_A = \sigma_{V_t} / \sigma_y^2, \tag{9}$$

where σ_{V_t} represents the standard deviation of the distribution of the sample statistic V_t .

Substituting (8) into (9) and rearranging yields the following result for σ_{V_t} :

$$\sigma_{V_t} = r_2 \sigma_y^2. \tag{10}$$

Substituting the estimate for σ_y^2 , given in (5), into (10), the estimate for σ_{V_t} is given as:

$$\hat{\sigma}_{V_t} = r_2 \bar{V}_t. \tag{11}$$

The expression in (11) is similar to the expression for $\hat{\sigma}_R$ of the R chart as provided in Alwan (2000, p. 394).

An approximation for σ_{V_t} , when (Y, X) follows a bivariate normal distribution, is given as (see Garcia and Cebrian 1996):

$$\sigma_{V_t} \simeq \sqrt{2\sigma_y^4(1 - \rho_{yx}^4)/n}. \quad (12)$$

Consequently,

$$r_2 \simeq \sqrt{2(1 - \rho_{yx}^4)/n}. \quad (13)$$

This approximation result (13) works very well asymptotically; however, for the case of very small values of n it does not provide a good approximation as can be seen from Table 1 in the Appendix.

Lastly, for the quantile points of the distribution of A , let A_a represents the a th quantile point of the distribution of A (i.e., the point where A completes $a\%$ area). The analytical results for A_a are difficult to obtain so simulation results are obtained for A_a . For a bivariate normal distribution of (Y, X) , the quantile points of the distribution of A entirely depend on ρ_{yx} and n . Using the same 10,000 simulated random samples, results of A_a have been obtained (such as the quantile points of $W = R/\sigma$ that determine the values of the control limits of the R chart and the power of the chart) following Pearson (1932), for different combinations of ρ_{yx} and n , 1000 times for each combination. Based on these results, the mean values of some commonly used quantile points, along with their respective standard errors, are provided for $n = 5, 6, \dots, 15, 20, 25, 30, 50, 100$ in Tables 2, 3, 4, 5, 6, 7, 8, 9, 10 and 11 at some representative values of ρ_{yx} , in the Appendix. Similar results can easily be obtained for any combination of ρ_{yx} and n . These quantile points help in determining the control limits and the power of the proposed V_t chart to detect shifts in the process variability σ_y^2 . The distributional behavior of A is not symmetrical at least for small values of n as is obvious from Tables 2, 3, 4, 5, 6, 7, 8, 9, 10 and 11 in the Appendix. Asymptotically, A is normally distributed, $N(1, 2(1 - \rho_{yx}^4)/n)$ (see Garcia and Cebrian 1996). The details regarding the asymptotic properties (mean, standard deviation and sampling distribution) of the sample statistic V_t (or A), reported and used in this paper, may be seen in Isaki (1983) and Garcia and Cebrian (1996).

It is observed that for a given value of ρ_{yx} the distributional behavior of A (at least in terms of its mean, standard error and quantile points) remains the same irrespective of the sign of ρ_{yx} . Thus, the design structure of the proposed V_t chart is function of $|\rho_{yx}|$ and n .

Now, based on the results obtained in Sect. 2, the parameters of the proposed V_t chart are discussed in the following section.

3 Parameters of the proposed chart

The central line (CL), lower control limit (LCL) and upper control limit (UCL) are the three parameters of any Shewhart type control chart. There are two approaches to express these parameters namely, the probability limits approach and the 3-sigma limits

approach. In case of an asymmetric distributional behavior of a relevant estimator the probability limits approach is preferred. If the distributional behavior of a relevant estimator is nearly symmetric then the 3-sigma limits approach is a good alternative. The parameters of the proposed V_t chart using both the approaches are expressed in the following two subsections. Following Shewhart’s recommendations, ideally 20–30 initial random samples for Phase-I are required to compute the parameters of the proposed chart.

3.1 Probability limits approach

The value \bar{V}_t corresponds to CL of the proposed V_t chart just like \bar{R} for the R chart provided in Alwan (2000, p. 347) and \bar{S} for the S chart provided in Alwan (2000, p. 362). Assuming the probability of making a Type-I error to be less than a specified value say α , then the control limits (which are actually true probability limits) for the proposed V_t chart are defined as:

$$\left. \begin{aligned} LCL &= V_{tl} \quad \text{with} \quad P_n(V_t = V_{tl}) \leq \alpha_l \\ UCL &= V_{tu} \quad \text{with} \quad P_n(V_t = V_{tu}) \geq 1 - \alpha_u \end{aligned} \right\}, \tag{14}$$

where $\alpha = \alpha_l + \alpha_u$ and P_n represents the cumulative distribution function for a given value of n .

Now, using (2) and (5) in (14) and with simplification gives the following control limits:

$$\left. \begin{aligned} LCL &= V_{tl} = A_l \bar{V}_t \quad \text{with} \quad P_n(A = A_l) \leq \alpha_l \\ UCL &= V_{tu} = A_u \bar{V}_t \quad \text{with} \quad P_n(A = A_u) \geq 1 - \alpha_u \end{aligned} \right\}, \tag{15}$$

Thus, the quantile points of the distribution of A and the average of sample V_t ’s (i.e., \bar{V}_t) allow the setting of true probability limits of the proposed V_t chart.

3.2 Three-sigma limits approach

If a normal approximation to the distribution of A is used then the parameters of the proposed V_t chart with the usual 3-sigma control limits are given as:

$$\left. \begin{aligned} UCL &= \bar{V}_t + 3\sigma_{V_t} \\ CL &= \bar{V}_t \\ LCL &= \bar{V}_t - 3\sigma_{V_t} \end{aligned} \right\}, \tag{16}$$

Using (10) in (16) and then substituting result (5) gives the following result:

$$\left. \begin{aligned} UCL &= \bar{V}_t + 3r_2\bar{V}_t \\ CL &= \bar{V}_t \\ LCL &= \bar{V}_t - 3r_2\bar{V}_t \end{aligned} \right\}, \quad (17)$$

where values of r_2 are provided in Table 1 in the Appendix.

The validity of these 3-sigma limits based parameters of the proposed V_t chart depends on how close the normal approximation is to the true distribution of A .

For small values of n , sometimes the LCL results in a negative value. A negative value for a variability measure has no realistic meaning. Therefore, in such situations the LCL is assigned the value of 0 (as is done for the range statistic in the R chart, see Alwan (2000, p. 355)).

After deciding the control structure, for a given significance level, by either the probability limit approach or the 3-sigma limit approach, the sample statistic V_t is plotted against the time order of the samples. If all of the sample V_t 's lie within the control limits, there is reasonable evidence to conclude that there is no shift in the process variability σ_y^2 and the process is stable at \bar{V}_t . Otherwise some assignable cause or causes are at work causing a shift in the process variability σ_y^2 .

To address specifically small and moderate shifts: (i) the runs rules [as discussed by Nelson (1984), Wheeler (1995), Quesenberry (1997)] may be supplemented to the basic structure of the V_t chart developed in this paper. As a result the risk of false alarms is increased, (ii) EWMA and CUSUM schemes may be developed based on the sample statistic V_t (an area for further research).

“The situations where knowledge of the properties of auxiliary population is lacking, a larger first phase sample is observed only on the auxiliary characteristic(s). The first phase sample is used to furnish good estimates for the characteristics of the auxiliary population. A sub-sample (also called second phase sample) from the initial sample is selected for observing the variable of interest. Information collected on the two samples is then used to construct estimators for the parameter under consideration” see Singh and Mangat (1996). This is known as two phase sampling or double sampling. In case of unknown properties of the auxiliary characteristic(s), the case of double sampling may also be seen in Yu and Lam (1997) and Singh et al. (2004).

Now in the following section, the performance of the developed design structure of the V_t chart is compared with those of the S^2 and V_r charts as given in Riaz (2008a).

4 Comparisons

In this section, comparisons of the means, standard errors of the random variables used in the V_t , V_r and S^2 charts and the power curves of these charts are provided. The random variable used for the V_t chart of this paper is A and its mean and standard error are r_0 and r_2 respectively. The corresponding random variable used for the V_r chart from Riaz (2008a) is D and its mean and standard error are v_0 and v_1 respectively, and the corresponding random variable used for the S^2 chart is J and its mean and standard error are u_0 and u_1 respectively (see Riaz 2008a).

