

Investigating the Quality Performance of Production of Some Selected Drinks using Hotelling T-square and Control Chart

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Abstract

Consumers make complaint about the state of home-made goods, in fact many claim that foreign goods are of high quality compared to home-made goods. We discovered that many of our indigenous industries are no more in existence and so this brought the desire to carry out this research work so as to find out whether products from our indigenous brewery industry fall within the lay-down acceptable standard that is devoid of the consumers' complaint

Index terms— control charts, cumulative sum technique charts, hotellings t-square, fill height,

Introduction n many manufacturing firms where there exists mass production, measurement made on each product is subject to error due to variation from one item to the other. Since there must be variations, it becomes important to study and determine when any observed variation is significant or not. This is the reason why the Federal Government of Nigeria came up with legislations to protect the buyers from buying inferior goods. Increase in consumer buying behaviour towards some selected drinks will directly affect the production of such drinks in our breweries industry. Quality control relies partly upon patronage and some other reliable factors, in beer production process, the measurement of attributes such as fill height and level of co 2 is of paramount important and that is the reason why quality control is evolving in developing systems to ensure standard products or services as well as meeting or exceeding customer's requirements. Walter Shewhart introduced the concept of statistical quality control thereby controlling quality of mass produced goods. Shewhart believed that variation always exists in manufactured products and that the variation can be studied, monitored and controlled using Statistics. Walter Shew hart explained the theories about using statistical quality control charts to improve quality and productivity in which case he developed fourte en points agenda for companies to improve quality and productivity, reduce costs and compete effectively in the world market.

1 II.

2 Literature Review

Reeves and Bednar (1994) define quality as excellence, value, conformance to specifications, and meeting or exceeding customers' expectation. The term "fitness for use" defined by ??uran (1974) is also included in the quality definition presented by ??eeves and Bednar (1994). Thus, the customer perspective with respect to quality is the master key that should be understood while determining any term for quality or definition of quality. Deming, W.E ??1986). worked on Quality and Productivity Improvement using acceptance sampling method, and he was able to obtain increase in quality and simultaneous reduction in the cost of reducing waste, re write staff attrition and litigation while increasing customer's loyalty. Farhat, B. A. and Al-Darrab, I. (1998). Total quality management is now established and widely used management process. One of its associated features is the application of statistical quality control techniques. A quality product or service is one that meets the customer's needs and provides the value that they want and expect. They are also of the opinion that quality management is a formal approach to management in which the overriding priority of the organization is to deliver a quality product or service and to work towards excellence and continuous improvement in everything it does.

Quality can be viewed from the perspectives of design and product in which case; design quality is the different grades or levels of performance, reliability, I serviceability and function that are the results of deliberate

be used in the multivariate normal distribution case. Suppose that we have p variables, given by $x_1, x_2, \dots, x_p$. Arrange these variables in a p -component vector $x' = [x_1, x_2, \dots, x_p]$. Let $\bar{x} = [\bar{x}_1, \bar{x}_2, \dots, \bar{x}_p]$ be the vector of the means of the x 's, and let the variances and covariances of the random variables in x be contained in a $p \times p$ covariance matrix Σ .

The main diagonal elements of Σ are the variances of the x 's and the off-diagonal elements are the covariances. Now the squared standardized (generalized) distance from x to $\bar{x}$ is $(x - \bar{x})' \Sigma^{-1} (x - \bar{x})$ (3)

The multivariate normal density function is obtained simply by replacing the standardized distance in equation (2) by the multivariate generalized distance in equation (3) and changing the constant term to a more general form that makes the area under the probability density function unity regardless of the value of p . Therefore, the multivariate normal probability density function is
$$f(x) = \frac{1}{(2\pi)^{p/2} |\Sigma|^{1/2}} \exp\left\{-\frac{1}{2}(x - \bar{x})' \Sigma^{-1} (x - \bar{x})\right\}$$
 (4) where $|\Sigma| < \infty$, $j = 1, 2, \dots, p$.

