# Variable Control Charts based on percentiles of The New Weighted Exponential Distribution 

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#### Abstract

The intention of this article is to evaluate the proposed new weighted exponential distribution's percentiles of mean, median, range, and standard deviation, as well as establish control limits for these parameters. We compare the competence and suitability of the derived control limits to those based on well-known Shewhart control limits by assuming that the variable quality characteristic follows the new weighted exponential distribution. In the simulation phase, the coverage probabilities were computed and compared.


Keywords: most probable, pdf, cdf, equi-tailed, percentiles, NWED

## I. Introduction

In statistical quality control, variable control charts play a vital role in evaluating variation in a process. The two types of variable control charts evaluate variation between the samples and within the samples as well. The processbehavior charts or Shewhart charts, which are based on the premise that the character variate pursues a Gaussian distribution, are also known as variable control charts. The control lines of a process-behavior control chart or Shewhart chart are $E\left(t_{n}\right) \pm 3^{*} S . E\left(t_{n}\right)$ if $x_{1}, x_{2}, \ldots, x_{n}$ is a list of $n$ measurements on a quality variable of a material and $t_{n}$ is a specific characteristic according to this study. Regardless the population model, to test the process quality we use Shewhart control limits. These limits involve Shewhart constants which are readily available in any SQC text book. The Shewhart control charts are constructed based on the assumption that the population model tend to normal distribution. This is valid from the central limit theorem which states that as the size $n$ of the sample gets larger, for almost all the distributions, the distribution of sample means approximates a normal distribution, assuming that all the samples are in same size. But, in quality assurance studies majority of the times, the observed information is consistently taken in limited samples and the sample size may not be the same. Furthermore, if evidence was provided that the sampled variable quality characteristic follows a distribution pattern other than the normal distribution, the online step of such a characteristic can be tracked using the corresponding distribution theory. In such situations, the inference of normality should be avoided unless a prior analysis of the goodness of fit test has been performed. Using
the central limit theorem, on the other hand, is not recommended since it provides asymptotic normality for a given statistic. As a consequence, when the population distribution is non-normal, an alternate method for the development of control charts is required.
Many authors have attempted to create statistical quality control approaches using skewed distributions. Some of them are "Edgeman(1989) [3]-Inverse Gaussion Distribution, Gonzalez and Viles(2000) [4]-Gamma Distribution, Kantam and Sriram(2001) [5]-Gamma Distribution, Chan and Cui (2003) [2] have developed control chart constants for skewed distributions where the constants are dependent on the coefficient of skewness of the distribution, Kantam et al(2006) [6]-Log logistic Distribution, Betul and Yaziki(2006) [1]-Burr Distribution, Subba Rao and Kantam(2008) [18]-Double exponential distribution, Kantam and Rao(2010) [7]-control charts for process variate, Rao and Sarath Babu (2012) [13]-Linear failure rate distribution, Rao and Kantam (2012) [17]-Half logistic distribution, Rao et al(2013) [11]-Inverse Rayleigh distribution, Rao et al(2014) [14]-Size biased lomax distribution, Rao and Kumar (2015) [15]-Exponential Gamma distribution, Rao et al(2016) [16]-Half normal distribution, Rosaiah et al(2018) [9]-Gumbel distribution, Rosaiah et al(2019) [10]-Exponentiated inverse Kumaraswamy distribution, Sricharani and Rao (2019) [12]Dagum distribution" and references there in.
In this paper, we made an assumption on process quality variable that it follows a new weighted exponential distribution and constructed control limits for simulated data in similar to the familiar Shewhart control limits. Consider $X$ as a variable drawn randomly from a new weighted
exponential distribution (NWED) whose distribution function (cdf) is,

$$
F(x ; \alpha, \lambda)=1-e^{-(1+\lambda) \alpha x} ; \quad x>0, \alpha>0, \lambda>0
$$

(1.1)

Where, $\alpha$ is a location parameter and $\lambda$ is a spread parameter.

The probability distribution function (pdf) corresponds to (1.1) is,
$f(x ; \alpha, \lambda)=(1+\lambda) \alpha \cdot e^{-(1+\lambda) \alpha x} ; \quad x>0, \alpha>0, \lambda>0$.