Firstly, the comparison of means reveals that $r_0 = u_0 = 1$ whereas v_0 deviates from 1 at least for small values of $|\rho_{yx}|$ and n . The deviation of the means (r_0 , v_0 and u_0) from 1 reveals the extent of biasness of the estimators used in V_t , V_r and S^2 charts respectively. Thus, the V_t and S^2 charts need no constant, such as v_0 for the V_r chart, for unbiased estimation of the process variability σ_y^2 .

Secondly, the values of standard errors r_2 , v_1 and u_1 differ depending on $|\rho_{yx}|$ and n . It is observed that: (i) r_2 remains smaller than u_1 for all combinations of $|\rho_{yx}|$ and n , and the difference keeps increasing with an increase in either $|\rho_{yx}|$ or n ; (ii) r_2 remains smaller than v_1 for all combinations of $|\rho_{yx}|$ and n , and the difference keeps decreasing with an increase in either $|\rho_{yx}|$ or n ; (iii) for larger values of $|\rho_{yx}|$ or n , r_2 and v_1 become very close to each other; (iv) for small values of $|\rho_{yx}|$, v_1 is larger than u_1 for a given value of n , and v_1 becomes smaller than u_1 with an increase in $|\rho_{yx}|$ as can be seen from Riaz (2008a).

Lastly, the efficiency of the V_t chart as compared to those of the V_r and S^2 charts has been examined using power curves as a performance measure. As the focus of the proposal is on Phase-I quality control, so power curves have been used as a performance measure (in contrast to Phase-II quality control where Average Run Length (ARL) is used as a performance measure) of the control charts following Albers and Kallenberg (2006) and Riaz (2008a,b). As the distributional behaviors of A , D and J are not symmetrical, at least for smaller values of n so we used the probability limits approach for the three charts to set the control limits for a given significance level (α). Using their respective control structures, the probability limits of the V_t , V_r and S^2 charts have been obtained for different combinations of $|\rho_{yx}|$ and n with different significance levels, and the power curves for the three charts have been constructed. The power curves of the three charts, for $n = 15$ and 25 , are produced here, for examination purposes, at a low, a moderate and a high value of $|\rho_{yx}|$ in the following Figs. 1a–c and 2a–c, respectively (using $\alpha = 0.002$).

In the above figures, the curves referred to as S^2 , V_r and V_t represent the power curve of the S^2 , V_r and V_t charts respectively. A similar behavior, as shown in the above figures, is observed for other combinations of $|\rho_{yx}|$ and n . The performance of the three charts differ depending on $|\rho_{yx}|$ and n as can be seen from above figures. In general, the following points are observed:

V_t Chart versus S^2 chart: The discriminatory power of the V_t chart is higher than that of the S^2 chart for all combinations of $|\rho_{yx}|$ and n as obvious from the above figures. The gain in terms of the discriminatory power for the V_t chart, as compare to the S^2 chart, keeps increasing with an increase in either $|\rho_{yx}|$ or n as obvious from the above figures. For $|\rho_{yx}|=0$ the power curves of the two charts coincide while for all $|\rho_{yx}| > 0$, the V_t chart remains superior to the S^2 chart for all values of n .

V_t Chart versus V_r chart: The discriminatory power of the V_t chart is higher than that of the V_r chart for all combinations of $|\rho_{yx}|$ and n as obvious from the above figures. Difference in the power curves of the two charts keeps decreasing with an increase in either $|\rho_{yx}|$ or n as obvious from the above figures. For very large values of $|\rho_{yx}|$ or n , power curves of the two charts almost coincide as obvious from the above figures.

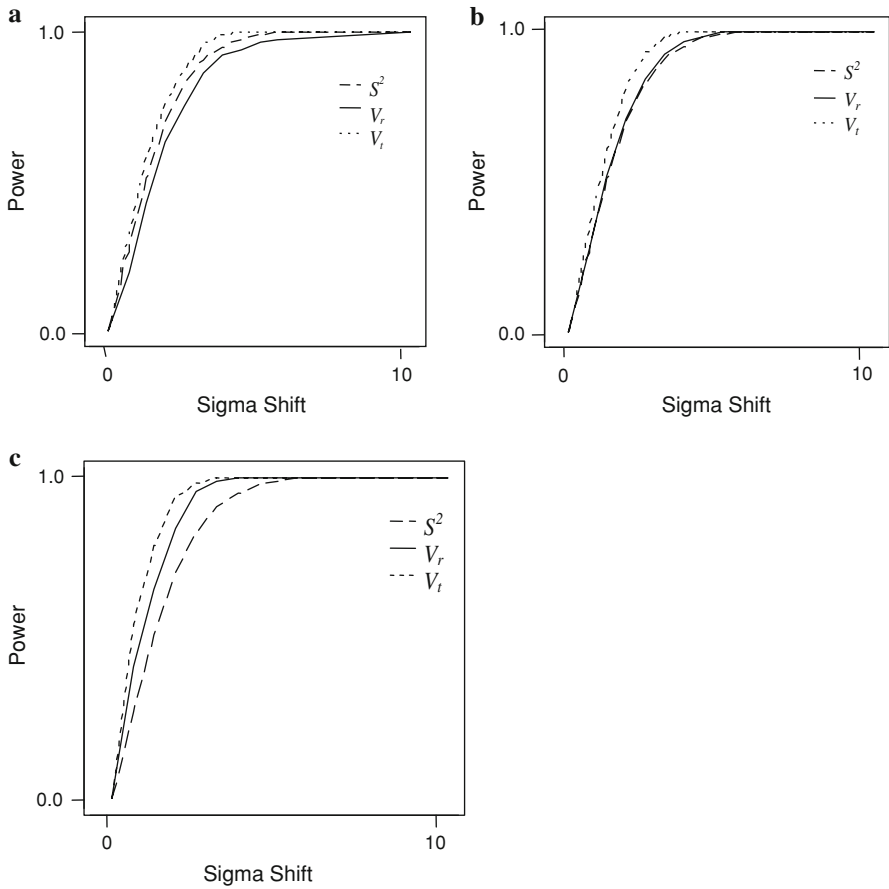


Fig. 1 **a** Power curves of the V_t V_r and S^2 charts for $n = 15$, $|\rho_{yx}| = 0.30$ and $\alpha = 0.002$. **b** Power Curves of the V_t , V_r and S^2 charts for $n = 15$, $|\rho_{yx}| = 0.70$ and $\alpha = 0.002$. **c** Power Curves of the V_t , V_r and S^2 charts for $n = 15$, $|\rho_{yx}| = 0.90$ and $\alpha = 0.002$

V_r Chart versus S^2 chart: The discriminatory power of the V_r chart is conditionally higher than that of the S^2 chart, conditioned on $|\rho_{yx}|$ and n . For each value of n there exists a value of $|\rho_{yx}|$ below which the V_r chart remains less powerful than the S^2 chart and above which the V_r chart becomes more powerful than the S^2 chart as obvious from the above figures. The smallest value of $|\rho_{yx}|$, for a sample of size n , above which V_r chart outperforms the S^2 chart for detecting shifts (especially moderate to large shifts) in process variability are given in Riaz (2008a).

Thus, the proposed chart of this paper (i.e., the V_t chart) outperforms the V_r and S^2 charts without any condition on $|\rho_{yx}|$ and n , as are required for the V_r chart to outperform the S^2 chart and are given in Riaz (2008a). This is a major advantage of the proposed V_t chart over the V_r chart, and hence the V_t chart is an improvement over the V_r chart.

