

Data cleaning – handling missing values and outlier analyses

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1

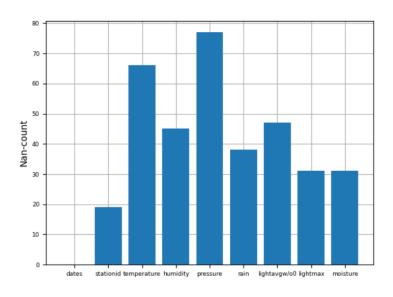


Figure 1 Number of missing values vs. attributes

Inferences:

- 1. Attribute 'pressure' has the maximum missing values and attribute 'dates' has minimum.
- 2. The frequency of missing values is very less compared to rows in dataframe.

2 a.

- 1. If 'station id' attribute is missing then there is no meaning of all its data as we don't know which station's data it is.
- 2. Total number of rows deleted is 19.
- 3. Percentage of rows deleted: 2.01%.



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b.

Inferences:

- 1. Total number of tuples deleted is 39 which have missing values more than 2.
- 2. Percentage of rows deleted: 4.12%.
- 3. Data loss is very less so we can use this data.
- 4. Since rows which are deleted have 3 or more missing values which makes that tuple less informative and it just cause problem for whole dataset so removing such row is better.

3

Table 1 Number of missing values per attribute after removing missing values

S. No	Attribute	Number of missing values			
1	dates	0			
2	stationid	0			
3	temperature (in °C)	34			
4	humidity (in g.m ⁻³)	13			
5	pressure (in mb)	41			
6	rain (in ml)	6			
7	lightavgw/o0 (in lux)	15			
8	lightmax (in lux)	1			
9	moisture (in %)	6			

- 1. Attribute 'pressure' has maximum missing values which is 41 and attribute 'dates' and 'Stationid' has minimum missing values which is 0.
- 2. "Dates" and "Stationid" have 0% missing data, "temperature" has 3.59% missing data, "humidity" has 1.37% missing data, "pressure" has 4.33% missing value, "rain" has 0.63% missing data, "lightavg" has 1.58% missing data, "lightmax" has 0.1% missing data and "moisture" has 0.63% missing data.
- 3. Total number of missing values: 116.



Data cleaning – handling missing values and outlier analyses

4 a. i.

Table 2 Mean, mode, median and standard deviation before and after replacing missing values by mean

S.	Attribute	Before				After			
No		Mean	Mode	Median	S.D.	Mean	Mode	Median	S.D.
1	dates	NA	NA	NAN	NA	NAN	NA	NA	NA
2	stationid	NA	NA	NA	NA	NA	NA	NA	NA
3	temperature (in °C)	21.21	12.72	22.27	4.35	21.07	21.07	21.80	4.24
4	humidity (in g.m ⁻³)	83.47	99.00	91.38	18.20	83.26	99.00	90.11	17.95
5	pressure (in mb)	1009.00	789.39	1014.67	46.95	1009.22	1009.22	1014.07	45.19
6	rain (in ml)	10701.53	0.00	18.00	24839.10	10942.72	0.00	24.75	24561.24
7	lightavgw/o0 (in lux)	4438.42	4488.91	1656.88	7569.15	4430.92	4488.91	1911.23	7396.66
8	lightmax (in lux)	21788.62	4000.00	6634.00	22053.31	21650.16	4000.00	7544.00	21666.72
9	moisture (in %)	32.38	0.00	16.70	33.63	32.67	0.00	17.72	33.39

- 1. Maximum change: (1) Mean: rain (2) Median: lightavgw/0 (3) Mode: pressure (4) S.D.: lightmax Minimum change: (1) Mean: temperature (2)Median: temperature (3)Mode: humidity (4) S.D.: temperature
- 2. There is very less change in data for attributes having less values and large change in data for attributes having large values.
- 3. Since for most of the attributes, change in values is very small, so data is still reliable.



Data cleaning – handling missing values and outlier analyses

ii.

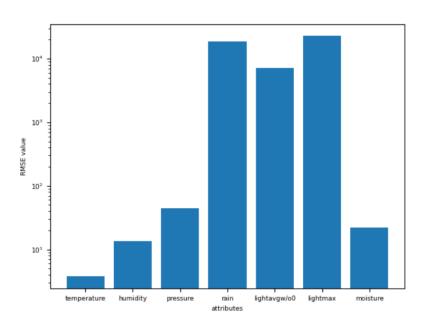


Figure 2 RMSE vs. attributes

- 1. Attribute 'lightmax' has maximum RMSE value which is 22711.61 and attribute 'temperature' has minimum RMSE value which is 3.7.
- 2. There is no specific relation between missing values and maximum RMSE and also for minimum missing values and minimum RMSE values.
- 3. Since RMSE values are quite high for almost all attributes, data is not reliable.



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b. i.