Plots of the $p d f_{s}$ and $c d f_{s}$ of NWED for selected parameter values are shown in Figures 1 and 2.


Fig2. CDFs of NWED


Figure 1 shows that the curve of the NWED is very close to that of a standard Exponential distribution with parameter $\alpha$ as $\lambda \rightarrow 0$.
The survival (or) reliability function of NWED is given by, $\bar{F}(x)=e^{-(1+\lambda) \alpha x}(1.3)$
The hazard function of NWED is given by,
$h(x)=(1+\lambda) \alpha(1.4)$
The distributional properties are:

Mean: $E(x)=\left(\frac{1}{(1+\alpha) \lambda}\right)(1.5)$
Median: $M d=\left(\frac{\ln (2)}{(1+\alpha) \lambda}\right)(1.6$
Variance: $V(x)=\left(\frac{1}{(1+\alpha) \lambda}\right)^{2}$
Quantile function: $Q(U)=\left(\frac{\ln \left(\frac{1}{1-U}\right)}{(1+\alpha) \lambda}\right)$
where, $U$ follows uniform distribution $[0,1]$.
The density function of the $i^{\text {th }}$ order statistic $X_{(i)}$ is provided by,
$\alpha_{(i)}=\frac{n!}{(i-1)!(n-i)!}(1+\lambda) \alpha \cdot e^{-(1+\lambda) \alpha x}\left(1-e^{-(1+\lambda) \alpha x}\right)^{i-1}\left(e^{-(1+\lambda) \alpha}\right.$ (1.9)

If $i=1$, the density function of the minimum order statistic
$X_{(1)}$ is provided by,
$\alpha_{(1)}=n(1+\lambda) \alpha \cdot e^{-(1+\lambda) \alpha x}\left(e^{-(1+\lambda) \alpha x}\right)^{n-1}$
If $i=n$, the density function of the maximum order statistic
$X_{(n)}$ is provided by,
$\alpha_{(n)}=n(1+\lambda) \alpha \cdot e^{-(1+\lambda) \alpha x}\left(1-e^{-(1+\lambda) \alpha x}\right)^{n-1}$
The other distributional properties are extensively explored by Oguntunde et al(2016)[8]. NWED is another skewed distribution situation that is given a lot of attention in this study when it comes to developing control charts. The hazard function implies a constant failure rate functions if $\alpha=0.2$ and $\lambda=0.6$, making the NWED ideal for modeling components that wears continuously over time. It is also one of the statistical models that can be used for life checking and research on reliability. As a result, if lifetime data is considered to be of high quality, practitioners will benefit from the creation of control charts for it. Since NWED is a skewed distribution, this paper aims to compare it to other distributions. This paper makes an effort to answer and solve this problem to the best of our ability. The remaining portion of the paper is laid out as follows. The fundamentals of statistics are covered in Section 2 as well as the development of process control charts for mean, median, range, and standard deviation. A comparison of the existing Shewhart control lines to the derived NWED control lines is presented in Section 3. A description and conclusions are found in Section 4.

## II. CONTROL CHART CONSTANTS THROUGH PERCENTILES

### 2.1. Mean-chart

Consider $x_{1}, x_{2}, \ldots, x_{n}$ be a $n$ sized arbitrary sample drawn from NWED with $\alpha=0.2$ and $\lambda=0.6$. If you target the average population, this is a sample of industrial process data. The statistic $\bar{x}$ determines whether the process average is similar to the desired average when using re-sampling. Statistically speaking, we need to determine the most likely limit value beyond the importance of a fall. The term is most likely used here to refer to a relative definition, which should be considered in the context of aggregation. Since the current method is only applicable to the normal distribution, the
determination of the $3 \sigma$ limit value is considered as the most probable limit value. As we all know, the $3 \sigma$ normal distribution limit gives 0.9973 probability content. As a result, we need to find the two sample distributions of the sample mean in NWED, with a probability level of 0.9973 . We must find an $\mathrm{L}, \mathrm{U}$

$$
\begin{equation*}
P(L \leq \bar{x} \leq U)=0.9973 \tag{2.1}
\end{equation*}
$$

Where, $\bar{x}$ is the mean of sample size $n$.
Using the equi-tailed concept, the percentiles of the sampling distribution of $\bar{X}$ are 0.00135 and 0.99865 , respectively. By simulating the empirical sampling distribution of $\bar{x}$, we were able to compute its percentiles. Table 1 shows this detail.

