Data Aggregation in WSN Using Robust Mean

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Abstract-As Environment scrutiny has risen and regulation has tightened in recent years, the importance of having accurate information on the environment performance of an industry has increased. India among other developing nations of the world, is facing the challenge of industrial pollution at an alarming rate. This has made the constant surveillance of environmental characteristics a necessary task. There is an urgent need to identify critically polluted areas and the pollutant levels. Air pollution monitoring is considered as a very complex task but nevertheless it is very important. One of the methods developed for the assessment of environmental consequences is Comprehensive Environmental Pollution Index(CEPI). A major problem for constructing the CEPI is the determination of an appropriate aggregation method to combine multi-dimensional environmental variables in to an overall index. The main aim of this paper to apply and analyses the robust statistical method namely Trimmed Mean and Winsorized Mean in air pollution data aggregation with the consideration of data accuracy and usefulness of data.

Keywords: Mean; Order Statistics; Trimmed mean; Winsorized mean

1. INTRODUCTION

Air quality is being a concern for all.The proper measurement and transmission of air quality data to the central control board has become a need for the society. The instantaneous data may also become relevant for depicting real-time generic information in the form of Air Quality Index.It has been taken as one of the priority of concern for Central Pollution Control Board. The air pollution level is increasing at a robust pace mainly in the cities such as Delhi,Lucknow,Kolkata,Bangalore,Chennai and Mumbai.[1]

Statistical methods have been used in almost every applied field to analyze the experimental data. Statistical methods are mathematical formulas, models and techniques that are used in statistical analysis of raw data. The application of statistical methods extracts information from data provides different ways to assess the outputs. Robust statistics provides an alternative approach to classical statistics methods. It is an extension of parametric statistics, taking into account that parametric models are at best only approximations to reality. Its primary goal is the development of procedures which are still reliable and reasonably efficient under small deviations from the model. Statistical procedures help to ensure that the information or observations are presented and interpreted in an accurate and informative way.[2]This paper attempts to apply two of the robustmean classifications namely TrimmedMean and WinsorizedMean. Section 1presents the literature survey regarding the importance of aggregation and various aggregation methods, Section 2 describes the category of air quality,Section 3 discusses mean and robust mean ,Section4 discusses simulation study and Section 5 concludes the paper.

2. LITERATURE SURVEY

Energy efficiency is a major concern in Wireless Sensor Network (WSN). One of the well known techniques to achieve energy efficiency is In-network aggregation.In In-network aggregation instead of sending multiple data items to the sink, data items are aggregated and forward to the network.Data aggregation is application dependent that is depending on the target application and, the appropriate data aggregation operator (or aggregator). So number of data aggregation algorithms by targeting different sensors network scenarios are necessary.[3]Although aggregation measures such as average and sum are sufficient in many applications, there are situations when they may not be enough.[4]Many effective type of data aggregation function is needed in wireless sensor network. These functions and performance measures are closely related to the sensor network application. The important performance measures

are energy efficiency, network lifetime, latency, communication, overhead and data accuracy. Duplicate sensitive, insensitive and lossy, lossless aggregation are the categories of approaches.[5]In sensor networks where the in-network processing of various aggregation quires is paramount ,data aggregation inside the network could drastically reduce the communication cost and ensure the desired bounds on the quality of data. The main property of a good aggregation algorithm is to extract the most representative data by using minimum resources[6]. A major problem for constructing the CEI is the determination of an appropriate aggregation method to combine multidimensional environment variable in to an overall index.Despite the existence of large number of CEI, there is a lack of objective criteria for choosing an appropriate aggregation method [7].Most wireless sensor network involve the collection of high amounts of data, for this reason in recent year's considerable research effort has been devoted data aggregation algorithms. There are several statistical methods exists to summarize a list of data. One example is the use of one example is three quartiles-lower, median and upper, since they are unaffected by extreme values.It is also pointed that, quartiles reduce the amount of data to only three values while still reflecting the original data in an accurate way [8].