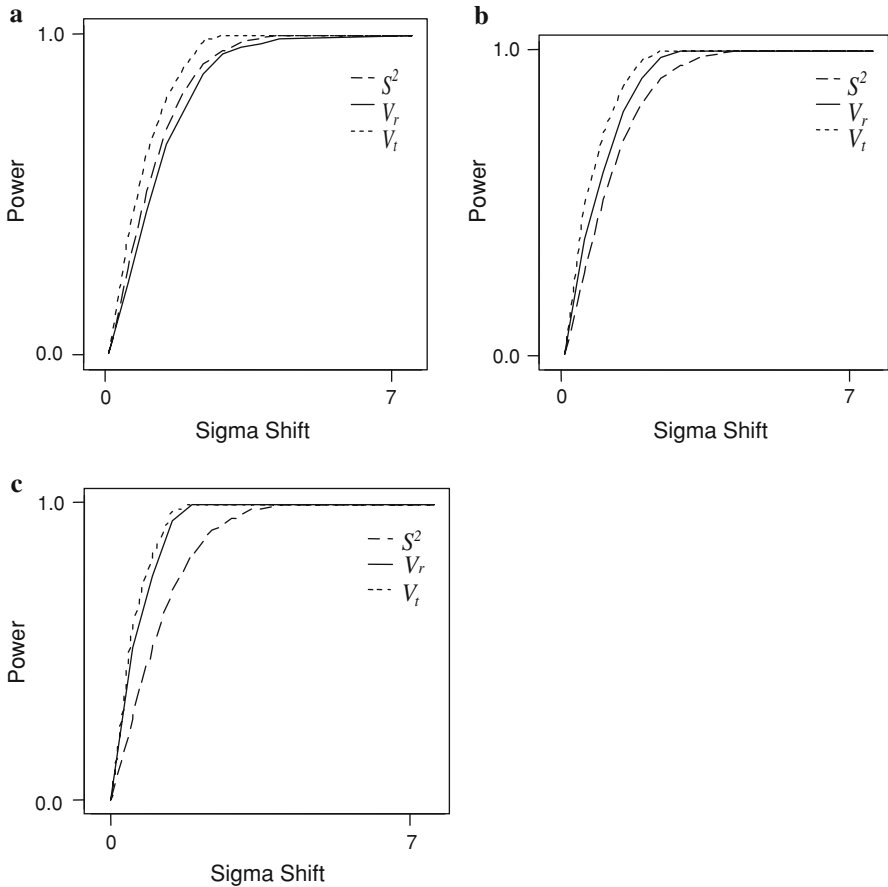


Fig. 2 a Power curves of the V_t , V_r and S^2 charts for $n = 25$, $|\rho_{yx}| = 0.30$ and $\alpha = 0.002$. b Power curves of the V_t , V_r and S^2 charts for $n = 25$, $|\rho_{yx}| = 0.70$ and $\alpha = 0.002$. c Power curves of the V_t , V_r and S^2 charts for $n = 25$, $|\rho_{yx}| = 0.90$ and $\alpha = 0.002$

5 Conclusions

The proposal of this article is a Shewhart type process variability control chart, focusing Phase-I quality control. The proposed V_t chart uses the information on a single auxiliary variable for monitoring the process variability of a quality characteristic of interest. The V_t chart outperforms the S^2 and V_r charts for detecting shifts (especially of moderate and larger magnitudes because the Shewhart control charts target such shifts) in the process variability σ_y^2 . It is observed that the performance of the V_t chart, in terms of discriminatory power, improves with an increase in either $|\rho_{yx}|$ or n . The proposed V_t chart is an improvement over the V_r chart in the sense that its design structure is free from the conditions on $|\rho_{yx}|$ and n to outperform the S^2 and V_r charts, as are required for the V_r chart to outperform the S^2 chart as given in Riaz (2008a).

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Appendix

Note: In the following tables the results (in bold) are reported up to five decimal places. The value reported below each result is the standard error (reported up to six decimal places) for the result of each cell, reported to show precision of the result of each cell.

Table 1 Control chart coefficient r_2 of the V_r chart

n	$ \rho_{yx} $									
	0.10	0.20	0.30	0.40	0.50	0.60	0.70	0.80	0.90	0.99
5	0.70753	0.70632	0.70558	0.70475	0.70582	0.70387	0.70486	0.70268	0.59524	0.24717
	0.000632	0.000906	0.000582	0.000166	0.000202	0.000470	0.000119	0.000274	0.000182	0.000135
6	0.63211	0.63188	0.63119	0.63078	0.63004	0.62924	0.61579	0.58160	0.57079	0.16863
	0.000213	0.000267	0.000299	0.000297	0.000653	0.000311	0.000320	0.000613	0.000217	0.000217
7	0.57579	0.57281	0.57127	0.57005	0.56931	0.56329	0.55442	0.50565	0.41209	0.14704
	0.000198	0.000338	0.000137	0.000872	0.000747	0.000242	0.000188	0.000112	0.000461	0.000149
8	0.53373	0.53217	0.53180	0.53004	0.52975	0.52583	0.50252	0.45140	0.36337	0.12418
	0.000543	0.000229	0.000441	0.000518	0.000524	0.000511	0.000407	0.000218	0.000398	0.000213
9	0.50006	0.49847	0.49523	0.49260	0.48988	0.48698	0.47361	0.41428	0.32915	0.11247
	0.000447	0.000558	0.000102	0.000774	0.000669	0.000663	0.000100	0.000493	0.000113	0.000138
10	0.47077	0.46809	0.46260	0.46193	0.46007	0.45767	0.43696	0.38489	0.30250	0.10539
	0.000369	0.000207	0.000235	0.000391	0.000238	0.000400	0.000326	0.000364	0.000416	0.000219
11	0.44561	0.44219	0.44067	0.43933	0.43556	0.42990	0.40589	0.36401	0.28475	0.09993
	0.000824	0.0001724	0.000696	0.000890	0.000134	0.000136	0.000812	0.000105	0.000635	0.000122
12	0.42465	0.42221	0.42182	0.42013	0.41970	0.40890	0.39020	0.34775	0.27647	0.09268
	0.000442	0.000668	0.000338	0.000614	0.000216	0.000129	0.000363	0.000544	0.000367	0.000214
13	0.40697	0.40447	0.40181	0.40001	0.39766	0.39090	0.37008	0.33552	0.25719	0.08828
	0.000943	0.000410	0.000581	0.000223	0.000365	0.000931	0.000456	0.000603	0.000169	0.000136
14	0.39123	0.39064	0.39001	0.38943	0.38397	0.37672	0.35071	0.31160	0.24882	0.08613
	0.000708	0.000317	0.000501	0.000450	0.000528	0.000287	0.000544	0.000714	0.000447	0.000299
15	0.37674	0.37532	0.37341	0.37165	0.37068	0.35931	0.34013	0.30702	0.23564	0.07995
	0.000286	0.000156	0.000228	0.000355	0.000458	0.000670	0.000258	0.000360	0.000278	0.000748
20	0.32357	0.32296	0.32115	0.32062	0.31449	0.30869	0.29145	0.25402	0.19621	0.06797
	0.000342	0.000765	0.000476	0.000713	0.000673	0.000409	0.000508	0.000499	0.000510	0.000395
25	0.28765	0.28498	0.28265	0.28009	0.28076	0.27319	0.25485	0.22524	0.17314	0.06072
	0.000275	0.000532	0.000826	0.000407	0.000257	0.000332	0.000287	0.000783	0.000333	0.000657
30	0.26255	0.26240	0.26195	0.26058	0.25470	0.24487	0.23115	0.20562	0.15572	0.05439
	0.000667	0.000237	0.000620	0.000215	0.000401	0.000636	0.000442	0.00448	0.000661	0.000225
50	0.20200	0.20189	0.20104	0.20042	0.19643	0.19244	0.17626	0.15456	0.11973	0.04064
	0.000124	0.000114	0.000188	0.000746	0.000887	0.000198	0.000746	0.00330	0.000209	0.000541
100	0.14209	0.14201	0.14193	0.13958	0.13837	0.13241	0.12460	0.11054	0.08436	0.02841
	0.000476	0.000366	0.000332	0.000638	0.000236	0.000817	0.000098	0.000687	0.000742	0.000713

Table 2 Quantile points of the distribution of A (when $|\rho_{yx}| = 0.10$)