Table 1: Percentiles of Mean in NWED

| n | 0.99865 | 0.9950 | 0.99 | 0.975 | 0.95 | 0.05 | 0.025 | 0.01 | 0.005 | 0.00135 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 2 | 10.1224 | 8.4359 | 7.1125 | 5.7094 | 4.6020 | 0.4987 | 0.4458 | 0.3998 | 0.3758 | 0.3529 |
| 3 | 8.4559 | 6.9699 | 6.0168 | 4.9388 | 4.144 | 0.5664 | 0.5141 | 0.4641 | 0.4393 | 0.4002 |
| 4 | 6.7542 | 5.8438 | 5.2652 | 4.3630 | 3.6551 | 0.6202 | 0.5715 | 0.5151 | 0.4813 | 0.4279 |
| 5 | 6.2198 | 5.3871 | 4.7866 | 4.0340 | 3.5009 | 0.6516 | 0.5991 | 0.5512 | 0.5188 | 0.4713 |
| 6 | 6.0049 | 5.0820 | 4.5851 | 3.9095 | 3.3762 | 0.6811 | 0.6293 | 0.5774 | 0.5480 | 0.4937 |
| 7 | 5.2979 | 4.5646 | 4.2024 | 3.6288 | 3.1799 | 0.7027 | 0.6564 | 0.6119 | 0.5778 | 0.5266 |
| 8 | 5.1336 | 4.4300 | 4.0545 | 3.4849 | 3.0881 | 0.7245 | 0.6739 | 0.6291 | 0.5952 | 0.5414 |
| 9 | 4.8645 | 4.2745 | 3.8867 | 3.3865 | 2.9910 | 0.7549 | 0.7032 | 0.6490 | 0.6253 | 0.5785 |
| 10 | 4.5500 | 3.9553 | 3.7207 | 3.2355 | 2.8649 | 0.7693 | 0.7135 | 0.6613 | 0.6400 | 0.5909 |

The sample mean control limits are calculated in the following way using the percentiles in the above table. Take a look at the $\bar{x}$ distribution,
$P\left(Z_{0.00135} \leq \bar{x} \leq Z_{0.99865}\right)=0.9973$
But $\bar{x}$ of sampling distribution when $\alpha=0.2$ and $\lambda=0.6$ is 3.125 for NWED.
From equation (2.2) over the course of repeated sampling, for the mean of $i^{\text {th }}$ subgroup, we are able to have
$P\left(Z_{0.00135} \cdot \frac{\bar{x}}{3.125} \leq \bar{x}_{i} \leq Z_{0.99865} \cdot \frac{\bar{x}}{3.125}\right)=0.9973$ (2.3)
This can be expressed as:
$P\left(\mathrm{~A}_{2 \mathrm{p}}^{*} \cdot \overline{\bar{x}} \leq \bar{x}_{i} \leq \mathrm{A}_{2 \mathrm{p}}^{* *} \cdot \overline{\bar{x}}\right)=0.9973$ (2.4)
Where, $\overline{\bar{x}}$ is grand mean, $\bar{x}_{i}$ is $i^{\text {th }}$ subgroup mean, $\mathrm{A}_{2 \mathrm{p}}^{*}=\frac{Z_{0.00135}}{3.125}, \mathrm{~A}_{2 \mathrm{p}}^{* *}=\frac{Z_{0.99865}}{3.125}$. Thus, $\mathrm{A}_{2 \mathrm{P}}^{*}$ and $\mathrm{A}_{2 \mathrm{P}}^{* *}$ are percentile constants. Table 2 shows the percentile constants for the $\bar{X}$ chart for NWED.

