# MP10-field-data-treatment-Diamante

February 13, 2024

# 1 Electrochemical PM10 sensor data preprocessing

Pollutant: Particulate MatterSensor: Alphasense OPC-N3

#### 1.1 Constants

#### 1.1.1 Sensors IDs

#### 1.1.2 API Constants

```
[]: HOST = "renovar.lcqar.ufsc.br"
    PORT = 8080
    RAW_DATA_DIR = "data/raw-data-monit-fix-2022-2023-Diamante/"
    RAW_FILE_NAME = "OPC_PM10.CSV"

SENSOR_FILE_DIR = 'data/input/'
    SENSOR_NAME = 'alpha_pm_10_conc'
    SENSOR_FILE_NAME = SENSOR_NAME + 'web_dataframe.csv'
```

# 1.1.3 Sensor Constants

```
[]: lower_limit=0.0
upper_limit=2e3
sampling_period = 15 * 60
```

# 2 Alphasense PM10 Sensor Data

```
[]: import locale locale.LC_TIME, 'pt_BR')
```

[]: 'pt\_BR'

```
sensor_data.head()
    /Users/Fernando/Documents/Projects/Github/lcgar-low-cost-monit-proc/data-pre-
    processing/GetSensorDataService.py:13: UserWarning: The argument
    'infer_datetime_format' is deprecated and will be removed in a future version. A
    strict version of it is now the default, see
    https://pandas.pydata.org/pdeps/0004-consistent-to-datetime-parsing.html. You
    can safely remove this argument.
      df['DateTime'] = (pd.to_datetime(df[date_time_col],
    infer_datetime_format=False, format='%d/%m/%Y/%H/%M/%S'))
[]:
        latitude longitude measuring
                                                  DateTime
    0 -28.456899 -48.972999
                              -9999.99 2022-11-21 10:44:06
    1 -28.456899 -48.972999 -9999.99 2022-11-21 10:59:51
    2 -28.456899 -48.972999 -9999.99 2022-11-21 11:15:41
    3 -28.456899 -48.972999 -9999.99 2022-11-21 11:31:25
    4 -28.456899 -48.972999 -9999.99 2022-11-21 11:47:05
```

# 2.1 Upload Data from File

```
[]: import pandas as pd

df = pd.read_csv(SENSOR_FILE_DIR + SENSOR_FILE_NAME)
    df.head()
```

```
[]:
       Unnamed: 0
                    latitude longitude measuring
                                                              DateTime
                0 -28.456899 -48.972999
                                         -9999.99 2022-11-21 10:44:06
    0
    1
                1 -28.456899 -48.972999
                                         -9999.99 2022-11-21 10:59:51
    2
                2 -28.456899 -48.972999
                                         -9999.99 2022-11-21 11:15:41
    3
                3 -28.456899 -48.972999 -9999.99 2022-11-21 11:31:25
                4 -28.456899 -48.972999
                                         -9999.99 2022-11-21 11:47:05
```

## 2.1.1 Create Sensor Dataframe as Pandas Series with a period of 15 mins

/var/folders/wc/\_83zcrx913j1dqwg4g90kbhh0000gp/T/ipykernel\_4959/166902210.py:3:

UserWarning: The argument 'infer\_datetime\_format' is deprecated and will be removed in a future version. A strict version of it is now the default, see https://pandas.pydata.org/pdeps/0004-consistent-to-datetime-parsing.html. You can safely remove this argument.

