

# Water Quality Analysis of Drinking Water by Identification of Their Distribution and Process Capability

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**Abstract:** The present study deals with the water quality management using Water Quality Index (WQI) and Distribution Identification model. The drinking water quality in Jaipur city has been used for the various physio-chemical parameters such as temperature, pH, turbidity, conductance, total hardness, alkalinity, magnesium, calcium, nitrate, chloride, fluoride, sodium, potassium. Drinking water suitability for domestic purposes was examined by WHO and BIS standards, which indicate the drinking water in almost all the areas, were not much suitable for drinking purpose. It is observed that Water Quality Index (WQI) is very high in Sanganer and VKI areas and Gamma distribution model fits on all the parameters for different sites of water data. Thus it was observed that potable water quality has degraded with respect to all the parameters and almost in all the Sites parameters.

**Keywords:** Water Physio-Chemical Parameters, WQI, Distribution Identification Model, Process Capability and Health Effects

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## 1. Introduction

Potable water or drinking water has always been an important and life sustaining drinks to humans and is essential to the survival of all living organisms. The potable water should be free from pathogenic agents and chemical constituents, pleasant to taste and usable for domestic purposes. Parameters for drinking water quality typically fall under two categories - Chemical and Physical. Chemical parameters causes chronic health risks through built up of heavy metals whereas physical parameters affect the aesthetics and taste of drinking water [10]. Water supply in Jaipur comes either from surface water or ground water sources. The ground water is characterized by multiple quality problems. Many hazardous pollutants viz., colored dyes, heavy metals, nitrates and fluoride, agriculture runoffs, waste water treatment plants, combined sewage overflows, storm water discharges, failed septic tanks, prescription drugs etc. Pollute it.

Water Quality Index may be defined as a rating, reflecting the composite influence on overall quality characteristics of water of individual quality parameters, which is being regarded as one of the most effective way to communicate

water quality.

This will give us only certain numerical value but for estimating exact quality of water an indexing system has been developed known as Water Quality Index, which gives us the idea of water of whole system [2].

Quality rating equal to zero means a complete absence of pollutants. While 0 to 100 implies that the pollutants are under consideration means within the prescribed standards. When quality rating is more than 100 it implies that the pollutants are above the standards. Similarly when WQI is 0 to 100 indicate that the water is under consideration and is safe for use.

Distribution Identification Model is used to evaluate the optimal distribution based on the probability plots and goodness-of-fit tests prior to conducting a capability analysis study [2].

Capability analysis study is used to produce a process capability report when data do not follow a normal distribution. The report includes a capability histogram overlaid with a distribution curve, and a table of overall capability statistics [7]. The report also includes statistics of the process data, such as the mean, distribution parameter estimates and process specifications; the actual overall

capability (Pp, Ppk, PPU, and PPL), and the observed and expected overall performance. The report can be used to visually assess the distribution of the process relative to the target, whether the data follow a specified distribution, and whether the process is capable of meeting the specifications consistently.

**Study Area:** The present study was carried out in Jaipur city which is situated in eastern part of Rajasthan surrounded on three sides by the rugged of Aravali hills. The city has total geographical area of 14,068 sq. km. (4.11 percent of the state). It is located at 26.92°N 75.82°E. It has an average elevation of 431 meters.

The climate of the city is hot semiarid with extremes of temperature (15-45°C) and rainfall 650 mm (26 inch.). In this city water samples of selected for 10 different areas such as VKI, Jhothwara, Chitrkoot, Vaishali Nagar, Transport Nagar, Kanota, Bari Chopar, Subhash Chowk, Gopalpura and Sanganer were used with the view to understand the quality of potable water.

## 2. Statistical Analysis

### 2.1. Analysis for Water Quality Index

The basic equation for Water Quality Index is  $W. Q. I. = \sum \frac{q_i w_i}{\sum w_i}$

Where  $q_i$  = quality ratings of parameters  
 $w_i$  = unit weight of different parameters  
 $q_i w_i$  = parameters sub-index

#### 2.1.1. Quality Ratings of the Parameters

Quality ratings  $q_i = \frac{[VO-VI]}{VS-VI} * 100$

Where VO = observed value of the parameter  
 VS = standard value of the parameter  
 VI = ideal value of the parameter  
 VI for pH = 7 and for dissolved oxygen = 14.6 mg/l  
 While VI is zero for other parameters.

#### 2.1.2. Unit Weight for Various Parameters

$W_i = K / S_i$   
 Where  $K = 1 \div 1/vs1 + 1/vs2 + 1/vs3 + \dots + 1/vsn$   
 $S_i$  = recommended standard value for the corresponding Parameter  
 $vsi$  = standard value for the corresponding parameter

### 2.2. Distribution Identification Model

By the help of Mini Tab (Software) Distribution Identification Model is used to evaluate the optimal distribution based on the probability plots and goodness-of-fit tests prior to conducting a capability analysis study.

#### 2.2.1. MLE

Maximum likelihood estimates of the parameters are calculated by maximizing the likelihood function with respect to the parameters. The likelihood function describes, for each set of distribution parameters, the chance that the

true distribution has the parameters based on the sample [5].

The Newton-Raphson algorithm is used to calculate maximum likelihood estimates of the parameters which define the distribution. The Newton-Raphson algorithm is a recursive method for computing the maximum of a function. The percentiles are calculated from that distribution.