n	$A_{0.01}$	$A_{0.05}$	$A_{0.10}$	$A_{0.20}$	$A_{0.25}$	$A_{0.75}$	$A_{0.80}$	$A_{0.90}$	$A_{0.95}$	$A_{0.99}$
5	0.06964	0.17109	0.26051	0.41101	0.47891	1.46960	1.63983	2.06038	2.51979	3.54982
	0.000319	0.000446	0.000618	0.000572	0.000444	0.000372	0.000215	0.000362	0.000167	0.000692
6	0.08036	0.22402	0.31829	0.47094	0.53108	1.39999	1.55877	1.94269	2.30969	3.11161
	0.000500	0.000079	0.000269	0.000623	0.000964	0.000806	0.000390	0.000226	0.000706	0.000558
7	0.14023	0.27146	0.37075	0.51501	0.57821	1.38968	1.48991	1.82010	2.17197	2.90139
	0.000412	0.000493	0.000647	0.000546	0.000485	0.000278	0.000669	0.000808	0.000915	0.000427
8	0.16000	0.30222	0.40007	0.53611	0.60008	1.37969	1.48989	1.81011	2.09040	2.73949
	0.000630	0.000279	0.000558	0.000643	0.000228	0.000573	0.000947	0.000097	0.000714	0.000395
9	0.19066	0.33097	0.42082	0.56213	0.62737	1.33106	1.43952	1.71235	1.98915	2.61001
	0.000444	0.000237	0.000664	0.000883	0.000554	0.000553	0.000862	0.000495	0.00076	0.000249
10	0.21973	0.36066	0.46154	0.59850	0.65846	1.29998	1.40008	1.65987	1.91983	2.45491
	0.000168	0.000294	0.000543	0.000711	0.000684	0.000826	0.000278	0.000394	0.000946	0.000273
11	0.24986	0.38859	0.49054	0.62010	0.67687	1.29875	1.38643	1.63985	1.86193	2.40157
	0.000616	0.000070	0.000517	0.000829	0.000644	0.000276	0.000693	0.000943	0.000642	0.000276
12	0.26459	0.41068	0.50007	0.62892	0.67884	1.28991	1.37788	1.61071	1.84113	2.30124
	0.000491	0.000246	0.000614	0.000887	0.000652	0.000764	0.000282	0.000470	0.000937	0.000372
13	0.28166	0.42297	0.52000	0.64690	0.69621	1.28167	1.35078	1.58822	1.79122	2.23504
	0.000210	0.000728	0.000331	0.000615	0.000853	0.000409	0.000337	0.000067	0.000492	0.000129
14	0.30766	0.44863	0.53952	0.66003	0.71011	1.26101	1.33876	1.55141	1.74109	2.17269
	0.000507	0.000669	0.000718	0.000473	0.000658	0.000284	0.000779	0.000334	0.000456	0.000123
15	0.32102	0.44937	0.54003	0.67020	0.72502	1.24701	1.32119	1.54684	1.73894	2.13209
	0.000093	0.000224	0.000723	0.000872	0.000470	0.000197	0.000446	0.000521	0.000515	0.000763
20	0.39106	0.53061	0.60892	0.71928	0.76005	1.21943	1.28842	1.46809	1.62260	2.0148
	0.000076	0.000810	0.000619	0.000445	0.000387	0.000429	0.000553	0.000664	0.000745	0.000392
25	0.44716	0.55917	0.64209	0.74895	0.79003	1.18995	1.24868	1.40994	1.55983	1.84151
	0.000422	0.000678	0.000582	0.000378	0.000816	0.000773	0.000661	0.000246	0.000972	0.000158
30	0.48004	0.59871	0.66369	0.76144	0.80176	1.17572	1.23323	1.36992	1.48845	1.77494
	0.000264	0.000337	0.000549	0.000668	0.000114	0.000383	0.000089	0.000427	0.000664	0.000216
50	0.58048	0.68017	0.72959	0.81005	0.84004	1.15995	1.19997	1.29411	1.38136	1.56080
	0.000336	0.000565	0.000526	0.000946	0.000349	0.000227	0.000078	0.000467	0.000615	0.000092
100	0.68001	0.76036	0.81427	0.86599	0.88947	1.10461	1.14720	1.20018	1.26978	1.38194
	0.000059	0.000467	0.000658	0.000472	0.000552	0.000873	0.000645	0.000794	0.000365	0.000229

Table 3 Quantile points of the distribution of A (when $|\rho_{yx}| = 0.20$)

n	$A_{0.01}$	$A_{0.05}$	$A_{0.10}$	$A_{0.20}$	$A_{0.25}$	$A_{0.75}$	$A_{0.80}$	$A_{0.90}$	$A_{0.95}$	$A_{0.99}$
5	0.07040	0.17403	0.26684	0.41282	0.48000	1.46827	1.63811	2.05925	2.51148	3.53358
	0.000148	0.000651	0.000357	0.000623	0.000509	0.000987	0.000513	0.000098	0.000386	0.000066
6	0.08985	0.23307	0.32737	0.47381	0.53937	1.39997	1.55644	1.94059	2.30057	3.10104
	0.000781	0.000763	0.000465	0.000249	0.000666	0.000528	0.000729	0.000545	0.000298	0.000147
7	0.14245	0.27333	0.37090	0.51523	0.58018	1.38896	1.48902	1.81989	2.16686	2.89452
	0.000257	0.000364	0.000576	0.000514	0.000872	0.000927	0.000429	0.000211	0.000375	0.000111
8	0.16106	0.30866	0.40163	0.54959	0.60728	1.37771	1.48714	1.80183	2.08968	2.72343
	0.000312	0.000421	0.000752	0.000628	0.000478	0.000951	0.000713	0.000654	0.000826	0.000249
9	0.19483	0.33162	0.42851	0.56880	0.62940	1.32897	1.43826	1.70965	1.98089	2.60288
	0.000336	0.000817	0.000654	0.000682	0.000413	0.000543	0.000387	0.000856	0.000289	0.000117
10	0.22986	0.36286	0.46203	0.60320	0.66021	1.29997	1.39942	1.65794	1.91040	2.44236
	0.000267	0.000126	0.000193	0.000085	0.000546	0.000967	0.000882	0.000254	0.000683	0.000246
11	0.25131	0.39321	0.49230	0.62338	0.68116	1.29660	1.38524	1.63780	1.85966	2.37942
	0.000283	0.000369	0.000583	0.000173	0.000542	0.000357	0.000224	0.000394	0.000764	0.000284
12	0.27060	0.41524	0.50082	0.63097	0.68125	1.28850	1.37187	1.60911	1.83484	2.29881
	0.000255	0.000659	0.000512	0.000456	0.000673	0.000937	0.000448	0.000372	0.000652	0.000713
13	0.28979	0.42971	0.52014	0.64998	0.70000	1.27874	1.34982	1.58548	1.78980	2.22255
	0.000429	0.000562	0.000389	0.000447	0.000538	0.000176	0.000277	0.000498	0.000531	0.000553
14	0.31001	0.45004	0.54001	0.66108	0.71157	1.25996	1.33763	1.55076	1.73991	2.16175
	0.000611	0.000289	0.000427	0.000392	0.000557	0.000492	0.000753	0.000628	0.000549	0.000941
15	0.32611	0.45023	0.54032	0.67115	0.72603	1.24512	1.32013	1.54321	1.73620	2.12960
	0.000578	0.000655	0.000283	0.000891	0.000354	0.000672	0.000477	0.000589	0.000861	0.000456
20	0.39420	0.53250	0.61006	0.72017	0.76154	1.21794	1.28614	1.46733	1.61934	1.99962
	0.001046	0.000245	0.000823	0.000775	0.000619	0.000436	0.000367	0.000429	0.000955	0.000714
25	0.44918	0.56000	0.64857	0.75006	0.79201	1.18981	1.24708	1.40704	1.54546	1.83287
	0.000111	0.000528	0.000349	0.000664	0.000752	0.000449	0.000611	0.000346	0.000770	0.000215
30	0.48098	0.60051	0.66870	0.76978	0.80863	1.17292	1.23195	1.36898	1.48077	1.76192
	0.000643	0.000485	0.000573	0.000862	0.000362	0.000714	0.000129	0.000521	0.000878	0.000366
50	0.58150	0.68215	0.73058	0.81160	0.84101	1.15918	1.19803	1.29066	1.37987	1.55337
	0.000147	0.000524	0.000673	0.000525	0.000470	0.000775	0.000818	0.000526	0.000287	0.000416
100	0.68035	0.76112	0.81991	0.86814	0.89008	1.10100	1.14122	1.19998	1.26848	1.37970
	0.000741	0.000573	0.000419	0.000546	0.000854	0.000919	0.000744	0.000122	0.000327	0.000618