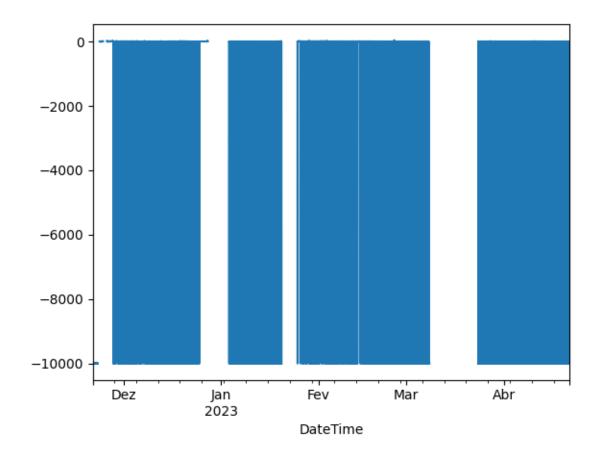
web\_dataframe['DateTime'] = (pd.to\_datetime(df['DateTime'],
infer\_datetime\_format=True))

```
[]:
                           latitude longitude measuring
    DateTime
     2022-11-21 10:30:00 -28.456899 -48.972999
                                                 -9999.99
     2022-11-21 10:45:00 -28.456899 -48.972999
                                                 -9999.99
     2022-11-21 11:00:00
                                NaN
                                                      NaN
     2022-11-21 11:15:00 -28.456899 -48.972999
                                                 -9999.99
     2022-11-21 11:30:00 -28.456899 -48.972999
                                                 -9999.99
     2023-04-21 20:30:00 -28.456899 -48.972999
                                                 -9999.99
     2023-04-21 20:45:00 -28.456899 -48.972999
                                                 -9999.99
     2023-04-21 21:00:00 -28.456899 -48.972999
                                                 -9999.99
     2023-04-21 21:15:00 -28.456899 -48.972999
                                                 -9999.99
     2023-04-21 21:30:00 -28.456899 -48.972999
                                                 -9999.99
```

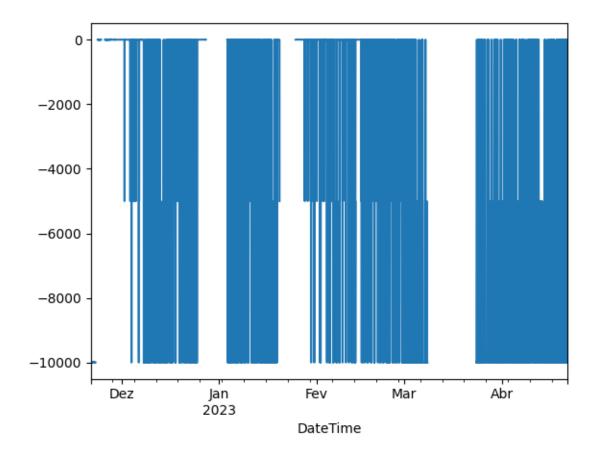
[14541 rows x 3 columns]

#### 2.1.2 Plot raw data

```
[]: sensor_dataframe['measuring'].plot()
```



# 2.1.3 Smooth data with a window of 4 samples, i.e.: an hour of data

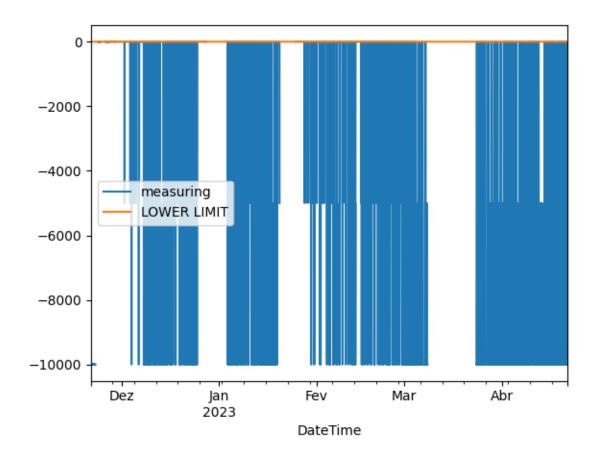


#### 2.2 Tag data

Apply tags to the data according to the quality control processes described above

```
[]: sensor_dataframe['Tag'] = 'VALID'
```

Tag values that are missing, are greater than sensor span or are lower than sensor resolution. The upper limit was too high that affected the graph resolution, therefore it is not plotted

