

paper05_data_aggregation

October 10, 2021

1 Data Aggregation Across Data Sources

We have 3 different sources of data:

1. Our sensor data: that has the Indoor Air Quality and Indoor Environmental Data.
2. SINAICA: Outdoor Air Quality Monitoring Data from the Government.
3. OpenWeatherData: Outdoor Environmental Data.

We need it to be available that data to the models we plan to train. In the following sections this process is detailed.

```
/home/jaa6766/.conda/envs/cuda/lib/python3.7/importlib/_bootstrap.py:219:
RuntimeWarning: numpy.ufunc size changed, may indicate binary incompatibility.
Expected 192 from C header, got 216 from PyObject
/home/jaa6766/.conda/envs/cuda/lib/python3.7/importlib/_bootstrap.py:219:
RuntimeWarning: numpy.ufunc size changed, may indicate binary incompatibility.
Expected 192 from C header, got 216 from PyObject
/home/jaa6766/.conda/envs/cuda/lib/python3.7/importlib/_bootstrap.py:219:
RuntimeWarning: numpy.ufunc size changed, may indicate binary incompatibility.
Expected 192 from C header, got 216 from PyObject
/home/jaa6766/.conda/envs/cuda/lib/python3.7/importlib/_bootstrap.py:219:
RuntimeWarning: numpy.ufunc size changed, may indicate binary incompatibility.
Expected 192 from C header, got 216 from PyObject
```

	CO	NO	NO2	NOx	O3	PM10	\
987	2.200000	0.205000	0.031000	0.207000	0.002000	45.000000	
988	2.200000	0.205000	0.031000	0.207000	0.002000	45.000000	
989	2.200000	0.205000	0.031000	0.207000	0.002000	45.000000	
990	2.200000	0.205000	0.031000	0.207000	0.002000	45.000000	
991	2.200000	0.205000	0.031000	0.207000	0.002000	45.000000	
...	
1582081	0.765217	0.009174	0.027826	0.039043	0.015304	20.304348	
1582082	0.765217	0.009174	0.027826	0.039043	0.015304	20.304348	
1582083	0.765217	0.009174	0.027826	0.039043	0.015304	20.304348	
1582084	0.765217	0.009174	0.027826	0.039043	0.015304	20.304348	
1582085	0.765217	0.009174	0.027826	0.039043	0.015304	20.304348	

	PM2.5	SO2	month	day	hour	datetime	\
987	22.000000	0.004000	2	12	6	2021-02-12 06:05:35.846304417	
988	22.000000	0.004000	2	12	6	2021-02-12 06:05:38.837326527	
989	22.000000	0.004000	2	12	6	2021-02-12 06:05:47.812360048	
990	22.000000	0.004000	2	12	6	2021-02-12 06:05:50.803695202	
991	22.000000	0.004000	2	12	6	2021-02-12 06:05:53.795462847	
...	
1582081	16.391304	0.001304	9	18	0	2021-09-18 00:59:47.142104626	
1582082	16.391304	0.001304	9	18	0	2021-09-18 00:59:50.136709690	
1582083	16.391304	0.001304	9	18	0	2021-09-18 00:59:53.131285429	
1582084	16.391304	0.001304	9	18	0	2021-09-18 00:59:56.125959396	
1582085	16.391304	0.001304	9	18	0	2021-09-18 00:59:59.120573282	

minute temperature pressure humidity gasResistance IAQ

987	35.0	21.51	777.41	44.04	152149.0	34.7
988	34.0	21.51	777.41	43.98	152841.0	33.6
989	32.0	21.54	777.41	43.73	153259.0	31.5
990	32.0	21.53	777.41	43.70	152841.0	31.5
991	30.0	21.52	777.41	43.70	153399.0	30.2
...
1582081	138.0	26.00	782.92	56.34	916837.0	138.2
1582082	138.0	26.00	782.92	56.33	917462.0	137.7
1582083	138.0	26.00	782.90	56.34	916837.0	137.6
1582084	136.0	26.00	782.92	56.35	921233.0	136.0
1582085	134.0	25.99	782.92	56.35	922497.0	134.5

[1581099 rows x 18 columns]