Table 4 Quantile points of the distribution of A (when $|\rho_{yx}|_x = 0.30$)

n	$A_{0.01}$	$A_{0.05}$	$A_{0.10}$	$A_{0.20}$	$A_{0.25}$	$A_{0.75}$	$A_{0.80}$	$A_{0.90}$	$A_{0.95}$	$A_{0.99}$
5	0.07699	0.18501	0.27293	0.41920	0.48937	1.46736	1.63720	2.05699	2.51004	3.52494
	0.000497	0.000668	0.000357	0.000510	0.000088	0.000987	0.000492	0.000226	0.000693	0.000645
6	0.09001	0.23584	0.33147	0.47806	0.54618	1.39991	1.55541	1.93907	2.29954	3.09185
	0.000321	0.000645	0.000763	0.000918	0.000482	0.000673	0.000557	0.000279	0.000545	0.000197
7	0.14990	0.27899	0.37346	0.51747	0.58128	1.38097	1.48832	1.81812	2.16136	2.88107
	0.000216	0.000335	0.000467	0.000589	0.000249	0.000876	0.000528	0.000667	0.000775	0.000149
8	0.16682	0.31146	0.40956	0.55588	0.61915	1.37504	1.48639	1.79994	2.08696	2.71775
	0.000524	0.000611	0.000288	0.000349	0.000476	0.000666	0.000683	0.000729	0.000645	0.000222
9	0.21035	0.34001	0.43983	0.58105	0.641209	1.32867	1.43678	1.70820	1.97117	2.55611
	0.000074	0.000519	0.000606	0.000465	0.000723	0.000829	0.000377	0.000573	0.000414	0.000298
10	0.23086	0.36959	0.46898	0.61076	0.66695	1.29942	1.39710	1.65178	1.90164	2.43462
	0.000258	0.000784	0.000283	0.000379	0.000645	0.000709	0.000642	0.000531	0.000918	0.000466
11	0.25884	0.40134	0.49459	0.62620	0.68274	1.29186	1.38018	1.63232	1.85877	2.37251
	0.000283	0.000447	0.000752	0.000492	0.000611	0.000498	0.000549	0.000416	0.000546	0.000620
12	0.27699	0.41919	0.50902	0.63455	0.68927	1.28797	1.37051	1.60875	1.83183	2.29072
	0.000450	0.00258	0.000209	0.000188	0.000476	0.000455	0.000675	0.000872	0.000912	0.000142
13	0.29083	0.43005	0.52779	0.65102	0.70007	1.27081	1.34147	1.58064	1.78790	2.21657
	0.000086	0.000716	0.000244	0.000369	0.000768	0.000582	0.000900	0.000706	0.000656	0.000289
14	0.31027	0.45165	0.54470	0.66727	0.71968	1.25987	1.33249	1.54974	1.73718	2.15721
	0.000677	0.000151	0.000346	0.000446	0.000842	0.000549	0.000674	0.000816	0.000048	0.000164
15	0.32747	0.45446	0.54541	0.67772	0.73010	1.24051	1.31999	1.54174	1.73225	2.11012
	0.000425	0.000486	0.000642	0.000747	0.000800	0.000546	0.000644	0.000794	0.000467	0.000444
20	0.40116	0.53571	0.61456	0.72173	0.76757	1.21803	1.28002	1.46082	1.61227	1.98119
	0.000545	0.000236	0.000287	0.000144	0.000761	0.000649	0.000346	0.000514	0.000410	0.000321
25	0.45006	0.56186	0.65175	0.75120	0.79998	1.18642	1.24021	1.40079	1.54078	1.82341
	0.000582	0.000673	0.000297	0.000336	0.000440	0.000411	0.000646	0.000285	0.000643	0.000252
30	0.48652	0.60806	0.67976	0.77004	0.81000	1.17001	1.23140	1.36294	1.47877	1.75656
	0.000078	0.000165	0.000263	0.000209	0.000648	0.000771	0.000669	0.000418	0.000523	0.000413
50	0.58877	0.68802	0.73867	0.81937	0.84856	1.15729	1.19088	1.28996	1.37282	1.54522
	0.000284	0.000645	0.000446	0.000278	0.000352	0.000361	0.000740	0.000526	0.000427	0.000197
100	0.68816	0.76964	0.82016	0.86897	0.89017	1.09999	1.13909	1.19885	1.26301	1.37288
	0.000415	0.000349	0.000527	0.000426	0.000658	0.000782	0.000453	0.000227	0.000674	0.000587

Table 5 Quantile points of the distribution of A (when $|\rho_{yx}| = 0.40$)

n	$A_{0.01}$	$A_{0.05}$	$A_{0.10}$	$A_{0.20}$	$A_{0.25}$	$A_{0.75}$	$A_{0.80}$	$A_{0.90}$	$A_{0.95}$	$A_{0.99}$
5	0.08074	0.18632	0.28386	0.42827	0.49977	1.46156	1.63632	2.05292	2.50341	3.51790
	0.000112	0.000254	0.000639	0.000417	0.000633	0.000543	0.000887	0.000323	0.000451	0.000237
6	0.09866	0.23863	0.33753	0.49367	0.56171	1.39975	1.55038	1.93137	2.29275	3.08463
	0.000365	0.000741	0.000258	0.000364	0.000419	0.000653	0.000742	0.000524	0.000448	0.000075
7	0.15447	0.28902	0.38528	0.53341	0.60050	1.36954	1.48153	1.81200	2.15489	2.87998
	0.000616	0.000070	0.000517	0.000829	0.000644	0.000276	0.000693	0.000943	0.000258	0.000811
8	0.17532	0.31780	0.41809	0.56599	0.63521	1.37106	1.48011	1.79945	2.08382	2.70861
	0.000407	0.000096	0.000448	0.000449	0.000623	0.000718	0.000462	0.000354	0.000744	0.000082
9	0.21322	0.34937	0.44076	0.58778	0.64494	1.32187	1.43021	1.70109	1.96290	2.53612
	0.000122	0.000652	0.000716	0.000663	0.000747	0.000613	0.000326	0.000841	0.000284	0.000441
10	0.23495	0.37690	0.47672	0.61868	0.67680	1.29983	1.39019	1.64837	1.89621	2.42481
	0.000645	0.000249	0.001501	0.000237	0.000412	0.000257	0.000473	0.000146	0.000477	0.000115
11	0.26010	0.40438	0.49962	0.63009	0.69086	1.28818	1.37252	1.63029	1.85159	2.35129
	0.000651	0.000491	0.000492	0.000897	0.000558	0.000149	0.000624	0.000773	0.000872	0.000914
12	0.28269	0.42277	0.51697	0.63873	0.69810	1.28281	1.36979	1.60171	1.82135	2.28684
	0.000364	0.000431	0.000829	0.000617	0.000767	0.000515	0.000258	0.000784	0.000283	0.000379
13	0.29898	0.43978	0.53000	0.65504	0.70570	1.26905	1.33816	1.57101	1.77306	2.20609
	0.000521	0.000369	0.000741	0.0006544	0.000801	0.000080	0.000245	0.000651	0.000527	0.000211
14	0.31440	0.45745	0.54739	0.67505	0.72623	1.25914	1.32999	1.54763	1.72980	2.14626
	0.000226	0.000419	0.000327	0.000369	0.000456	0.000966	0.000229	0.000368	0.000248	0.000521
15	0.33079	0.46893	0.55696	0.68294	0.73259	1.23987	1.31899	1.53262	1.72390	2.09970
	0.000641	0.000710	0.000462	0.000253	0.000462	0.000279	0.000645	0.000514	0.000333	0.000141
20	0.41460	0.54686	0.62935	0.72879	0.77806	1.21092	1.27990	1.45373	1.60543	1.96382
	0.000643	0.000228	0.000573	0.000947	0.000097	0.000714	0.000395	0.000087	0.000215	0.000622
25	0.45806	0.57886	0.65762	0.75533	0.80005	1.18041	1.23873	1.38895	1.53041	1.81796
	0.000245	0.000877	0.000366	0.000419	0.000552	0.000872	0.000673	0.000323	0.000416	0.000088
30	0.49038	0.61438	0.68402	0.77663	0.81377	1.16991	1.22300	1.35223	1.46860	1.74442
	0.000665	0.000587	0.000192	0.000084	0.000453	0.000216	0.000471	0.000615	0.000212	0.000346
50	0.59783	0.70076	0.74768	0.82539	0.85163	1.15157	1.18643	1.28374	1.36070	1.53734
	0.000752	0.000628	0.000478	0.000951	0.000713	0.000654	0.000524	0.000114	0.000347	0.000310
100	0.69573	0.77455	0.82660	0.87316	0.89742	1.09979	1.13642	1.19261	1.25121	1.36893
	0.000349	0.000664	0.000752	0.000449	0.000611	0.000346	0.000770	0.000412	0.000324	0.000542