#### 2.2.2. Pp

Pp is calculated using the parameter values estimated using maximum likelihood method for a distribution used in the analysis.

$$Pp = \frac{Zucl - Zlcl}{6}$$

Where:  $Zlsl = F^{-1}(p_1)$   
 $Zucl = F^{-1}(p_2)$   
 $F^{-1}(p) = p * 100$ th percentile of a standard normal distribution  
 $p_1 = \text{Prob}(X \leq LSL)$   
 $p_2 = \text{Prob}(X \leq USL)$

X follows the distribution used in the analysis

#### 2.2.3. PPL

PPL is calculated using the parameter values estimated using maximum likelihood method for a distribution used in the analysis.

$$PPL = \frac{-F^{-1}(p_1)}{3}$$

where:  $F^{-1}(p) = p * 100$ th percentile of a standard normal distribution

$p_1 = \text{Prob}(X \leq LSL)$   
 X follows the distribution used in the analysis

#### 2.2.4. PPU

PPU is calculated using the parameter values estimated using maximum likelihood method for a distribution used in the analysis.

$$PPU = \frac{F^{-1}(p_2)}{3}$$

Where:  $F^{-1}(p) = p * 100$ th percentile of a standard normal distribution

$p_2 = \text{Prob}(X \leq USL)$   
 X follows the distribution used in the analysis

#### 2.2.5. Ppk

The minimum of PPU and PPL

## 3. Result and Discussion

Table 1 and Table 2 give the detail of Quality Ratings, unit Weight and Water Quality Index of different areas for all quality parameters. It is observed that WQI is not in the prescribed limit for all the areas and for Sanganer and VKI, WQI is so high.

*Table 1. Quality Ratings and Unit Weights for different Stations.*

VKI	JOTHWARA	CHITRAKOOT	VAISHALI	TRANSPORT	KANOTA
266.66	283.33	166.66	100	250	133.33
100	126.66	60	33.33	33.33	46.66
76.66	80	98.86	96.13	65.2	43.33
218	178	156	194	178	184
625	592.5	605	625	870	630
275	325	265	255	215	280
101.16	102.32	96.51	101.16	100	108.13
305	240	265	180	270	280
216.66	466.66	366.66	306.66	483.33	326.66
280	284	186.66	244	292	230.66
90	84.5	88	105	107.5	100
266.66	237.77	188.88	197.77	248.88	260

*Table 1. Continued.*

VKI	BARI CHOPAD	SUBHASH	GOPALPURA	SANGANER	UNIT WEIGHT
266.66	116.66	133.33	100	466.66	0.7341
100	100	80	93.33	100	0.0518
76.66	76.66	85.33	68	73.33	0.00029
218	176	200	196	192	0.0881
625	800	765	550	880	0.0022
275	259.5	250	295	337.5	0.0022
101.16	100	103.48	100	119.76	0.0734
305	303.5	295	310	355	0.0022
216.66	183.33	216.66	316.66	543.33	0.0146
280	326.66	284	240	284	0.0058
90	95	91	101.5	110	0.0022
266.66	217.77	197.77	168.88	300	0.0097

*Table 2. Water Quality Index for different Stations.*

Name of Area	Water Quality Index
VKI	237.811
Jhotwara	251.268
Chirkoot	157.268
Vaishali Nagar	110.055
Transport Nagar	222.674
Kanota	135.921
Bari Chopar	123.579
Subhash Chowk	137.073
Gopal Pura	113.289
Sanganer	389.673

Identification of distribution and process capability for different parameters:

Ph

Descriptive statistics The table of descriptive statistics provides the summary information for the whole column of

data. All the statistics are based on the non-missing ( $N = 10$ ) values. For these data,  $\mu = 8.16$  and  $\sigma = 0.483506$ .

Goodness-of-fit test The table includes Anderson-Darling (AD) statistics and the corresponding p-value for a distribution. For a critical value  $\alpha$  a p-value greater than  $\alpha$  suggests that the data follow that distribution. The p-value 0.259, 0.003, 0.250, and 0.238 indicate that the gamma distributions fit the data well because this has the largest p value which is greater than  $\alpha$

Probability plot: The probability plot includes percentile points for corresponding probabilities of an ordered data set. The middle line is the expected percentile from the distribution based on maximum likelihood parameter estimates. The left and right line represents the lower and upper bounds for the confidence intervals of each percentile.

The probability plots shows that the data points fall approximately on a straight line and within the confidence intervals for Normal, Weibull and gamma distribution.

*Table 3. Descriptive Statistics for the parameter Ph.*

N	N*	Mean	StDev	Median	Minimum	Maximum	Skewness	Kurtosis
10	0	8.16	0.483506	8.3	7.5	8.9	-0.187851	-1.25583

*Table 4. Goodness of Fit Test for the Parameter Ph.*

Distribution	AD	P
Normal	0.420	0.259
Exponential	4.104	0.003
Weibull	0.397	0.250
Gamma	0.483	0.238

Table 5. ML Estimates of Distribution for Parameters Ph

Distribution	Location	Shape	Scale	Threshold
Normal*	8.16000		0.48351	
Exponential			8.16000	
Weibull		20.63355	8.37381	
Gamma		313.90323	0.02600	