Table 2: Percentile Constants of Mean chart.

| n | $\mathrm{A}_{2 \mathrm{P}}^{*}$ | $\mathrm{~A}_{2 \mathrm{P}}^{* *}$ |
| :---: | :---: | :---: |
| 2 | 0.1753 | 5.0282 |
| 3 | 0.2009 | 4.2447 |
| 4 | 0.2133 | 3.3676 |
| 5 | 0.2379 | 3.1400 |
| 6 | 0.2528 | 3.0747 |
| 7 | 0.2642 | 2.6579 |
| 8 | 0.2737 | 2.5949 |
| 9 | 0.2939 | 2.4710 |
| 10 | 0.2988 | 2.3005 |

### 2.2. Median-chart.

We must find two sampling distribution limits of sample median in NWED that have a probability content of 0.9973. Symbolically, we must locate L, U in a manner that $P(L \leq m \leq U)=0.9973$ (2.5)

Where, $m$ is the sample median of size $n$. Table 3 shows the percentiles discovered by simulation.
Table 3: Percentiles of Median in NWED

| n | 0.99865 | 0.9950 | 0.99 | 0.975 | 0.95 | 0.05 | 0.025 | 0.01 | 0.005 | 0.00135 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 2 | 10.1224 | 8.4359 | 7.1125 | 5.7094 | 4.6020 | 0.4987 | 0.4458 | 0.3998 | 0.3758 | 0.3529 |
| 3 | 8.4192 | 6.3327 | 5.4504 | 4.0038 | 2.9059 | 0.4684 | 0.4260 | 0.3893 | 0.3700 | 0.3466 |
| 4 | 6.0579 | 4.9053 | 4.2368 | 3.2116 | 2.5132 | 0.5423 | 0.4924 | 0.4453 | 0.4207 | 0.3970 |
| 5 | 5.4972 | 4.3851 | 3.7097 | 2.6410 | 1.8603 | 0.5236 | 0.4760 | 0.4323 | 0.4118 | 0.3838 |
| 6 | 4.9894 | 3.6335 | 3.1559 | 2.3781 | 1.8493 | 0.5679 | 0.5253 | 0.4772 | 0.4477 | 0.4181 |
| 7 | 4.1172 | 3.3842 | 2.7356 | 1.8569 | 1.2365 | 0.5613 | 0.5152 | 0.4679 | 0.4458 | 0.4115 |
| 8 | 3.5811 | 2.8979 | 2.4114 | 1.8454 | 1.4244 | 0.6014 | 0.5603 | 0.5079 | 0.4806 | 0.4480 |
| 9 | 3.5471 | 2.6732 | 2.1811 | 1.4688 | 1.2098 | 0.5968 | 0.5497 | 0.5106 | 0.4855 | 0.4432 |
| 10 | 2.8548 | 2.2435 | 1.9305 | 1.4849 | 1.2145 | 0.6214 | 0.5781 | 0.5338 | 0.5062 | 0.4710 |

The median control limits are determined in the following way using the percentiles in the above table. Take a look at the $m$ distribution
$P\left(Z_{0.00135} \leq m \leq Z_{0.99865}\right)=0.9973$ (2.6)
But median of sampling distribution when $\alpha=0.2$ and $\lambda=0.6$ is 2.1661 for NWED.
From equation(2.6) over repeated sampling, for the $i^{\text {th }}$ subgroup median we can have
$P\left(Z_{0.00135} \cdot \frac{\bar{m}}{2.1661} \leq m_{i} \leq Z_{0.99865} \cdot \frac{\bar{m}}{2.1661}\right)=0.9973$ (2.7)
This can be written as
$P\left(\mathrm{~A}_{7 \mathrm{p}}^{*} \cdot \bar{m} \leq m_{i} \leq \mathrm{A}_{7 \mathrm{p}}^{* *} \cdot \bar{m}\right)=0.9973$
Where, $\bar{m}$ is mean of subgroup medians. Thus $\mathrm{A}_{7 \mathrm{p}}^{*}=\frac{Z_{0.00135}}{2.1661}$ and $\mathrm{A}_{7 \mathrm{p}}^{* *}=\frac{Z_{0.99865}}{2.1661}$ are the percentile constants of median chart and are given in table 4.
Table 4: Percentile Constants of Median chart