	CO	NO	NO2	NOx	O3	PM10 \
0	2.500000	0.244000	0.035000	0.205000	0.002000	57.000000
1	2.500000	0.244000	0.035000	0.205000	0.002000	57.000000
2	2.500000	0.244000	0.035000	0.205000	0.002000	57.000000
3	2.500000	0.244000	0.035000	0.205000	0.002000	57.000000
4	2.500000	0.244000	0.035000	0.205000	0.002000	57.000000
...
1605252	0.765217	0.009174	0.027826	0.039043	0.015304	20.304348
1605253	0.765217	0.009174	0.027826	0.039043	0.015304	20.304348
1605254	0.765217	0.009174	0.027826	0.039043	0.015304	20.304348
1605255	0.765217	0.009174	0.027826	0.039043	0.015304	20.304348
1605256	0.765217	0.009174	0.027826	0.039043	0.015304	20.304348

	PM2.5	S02	month	day	...	pressure_outdoor \
0	25.000000	0.005000	2	12	...	1020
1	25.000000	0.005000	2	12	...	1020
2	25.000000	0.005000	2	12	...	1020
3	25.000000	0.005000	2	12	...	1020
4	25.000000	0.005000	2	12	...	1020
...
1605252	16.391304	0.001304	9	18	...	1015
1605253	16.391304	0.001304	9	18	...	1015
1605254	16.391304	0.001304	9	18	...	1015
1605255	16.391304	0.001304	9	18	...	1015
1605256	16.391304	0.001304	9	18	...	1015

	humidity_outdoor	wind_speed	wind_deg	rain_1h	rain_3h	clouds_all \
0	44	0.00	0	0.0	0.0	1
1	44	0.00	0	0.0	0.0	1
2	44	0.00	0	0.0	0.0	1
3	44	0.00	0	0.0	0.0	1
4	44	0.00	0	0.0	0.0	1
...
1605252	93	1.37	199	1.0	0.0	0
1605253	93	1.37	199	1.0	0.0	0
1605254	93	1.37	199	1.0	0.0	0
1605255	93	1.37	199	1.0	0.0	0
1605256	93	1.37	199	1.0	0.0	0

	weather_id	weather_main	year
0	800	Clear	2021
1	800	Clear	2021
2	800	Clear	2021
3	800	Clear	2021
4	800	Clear	2021

```

...
1605252      500      Rain  2021
1605253      500      Rain  2021
1605254      500      Rain  2021
1605255      500      Rain  2021
1605256      500      Rain  2021

```

```
[1605257 rows x 32 columns]
```

```

CO                float64
NO                float64
NO2              float64
NOx              float64
O3               float64
PM10             float64
PM2.5            float64
SO2              float64
month            int64
day              int64
hour             int64
datetime          datetime64[ns]
minute           float64
temperature       float64
pressure          float64
humidity          float64
gasResistance     float64
IAQ              float64
temperature_outdoor float64
feels_like        float64
temp_min          float64
temp_max          float64
pressure_outdoor  int64
humidity_outdoor  int64
wind_speed        float64
wind_deg          int64
rain_1h           float64
rain_3h           float64
clouds_all        int64
weather_id        int64
weather_main      object
year              int64
dtype: object

```

1.1 Interpolation of Hourly Data

We found that the dataframe contains repeated records on the columns of hourly data: SINAICA Gov't Air Quality Monitoring and OpenWeatherData.

We think that the repeated data can be an issue, as the data moves very abruptly from a record call it at 10:57 and 11:00.

We propose an approach similar to the imputations using the interpolation incorporating noise, that could avert the overfitting issue on our machine learning and deep learning training.

1.2 Resampling

To reduce training time we propose to have a resampling of the data.

In the following subsections we create those resampled-data dataframes.

1.2.1 1 Minute Resampling

1.2.2 2 Minute Resampling

1.2.3 3 Minute Resampling

1.3 References

- https://scikit-learn.org/stable/modules/linear_model.html#generalized-linear-regression
- <https://pythonhealthcare.org/2018/05/03/81-distribution-fitting-to-data/>
- <https://medium.com/@amirarsalan.rajabi/distribution-fitting-with-python-scipy-bb70a42c0aed>
- <https://scikit-learn.org/stable/modules/generated/sklearn.neighbors.KernelDensity.html?highlight=kernel%20density#sklearn.neighbors.KernelDensity>