3. ACCURATE AIR QUALITY ANALYSIS

Air Quality Index Summary Report displays an annual summary of Air Quality Index (AQI) values for countries or Core Based Statistical Areas (CBSA). To compute accurate air quality, AQI with all pollutant in a geographicarea is to be measured.

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| AQI CATEGORY | RANGE |
|--------------|-----------------|
| 0 TO < 50 | GOOD |
| 50 TO < 100 | MODERATE |
| 100 TO < 150 | UNHEALTHY FOR |
| | SENSITIVE GROUP |
| 150 TO < 200 | UNHEALTHY |
| 200 TO < 300 | VERY UNHEALTHY |

Table(1) shows the AQI category and the corresponding air quality values in ,parts per million (ppm) and parts per billion (ppb).

The point to be considered here is, the proper aggregation method used in industrial pollution monitoring. Mostly industries are sending average of air pollution data to the state Pollution Control Board. (TamilNadu Pollution Control Board). This may lead to the false analysis.

4. COMPARISION OF MEAN

Classical Mean is the most commonly used measure of central tendency Mean is nothing but the average. It is computed by adding all the values in the data set divided by the number of observations in it.

Formula for Mean

$\mathbf{\bar{x}} = (\Sigma \mathbf{x}_i) / \mathbf{n}$

- \bar{x} just stands for the "sample mean"
- Σ means "add up"
- x_i "all of the x-values"
- n means "the number of items in the sample"

The major disadvantage of mean is that one or two very small or very large items either increase its value or reduce its value. The average cannot be really typical of the entire series.

In the robust literature, several robust methods of estimation have been proposed to reduce the influence of outliers in the data, on the estimates [9]. Robust mean such as Trimmed mean and Winsorized mean provides an alternative approach to classical statistics method. Trimmed mean is a statistical measure of central tendency much like the mean and median. It involves the calculation of the mean after discarding given parts of a sample at the beginning and the end of the whole data, and typically discarding an equal amount of both one. For most statistical applications, 5 to 25 percent of the ends are discarded. The trimmed mean is a useful estimator because it is less sensitive to outliers than the mean. In this regard it is referred to as a robust estimator.

Formula for Trimmed Mean

$$\overline{y}_{tk} = \frac{1}{n-2k} \sum_{i=k+1}^{n-k} y_{(i)}$$

The winsorized mean is a useful estimator because it is less sensitive to outliers than the mean but will still give a reasonable estimate of central tendency or mean for almost all statistical models. In this regard it is referred to as a robust estimator. [10]

Formula for Winsorized Mean as:

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$$\overline{y}_{wk} = \frac{1}{n} \{ (k+1)y_{(k+1)} + \sum_{i=k+2}^{n-k-1} y_{(i)} + (k+1)y_{(k+1)} \} \}$$

- n = number of observations.
- Reorder them as "order statistics" Xi from the smallest to the largest.
- Find lower case p=P/100

• Compute np.

5. SIMULATION STUDY

This section presents the performance of two robust mean by providing numerical illustration of air pollution data.For the given data set estimates the value of scale and location parameters under various method of estimation by using NS2.Table(2)shows independent measurement of a Air Quality Index data obtained from UCI Repository. [11]

| Table 2 Sample Air | Quality Data | (NO_2) |
|--------------------|--------------|----------|
|--------------------|--------------|----------|