| n | $\mathrm{A}_{7 \mathrm{P}}^{*}$ | $\mathrm{~A}_{7 \mathrm{P}}^{* *}$ |
| :---: | :---: | :---: |
| 2 | 0.1629 | 4.6731 |
| 3 | 0.1600 | 3.8868 |
| 4 | 0.1833 | 2.7967 |
| 5 | 0.1772 | 2.5379 |
| 6 | 0.1930 | 2.3034 |
| 7 | 0.1900 | 1.9008 |
| 8 | 0.2068 | 1.6533 |
| 9 | 0.2046 | 1.6376 |
| 10 | 0.2174 | 1.3180 |

### 2.3. R-chart

In NWED, we must find two sampling distribution sample range limits with a probability content of 0.9973 . We must symbolically place L and U in such a way that $P(L \leq R \leq U)=0.9973$ (2.9)
where, $R$ is the sample range of size $n$. The percentiles discovered by simulation are shown in Table 5 .
Table 5: Percentiles of Range in NWED

| n | 0.99865 | 0.9950 | 0.99 | 0.975 | 0.95 | 0.05 | 0.025 | 0.01 | 0.005 | 0.00135 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 2 | 17.1603 | 13.8253 | 11.4034 | 8.6463 | 6.6313 | 0.0392 | 0.0916 | 0.0078 | 0.0040 | 0.0015 |
| 3 | 19.0499 | 15.5295 | 13.0925 | 10.4110 | 8.3303 | 0.1877 | 0.1267 | 0.0800 | 0.0533 | 0.0280 |
| 4 | 19.8896 | 16.9107 | 14.5485 | 11.5053 | 9.3612 | 0.3294 | 0.2588 | 0.1843 | 0.1440 | 0.0876 |
| 5 | 20.4410 | 17.2470 | 15.1729 | 12.2040 | 10.2067 | 0.4447 | 0.3609 | 0.2745 | 0.2241 | 0.1519 |
| 6 | 20.6521 | 17.8227 | 15.9946 | 12.8092 | 10.7383 | 0.5391 | 0.4561 | 0.3738 | 0.3174 | 0.2443 |
| 7 | 21.0641 | 18.3910 | 16.4794 | 13.5350 | 11.2981 | 0.6213 | 0.5390 | 0.4488 | 0.3976 | 0.3142 |
| 8 | 21.5317 | 18.5142 | 16.7579 | 13.7757 | 11.6490 | 0.6820 | 0.6028 | 0.5141 | 0.4471 | 0.3576 |
| 9 | 21.9174 | 19.0499 | 17.0796 | 14.1892 | 12.0253 | 0.7452 | 0.6529 | 0.5597 | 0.5068 | 0.4228 |
| 10 | 22.2275 | 19.2366 | 17.3187 | 14.5794 | 12.3440 | 0.7988 | 0.7074 | 0.6108 | 0.5641 | 0.4801 |

The sample range's control limits are calculated in the following way, using the percentiles from the above table. Take a look at the R distribution

$$
\begin{equation*}
P\left(Z_{0.00135} \leq R \leq Z_{0.99865}\right)=0.9973 \tag{2.10}
\end{equation*}
$$

From equation (2.10), for the $i^{\text {th }}$ subgroup range we can have

$$
P\left(Z_{0.00135} \cdot \frac{\bar{R}}{\alpha_{(n)}-\alpha_{(1)}} \leq R_{i} \leq Z_{0.99865} \cdot \frac{\bar{R}}{\alpha_{(n)}-\alpha_{(1)}}\right)=0.9973
$$