A multivariate normal distribution for $p = 2$ variables (called a bivariate normal) is
$$f(x_1, x_2) = \frac{1}{2\pi |\Sigma|^{1/2}} \exp\left\{-\frac{1}{2}(x - \bar{x})' \Sigma^{-1} (x - \bar{x})\right\}$$
 (5)

6 c) The Sample Mean Vector and Covariance Matrix

Suppose that we have a random sample from a multivariate normal distribution—say, where the i th sample vector contains observations on each of the p variables $x_{i1}, x_{i2}, \dots, x_{ip}$. Then the sample mean vector is
$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$
 (6)

and the sample variance is
$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$
 (7)

and the sample covariance is The Hotelling T^2 chart is the analog of the Shewhart $\bar{x}$ chart. Multivariate control charts work well when the number of process variables is not too large—say, 10 or fewer. As the number of variables grows, however, traditional multivariate control charts lose efficiency with regard to shift detection. A multivariate approach should be used to monitor process stability with more than one important characteristic. This approach can account for correlations between characteristics and will control the overall probability of falsely signaling a special cause of variation when one is not present. The most common multivariate chart is the T^2 chart. There are many situations in which the simultaneous monitoring or control of two or more $\bar{x}$ charts is required. For example, if $\bar{x}_1$ and $\bar{x}_2$ are the means of two quality characteristics, then the simultaneous monitoring or control of $\bar{x}_1$ and $\bar{x}_2$ is required. This is done by using the Hotelling T^2 chart. There are two versions of the Hotelling T^2 charts which are Subgrouped data and Individual observations.

7 e) Subgrouped Data

Suppose that p quality characteristics $x_1, x_2, \dots, x_p$ are jointly distributed according to the multivariate normal distribution (see equation 3.6.4). Let $\mu_1, \mu_2, \dots, \mu_p$ be the mean values of the quality characteristics and let σ_{jk} 's represent the variance-covariance values of the p characteristics. In practice, it is usually necessary to estimate μ and Σ from the preliminary samples of size n , taken when the process is assumed to be in control. Suppose that m such samples are available. The sample means and variances are calculated from each sample as usual; that is,
$$\bar{x}_j = \frac{1}{n} \sum_{i=1}^n x_{ij}$$
 (9)
$$s_j^2 = \frac{1}{n-1} \sum_{i=1}^n (x_{ij} - \bar{x}_j)^2$$
 (10)

where x_{ij} is the i th observation on the j th quality characteristics in the k th sample. The covariance between quality characteristic j and quality characteristic h in the k th sample is
$$s_{jh} = \frac{1}{n-1} \sum_{i=1}^n (x_{ij} - \bar{x}_j)(x_{ih} - \bar{x}_h)$$
 (11)

8 $\bar{\bar{x}} = 1$

The statistics $\bar{\bar{x}}_j$, s_j^2 and s_{jh} are then averaged over all m samples to obtain
$$\bar{\bar{x}}_j = \frac{1}{m} \sum_{k=1}^m \bar{x}_{jk}$$
 (12)
$$\bar{s}_j^2 = \frac{1}{m} \sum_{k=1}^m s_{jk}^2$$

and
$$\bar{s}_{jh} = \frac{1}{m} \sum_{k=1}^m s_{jhk}$$
 (13)

The $\bar{\bar{x}}_j$ are the elements of the vector $\bar{\bar{x}}$, and the $p \times p$ average of sample covariance matrices S is formed as
$$\bar{S} = \frac{1}{m} \sum_{k=1}^m S_k$$
 (15)

To use the T^2 Control Chart, we will use the test statistics; $T^2 = n(\bar{\bar{x}} - \mu)' \bar{S}^{-1} (\bar{\bar{x}} - \mu)$

Control chart. This is a directionally invariant control chart; that is, its ability to detect a shift in the mean vector only depends on the magnitude of the shift, and not in its direction. There are two distinct phases of control chart usage. PHASE I is the use of the charts for establishing control; that is, testing whether the process is in control when the m preliminary subgroups are drawn. The control limit for T^2 control chart are given by
$$T^2_{\alpha} = \frac{p(n-1)}{n-p} F_{\alpha, p, n-p}$$
 (17)

Phase II is the use of the chart for monitoring future production, sample size of at least $n = 200$ is needed. The control limits are as follows: f) Individual Observation Here, multivariate control charts with subgroup size, $n = 1$ is of interest. Suppose that m samples, each of size $n = 1$, are available and that p is the number of quality characteristics observed in each sample. Let $\bar{x}$ and S be the sample mean vector and covariance matrix, respectively, of these observations. The Hotelling T^2 statistic in equation becomes
$$T^2 = (n-1) \bar{x}' S^{-1} \bar{x}$$
 (19)

The phase II control limits for this statistic are $\bar{X} \pm 3\sigma/\sqrt{n} = \bar{X} \pm 3\sigma/\sqrt{m+1} \approx \bar{X} \pm 3\sigma/\sqrt{m}$ (20)

When the number of preliminary samples m is large, say $m > 100$, most practitioners use an approximate control limit, either $\bar{X} \pm 3\sigma/\sqrt{m}$ (21) or $\bar{X} \pm 3\sigma/\sqrt{m+1}$ (22)

However, for $m > 100$, equation (21) is a reasonable approximation.