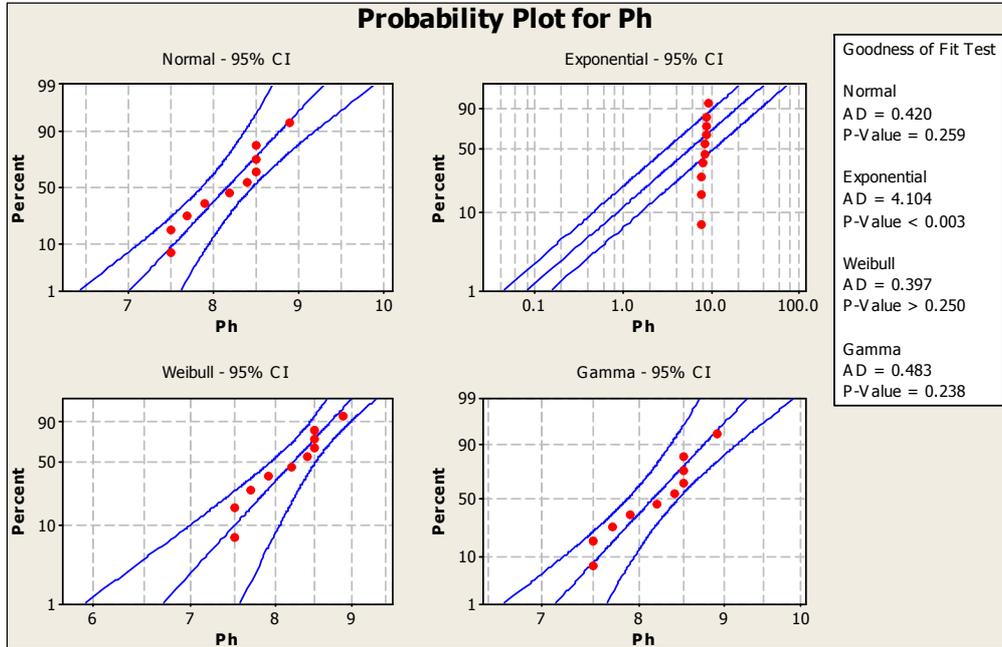


Figure 1. Probability Plot for the Parameter Ph.

The histogram shows the evidence of serious discrepancies between the assumed model and the data.

WE can see that the right tail of the distribution falls outside the upper specification limit. This means at all the areas the parameter Ph shows the values higher than the upper specification.

The Ppk and PPU indices tell us whether the process

within the tolerance limits. Both indices are -0.79, below the guideline. Thus this process of Ph cannot be considerable.

Likewise, the observed PPM > UCL is 1000000. This value means that 1000,000 out of 1,000,000 points more than the upper specification limit that is in all areas Ph has more than the upper specification limit of 7.

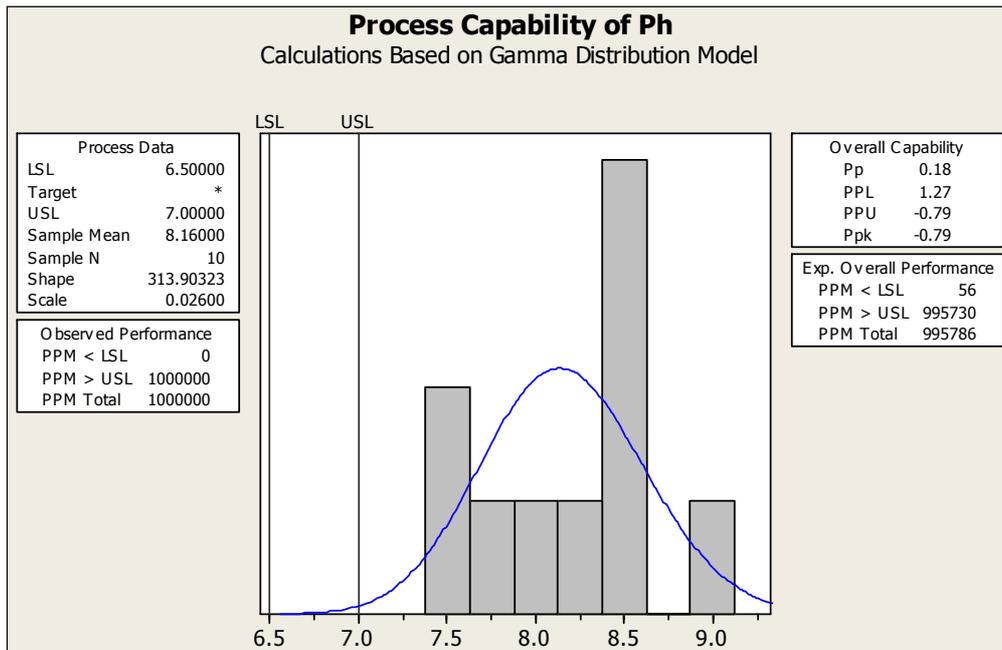


Figure 2. Process Capability for the Parameter Ph.

**Turbidity**

Descriptive statistics The table of descriptive statistics provides the summary information for the whole column of data. All the statistics are based on the non-missing (N = 10) values. For these data,  $\mu = 9.36$  and  $\sigma = 0.839577$ .

Goodness-of-fit test: The table includes Anderson-Darling (AD) statistics and the corresponding p-value for a distribution. For a critical value  $\alpha$  a p-value greater than  $\alpha$  suggests that the data follow that distribution. The p-value 0.68, 0.003, 0.250, and 0.250 indicate that the Normal, Weibull and gamma distributions fit the data well because

this has the largest p value which is greater than  $\alpha$

Probability plot: The probability plot includes percentile points for corresponding probabilities of an ordered data set. The middle line is the expected percentile from the distribution based on maximum likelihood parameter estimates. The left and right line represents the lower and upper bounds for the confidence intervals of each percentile.

The probability plots shows that the data points fall approximately on a straight line and within the confidence intervals for Normal, Weibull and gamma distribution.