Table 3 Mean, mode, median and standard deviation before and after replacing missing values by linear interpolation technique

S.	Attribute	Before				After			
No		Mean	Mode	Median	S.D.	Mean	Mode	Median	S.D.
1	dates	NA	NA	NA	NA	NA	NA	NA	NA
2	stationid	NA	NA	NA	NA	NA	NA	NA	NA
3	temperature (in °C)	21.196	12.727	22.169	4.327	21.214	12.727	22.272	4.353
4	humidity (in g.m ⁻³)	83.538	99.00	91.380	18.197	83.479	99.00	91.380	18.20
5	pressure (in mb)	1009.264	789.392	1014.677	45.974	1009.008	789.392	1014.677	46.955
6	rain (in ml)	10651.638	0	22.500	24766.397	10701.538	0	18.00	24839.102
7	lightavgw/o0 (in lux)	4486.340	4488.910	1623.494	7569.787	4438.428	4488.910	1656.880	7569.154
8	lightmax (in lux)	21517.191	4000.00	6569.00	21923.55	21788.623	4000.00	6634.00	22053.315
9	moisture (in %)	32.327	0	16.306	33.584	32.386	0	16.704	33.635

- 1. Attribute "lightmax" has maximum change in values and attributes "dates" and "stationed" have minimum change.
- 2. There is very less change in data for attributes having less values and large change in data for attributes having large values.
- 3. Since for most of the attributes, change in values is very small, so data is still reliable.



Data cleaning – handling missing values and outlier analyses

ii.

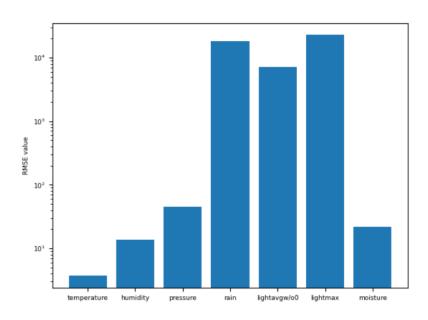


Figure 3 RMSE vs. attributes

- 1. Attribute 'lightmax' has maximum RMSE value which is 22736.704 and attribute 'temperature' has minimum RMSE value which is 3.695
- 2. There is no specific relation between missing values and maximum RMSE and also for minimum missing values and minimum RMSE values.
- 3. Since RMSE values are quite high for almost all attributes, data is not reliable.



Data cleaning – handling missing values and outlier analyses

5 a.

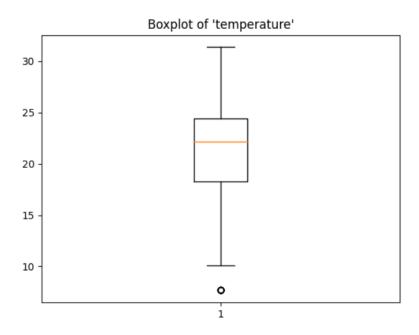


Figure 4 Boxplot for attribute temperature (in °C)

- 1. Value of outliers is 7.6729 and it's count is 10.
- 2. Inter quartile range is approximately 6.
- 3. Due to very few outliers present in this data set, variance/spread is very low.
- 4. Since the median line lies above the middle point so it is negatively skewed data.



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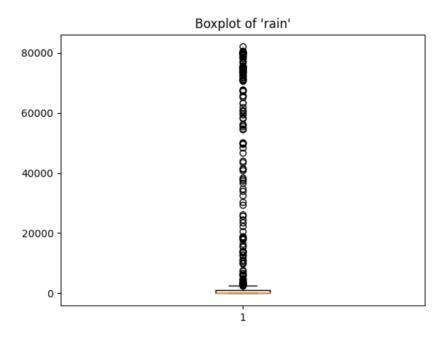


Figure 5 Boxplot for attribute rain (in ml)

- 1. There are total 185 outliers present in this data set ranging from 82037.25 to 2470.5.
- 2. Inter quartile range is around 100.
- 3. Since the number of outliers are 185, which is very high so it has quite high spread.
- 4. Since the median line lies below the middle point of iqr, it is positively skewed dataset.



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b.

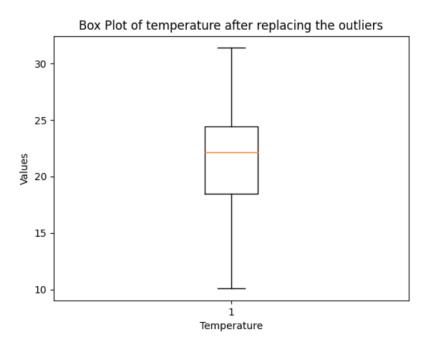


Figure 6 Boxplot for attribute temperature (in °C) after replacing median with outliers

- 1. No outliers are present in this boxplot.
- 2. IQR is same as before which his around 6.
- 3. Variance is also almost same as before.
- 4. It is still negatively skewed same as before.



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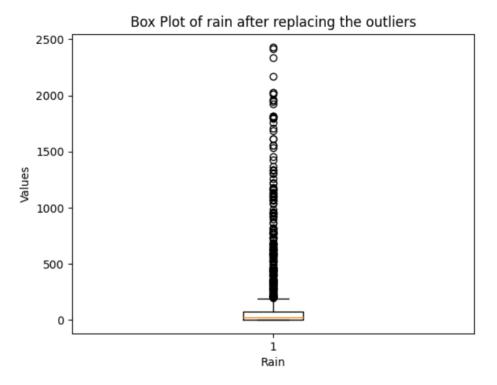


Figure 7 Boxplot for attribute rain (in ml) after replacing median with outliers

- 1. There are total 193 outliers present in this data set which has increased from previous case ranging from 200.25 to 2427.75.
- 2. Inter quartile range is around 70 which get decreased from previous case.
- 3. Since the number of outliers has increased, variance is still high.
- 4. Skewness of data set does not change and is same as before.