

FACULTY OF APPLIED SCIENCE

MATH2021-1 HIGH-DIMENSIONAL STATISTICS

Project 1 : Exploratory Data Analysis

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1 Presentation of the data

1.1 Discussion on the data

The data we used in this project is a subset of the data collected by the ENEA (National Agency for New Technologies, Energy and Sustainable Economic Development) alonside a road in a polluted area of Italy. This dataset is available on the UCI repository¹. The data was harvested using a multicensor device and reference analyzers. The data was collected between March 2004 and February 2005.

There are 5 couples of variables in the dataset, each couple is composed of a variable measured by the reference analyzer and the corresponding variable measured by the multicensor device. The values represent the hourly average concentration of each variable. In addition to the variables couples, we have 3 other variables representing the temperature and the humidity (both relative and absolute). The different values are stored in the following columns of our dataset:

- CO(GT): concentration of CO in the air (in mg/m³)
- PT08.S1(C0): average sensor response (nominally CO targeted)
- NMHC(GT): concentration of non-methane hydrocarbons in the air (in $\mu g/m^3$)
- C6H6(GT): concentration of benzene in the air (in $\mu g/m^3$)
- PT08.S2(NMHC): average sensor response (nominally NMHC targeted)
- NOx(GT): concentration of NOx in the air (in parts per billion)
- PT08.S3(NOx): average sensor response (nominally NOx targeted)
- NO2(GT) : concentration of NO₂ in the air (in $\mu g/m^3$)
- PT08.S4(NO2): average sensor response (nominally NO₂ targeted)
- PT08.S5(03): average sensor response (nominally O₃ targeted)
- T : temperature (in °C)
- RH: relative humidity (in %)
- AH: absolute humidity

On top of that, we created a binary indicator per variable measured by the reference analyzers which is equal to 1 when the measured value is above the median of the variable and 0 otherwise. The binary values are stored in the following columns:

- HIGH_CO: binary indicator for CO (above/under median)
- HIGH_NMHC: binary indicator for NMHC (above/under median)
- HIGH_C6H6: binary indicator for benzene (above/under median)
- HIGH_NOx: binary indicator for NOx (above/under median)
- HIGH_NO2: binary indicator for NO2 (above/under median)

1.2 Link between the variables

The values for each couple are obviously going to be quite correlated, as they are measuring the same thing. Furthermore, the values for the binary indicators are going to be highly correlated with the values of the corresponding measurements from the reference analyzers (as the binary indicator is equal to 1 when the value is above the median and 0 otherwise).

¹https://archive.ics.uci.edu/ml/datasets/air+quality

2 Information about missing data

We have a total of 2.1% of missing values but this number is overestimated because the binary indicators are taken into account. The real ratio is 1.7% without this indicators. The missing values are due to hardware problems related to the measuring instruments and to the fact that the data was collected in a real environment.

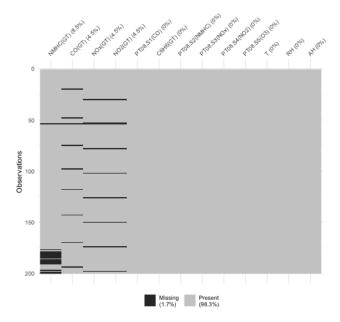


Figure 1: Missing data

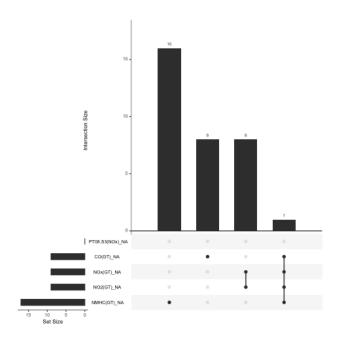


Figure 2: Missing data heatmap

To handle the missing values we will use for the rest of the project the complete case strategy. Indeed,

replacing missing data by the mean would not be a good idea because it would change the correlation structure of the data which is important for the next steps. We only have 33 lines with missing values so we can afford to remove them keeping the ratio $\frac{n}{p} > 5$ as we have a total of

3 Exploratory data analysis

3.1 Univariate statistical analysis

| | n | mean | sd | median | trimmed | mad |
|---------------|-----|-------------|------------|---------|-------------|-----------|
| CO(GT) | 191 | 2.748691 | 1.596801 | 2.50 | 2.560784 | 1.33434 |
| PT08.S1(CO) | 200 | 1339.000000 | 255.446559 | 1332.50 | 1328.356250 | 233.50950 |
| NMHC(GT) | 183 | 160.158470 | 139.745774 | 122.00 | 138.448980 | 118.60800 |
| C6H6(GT) | 200 | 12.254000 | 8.274006 | 11.05 | 11.318750 | 7.33887 |
| PT08.S2(NMHC) | 200 | 1016.950000 | 281.940276 | 1017.50 | 1005.556250 | 278.72880 |
| NOx(GT) | 191 | 175.842932 | 94.999980 | 161.00 | 168.830065 | 85.99080 |
| PT08.S3(NOx) | 200 | 1003.195000 | 278.431170 | 945.00 | 976.950000 | 234.99210 |
| NO2(GT) | 191 | 115.612565 | 34.357971 | 119.00 | 116.549020 | 35.58240 |
| PT08.S4(NO2)9 | 200 | 1671.040000 | 305.901187 | 1622.50 | 1641.750000 | 237.21600 |