*Table 6. Descriptive Statistics for Turbidity.*

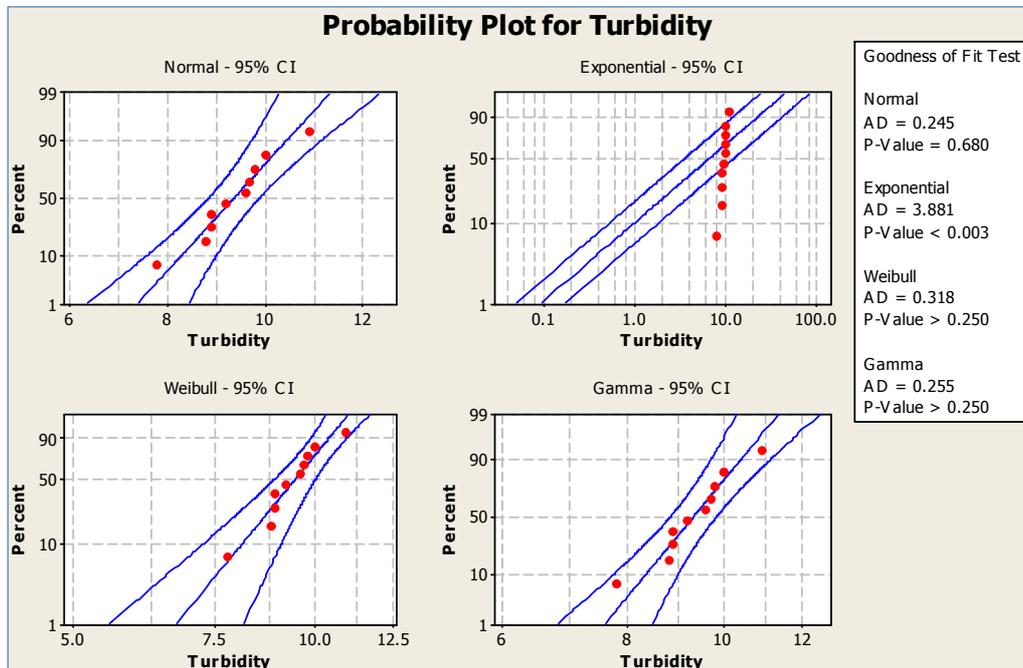
N	N*	Mean	StDev	Median	Minimum	Maximum	Skewness	Kurtosis
10	0	9.36	0.839577	9.4	7.8	10.9	-0.0277117	0.913103

*Table 7. Goodness of Fit Test for Turbidity.*

Distribution	AD	P
Normal	0.245	0.680
Exponential	3.881	<0.003
Weibull	0.318	>0.250
Gamma	0.255	>0.250

*Table 8. ML Estimates of Distribution Parameters for Turbidity..*

Distribution	Location	Shape	Scale	Threshold
Normal*	9.36000		0.83958	
Exponential			9.36000	
Weibull		12.50797	9.72588	
Gamma		136.58407	0.06853	



*Figure 3. Probability Plot for the Parameter Turbidity.*

The histogram show the evidence of serious discrepancies between the assumed model and the data.

WE can see that the right tail of the distribution falls outside the upper specification limit. This means at all the

areas the parameter Turbidity shows the values higher than the upper specification.

The Ppk and PPU indices tell us whether the process within the tolerance limits. Both indices are -1.66, below the

guideline. Thus this process cannot be considerable.

Likewise, the observed PPM > UCL is 1000000. This value means that 1000,000 out of 1,000,000 points more than

the upper specification limit that is in all areas Turbidity has more than the upper specification limit of 5.00.

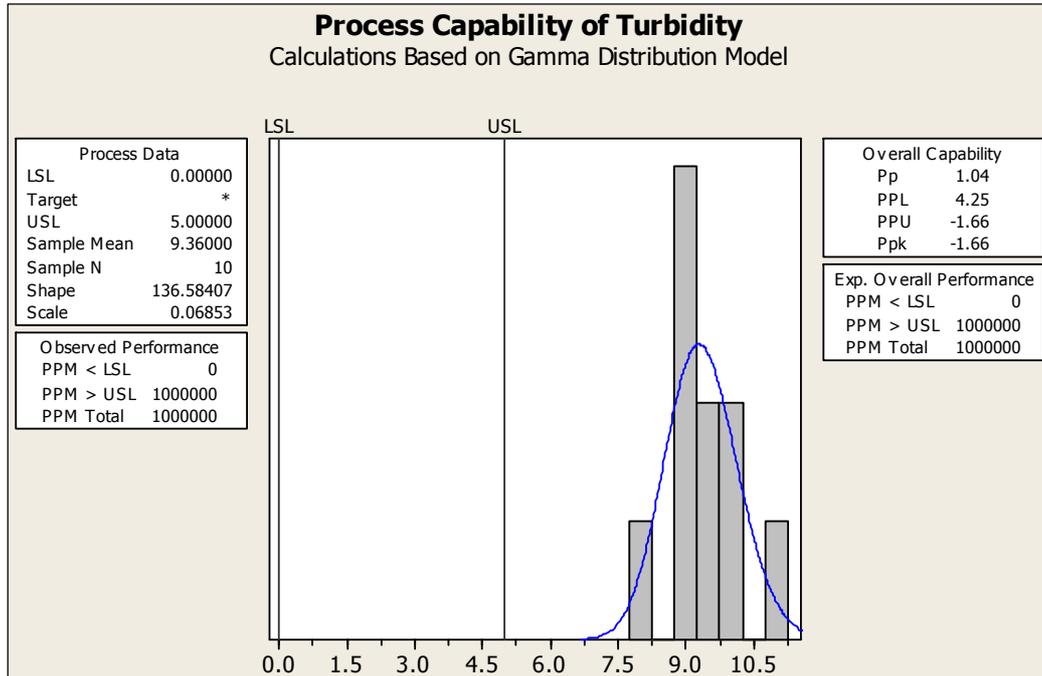


Figure 4. Process Capability for the Parameter Turbidity.