Table 6 Quantile points of the distribution of A (when $|\rho_{yx}|x = 0.50$)

n	$A_{0.01}$	$A_{0.05}$	$A_{0.10}$	$A_{0.20}$	$A_{0.25}$	$A_{0.75}$	$A_{0.80}$	$A_{0.90}$	$A_{0.95}$	$A_{0.99}$
5	0.08424	0.19630	0.28964	0.44282	0.52005	1.45970	1.62056	2.04143	2.49311	3.44925
	0.000182	0.000258	0.000348	0.000413	0.000665	0.000447	0.000632	0.000224	0.000642	0.000227
6	0.10559	0.24484	0.34284	0.50025	0.57091	1.39924	1.54183	1.92065	2.28275	3.07622
	0.000485	0.000631	0.000774	0.000334	0.000527	0.000519	0.000463	0.000778	0.000645	0.000224
7	0.15914	0.29655	0.39200	0.54164	0.61229	1.36278	1.47799	1.80179	2.14659	2.85765
	0.000493	0.000417	0.000558	0.000643	0.000224	0.000361	0.000549	0.000221	0.000518	0.00294
8	0.18057	0.32652	0.42808	0.57603	0.64010	1.36086	1.47238	1.79509	2.07080	2.69540
	0.000531	0.000813	0.000365	0.000441	0.000541	0.000872	0.000654	0.000980	0.000128	0.000521
9	0.21818	0.35442	0.46346	0.60598	0.66207	1.31915	1.42116	1.69570	1.95302	2.52168
	0.000652	0.002107	0.000568	0.000612	0.000830	0.000912	0.000887	0.000923	0.000102	0.000574
10	0.25143	0.38858	0.48374	0.62071	0.68037	1.29125	1.38099	1.64306	1.89001	2.39381
	0.000216	0.000529	0.000441	0.000619	0.000334	0.000249	0.000128	0.000574	0.000647	0.000553
11	0.27345	0.41576	0.50893	0.63914	0.69273	1.27984	1.36932	1.62576	1.83288	2.30198
	0.000665	0.000743	0.000412	0.000821	0.000211	0.000576	0.000612	0.000867	0.000090	0.000218
12	0.29597	0.43481	0.52853	0.65587	0.70626	1.27170	1.36247	1.59436	1.81118	2.27770
	0.000284	0.000764	0.000561	0.000643	0.000886	0.000586	0.000873	0.000691	0.000369	0.000473
13	0.31341	0.44604	0.53831	0.66561	0.71943	1.26144	1.33113	1.54893	1.75420	2.19321
	0.001361	0.000648	0.000294	0.000403	0.000114	0.000099	0.000351	0.000524	0.000364	0.000149
14	0.33133	0.46642	0.55757	0.68323	0.73288	1.25095	1.32696	1.52711	1.72234	2.13094
	0.000321	0.000447	0.000225	0.000576	0.000287	0.000741	0.000643	0.000542	0.000673	0.000249
15	0.35594	0.49262	0.57913	0.69881	0.74827	1.23584	1.31057	1.52479	1.71649	2.08344
	0.000255	0.000673	0.000362	0.00120	0.000369	0.0004160	0.000417	0.000632	0.000411	0.000366
20	0.42031	0.55379	0.63350	0.73874	0.78203	1.20444	1.27279	1.43100	1.58178	1.88486
	0.000441	0.000887	0.000371	0.000517	0.000214	0.000228	0.000337	0.000356	0.000642	0.000642
25	0.46663	0.58548	0.66125	0.76145	0.80039	1.17762	1.23296	1.38077	1.50749	1.76363
	0.000458	0.000125	0.000701	0.000646	0.000809	0.000132	0.000258	0.000361	0.000746	0.000858
30	0.50629	0.62127	0.69643	0.79442	0.82870	1.16627	1.21461	1.34559	1.45279	1.70155
	0.000255	0.000446	0.000336	0.000745	0.000528	0.000361	0.000887	0.000684	0.000972	0.000336
50	0.60357	0.70214	0.75805	0.83421	0.86343	1.14183	1.17398	1.2737	1.35719	1.52524
	0.000616	0.00209	0.000721	0.000883	0.000553	0.000336	0.001023	0.000329	0.000443	0.000264
100	0.70351	0.78513	0.83060	0.88208	0.90330	1.09463	1.12038	1.18727	1.24558	1.35774
	0.000364	0.000112	0.000775	0.000987	0.00547	0.000261	0.000264	0.000462	0.000228	0.000349

Table 7 Quantile points of the distribution of A (when $|\rho_{yx}| = 0.60$)

n	$A_{0.01}$	$A_{0.05}$	$A_{0.10}$	$A_{0.20}$	$A_{0.25}$	$A_{0.75}$	$A_{0.80}$	$A_{0.90}$	$A_{0.95}$	$A_{0.99}$
5	0.09253	0.20776	0.30678	0.46457	0.54039	1.45570	1.61647	2.03974	2.48324	3.43849
	0.000418	0.000221	0.000843	0.000902	0.000342	0.000411	0.000643	0.000942	0.000764	0.000091
6	0.11765	0.25643	0.35792	0.51824	0.58726	1.39744	1.53375	1.90131	2.27435	3.06561
	0.000116	0.000264	0.000461	0.000392	0.000543	0.000661	0.000916	0.000506	0.000143	0.000266
7	0.17661	0.31949	0.42095	0.57036	0.63378	1.35820	1.46955	1.79085	2.13916	2.83694
	0.000521	0.000072	0.000987	0.000367	0.000492	0.000114	0.000099	0.000351	0.000364	0.000431
8	0.20004	0.34571	0.44684	0.60036	0.66085	1.34082	1.44146	1.72996	2.01856	2.63842
	0.000369	0.000411	0.000873	0.000655	0.000392	0.000411	0.000747	0.000564	0.000336	0.000065
9	0.23329	0.37253	0.47337	0.61796	0.68002	1.31624	1.41424	1.68791	1.92493	2.49204
	0.000856	0.000364	0.000247	0.000336	0.000452	0.000913	0.000469	0.000714	0.000649	0.000357
10	0.25823	0.41299	0.50786	0.63974	0.69833	1.28960	1.37776	1.63969	1.88249	2.34757
	0.000229	0.000448	0.000114	0.000557	0.000641	0.000812	0.000643	0.000249	0.000944	0.000643
11	0.27755	0.43206	0.52792	0.66223	0.71616	1.27757	1.36343	1.61392	1.81961	2.26617
	0.000278	0.000669	0.000808	0.000254	0.000557	0.000619	0.000669	0.000441	0.000673	0.000543
12	0.31188	0.45503	0.54419	0.67318	0.72407	1.26940	1.35114	1.58369	1.77997	2.19720
	0.000573	0.000854	0.000919	0.000450	0.00258	0.000416	0.000546	0.000482	0.000673	0.000557
13	0.32863	0.47025	0.56149	0.68407	0.73389	1.25473	1.32859	1.54276	1.73373	2.12238
	0.000632	0.000552	0.000966	0.000749	0.000546	0.000720	0.000638	0.000467	0.000228	0.000173
14	0.35404	0.49103	0.57783	0.69985	0.75282	1.24920	1.31759	1.51645	1.70642	2.11915
	0.000081	0.000349	0.000542	0.000873	0.000558	0.000419	0.000673	0.000581	0.000664	0.000246
15	0.36812	0.50477	0.59331	0.70971	0.75928	1.22984	1.30029	1.49272	1.66649	2.04678
	0.000229	0.000416	0.000744	0.000813	0.000681	0.000212	0.000517	0.000421	0.000244	0.000645
20	0.43222	0.55514	0.63994	0.74404	0.78682	1.20138	1.26177	1.42613	1.56148	1.83603
	0.000442	0.000576	0.000558	0.000643	0.000224	0.000673	0.000576	0.000883	0.000649	0.000246
25	0.48472	0.60170	0.67522	0.77239	0.81172	1.17079	1.22408	1.37260	1.49763	1.75086
	0.000471	0.000622	0.000746	0.000511	0.000228	0.000376	0.000458	0.000512	0.000642	0.000097
30	0.51572	0.63953	0.70815	0.80080	0.83779	1.16303	1.20727	1.33314	1.43737	1.66238
	0.000673	0.000472	0.000673	0.000882	0.000919	0.000726	0.000573	0.000083	0.000128	0.0000149
50	0.60970	0.71275	0.76999	0.84528	0.87318	1.13026	1.16802	1.26263	1.34648	1.51069
	0.000241	0.000627	0.000891	0.000349	0.000641	0.000549	0.000603	0.000249	0.000349	0.000549
100	0.72315	0.79422	0.83617	0.88981	0.90973	1.08921	1.11218	1.17407	1.23061	1.33886
	0.000543	0.000417	0.000255	0.000426	0.000559	0.000749	0.000409	0.000667	0.000497	0.000254