For phase I, the limits are based on a beta distribution, $\bar{X} \pm 3\sigma/\sqrt{m} = \bar{X} \pm 3\sigma/\sqrt{m+1} \approx \bar{X} \pm 3\sigma/\sqrt{m}$ (23)

Where $1 - \alpha/2$, $(1 - \alpha/2)$ is the upper α percentage point of a beta distribution with parameters $p/2$ and $(m-p-1)/2$. Approximations to the phase I limit based on the F and chi-square distributions are likely to be inaccurate. Basically, the focus will be on the Sub grouped data because it suits the type of data that was collected.

9 Control Chart for Monitoring Variability

Monitoring multivariate process are in two levels, which are to monitor the process mean vector μ and to monitor process variability. Process variability is summarized by the $p \times p$ covariance matrix Σ . The main diagonal elements of this matrix are the variances of the individual process variables, and the off-diagonal elements are the covariances. We can use the approach based on the sample generalized variance, $|S|$. This statistic, which is the determinant of the sample covariance matrix, is a widely used measure of multivariate dispersion. Another method would be to use the mean and variance of $|S|$, that is, $E(|S|)$ and $V(|S|)$, and the property that most of the probability distribution of $|S|$ is contained in the interval

10 $|S| \pm 3\sqrt{V(|S|)}$.

It can be shown that $E(|S|) = \frac{1}{n} \sum_{i=1}^n \sigma_{ii}$ (24)

where $\sigma_{ii} = \frac{1}{n} \sum_{j=1}^n \sigma_{ij}^2$ (25)

and $\sigma_{ij} = \frac{1}{n} \sum_{k=1}^n (x_{ik} - \bar{x}_i)(x_{jk} - \bar{x}_j)$ (26)

Therefore, the parameters of the control charts for $|S|$ would be Investigating the Quality Performance of Production of Some Selected Drinks using Hotelling T-square and Control Chart (g) $|S| \pm 3\sqrt{V(|S|)}$ (27)

The lower control limit in equation (27) is replaced with zero if the calculated value is less than zero. In practice, σ usually will be estimated by a sample covariance matrix S , based on the analysis of preliminary samples. If this is the case, we should replace $|S|$ in equation (27) In this study, two measurement quality characteristics are being analyzed using Multivariate statistical quality control.

Fill height: It measures the level of liquid in a bottle of drink. The products under study are STAR, MALTINA and GOLDBERG from Nigerian Breweries plc. The standard is always at 60cl. Co 2 level: It measures the level of co 2 in each bottle. The target for corking a bottle of STAR is between (0.52-0.54%wt/wt), that of MALTINA is (0.59-0.61%wt/wt) and GOLDBERG is (0.62-0.64%wt/wt) where %wt/wt means weight per weight.

Data Presentation: The data used for this analysis is shown in the appendix 'A to appendix F.

11 IV.

12 Data Analysis and Results

In this chapter, the Hotelling T 2 control chart is used for the analysis of fill height and level of co 2 measurements of Star, Maltina and Goldberg using R.

13 Analysis on the Fill Height Measurement and co 2 Level of Star

The fill height of STAR refers to the height of the liquid content in a bottle of a STAR. And the co 2 level refers to the level of co 2 in each bottle of STAR. There can be cases of low fill, high fill and normal fill. The normal or standard fill height of STAR of the company is 60cl. And the standard co 2 level of STAR is between (0.52-0.54%wt/wt). The tables below display analysis carried out using R on various readings on fill height and co 2 level that was observed at different times. Table1.0 shows the Variances and Covariances of the fill height and level of CO 2 of STAR and also the Hotelling T 2 and Variability of each of the 20 samples. The Grand mean, Variance-Covariance Matrix (s) for the control limit used in the Variability plot, and the control Limits for the Hotelling T 2 and Variability plot are represented in the table(s) below.

14 Grand Mean

15 h) Interpretation of Star Chart

From the Variability plot above, most of the sample variances are on or close to the lower control limit (LCL) while they are very far from the upper control limit, which means that the variability (the variances of the observation from the mean) is in control. Thus, the Hotelling T 2 can be plotted to see if the process is actually

in control. From the Hotelling T 2 plotted above also, it can be seen that all the plot point fall within the UCL and LCL, which means that it can be concluded that the fill height and level of Co 2 of STAR is under control. The R code was used for the analysis of STAR.