TDS

Descriptive statistics. The table of descriptive statistics provides the summary information for the whole column of data. All the statistics are based on the non-missing (N = 10) values. For these data,  $\mu = 1388.5$  and  $\sigma = 244.427$ .

Goodness-of-fit test. The table includes Anderson-Darling (AD) statistics and the corresponding p-value for a distribution. For a critical value  $\alpha$  a p-value greater than  $\alpha$  suggests that the data follow that distribution. The p-value 0.054, 0.003, 0.045, and 0.071 indicate that the gamma distributions fit the data well because this has the largest p

value which is greater than  $\alpha$

Probability plot The probability plot includes percentile points for corresponding probabilities of an ordered data set. The middle line is the expected percentile from the distribution based on maximum likelihood parameter estimates. The left and right line represents the lower and upper bounds for the confidence intervals of each percentile.

The probability plots shows that the data points fall approximately on a straight line and within the confidence intervals for Normal, Weibull and gamma distribution.

Table 9. Descriptive Statistics for TDS.

N	N*	Mean	StDev	Median	Minimum	Maximum	Skewness	Kurtosis
10	0	1388.5	244.427	1255	1100	1760	0.584265	-1.44218

Table 10. Goodness of Fit Test for TDS.

Distribution	AD	P
Normal	0.674	0.054
Exponential	3.286	<0.003
Weibull	0.733	0.045
Gamma	0.700	0.071

Table 11. ML Estimates of Distribution Parameters for TDS.

Distribution	Location	Shape	Scale	Threshold
Normal*	1388.50000		244.42734	
Exponential			1388.50000	
Weibull		6.39721	1489.81653	
Gamma		37.28603	37.23915	

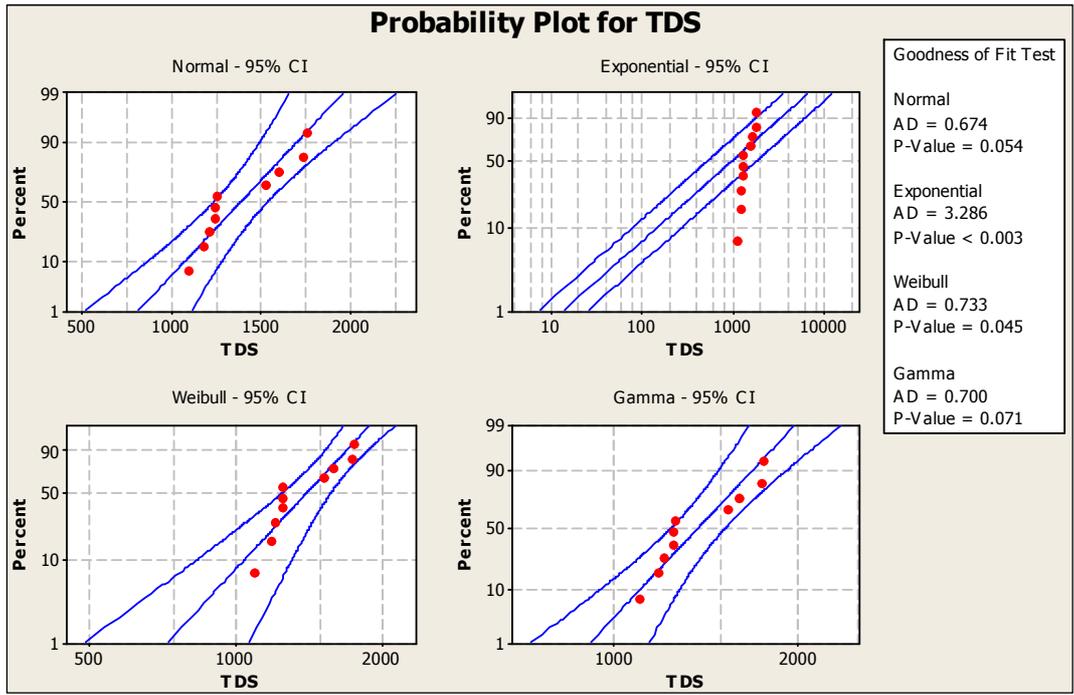


Figure 5. Probability Plot for the Parameter TDS.

The histogram show the evidence of serious discrepancies between the assumed model and the data.

WE can see that the right tail of the distribution falls outside the upper specification limit. This means at all the areas the parameter TDS shows the values higher than the upper specification.

The Ppk and PPU indices tell us whether the process

within the tolerance limits. Both indices are -1.10, below the guideline. Thus this process cannot be considerable.

Likewise, the observed PPM > USL is 1000000. This value means that 1000,000 out of 1,000,000 points more than the upper specification limit that is in all areas Turbidity has more than the upper specification limit of 500.