| | min | max | range | skew | kurtosis | se |
|---------------|--------|--------|--------|------------|------------|------------|
| CO(GT) | 0.5 | 8.1 | 7.6 | 1.0755547 | 0.9704424 | 0.1155405 |
| PT08.S1(CO) | 831.0 | 2040.0 | 1209.0 | 0.3506958 | -0.0955406 | 18.0627994 |
| NMHC(GT) | 7.0 | 685.0 | 678.0 | 1.3200143 | 1.4422105 | 10.3303049 |
| C6H6(GT) | 1.0 | 39.2 | 38.2 | 1.0086394 | 0.8567735 | 0.5850606 |
| PT08.S2(NMHC) | 501.0 | 1754.0 | 1253.0 | 0.3211184 | -0.3339212 | 19.9361881 |
| NOx(GT) | 16.0 | 478.0 | 462.0 | 0.6699740 | -0.0263597 | 6.8739573 |
| PT08.S3(NOx) | 537.0 | 1918.0 | 1381.0 | 0.9745243 | 0.9172654 | 19.6880569 |
| NO2(GT) | 28.0 | 194.0 | 166.0 | -0.2179968 | -0.4405303 | 2.4860555 |
| PT08.S4(NO2) | 1134.0 | 2679.0 | 1545.0 | 0.9116828 | 0.7869557 | 21.6304804 |

| | n | | mean | sd | median | trimmed | mad |
|-------------|-----------------|------------------------|-----------|-------------|-------------|--------------|------------|
| PT08.S5(O3) | PT08.S5(O3) 200 | | 3.2450000 | 389.2906253 | 1204.5000 | 1222.0375000 | 384.734700 |
| T | 200 | 15 | .1965000 | 5.5702402 | 14.3000 | 14.8068750 | 5.189100 |
| RH | 200 | 49 | .8030000 | 15.1352426 | 53.9000 | 50.6387500 | 15.270780 |
| AH | 200 | 0. | 8085450 | 0.1059962 | 0.8125 | 0.8092338 | 0.104375 |
| HIGH_CO | 191 | 0. | 7225131 | 0.4489355 | 1.0000 | 0.7777778 | 0.000000 |
| HIGH_NMHC | 183 | 0. | 9945355 | 0.0739221 | 1.0000 | 1.0000000 | 0.000000 |
| HIGH_C6H6 | 200 | 0.6500000 0.9947644 | | 0.4781665 | 1.0000 | 0.6875000 | 0.000000 |
| HIGH_NOx | 191 | | | 0.0723575 | 1.0000 | 1.0000000 | 0.000000 |
| HIGH_NO2 | 191 | 0. | 5968586 | 0.4918179 | 1.0000 | 0.6209150 | 0.000000 |
| | min | | max | range | skew | kurtosis | se |
| PT08.S5(O3) | 384.0 | 0000 | 2359.0000 | 1975.0000 | 0.2942286 | -0.1157893 | 27.5270041 |
| Т | 6.10 | 000 | 29.3000 | 23.2000 | 0.6005822 | -0.4670720 | 0.3938755 |
| RH | 14.9 | 000 | 81.1000 | 66.2000 | -0.4375140 | -0.8600569 | 1.0702233 |
| AH | 0.52 | 237 | 1.0945 | 0.5708 | 0.0279583 | -0.2727186 | 0.0074951 |
| HIGH_CO | 0.00 | 000 | 1.0000 | 1.0000 | -0.9861019 | -1.0329289 | 0.0324838 |
| HIGH_NMHC | 0.00 | 000 | 1.0000 | 1.0000 | -13.3067908 | 176.0326973 | 0.0054645 |
| HIGH_C6H6 | 0.00 | 000 | 1.0000 | 1.0000 | -0.6242595 | -1.6183168 | 0.0338115 |
| HIGH_NOx | 0.00 | 000 | 1.0000 | 1.0000 | -13.6039602 | 184.0313314 | 0.0052356 |
| HIGH_NO2 | 0.00 | 000 | 1.0000 | 1.0000 | -0.3918179 | -1.8561145 | 0.0355867 |

Table 1: Summary statistics

3.2 Multivariate graphical analysis

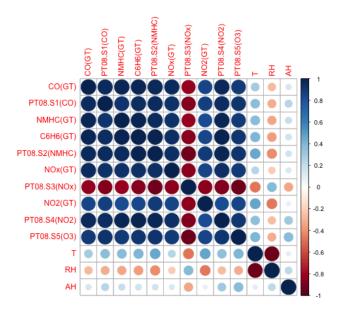


Figure 3: Correlation matrix

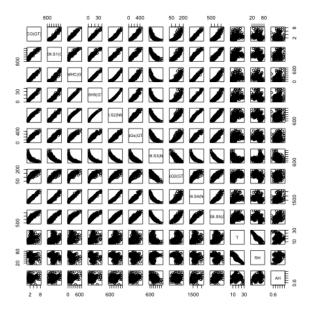


Figure 4: Scatter matrix

- 3.3 Correlation structure of the data
- 3.4 Outlying observations using Mahalanobis distance
- 4 Correlation analysis with data reduction
- 4.1 Choice between PCA and t-SNE
- 4.2 2D plot of the data