16 Analysis on the Fill Height Measurement and co 2 Level of Maltina.

The fill height of MALTINA refers to the height of the liquid content in a bottle of a MALTINA. And the co 2 level refers to the level of co 2 in each bottle of MALTINA. There can be cases of low fill, high fill and normal fill. The normal or standard fill height of MALTINA of the company is 60cl. And the standard co 2 level of MALTINA is between (0.59-0.61%wt/wt). The tables below display analysis carried out using R on various readings on fill height and co 2 level that was observed at different times. Table2 shows the Variances and Covariances of the fill height and level of Co 2 of MALTINA and also the Hotelling T 2 and Variability of each of the 20 samples. The Grand mean, Variance-Covariance Matrix (s) for the control limit used in the Variability plot, and the control Limits for the Hotelling T 2 and Variability plot are represented in the table(s) below. From the Hotelling T 2 plotted above, it can be seen that all the plot point fall within the UCL and LCL, which means that the fill height and level of co 2 of MALTINA is under control.

17 Grand Mean

18 Variance-Covariance Matrix (S) for the control limit used in the variability plot

19 Analysis on the Fill Height Measurement and Co 2 Level of Goldberg

The fill height of GOLDBERG refers to the height of the liquid content in a bottle of a GOLDBERG. And the co 2 level refers to the level of co 2 in each bottle of GOLDBERG. There can be cases of low fill, high fill and normal fill. The normal or standard fill height of GOLDBERG of the company is 60cl. And the standard level of co 2 GOLDBERG is between (0.62-0.64%wt/wt). The tables below display analysis carried out using R on various readings on fill height and co 2 level that was observed at different times. From the Hotelling T2 plotted above also, it shows that all the plot point fall within the UCL and LCL, which means that the fill height and level of Co of GOLDBERG is under control.

Based on the results obtained from the analysis so far for all the drinks considered, none of the characteristics examined and analyzed fall within the control which invariably means we do not have sufficient evidence to reject the null hypothesis hence we Accept the null hypothesis for both the fill height and the level of Co2.

V.

20 Conclusion

The results obtained from the method used show that the components for the production of the beer under consideration(fill height and level of Co2) shows that the variability of the three products are in control, and this information helped in proceeding to check if the two quality characteristics are in control, also, using the Hotelling T2 control chart of Sub grouped data, the values were all within the lower and upper control limit for the three products, which helps to affirm the fact that the quality characteristics of STAR, MALTINA AND GOLDBERG are in control. This shows that the Quality Control Unit of the Company should not relent in carrying out their test on the products, all these will help the company to maintain the required standard and survive competition with other likely products from other company.

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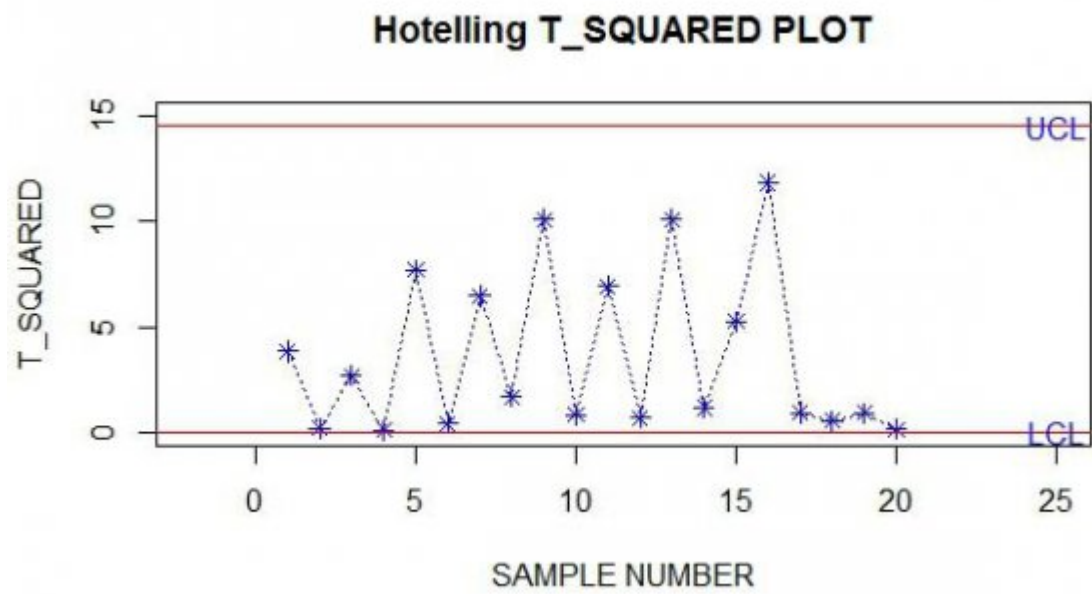