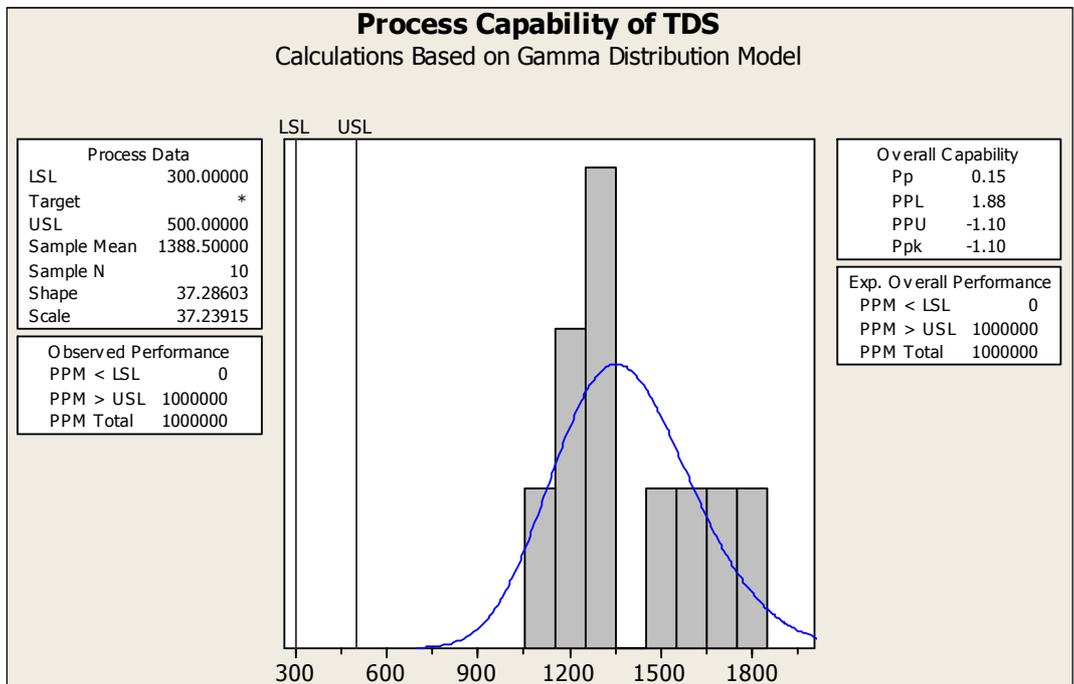


Figure 6. Process Capability for the parameter TDS.

DO

Descriptive statistics The table of descriptive statistics provides the summary information for the whole column of

data. All the statistics are based on the non-missing (N = 10) values. For these data,  $\mu = 5.75$  and  $\sigma = 0.479004$ .

Goodness-of-fit test The table includes Anderson-Darling

(AD) statistics and the corresponding p-value for a distribution. For a critical value  $\alpha$  a p-value greater than  $\alpha$  suggests that the data follow that distribution. The p-value 0.016, 0.003, 0.138, and 0.017 indicate that the Weibull distributions fit the data well because this has the largest p value which is greater than  $\alpha$

Probability plot The probability plot includes percentile points for corresponding probabilities of an ordered data set.

The middle line is the expected percentile from the distribution based on maximum likelihood parameter estimates. The left and right line represents the lower and upper bounds for the confidence intervals of each percentile.

The probability plots shows that the data points fall approximately on a straight line and within the confidence intervals for Normal, Weibull and gamma distribution.

Table 12. Descriptive Statistics for DO.

N	N*	Mean	StDev	Median	Minimum	Maximum	Skewness	Kurtosis
10	0	5.75	0.479004	5.9	4.6	6.3	-1.75910	3.51392

Table 13. Goodness of Fit Test for DO.

Distribution	AD	P
Normal	0.871	0.016
Exponential	3.988	<0.003
Weibull	0.561	0.138
Gamma	0.960	0.017

Table 14. ML Estimates of Distribution Parameters for DO.

Distribution	Location	Shape	Scale	Threshold
Normal*	5.75000		0.47900	
Exponential			5.75000	
Weibull		19.15438	5.93017	
Gamma		146.45689	0.03926	

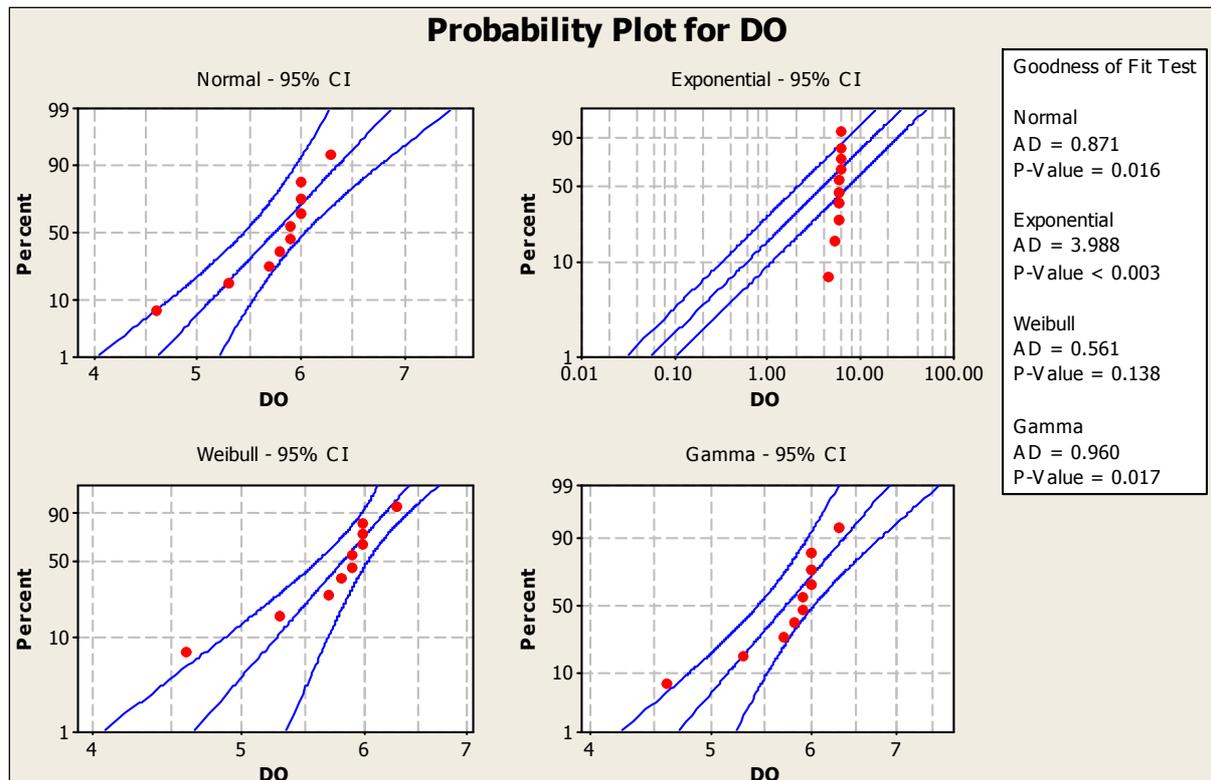


Figure 7. Probability Plot for the Parameter DO.