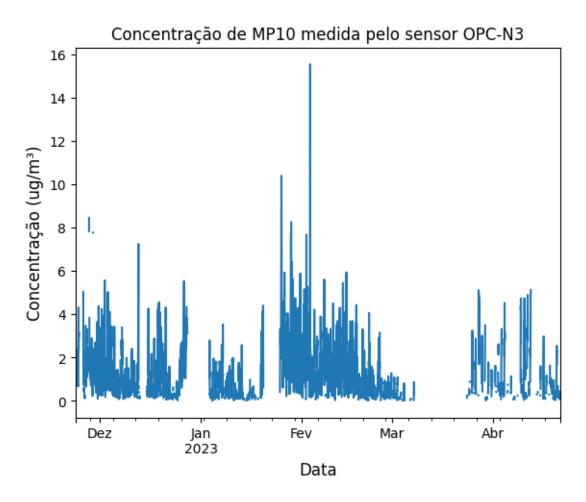


```
[]: import numpy as np
     from SensorDataAnalysisService import SensorDataAnalysisService as \square
      ⇔sensor_analysis
     def tag_by_sensor_limits(value, tag, lower_limit, upper_limit):
      if (tag != 'VALID'): return tag
      return sensor_analysis.get_tags_from_series(value=value,
                                                   lower_limit=lower_limit,
                                                   upper_limit=upper_limit)
     sensor_dataframe['Tag'] = (sensor_dataframe[['measuring', 'Tag']]
                                              .apply(lambda df:
      →tag_by_sensor_limits(value=df[0], tag=df[1],
                                                                                    Ш
      →lower_limit=lower_limit,
                                                                                    Ш

¬upper_limit=upper_limit),
                                                      axis=1))
```

Plot valid data after removing samples bellow lower limit

[]: Text(0, 0.5, 'Concentração (ug/m³)')

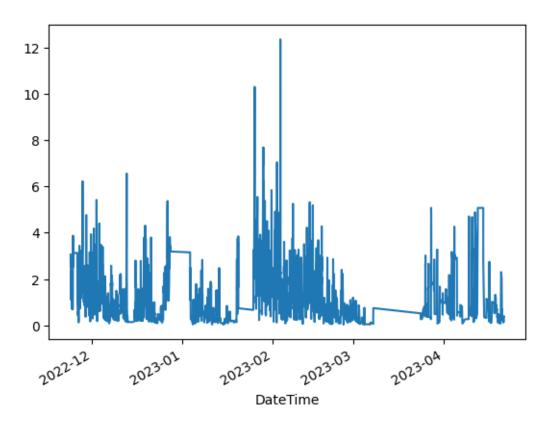


# 2.3 Analyse valid data

The graph shows the time series of the valid data, box plots of the data grouped by hour of the day, and histogram of the data.

#### 2.3.1 Smooth data with a window of 4 samples, i.e.: an hour of data

[]: <Axes: xlabel='DateTime'>

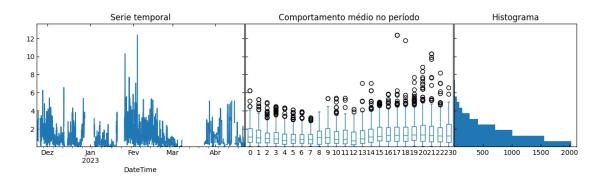


```
[]: valid_dataframe = valid_dataframe.resample('15T').mean()
sensor_analysis.plot_box_hist(df=valid_dataframe, bins=20)
```

/Users/Fernando/Documents/Projects/Github/lcqar-low-cost-monit-proc/data-pre-processing/SensorDataAnalysisService.py:111: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.

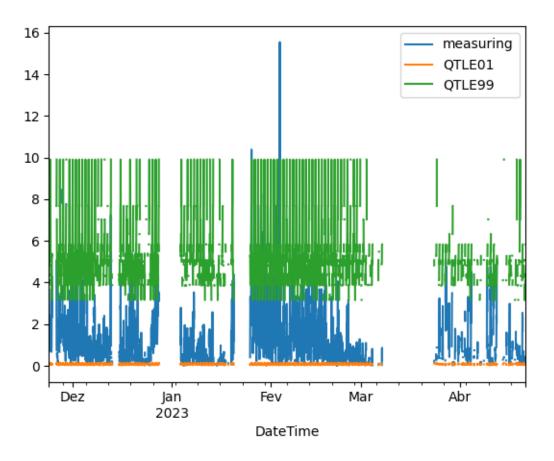
Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy df['Hour'] = df['Hour'].astype('int64')