This can be expressed as:
$P\left(\mathrm{D}_{3 \mathrm{p}}^{*} \cdot \bar{R} \leq R_{i} \leq \mathrm{D}_{4 \mathrm{p}}^{* *} \cdot \bar{R}\right)=0.9973$
where, $\bar{R}$ is mean of ranges, $R_{i}$ is $i^{\text {th }}$ subgroup range. Thus $\mathrm{D}_{3 \mathrm{p}}^{*}=\frac{Z_{0.00135}}{\alpha_{(n)}-\alpha_{(1)}}, \mathrm{D}_{4 \mathrm{p}}^{* *}=\frac{Z_{0.99865}}{\alpha_{(n)}-\alpha_{(1)}}$ are the percentile
constants of $R$ chart for NWED process data and are given in table 6.
Table 6: Percentile Constants of Range chart

| $n$ | $D_{3 P}^{*}$ | $D_{4 \mathrm{P}}^{* *}$ |
| :---: | :---: | :---: |
| 2 | 0.0027 | 184.3956 |
| 3 | 0.0611 | 209.2084 |
| 4 | 0.1730 | 43.8671 |
| 5 | 0.2256 | 47.9693 |
| 6 | 0.3245 | 37.3675 |
| 7 | 0.3718 | 56.5452 |
| 8 | 0.4158 | 39.7781 |
| 9 | 0.4551 | 60.6146 |
| 10 | 0.5033 | 41.8627 |

## 2.4. $\sigma$-chart.

We would find two sampling distribution limits of sample standard deviation in NWED that have a probability content of 0.9973. Symbolically, we must locate L and U in such a way that
$P(L \leq s \leq U)=0.9973$ (2.13)
where, $s$ denotes the standard deviation of an $n$ sized sample. Table 7 shows the percentiles discovered by simulation.
Table 7: Percentiles of Standard Deviation in NWED

| n | 0.99865 | 0.9950 | 0.99 | 0.975 | 0.95 | 0.05 | 0.025 | 0.01 | 0.005 | 0.00135 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 2 | 8.5801 | 6.9126 | 5.7017 | 4.3232 | 3.3157 | 0.0196 | 0.0098 | 0.0039 | 0.0020 | 0.0008 |


| 3 | 9.0695 | 7.1137 | 6.1452 | 4.8797 | 3.9081 | 0.0759 | 0.0546 | 0.0365 | 0.0271 | 0.0137 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 4 | 8.4848 | 6.8710 | 6.0247 | 4.8608 | 3.9407 | 0.1297 | 0.1000 | 0.0731 | 0.0588 | 0.0400 |
| 5 | 8.2370 | 6.7509 | 5.8609 | 4.8732 | 3.9621 | 0.1612 | 0.1310 | 0.1029 | 0.0863 | 0.0575 |
| 6 | 7.9674 | 6.6951 | 5.8930 | 4.9713 | 4.1353 | 0.1928 | 0.1622 | 0.1275 | 0.1068 | 0.0819 |
| 7 | 7.1379 | 6.3462 | 5.6742 | 4.4920 | 3.9511 | 0.2083 | 0.1787 | 0.1504 | 0.1276 | 0.1063 |
| 8 | 7.2090 | 6.2343 | 5.6070 | 4.7051 | 3.9760 | 0.2282 | 0.2013 | 0.1714 | 0.1493 | 0.1209 |
| 9 | 7.2138 | 6.0690 | 5.5177 | 4.5954 | 3.9098 | 0.2464 | 0.2142 | 0.1875 | 0.1658 | 0.1364 |
| 10 | 6.7540 | 5.8264 | 5.2779 | 4.4744 | 3.8730 | 0.2593 | 0.2273 | 0.1969 | 0.1834 | 0.1519 |

The sample standard deviation control limits are determined in the following way using the percentiles from the above table. Consider the distribution of $s$
$P\left(Z_{0.00135} \leq s \leq Z_{0.99865}\right)=0.9973$
But standard deviation of sampling distribution when $\alpha=0.2$ and $\lambda=0.6$ is 3.125 for NWED. From equation (2.14), for the $i^{\text {th }}$ subgroup standard deviation we can have
$P\left(Z_{0.00135} \cdot \frac{\bar{s}}{3.125} \leq s_{i} \leq Z_{0.99865} \cdot \frac{\bar{s}}{3.125}\right)=0.9973$
This can be written as
$P\left(\mathrm{~B}_{3 \mathrm{p}}^{*} \cdot \bar{s} \leq s_{i} \leq \mathrm{B}_{4 \mathrm{p}}^{* *} \cdot \bar{s}\right)=0.9973$
where, $\bar{s}$ is mean of standard deviations, $s_{i}$ is $i^{\text {th }}$ subgroup standard deviation. Thus $\mathrm{B}_{3 \mathrm{p}}^{*}=\frac{Z_{0.00135}}{3.125}$ and $\mathrm{B}_{4 \mathrm{p}}^{* *}=\frac{Z_{0.99865}}{3.125}$ are
the constants of standard deviation chart for NWPD process data given in table 8.
Table 8: Percentile Constants of SD chart