| TIME | DAY1 | DAY4 | DAY5 | DAY6 | DAY7 | DAY9 | DAY20 | DAY27 |
|----------|------|------|------|------|------|------|-------|-------|
| 0:00:00 | 10 | 30 | 50 | 212 | 163 | 80 | 15 | 10 |
| 1:00:00 | 15 | 45 | 55 | 212 | 198 | 85 | 18 | 11 |
| 2:00:00 | 37 | 60 | 52 | 215 | 169 | 146 | 19 | 27 |
| 3:00:00 | 69 | 65 | 20 | 215 | 169 | 154 | 21 | 16 |
| 4:00:00 | 72 | 67 | 23 | 75 | 150 | 155 | 33 | 22 |
| 5:00:00 | 75 | 70 | 125 | 85 | 150 | 231 | 45 | 26 |
| 6:00:00 | 86 | 73 | 130 | 128 | 155 | 242 | 17 | 26 |
| 7:00:00 | 87 | 75 | 133 | 207 | 179 | 245 | 39 | 25 |
| 8:00:00 | 94 | 90 | 134 | 211 | 199 | 260 | 60 | 30 |
| 9:00:00 | 97 | 100 | 140 | 216 | 223 | 261 | 70 | 39 |
| 10:00:00 | 99 | 100 | 143 | 217 | 225 | 263 | 76 | 40 |
| 11:00:00 | 99 | 105 | 154 | 230 | 228 | 160 | 78 | 48 |
| 12:00:00 | 106 | 105 | 160 | 252 | 252 | 161 | 80 | 52 |
| 13:00:00 | 109 | 110 | 172 | 253 | 269 | 165 | 95 | 68 |
| 14:00:00 | 113 | 110 | 180 | 290 | 270 | 171 | 97 | 70 |
| 15:00:00 | 115 | 113 | 184 | 291 | 271 | 211 | 98 | 71 |
| 16:00:00 | 126 | 122 | 189 | 292 | 190 | 156 | 10 | 73 |
| 17:00:00 | 134 | 123 | 190 | 134 | 197 | 218 | 120 | 74 |
| 18:00:00 | 137 | 130 | 200 | 138 | 178 | 227 | 78 | 74 |
| 19:00:00 | 147 | 140 | 250 | 167 | 180 | 227 | 88 | 80 |
| 20:00:00 | 150 | 150 | 251 | 169 | 192 | 227 | 21 | 89 |
| 21:00:00 | 150 | 155 | 240 | 172 | 189 | 228 | 10 | 87 |
| 22:00:00 | 160 | 156 | 252 | 200 | 198 | 228 | 10 | 90 |

Table 3 Parameters considered in this study

| Network Dimension | 700 m x 700 m |
|---------------------|---------------|
| Simulation Duration | 25 s |
| No of Nodes | 32(1to 32) |

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| Node Location | Predefined Location | | | | |
|------------------|---|--|--|--|--|
| No of Leaf Nodes | 1 to 31 | | | | |
| Data sensed | Numeric data may be Air Water, Noise pollution level, Temperature | | | | |
| Network Model | Tree Structure | | | | |
| Protocol Used | Destination Sequenced Distance Vector Routing Protocol | | | | |

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Table

(3) shows the WSN setup parameters used in this study. The 23 values are assigned to each node to signify the pollutant values of a day in one hour duration. To supply pollution data for 31 days 31 leaf nodes are created and values are assigned.