#### 2.3.2 Calculate quantiles

We calculated the 1% and 99% quantiles of every hour of the day. The values greater outside the 1% - 99% where tagged as Greater that Quantile 99 (GTQTLE99) and Lower than Quantile 1 (LTQTLE01)



#### 2.3.3 Tag data according to quantiles

6481 44.570525

[]:

MISSING

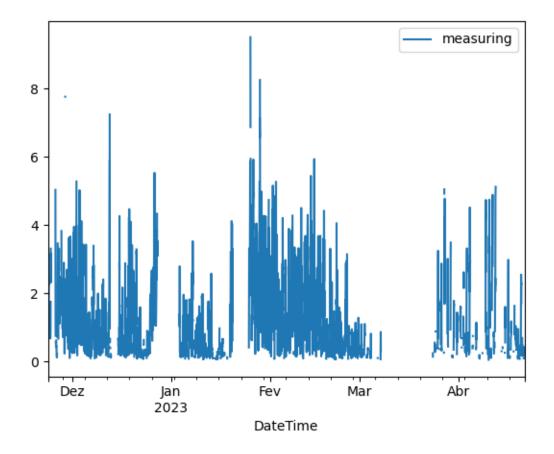
```
LTLL
            1759
                   12.09683
GTUL
               0
                         0.0
                         0.0
BADSPIKE
               0
VALID
            6098
                  41.936593
LTQTLE01
             117
                   0.804621
GTQTLE99
              86
                   0.591431
TOTAL
           14541
                       100.0
```

#### Plot valid data

```
[]: sensor_dataframe[sensor_dataframe['Tag'] == 'VALID'][['measuring']].

→resample('15T').mean().plot()
```

[]: <Axes: xlabel='DateTime'>

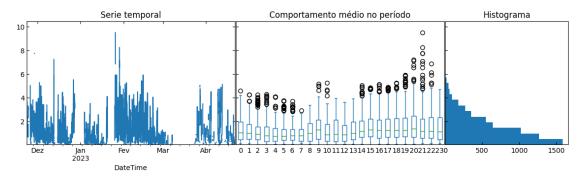


# Analyse data after removing quantiles

/Users/Fernando/Documents/Projects/Github/lcqar-low-cost-monit-proc/data-pre-

processing/SensorDataAnalysisService.py:111: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row\_indexer,col\_indexer] = value instead

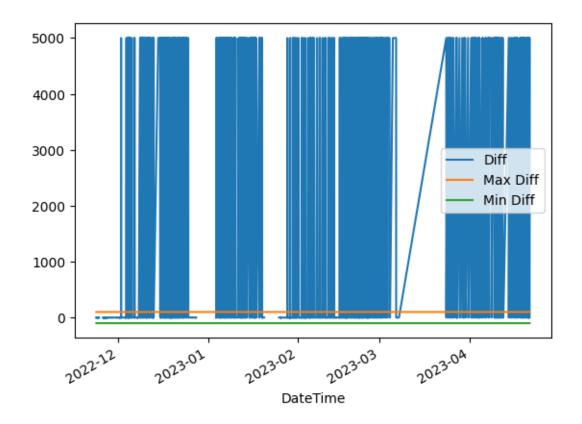
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy df['Hour'] = df['Hour'].astype('int64')



# 2.4 Analyse data derivatives

Analyse the derivatives of the data for removing abrupt changes in the data (BADSPIKE)

Tag values with derivatives faster than sensor response time The value of 100 ug/m3 was selected based on the maximum derivative found in the reference data. The maximum derivative (with measuring period of 1 hour) found was 150 ug/m3 ppb. Therefore, a maximum of 100 ug/m3 was considered appropriate for a 15 mins period.