Figure 1:

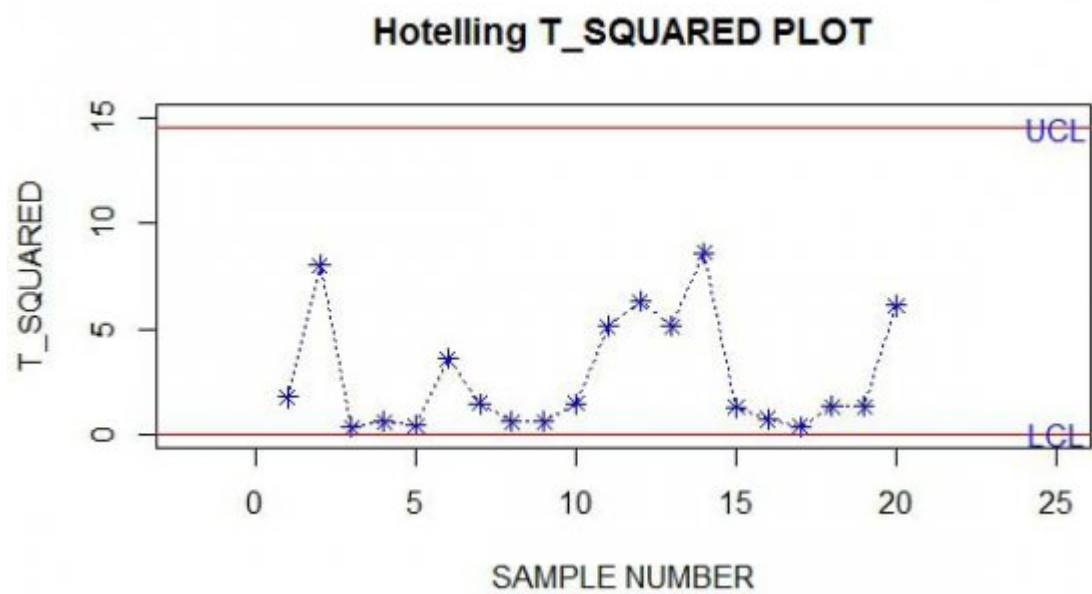


Figure 2:

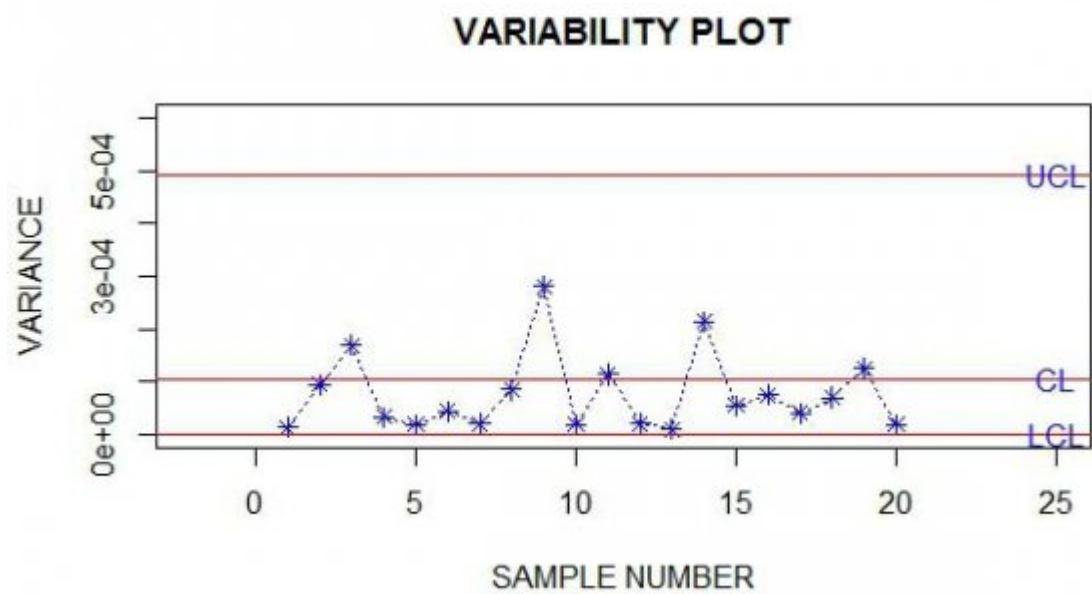
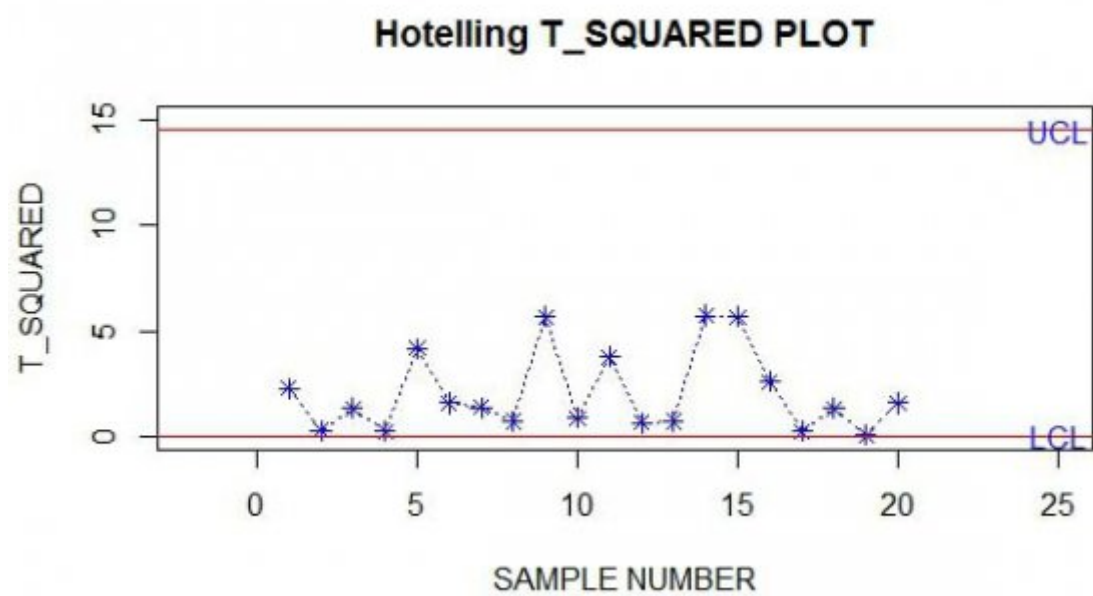


Figure 3:



1

Figure 4: ControlFigure 1 :

1

Sample Number	Means	Level of CO 2	Variance and Covariances		Control Chart Statistics		S k
	Fill Height (?? ? 1k)		S 2 1k	S S 12k 2 2k	Hotelling T 2 k		
1	60.0	0.518	1.5 0.00037	0.0200	3.8339370 0.00015500		
2	59.6	0.530	1.3 0.00035	0.0100	0.1962245 0.00035500		
3	59.8	0.520	0.7 0.00005	0.0000	2.6772350 0.00003500		
4	60.0	0.530	1.0 0.00010	0.0025	0.1181133 0.00009375		
5	59.8	0.546	2.2 0.00013	0.0015	7.7073474 0.00028375		
6	60.0	0.526	1.0 0.00013	-0.0025	0.4357597 0.00012375		
7	59.4	0.516	1.3 0.00008	-0.0005	6.4532623 0.00010375		
8	60.0	0.522	1.5 0.00007	0.0075	1.6743676 0.00004875		
9	60.0	0.548	2.5 0.00012	0.0000	10.0856025 0.00030000		
10	59.4	0.528	0.3 0.00017	-0.0065	0.8187689 0.00000875		
11	60.0	0.514	1.0 0.00013	-0.0025	6.9144678 0.00012375		
12	59.6	0.534	0.8 0.00003	-0.0030	0.6801556 0.00001500		
13	60.0	0.548	1.5 0.00017	0.0000	10.0856025 0.00025500		
14	59.8	0.536	2.7 0.00053	-0.0285	1.1678815 0.00061875		
15	60.0	0.516	1.0 0.00023	-0.0125	5.2590822 0.00007375		
16	59.6	0.550	0.3 0.00005	-0.0025	11.8254949 0.00000875		
17	60.0	0.524	2.5 0.00003	0.0025	0.9399435 0.00006875		
18	59.8	0.534	1.7 0.00008	-0.0015	0.5507094 0.00013375		
19	59.8	0.524	0.7 0.00013	0.0035	0.9184544 0.00007875		
20	60.0	0.528	1.5 0.00057	-0.0225	0.1618163 0.00034875		
Average	59.83	0.5296	1.5 0.00057	-0.0225			