Table 8 Quantile points of the distribution of A (when $|\rho_{yx}|=0.70$)

n	$A_{0.01}$	$A_{0.05}$	$A_{0.10}$	$A_{0.20}$	$A_{0.25}$	$A_{0.75}$	$A_{0.80}$	$A_{0.90}$	$A_{0.95}$	$A_{0.99}$
5	0.10424	0.23205	0.33079	0.49787	0.58353	1.44860	1.59072	2.02327	2.45141	3.41911
	0.0002347	0.0008916	0.000269	0.000913	0.000527	0.000369	0.000491	0.000098	0.000294	0.000162
6	0.15378	0.28914	0.39882	0.55783	0.62929	1.38124	1.50641	1.85509	2.19933	3.03838
	0.000624	0.000369	0.000189	0.000729	0.000509	0.000651	0.000444	0.000608	0.000897	0.000114
7	0.18708	0.33467	0.42915	0.58632	0.65449	1.35515	1.46545	1.77039	2.05662	2.80257
	0.000148	0.000258	0.000643	0.000510	0.000664	0.000806	0.000099	0.000226	0.000319	0.000080
8	0.23625	0.37027	0.47556	0.61772	0.67978	1.33087	1.43097	1.71811	1.96779	2.55939
	0.000058	0.000668	0.000582	0.000669	0.000068	0.000987	0.000367	0.000492	0.000686	0.000558
9	0.26328	0.40551	0.50342	0.64578	0.70474	1.30831	1.40120	1.66481	1.91193	2.48411
	0.0004489	0.000651	0.000887	0.000623	0.000088	0.000567	0.000492	0.000332	0.000386	0.000408
10	0.28730	0.42995	0.52809	0.66502	0.71646	1.28785	1.36942	1.61043	1.83518	2.30553
	0.000500	0.000293	0.000461	0.000339	0.000065	0.000478	0.000658	0.000587	0.000693	0.000109
11	0.30966	0.46530	0.55194	0.67786	0.73122	1.27095	1.34316	1.56306	1.75764	2.21297
	0.000049	0.000159	0.000357	0.000846	0.000964	0.000555	0.000390	0.000861	0.000706	0.000645
12	0.33011	0.47440	0.56544	0.69007	0.74621	1.24675	1.33152	1.53493	1.71827	2.18834
	0.000147	0.000079	0.000610	0.000578	0.000225	0.002104	0.000513	0.000574	0.000941	0.000066
13	0.36433	0.50171	0.59290	0.71235	0.76411	1.24658	1.32341	1.51743	1.70589	2.07515
	0.000236	0.000486	0.000743	0.000891	0.000459	0.001173	0.000298	0.000447	0.000309	0.000119
14	0.37421	0.51480	0.60460	0.72664	0.77118	1.24092	1.30380	1.49457	1.66006	2.02655
	0.000871	0.000432	0.000369	0.000130	0.000554	0.000834	0.000076	0.000673	0.000418	0.000198
15	0.38382	0.52314	0.61516	0.72739	0.77595	1.22814	1.28793	1.46930	1.62245	1.96856
	0.000227	0.00364	0.000440	0.000512	0.000809	0.000147	0.000387	0.00113	0.000095	0.000108
20	0.45628	0.58486	0.66182	0.76425	0.80685	1.19424	1.25060	1.39692	1.53788	1.81301
	0.000497	0.000257	0.000449	0.000364	0.000431	0.000829	0.000617	0.000767	0.000515	0.000126
25	0.50338	0.62569	0.69054	0.78907	0.82606	1.16700	1.21690	1.34632	1.45022	1.68131
	0.000086	0.000376	0.000461	0.000612	0.000069	0.000336	0.000472	0.001640	0.000486	0.000418
30	0.54497	0.65435	0.72136	0.80753	0.84440	1.15279	1.19454	1.30951	1.41009	1.60982
	0.000504	0.000072	0.000249	0.001361	0.000648	0.000294	0.000403	0.000114	0.000099	0.000351
50	0.62952	0.73296	0.78559	0.85336	0.88236	1.11777	1.14708	1.23123	1.31118	1.45303
	0.000077	0.000469	0.000716	0.000858	0.000767	0.000431	0.000349	0.000228	0.000379	0.000901
100	0.73854	0.80946	0.84783	0.89838	0.91807	1.08507	1.10685	1.16750	1.21870	1.31350
	0.000156	0.000237	0.000643	0.000527	0.000887	0.000394	0.000495	0.000147	0.000532	0.000419

Table 9 Quantile points of the distribution of A (when $|\rho_{yx}| = 0.80$)

n	$A_{0.01}$	$A_{0.05}$	$A_{0.10}$	$A_{0.20}$	$A_{0.25}$	$A_{0.75}$	$A_{0.80}$	$A_{0.90}$	$A_{0.95}$	$A_{0.99}$
5	0.12760	0.28841	0.39975	0.56940	0.63878	1.41808	1.53796	1.90934	2.32119	3.29786
	0.000081	0.000553	0.000336	0.001023	0.000559	0.000327	0.000638	0.000443	0.0008839	0.0001279
6	0.18834	0.33301	0.44612	0.61027	0.67685	1.35851	1.46952	1.79955	2.13583	2.94291
	0.000112	0.000349	0.000243	0.000564	0.000493	0.000587	0.000463	0.000554	0.000199	0.000264
7	0.22760	0.39054	0.49772	0.63384	0.69949	1.32017	1.41303	1.68683	1.94888	2.59519
	0.000664	0.000069	0.000333	0.000443	0.000980	0.001183	0.000337	0.000873	0.000492	0.000146
8	0.26282	0.42425	0.52562	0.66991	0.72573	1.29293	1.37956	1.62570	1.85220	2.35723
	0.000813	0.000546	0.000237	0.000615	0.000543	0.000329	0.000443	0.000506	0.000447	0.000118
9	0.30571	0.46255	0.55397	0.69242	0.74858	1.28245	1.36366	1.57725	1.78672	2.24067
	0.000440	0.000728	0.000981	0.000381	0.000238	0.000119	0.000441	0.000665	0.000873	0.000243
10	0.32731	0.48501	0.57467	0.70030	0.75381	1.25290	1.32744	1.52305	1.70744	2.17709
	0.000776	0.000321	0.000449	0.000834	0.001023	0.000409	0.000734	0.000493	0.000654	0.000555
11	0.35820	0.50426	0.59668	0.71791	0.76907	1.23761	1.30845	1.50528	1.67926	2.10165
	0.000077	0.000444	0.000332	0.000965	0.000853	0.000678	0.000195	0.001132	0.000765	0.000065
12	0.36747	0.51731	0.60870	0.72647	0.77378	1.23216	1.29525	1.47356	1.64414	1.97034
	0.000617	0.000447	0.000331	0.000219	0.000664	0.000987	0.000888	0.000067	0.000553	0.000446
13	0.39388	0.53966	0.62847	0.73960	0.78566	1.21811	1.28135	1.45056	1.61294	1.96986
	0.000887	0.000616	0.00209	0.000336	0.000489	0.000721	0.000883	0.000381	0.000459	0.000228
14	0.41966	0.55855	0.64399	0.75382	0.79864	1.21084	1.26531	1.43116	1.58240	1.90339
	0.000210	0.000529	0.000873	0.000919	0.000889	0.000694	0.000087	0.001123	0.000556	0.000129
15	0.42497	0.57242	0.65493	0.76315	0.80708	1.20153	1.26120	1.42446	1.57170	1.89125
	0.000819	0.000239	0.000132	0.000089	0.001143	0.000553	0.000646	0.000236	0.000476	0.000369
20	0.50252	0.63058	0.70189	0.79165	0.83098	1.17480	1.22160	1.34186	1.46084	1.68477
	0.000568	0.000719	0.001006	0.000090	0.000687	0.000813	0.000147	0.000623	0.000049	0.000238
25	0.53893	0.65786	0.72923	0.81468	0.84746	1.14945	1.19313	1.29974	1.39123	1.58336
	0.000777	0.000159	0.000398	0.000858	0.000903	0.001225	0.000714	0.000342	0.000369	0.000200
30	0.58633	0.69327	0.75731	0.83688	0.86895	1.13997	1.17633	1.28460	1.36762	1.54026
	0.000462	0.000987	0.000225	0.000630	0.000701	0.000890	0.000978	0.000431	0.000809	0.000210
50	0.67413	0.76400	0.81535	0.87490	0.89816	1.10330	1.13112	1.20826	1.27396	1.40340
	0.000413	0.000654	0.002107	0.000568	0.000612	0.000830	0.000912	0.000887	0.000923	0.000102
100	0.75698	0.82864	0.86403	0.90762	0.92458	1.07610	1.09432	1.14616	1.19063	1.27516
	0.000236	0.000458	0.000248	0.003698	0.001097	0.000489	0.000721	0.000111	0.000874	0.000125