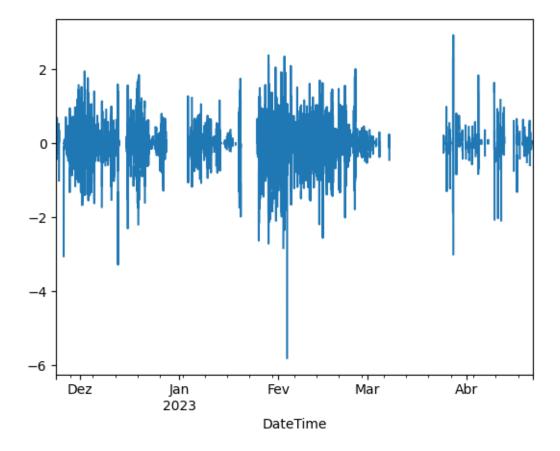


```
[]: import numpy as np
    def tag_data_with_derivatives(tagged_df, max_diff_value):
            current_tag = tagged_df[0]
            value = tagged_df[1]
            if ((current_tag != 'VALID') or (np.isnan(value))): return current_tag
            if ((value > max_diff_value) or (value < -max_diff_value)): return_
      →'BADSPIKE'
            return 'VALID'
    sensor_dataframe['Tag'] = (sensor_dataframe[['Tag', 'Diff', 'Max Diff']]
                              .apply(lambda df:
     →tag_data_with_derivatives(tagged_df=df, max_diff_value=df[2]),
                                      axis=1))
    tags = ['MISSING', 'LTLL', 'GTUL', 'STABILIZING', 'BADSPIKE', 'VALID',
     data_contabilization = sensor_analysis.count_tags(tags, sensor_dataframe)
    data_contabilization
```

```
[]:
     MISSING
                    6481
                          44.570525
     LTLL
                    1759
                           12.09683
     GTUL
                       0
                                0.0
                                0.0
     STABILIZING
                       0
     BADSPIKE
                     427
                           2.936524
     VALID
                    5671
                          39.000069
     LTQTLE01
                     117
                           0.804621
     GTQTLE99
                      86
                           0.591431
     REBASE
                       0
                                0.0
     TOTAL
                   14541
                              100.0
```

```
[]: sensor_dataframe[sensor_dataframe['Tag'] == 'VALID']['Diff'].resample('15T').

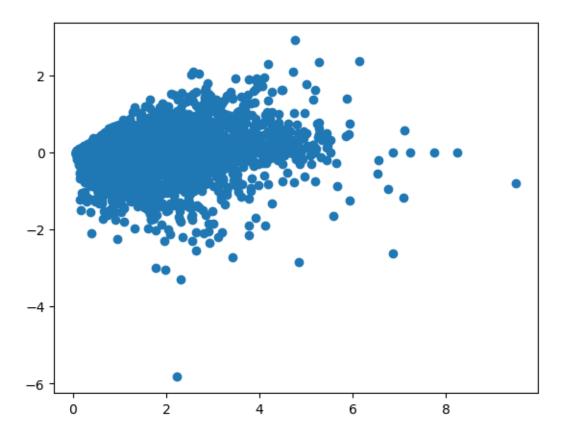
⇔mean().plot()
```



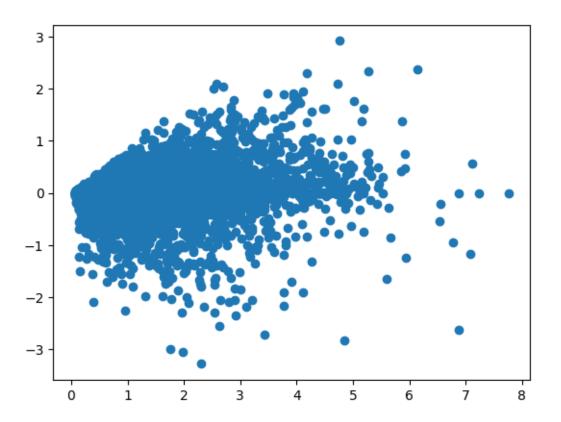
```
Plot data derivatives vs data
[]: import matplotlib.pyplot as plt
```

```
valid_dataframe = sensor_dataframe[sensor_dataframe['Tag'] == 'VALID']
fig = plt.figure(figsize=(1.3*5,5))
plt.scatter(valid_dataframe['measuring'], valid_dataframe['Diff'])
```