## III. Comparative Study

The population represented by NWED is used to plot the average, median, range and standard deviation constants of the indicators previously described in Section 2 in the control chart. To use it for data, you need to ensure that the data is correctly compliant with NWED. As a result, the strength of the control limit can be estimated by comparing the use of the control limit with the Shewhart limit of the actual NWED data. With this in mind, we simulate the random subset of size $\mathrm{n}=2(1) 10$ in NWED and use the constants in Section 2 to calculate the control limits of the average, median, range, and standard deviation throughout the process. To create this comparative study, the number of readings within each control limit is used to calculate the probability of NWED coverage. Similar to NWED coverage probability, Shewhart coverage probability is calculated by counting the number of readings in certain individual control ranges by using Shewhart constants in quality tables $9,10,11$, and 12 that indicate coverage probability. For the two scenarios, NWED and Shewhart's real limitations.

Table 9: Probabilities of Coverage for a Mean-chart

|  | Shewart Limits |  |  | NWED Percentile Limits |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| n | $\bar{x}-\mathrm{A}_{2} \overline{\mathrm{R}}$ | $\bar{x}+\mathrm{A}_{2} \overline{\mathrm{R}}$ | Coverage probability | $\mathrm{A}_{2 \mathrm{P}}^{*} \times \bar{x}$ | $\mathrm{~A}_{2 \mathrm{P}}^{* *} \times \bar{x}$ | Coverage probability |
| 2 | 0 | 4.4734 | 0.9458 | 0.1753 | 5.0282 | $\mathbf{0 . 9 6 1 2}$ |
| 3 | 0 | 4.0013 | 0.9448 | 0.2990 | 4.2447 | $\mathbf{0 . 9 5 3 9}$ |
| 4 | 0 | 3.7133 | 0.9521 | 0.2133 | 3.3676 | 0.9335 |


| 5 | 0 | 3.6134 | 0.9566 | 0.2379 | 3.1400 | 0.9261 |
| :---: | :--- | :--- | :--- | :--- | :--- | :--- |
| 6 | 0 | 3.5620 | 0.9611 | 0.2528 | 3.0747 | 0.9263 |
| 7 | 0 | 3.3812 | 0.9631 | 0.2642 | 2.6579 | 0.8988 |
| 8 | 0 | 3.3505 | 0.9684 | 0.2737 | 2.5949 | 0.8980 |
| 9 | 0 | 3.3103 | 0.9710 | 0.2939 | 2.4710 | 0.8845 |
| 10 | 0 | 3.2339 | 0.9749 | 0.2988 | 2.3005 | 0.8655 |

Table 10: Probabilities of Coverage for a Median-chart

|  | Shewart limits |  | NWED Percentile limits |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| n | $\bar{m}-\mathrm{A}_{7} \overline{\mathrm{R}}$ | $\bar{m}+\mathrm{A}_{7} \overline{\mathrm{R}}$ | Coverage probability | $\mathrm{A}_{7 \mathrm{P}}^{*} \times \bar{m}$ | $\mathrm{~A}_{7 \mathrm{P}}^{*} \times \bar{m}$ | Coverage probability |
| 2 | 0 | 4.4734 | 0.9458 | 0.2529 | 7.2541 | $\mathbf{0 . 9 9 0 5}$ |
| 3 | 0 | 3.6426 | 0.9680 | 0.1769 | 4.4965 | $\mathbf{0 . 9 7 9 2}$ |
| 4 | 0 | 3.4482 | 0.9792 | 0.2007 | 3.0621 | 0.9704 |
| 5 | 0 | 3.3108 | 0.9856 | 0.1740 | 2.4927 | 0.9708 |
| 6 | 0 | 3.3417 | 0.9917 | 0.1903 | 2.2705 | 0.9712 |
| 7 | 0 | 3.1881 | 0.9938 | 0.1773 | 1.7738 | 0.9726 |
| 8 | 0 | 3.1984 | 0.9974 | 0.1931 | 1.5436 | 0.9590 |
| 9 | 0 | 3.1889 | 0.9970 | 0.1880 | 1.5048 | 0.9763 |
| 10 | 0 | 3.1642 | 0.9994 | 0.1988 | 1.2050 | 0.9475 |