Figure 5: Table 1 :

2

Sample Number	Means	Variances and Covariances		Control Chart Statistics		S k
	Fill Height (?? ? 1k)	Level of CO 2	(?? ? 2k)	S 2 1k S S 12k 2 2k	Hotelling T 2 k	
1	60.0	0.596	1.0 0.00003	0.0025	1.7606748	0.0
2	59.6	0.590	1.3 0.00010	0.0100	8.0208486	0.0
3	59.8	0.598	1.2 0.00007	0.0070	0.3296529	0.0
4	60.0	0.598	0.5 0.00017	0.0075	0.6161894	0.0
5	59.6	0.598	1.3 0.00002	-0.0010	0.4235547	0.0
6	60.0	0.594	0.5 0.00003	0.0025	3.5875338	0.0
7	59.4	0.602	1.3 0.00002	-0.0035	1.4622119	0.0

Figure 6: Table 2 :

Sample Number	Fill Height (?? ? k 1k)	Means	Variances and Covariances			Control Chart Statistics		
		Level of CO ₂ (?? ? 2k)	S ²	S	S	Hotelling's T ²	Sing	k
1	60.0	0.642	1.0	0.00007	0.0075000	2.28910604	0.00001375	
2	59.6	0.636	1.3	0.00008	0.0030000	0.28786007	0.00009500	
3	59.8	0.630	0.7	0.00025	-0.0025000	1.31283293	0.00016875	
4	60.0	0.634	0.5	0.00008	0.0025000	0.25264057	0.00003375	
5	59.6	0.626	0.3	0.00013	-0.0045000	4.15963167	0.00001875	
6	60.0	0.630	1.0	0.00010	-0.0075000	1.58110868	0.00004375	
7	59.4	0.638	0.8	0.00007	0.0060000	1.37548756	0.00002000	
8	60.0	0.632	0.5	0.00037	0.0010000	0.72131622	0.00008500	
9	60.0	0.646	3.5	0.00008	0.0000000	5.65403962	0.00028000	
10	59.4	0.634	1.3	0.00003	-0.0045000	0.88317403	0.00001875	
11	60.0	0.644	1.5	0.00008	-0.0025000	3.77601443	0.00011375	
12	59.6	0.632	0.3	0.00007	0.0010000	0.66321829	0.00002000	
13	60.0	0.632	0.5	0.00002	0.0000000	0.72131622	0.00001000	
14	59.8	0.646	2.7	0.00008	0.0015000	5.72751642	0.00021375	
15	60.0	0.646	1.0	0.00008	0.0050000	5.65403962	0.00005500	
16	59.6	0.628	1.3	0.00007	0.0040000	2.60304373	0.00007500	
17	60.0	0.634	0.5	0.00008	0.0000000	0.25264057	0.00004000	
18	59.8	0.630	0.7	0.00010	-1.1 x 10 ⁻²²	1.31283293	0.00007000	
19	59.8	0.634	0.7	0.00018	0.0010000	0.06980296	0.00012500	
20	60.0	0.630	0.5	0.00005	0.00250000	1.58110868	0.00001875	

Figure 7: Table 3 :

3

Investigating the Quality Performance of Production of Some Selected Drinks using Hotelling T-square and Control Chart

Grand Mean Variance-Covariance Matrix (S) for the control limit used in the variability plot Fill Height Lev

Level of CO ₂	0.001125	0.0001035
Control Limits for the Hotelling T ² and Variability plot		
	LCL	UCL
Hotelling T ²	0 -	14.52384
Variability	0 0.0001053	0.000492
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	nals	