#### []: <matplotlib.collections.PathCollection at 0x168e1c220>



#### Remove sample with derivative < -4 and value 8



# Data Tag contabilization

```
[]: tags = ['MISSING','LTLL', 'GTUL', 'STABILIZING', 'BADSPIKE', 'VALID', GOVERNOUS COUNTLEONS ('ACCOUNT TO THE STABILIZING', 'BADSPIKE', 'VALID', GOVERNOUS COUNTLEONS ('ACCOUNT TO THE STABILIZING') ('ACCOUNT TO THE STABILIZING') ('ACCOUNT TO THE STABILIZING', 'BADSPIKE', 'VALID', GOVERNOUS COUNTLEONS ('ACCOUNT TO THE STABILIZING') ('ACCOUNT TO THE STAB
```

[]:		#	%
	MISSING	6481	44.570525
	LTLL	1759	12.09683
	GTUL	0	0.0
	STABILIZING	0	0.0
	BADSPIKE	430	2.957156
	VALID	5668	38.979437
	LTQTLE01	117	0.804621
	GTQTLE99	86	0.591431
	REBASE	0	0.0
	TOTAL	14541	100.0

# Plot valid data

```
[]: fig, ax = plt.subplots(figsize=(1.3*5,5))
```

```
sensor_dataframe[sensor_dataframe['Tag'] == 'VALID']['measuring'].

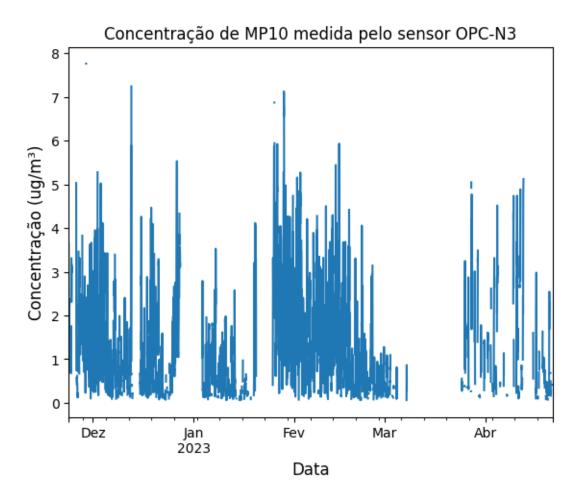
→rename('Concentração de MP10').resample('15T').mean().plot()

plt.title('Concentração de MP10 medida pelo sensor OPC-N3')

ax.set_xlabel('Data', fontsize=12)

ax.set_ylabel('Concentração (ug/m\N{SUPERSCRIPT THREE}))', fontsize=12)
```

[]: Text(0, 0.5, 'Concentração (ug/m³)')



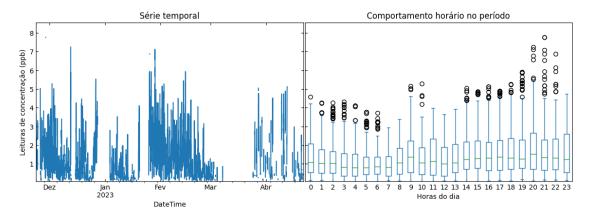
```
[]: valid_dataframe = sensor_dataframe[sensor_dataframe['Tag'] == 'VALID'].

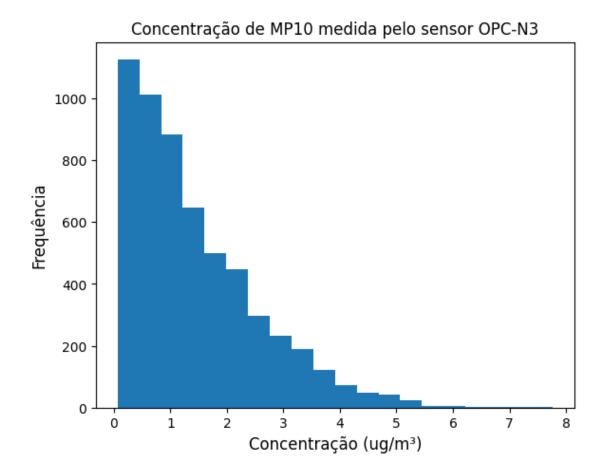
drop(columns=['Tag']).resample('15T').mean()
sensor_analysis.plot_box(df=valid_dataframe)
```