Table 11: Probabilities of Coverage for a Range-chart

|  | Shewart limits |  |  |  | NWED Percentile limits |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| n | $D_{3} \bar{R}$ | $D_{4} \bar{R}$ | Coverage probability | $D_{3 \mathrm{P}}^{*} \times \bar{R}$ | $D_{4 \mathrm{P}}^{* *} \times \bar{R}$ | Coverage probability |
| 2 | 0 | 5.0762 | 0.9136 | 0.0041 | 286.5137 | $\mathbf{0 . 9 9 5 0}$ |
| 3 | 0 | 6.0636 | 0.8902 | 0.1438 | 492.6438 | $\mathbf{0 . 9 6 6 5}$ |
| 4 | 0 | 6.7533 | 0.8819 | 0.5119 | 129.8202 | 0.8513 |
| 5 | 0 | 7.2918 | 0.8690 | 0.7778 | 165.3836 | 0.7501 |
| 6 | 0 | 7.8064 | 0.8611 | 1.2640 | 145.5650 | 0.7283 |
| 7 | 0.3281 | 8.3078 | 0.8655 | 1.6054 | 244.1620 | 0.7304 |
| 8 | 0.6326 | 8.6707 | 0.8330 | 1.9341 | 185.0357 | 0.7465 |
| 9 | 0.9166 | 9.0473 | 0.7572 | 2.2672 | 301.9819 | 0.7489 |
| 10 | 1.1789 | 9.3948 | 0.7619 | 2.6608 | 221.3238 | 0.7 |

Table 12: Probabilities of Coverage for a SD-chart

|  | Shewart limits |  |  | NWED Percentile limits |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| n | $B_{3} \bar{S}$ | $B_{4} \bar{S}$ | Coverage probability | $B_{3 \mathrm{P}}^{*} \times \bar{S}$ | $B_{4 \mathrm{P}}^{* *} \times \bar{S}$ | Coverage probability |
| 2 | 0 | 2.5380 | 0.9136 | 0.0001 ? | 2.1329 | 0.8865 |
| 3 | 0 | 2.7207 | 0.8896 corch in | 0.0046 | 3.0748 | 0.9128 |
| 4 | 0 | 2.7412 | 0.8784 | 0.0154 | 3.2845 | 0.9176 |
| 5 | 0 | 2.8183 | 0.8689 | 0.0248 | 3.5560 | 0.9318 |
| 6 | 0.0440 | 2.8928 | 0.8642 | 0.0384 | 3.7439 | 0.9297 |
| 7 | 0.1752 | 2.7954 | 0.8346 | 0.0505 | 3.3927 | 0.9165 |
| 8 | 0.2895 | 2.8403 | 0.7409 | 0.0605 | 3.6100 | 0.9290 |
| 9 | 0.3870 | 2.8515 | 0.7176 | 0.0707 | 3.7381 | 0.9405 |
| 10 | 0.4679 | 2.8275 | 0.7116 | 0.0800 | 3.5612 | 0.9311 |

## IV. SumMARY \& CONCLUSIONS

In statistical quality control problems the Shewhart constants are used irrespective of the data follows a normal distribution or else. If the given data does not follow normal distribution, the applications of Shewhart constants for any statistical quality control problems are at stake.

The above tables show that when the Shewhart limits are used in a mechanical way and if the data follows a true NWED, the decision of process variation coefficient for mean, median, range and standard deviation charts will be
reduced. So, in order to overcome such a problem, the control constants given at various sample sizes in the above tables are more preferable to Shewhart constants in statistical quality control.

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