| | MEAN | TRIMMED MEAN | | | | WINZORIZED MEAN | | | |
|-----|--------|--------------|--------|--------|--------|-----------------|--------|--------|--------|
| DAY | | 5% | 10% | 15% | 20% | 5% | 10% | 15% | 20% |
| S | | | | | | | | | |
| 1 | 99.43 | 100.80 | 102.73 | 103.82 | 103.26 | 99.21 | 101.13 | 104.91 | 103.69 |
| | (M) | (US) | (US) | (US) | (US) | (M) | (US) | (US) | (US) |
| 4 | 99.73 | 100.38 | 100.42 | 99.88 | 99.53 | 100.34 | 101.21 | 100.56 | 99.17 |
| | (M) | (US) | (US) | (M) | (M) | (US) | (US) | (US) | (M) |
| 5 | 149.0 | 150.23 | 151.63 | 151.82 | 152.59 | 149.08 | 151.34 | 150.30 | 143.86 |
| | (US) | (UH) | (UH) | (UH) | (UH) | (US) | (UH) | (UH) | (US) |
| 6 | 199.17 | 200.66 | 202.0 | 201.17 | 202.19 | 199.56 | 203.21 | 199.17 | 199.69 |
| | (UH) | (VUH) | (VUH) | (VUH) | (VUH) | (UH) | (VUH) | (UH) | (UH) |
| 7 | 199.73 | 198.71 | 197.52 | 195.82 | 194.26 | 199.69 | 200.04 | 198.86 | 195.73 |
| | (UH) | (UH) | (UH) | (UH) | (UH) | (UH) | (VUH) | (UH) | (UH) |
| 9 | 195.69 | 198 | 200.63 | 200.35 | 200.46 | 195.82 | 201.04 | 200.13 | 199.78 |
| | (UH) | (UH) | (VUH) | (VUH) | (VUH) | (UH) | (VUH) | (VUH) | (UH) |
| 20 | 49.916 | 50.85 | 50.52 | 50.17 | 49.53 | 51.13 | 51.04 | 51.43 | 50.56 |
| | (G) | (M) | (M) | (M) | (G) | (M) | (M) | (M) | (M) |
| 27 | 49.913 | 49.90 | 49.89 | 49.70 | 49.53 | 49.91 | 50.17 | 50.04 | 49.52 |
| | (G) | (G) | (G) | (G) | (G) | (G) | (M) | (M) | (G) |

Table 4 Estimates of Mean, Trimmed Mean and Winsorized Mean

The calculation of Mean value, Trimmed Mean (5%, 10%, 15%, 20%) and Winsorized Mean (5%,10%,15%,20%) using the data in Table [2] is tabulated in Table [4]

Table 5 Comparision of Mean, Trimmed Mean(20%), Winsorized Mean(20%)

| Estimators/Day | 'S | Day1 | Day4 | Day5 | Day6 | Day7 | Day9 | Day20 | Day27 |
|----------------|------|--------|-------|--------|--------|--------|--------|--------|--------|
| Mean | | 99.43 | 99.73 | 149.0 | 199.17 | 199.73 | 195.69 | 49.916 | 49.913 |
| | | (M) | (M) | (US) | (UH) | (UH) | (UH) | (G) | (G) |
| Trimmed M | lean | 103.26 | 99.53 | 152.59 | 202.19 | 194.26 | 200.46 | 49.53 | 49.53 |
| 20% | | (US) | (M) | (UH) | (VUH) | (UH) | (VUH) | (G) | (G) |
| Winsorized M | lean | 103.69 | 99.17 | 143.86 | 199.69 | 195.73 | 199.78 | 50.56 | 49.52 |
| 20% | | (US) | (M) | (US) | (UH) | (UH) | (UH) | (M) | (G) |

Table [5] shows the Comparision of Trimmed Mean(20%), Winsorized Mean, Mean(20%). From Table [V]it is observed that in

Day(4 & 27) all the estimates are close to each other and that the data seem to be around 99 and 49. In Day(7) even though there is a minimum

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variation in estimates the range same as UnHealthy. In Day(20) Mean and Trimmed Mean estimates value are very close to each other and the Winsorized Mean value is slightly different. In Day (1,5,6,9) the Mean values are different from Trimmed mean values. The data are applied to the DSDV protocol under the nodes and the results obtained are tabulated in Table [5].



Figure 1 Graphical Representation of Table 5

Figure1 shows the comparision of pollution category using estimates mean, Trimmed mean 20% and Winsorized mean 20%. Only after 20% elimination of data, the pollution category falls on similar type or some higher category of pollution range.

6. CONCLUSION

One of the essential techniques in WSN is Data aggregation. The simple and mostly used aggregation is mean. The major drawback of mean is that one or two very small or very large items underlie affect the average. To study the performance of robust estimators with classical mean the numerical illustrations are carried out by using NS2. If the data deviate from the Air quality Index the estimate based on the Mean are getting affected since they are highly non robust.Hence it is concluded that the robust estimators are not affected and provide the better results when outliers are present in the data. The next work is to apply various distribution methods that may be used for data aggregation in WSN.

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