Figure 8: Table 3

B

3	9:00am	59	60	60	61	59
4	10:00am	60	61	61	59	59
5	11:00am	60	59	60	58	62
6	12 noon	61	59	60	59	61
7	1:00pm	60	59	59	58	61
8	2:00pm	62	60	59	60	59
9	3:00pm	58	62	60	59	61
10	4:00pm	60	59	59	60	59
11	5:00pm	61	59	61	60	59
12	6:00pm	60	59	59	61	59
13	7:00pm	58	60	60	61	61
14	8:00pm	57	60	61	61	60
15	9:00pm	59	61	59	60	61
16	10:00pm	60	59	60	60	59
17	11:00pm	58	60	62	59	61
18	12:00am	59	60	59	62	59
19	1:00am	59	60	59	60	61
20	2:00am	60	60	59	59	62
Sample No	Time	A	B	C	D	E
1	7:00am	0.52	0.51	0.50	0.55	0.51
2	8:00am	0.55	0.50	0.54	0.53	0.53
3	9:00am	0.53	0.52	0.52	0.52	0.51
4	10:00am	0.54	0.54	0.52	0.52	0.53
5	11:00am	0.53	0.56	0.55	0.54	0.55
6	12noon	0.51	0.52	0.54	0.53	0.53
7	1:00pm	0.52	0.51	0.53	0.51	0.51
8	2:00pm	0.53	0.52	0.52	0.53	0.51
9	3:00pm	0.54	0.54	0.54	0.56	0.56
10	4:00pm	0.52	0.54	0.53	0.51	0.54
11	5:00pm	0.51	0.53	0.51	0.52	0.50
12	6:00pm	0.53	0.54	0.54	0.53	0.53
13	7:00pm	0.55	0.56	0.53	0.54	0.56
14	8:00pm	0.56	0.55	0.53	0.50	0.54
15	9:00pm	0.54	0.51	0.52	0.51	0.50
16	10:00pm	0.55	0.56	0.55	0.54	0.55
17	11:00pm	0.52	0.52	0.53	0.53	0.52
18	12:00am	0.54	0.54	0.52	0.53	0.54
19	1:00am	0.52	0.51	0.52	0.54	0.53
20	2:00am	0.51	0.54	0.53	0.56	0.50

[Note: Note: A, B, C, D, and E are the numbers of observations for each samples respectively.]

Figure 9: Table B :

D

Sample No.	Time	A	B	C	D	E
1	7:00am	0.60	0.60	0.59	0.60	0.59
2	8:00am	0.58	0.58	0.59	0.60	0.60
3	9:00am	0.59	0.60	0.60	0.61	0.59
4	10:00am	0.59	0.61	0.61	0.60	0.58
5	11:00am	0.60	0.60	0.60	0.60	0.59
6	12noon	0.59	0.59	0.59	0.60	0.60
7	1:00pm	0.60	0.60	0.61	0.60	0.60
8	2:00pm	0.60	0.61	0.60	0.59	0.59
9	3:00p m	0.59	0.61	0.60	0.59	0.60
10	4:00pm	0.60	0.60	0.60	0.60	0.61
11	5:00pm	0.62	0.61	0.61	0.60	0.60
12	6:00pm	0.61	0.61	0.61	0.60	0.61
13	7:00pm	0.61	0.62	0.60	0.61	0.60
14	8:00pm	0.62	0.60	0.60	0.61	0.62
15	9:00pm	0.60	0.61	0.60	0.61	0.60
16	10:00pm	0.61	0.59	0.60	0.61	0.60
17	11:00pm	0.60	0.60	0.60	0.60	0.61
18	12:00am	0.60	0.61	0.59	0.59	0.59
19	1:00am	0.59	0.60	0.58	0.60	0.61

Figure 10: Table D :

F

Year 2020
22
Volume XX
Issue X
Version I
() B

Global	Sample	Time	7:00am	A	0.65	B	0.64	C	0.64	D	0.65	E	0.63
Journal of	No 1	8:00am		0.63	0.61	0.63	0.63	0.64	0.65	0.65	0.62	0.63	0.64
Manage-	2 3 4	9:00am		0.64	0.63	0.64	0.62	0.64	0.61	0.63	0.64	0.62	0.63
ment and	5 6 7	10:00am		0.62	0.63	0.64	0.65	0.64	0.64	0.63	0.63	0.62	0.64
Business	8 9	11:00am		0.64	0.65	0.66	0.66	0.62	0.64	0.63	0.64	0.61	0.64
Research	10 11	12noon		0.64	0.64	0.64	0.63	0.63	0.65	0.63	0.65	0.63	0.65
	12 13	1:00pm		0.62	0.63	0.64	0.63	0.63	0.63	0.64	0.63	0.63	0.64
	14	2:00pm		0.64		0.66		0.64		0.64		0.65	
		3:00pm											
		4:00pm											
		5:00pm											
		6:00pm											
		7:00pm											
		8:00pm											
	15	9:00pm		0.64		0.66		0.64		0.65		0.64	
	16	10:00pm		0.63		0.62		0.62		0.63		0.64	
	17	11:00pm		0.64		0.64		0.64		0.62		0.63	
	18	12:00am		0.62		0.64		0.64		0.62		0.63	
	19	1:00am		0.62		0.64		0.65		0.62		0.64	
	20	2:00am		0.62		0.63		0.63		0.63		0.64	

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Figure 11: Table F :

.1 Appendix

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