Table 10 Quantile points of the distribution of A (when $|\rho_{yx}| = 0.90$)

n	$A_{0.01}$	$A_{0.05}$	$A_{0.10}$	$A_{0.20}$	$A_{0.25}$	$A_{0.75}$	$A_{0.80}$	$A_{0.90}$	$A_{0.95}$	$A_{0.99}$
5	0.19031	0.37981	0.49410	0.65492	0.71963	1.32630	1.42068	1.71968	2.05616	2.99270
	0.000140	0.000613	0.000085	0.000467	0.000349	0.000278	0.000821	0.000467	0.000594	0.000746
6	0.26314	0.44880	0.55281	0.69442	0.74955	1.27500	1.35618	1.60964	1.83389	2.49855
	0.000216	0.000649	0.000526	0.000348	0.000456	0.000627	0.000919	0.000278	0.000446	0.000290
7	0.30918	0.48467	0.58568	0.72612	0.77660	1.25438	1.32331	1.53593	1.73640	2.30705
	0.000336	0.000673	0.000246	0.000081	0.000097	0.000592	0.000249	0.000462	0.000074	0.000056
8	0.35150	0.51813	0.61600	0.73896	0.78776	1.22601	1.29113	1.47426	1.66615	2.10205
	0.000552	0.000629	0.000527	0.000479	0.000697	0.000227	0.001462	0.000376	0.000916	0.000179
9	0.39752	0.55700	0.64318	0.75811	0.80072	1.22242	1.28047	1.44535	1.59290	1.96710
	0.000318	0.000665	0.000587	0.000192	0.000084	0.000453	0.000216	0.000471	0.000615	0.000334
10	0.43652	0.58053	0.66296	0.77073	0.81538	1.19784	1.25482	1.41366	1.55279	1.87735
	0.000292	0.000565	0.000413	0.000946	0.000729	0.000183	0.000078	0.000824	0.000561	0.000091
11	0.45717	0.59920	0.67945	0.78104	0.82524	1.19087	1.24014	1.38008	1.51322	1.82994
	0.000468	0.000409	0.000273	0.000694	0.000592	0.000774	0.000816	0.000087	0.000412	0.000345
12	0.48045	0.61447	0.70043	0.80068	0.83242	1.18253	1.22862	1.36613	1.48909	1.79328
	0.000447	0.000613	0.000276	0.000489	0.000625	0.000714	0.000885	0.000215	0.000645	0.000089
13	0.48976	0.62854	0.70669	0.80201	0.83719	1.16927	1.21266	1.33232	1.44752	1.73246
	0.000236	0.000378	0.000573	0.000824	0.000442	0.000334	0.000496	0.000547	0.000771	0.000464
14	0.50880	0.63869	0.70883	0.80547	0.84072	1.16534	1.20953	1.32777	1.43774	1.67538
	0.000417	0.000168	0.000211	0.000467	0.000555	0.000743	0.000698	0.000549	0.000628	0.000362
15	0.52828	0.65868	0.72708	0.81693	0.84924	1.15741	1.19817	1.31441	1.42044	1.65283
	0.000064	0.000264	0.000728	0.000341	0.000673	0.000483	0.000607	0.000894	0.000761	0.000259
20	0.59635	0.70751	0.76927	0.84372	0.87183	1.13026	1.16700	1.26102	1.34918	1.53386
	0.000164	0.000647	0.000275	0.000349	0.000543	0.000673	0.000916	0.000783	0.000662	0.000805
25	0.64662	0.73617	0.79215	0.86170	0.88928	1.11688	1.14531	1.23331	1.30798	1.46080
	0.000111	0.000267	0.000790	0.000463	0.000583	0.000918	0.000664	0.000785	0.000553	0.000449
30	0.65662	0.76113	0.81080	0.87393	0.89996	1.10272	1.13057	1.20550	1.26921	1.39693
	0.000330	0.000453	0.000576	0.000918	0.000557	0.000246	0.000776	0.000076	0.000162	0.000238
50	0.73552	0.81359	0.85080	0.90160	0.91985	1.08025	1.10067	1.15688	1.20589	1.29695
	0.000444	0.000349	0.000194	0.000853	0.000627	0.000435	0.000864	0.000761	0.000565	0.000068
100	0.81336	0.86827	0.89651	0.93144	0.94546	1.05754	1.07180	1.11188	1.14567	1.20903
	0.000301	0.000462	0.000879	0.000276	0.000764	0.000345	0.000682	0.000943	0.000073	0.000143

Table 11 Quantile points of the distribution of A (when $|\rho_{yx}| = 0.99$)

n	$A_{0.01}$	$A_{0.05}$	$A_{0.10}$	$A_{0.20}$	$A_{0.25}$	$A_{0.75}$	$A_{0.80}$	$A_{0.90}$	$A_{0.95}$	$A_{0.99}$
5	0.56350	0.72674	0.80010	0.87310	0.89953	1.10780	1.13908	1.23118	1.33179	1.65200
	0.000080	0.000094	0.000157	0.000346	0.000662	0.000580	0.000328	0.000463	0.000846	0.000393
6	0.64227	0.77008	0.82873	0.88905	0.91062	1.09573	1.12099	1.19559	1.26814	1.49911
	0.000680	0.000293	0.000430	0.000367	0.000591	0.000440	0.000442	0.000407	0.000673	0.000219
7	0.69046	0.79629	0.84254	0.89774	0.91774	1.08682	1.10868	1.17517	1.23935	1.43276
	0.000413	0.000529	0.000225	0.000919	0.000612	0.000553	0.000883	0.000431	0.000459	0.000210
8	0.72553	0.81481	0.85860	0.90760	0.92548	1.07617	1.09633	1.15389	1.21209	1.35511
	0.000428	0.000672	0.000816	0.000179	0.000762	0.0000864	0.000963	0.000426	0.000492	0.000195
9	0.74525	0.82732	0.86778	0.91467	0.93079	1.07259	1.09175	1.14453	1.19351	1.30152
	0.000078	0.000117	0.000645	0.000531	0.000276	0.000582	0.000919	0.000726	0.000573	0.000372
10	0.75844	0.84038	0.87283	0.92070	0.93642	1.06876	1.08619	1.13484	1.18005	1.29255
	0.000243	0.000760	0.000314	0.000417	0.000558	0.000643	0.000224	0.000361	0.000549	0.000221
11	0.77049	0.84236	0.87868	0.92220	0.93827	1.06267	1.07894	1.12522	1.16809	1.25977
	0.000406	0.00072	0.000246	0.000664	0.000289	0.000643	0.000592	0.000642	0.000793	0.000415
12	0.78935	0.85588	0.88870	0.92703	0.94063	1.05929	1.07558	1.11640	1.15397	1.24467
	0.000064	0.000183	0.000675	0.000642	0.000556	0.000462	0.000190	0.000867	0.000473	0.000276
13	0.79773	0.86276	0.89396	0.93103	0.94149	1.05739	1.07163	1.11176	1.14785	1.23078
	0.000493	0.000671	0.000463	0.000512	0.000793	0.000474	0.000776	0.000649	0.000379	0.000226
14	0.80353	0.86583	0.89472	0.93120	0.94494	1.05614	1.07038	1.11104	1.14653	1.22454
	0.000713	0.000446	0.000573	0.000876	0.000413	0.000997	0.000408	0.000673	0.000576	0.000245
15	0.81570	0.87378	0.90291	0.93660	0.94868	1.05142	1.06376	1.10066	1.13454	1.20833
	0.000346	0.000442	0.000576	0.000461	0.000911	0.000567	0.000605	0.000777	0.000294	0.000090
20	0.84455	0.89200	0.91566	0.94520	0.95600	1.04595	1.05723	1.08673	1.11433	1.16948
	0.000147	0.000493	0.000542	0.000708	0.000867	0.000962	0.000285	0.000556	0.000673	0.000344
25	0.85569	0.90366	0.92610	0.95179	0.96121	1.03992	1.05076	1.07772	1.10301	1.15087
	0.000522	0.000673	0.000472	0.000673	0.000882	0.000503	0.000762	0.000448	0.000562	0.000440
30	0.87793	0.91429	0.93367	0.95709	0.96544	1.03680	1.04654	1.07211	1.09294	1.13516
	0.000415	0.000649	0.000751	0.000906	0.000683	0.000286	0.000794	0.000664	0.000074	0.000675
50	0.90553	0.93419	0.94849	0.96651	0.97325	1.02718	1.03397	1.05183	1.06722	1.09732
	0.000276	0.000527	0.000667	0.000750	0.000883	0.000649	0.000246	0.000083	0.000128	0.000248
100	0.93339	0.95390	0.96410	0.97630	0.98097	1.01893	1.02373	1.03604	1.04695	1.06818
	0.000312	0.000466	0.000617	0.000883	0.000529	0.000382	0.000667	0.000773	0.000109	0.000627

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