The histogram shows the evidence of serious discrepancies between the assumed model and the data.

WE can see that the right tail of the distribution falls within the upper specification limit. This means at all the areas the parameter DO shows the values lower than the

upper specification.

Likewise, the observed PPM > UCL is 0 and PPM < LCL is 600000. This value means that 6000,00 out of 1,000,000 points less than the lower specification limit that is in almost all areas DO has less than the lower specification limit of 6.

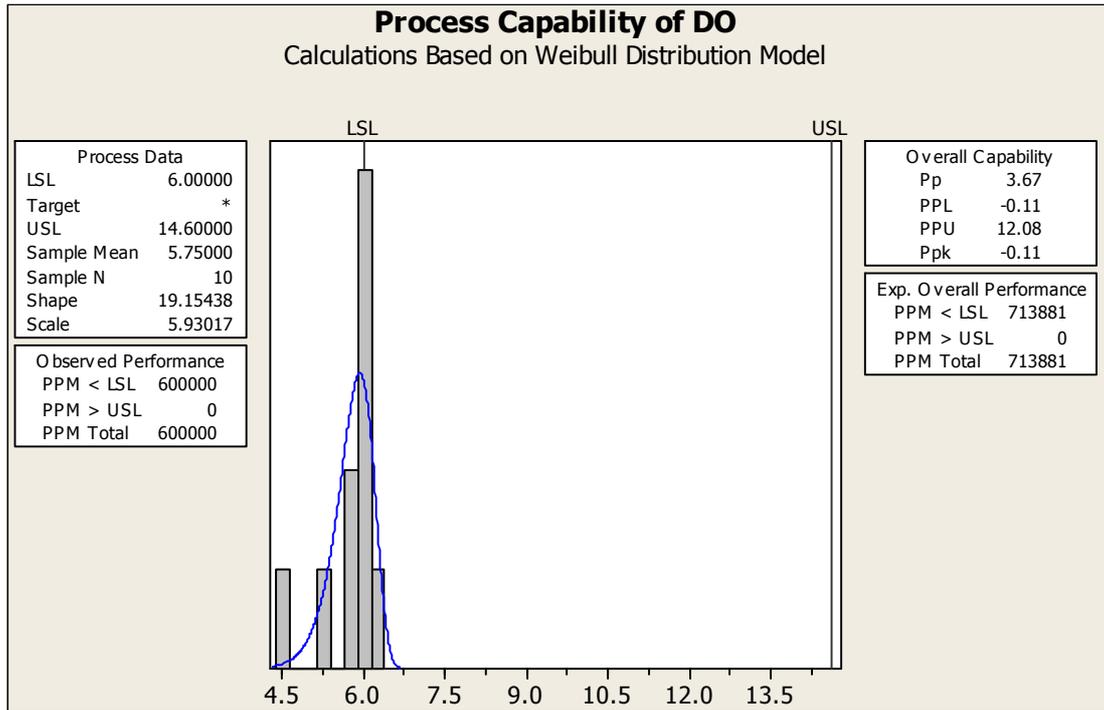


Figure 8. Process Capability for the Parameter DO.

Nitrate

Descriptive statistics The table of descriptive statistics provides the summary information for the whole column of data. All the statistics are based on the non-missing (N = 10) values. For these data,  $\mu = 102.8$  and  $\sigma = 18.4860$ .

Goodness-of-fit test The table includes Anderson-Darling (AD) statistics and the corresponding p-value for a distribution. For a critical value  $\alpha$  a p-value greater than  $\alpha$  suggests that the data follow that distribution. The p-value 0.823, 0.003, 0.250, and 0.250 indicate that the Normal, Weibull and gamma distributions fit the data well because

this has the largest p value which is greater than  $\alpha$

Probability plot The probability plot includes percentile points for corresponding probabilities of an ordered data set. The middle line is the expected percentile from the distribution based on maximum likelihood parameter estimates. The left and right line represents the lower and upper bounds for the confidence intervals of each percentile.

The probability plots shows that the data points fall approximately on a straight line and within the confidence intervals for Normal, Weibull and gamma distribution.

Table 15. Descriptive Statistics for Nitrate.

N	N*	Mean	StDev	Median	Minimum	Maximum	Skewness	Kurtosis
10	0	102.8	18.4860	102.5	76	135	0.262407	-0.788338

Table 16. Goodness of Fit Test for Nitrate.

Distribution	AD	P
Normal	0.205	0.823
Exponential	3.185	<0.003
Weibull	0.239	>0.250
Gamma	0.227	>0.250

Table 17. ML Estimates of Distribution Parameters for Nitrate.