/Users/Fernando/Documents/Projects/Github/lcqar-low-cost-monit-proc/data-pre-processing/SensorDataAnalysisService.py:143: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-

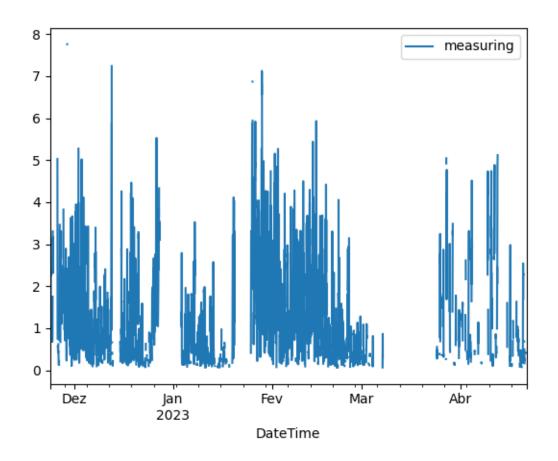
# docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy df['Hour'] = df['Hour'].astype('int64')





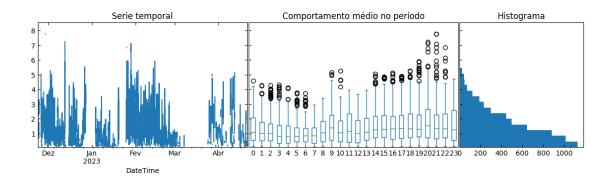
```
[]: sensor_dataframe[sensor_dataframe['Tag'] == 'VALID'][['measuring']].

resample('15T').mean().plot()
```

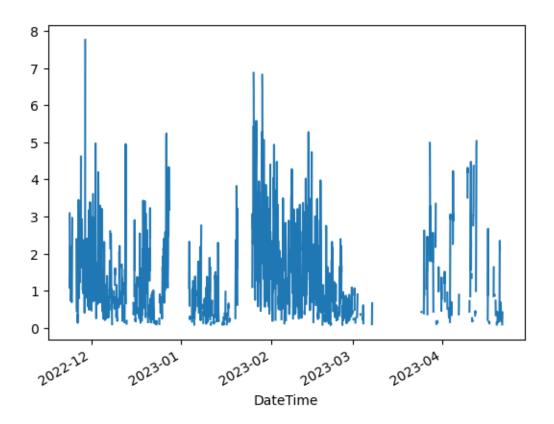


/Users/Fernando/Documents/Projects/Github/lcqar-low-cost-monit-proc/data-pre-processing/SensorDataAnalysisService.py:111: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy df['Hour'] = df['Hour'].astype('int64')

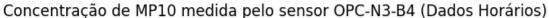


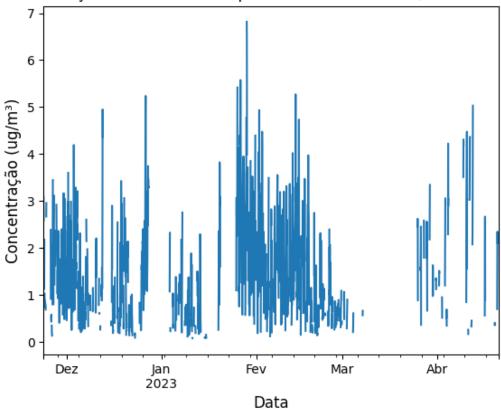
# 2.5 Resample valid data to 1 HR



Tag hourly data according to the number of samples in an hour At least 3 samples must be valid in an hour (75 %) for the hourly data be considered as valid

```
[]: Text(0, 0.5, 'Concentração (ug/m³)')
```