Distribution	Location	Shape	Scale	Threshold
Normal*	102.80000		18.48603	
Exponential			102.80000	
Weibull		6.40291	110.30535	
Gamma		34.42168	2.98649	

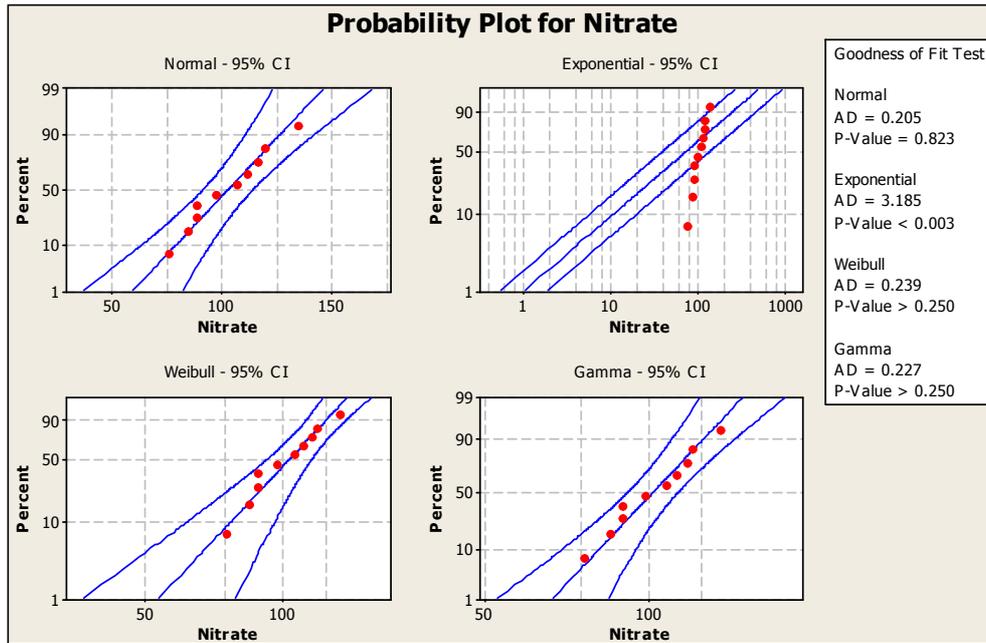


Figure 9. Probability Plot for the Parameter Nitrate.

The histogram shows the evidence of serious discrepancies between the assumed model and the data.

WE can see that the right tail of the distribution falls outside the upper specification limit. This means at all the areas the parameter Nitrate shows the values higher than the upper specification.

The Ppk and PPU indices tell us whether the process

within the tolerance limits. Both indices are -0.84, below the guideline. Thus this process cannot be considerable.

Likewise, the observed PPM > UCL is 1000000. This value means that 1000,000 out of 1,000,000 points more than the upper specification limit that is in all areas Turbidity has more than the upper specification limit of 50.

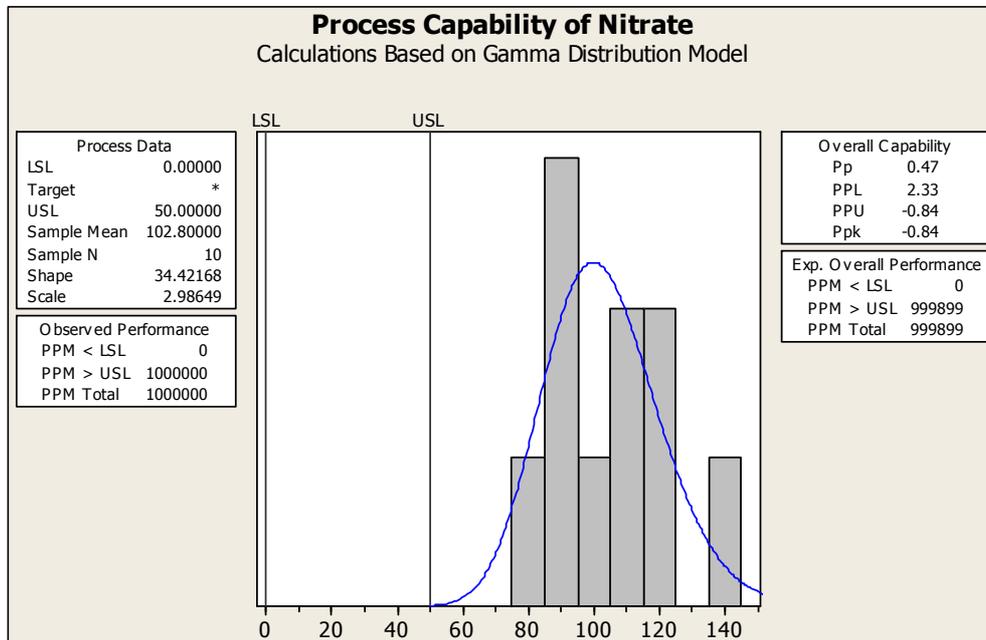


Figure 10. Process Capability for the Parameter Nitrate.

### 4. Conclusion

Statistical analysis of water quality parameters has been studied by using MINI TAB. It is observed that WQI is so high in all the areas of Jaipur under study and for Sanganer it

is highest.

The textile industry is one of the oldest and largest organized sectors in Sanganer area with geographical coordination 26° 49' 0" North, 75° 47' 0" East. Which houses almost 400 small-scale textile printing units, a large volume

of textile wastewater (>10 000 KL/day) is discharged. Further, untreated and sometimes even treated effluents from these industries are released into surface waters of Amanishah drainage or through the drainage systems, seep into the ground water and adjoining water bodies. Since many textile dyes are known carcinogens and mutagens<sup>12</sup>. These disposal practices have contaminated the environment and caused adverse effects on the flora, crops and fauna in Sanganer town<sup>13</sup> and had higher fluoride (2.8 mg/L) compared to other areas.

We study that all the parameters are responsible for this increased WQI in the city. Since all the parameters show the higher values than UCL in Process Capability.

DO is essential for living organism and we have been observed that in all the areas DO has near to its minimum value.

Now we have to think that what we are getting by this urbanization, industries, textile mills etc.

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