# Valid data contabilization

```
[]: tags = ['LOWSAMPLES','VALID']
data_contabilization = sensor_analysis.count_tags(tags, resampled_dataframe)
data_contabilization
```

[]: # % LOWSAMPLES 2269 63.3445 VALID 1313 36.6555 TOTAL 3582 100.0

## 2.6 Analyse the mean and standard deviation of the resampled data

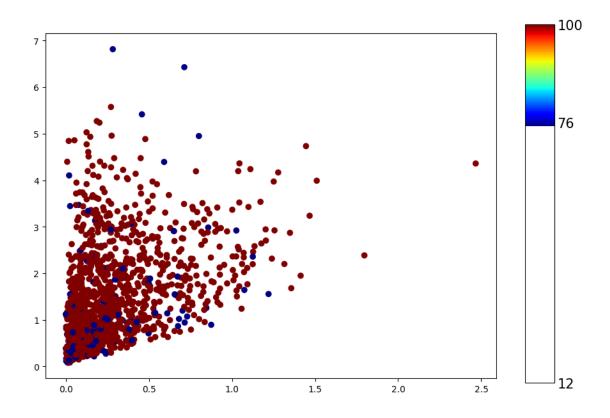
The color of each data point represents the percentage of valid 15 mins samples found in an hour

Plot mean (Y axis) vs. standard deviation (X axis) for valid hourly data colored by valid percentage of valid samples

```
[]: valid_resampled_dataframe = resampled_dataframe[resampled_dataframe['Tag'] == ∪

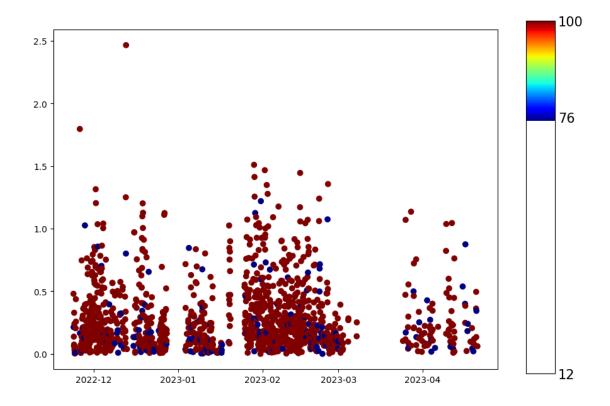
→'VALID']

sensor_analysis.plot_mean_vs_std(valid_resampled_dataframe)
```



2.6.1 Plot standard deviation in time colored according to the percentage of valid samples

[]: sensor\_analysis.plot\_std\_in\_time(valid\_resampled\_dataframe)



# 3 Save Data

```
valid_processing_file_path = processing_directory_path + SENSOR_NAME +_
 ⇔valid_filename
processing_dataframe_path_1HR = processing_directory_path + SENSOR_NAME +_

dataframe_1HR_filename

processing dataframe path = processing directory path + SENSOR_NAME + L

→dataframe_filename

if not os.path.exists(output_directory_path):
    os.makedirs(output_directory_path)
if not os.path.exists(processing_directory_path):
   os.makedirs(processing_directory_path)
sensor_dataframe[sensor_dataframe['Tag'] == 'VALID']['measuring'].
 →to_csv(valid_file_path_output)
resampled_dataframe[resampled_dataframe['Tag'] == 'VALID']['measuring'].
 oto_csv(valid_1HR_file_path_output)
sensor_dataframe.to_csv(dataframe_path_output)
resampled_dataframe.to_csv(dataframe_path_1HR_output)
sensor_dataframe[sensor_dataframe['Tag'] == 'VALID']['measuring'].
 sto_csv(valid_processing_file_path)
resampled_dataframe[resampled_dataframe['Tag'] == 'VALID']['measuring'].
 sto_csv(valid_processing_1HR_file_path)
sensor_dataframe.to_csv(processing_dataframe_path)
resampled_dataframe.to_csv(processing_dataframe_path_1HR)
```