1.Try to Define Business Problem and Metrics then Machine Learning Metrics

Business Problem:

Significant Customer Complaints towards low performances of Conventional Smoke Detector in detect the true fire trigger Sales softening trend for Months.

Business Metrics:

Customer Satisfaction Survey Score towards performances of Smoke Detector Product

Al Solution:

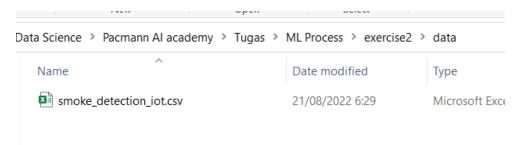
AI based Smoke Detector

Machine Learning Metrics

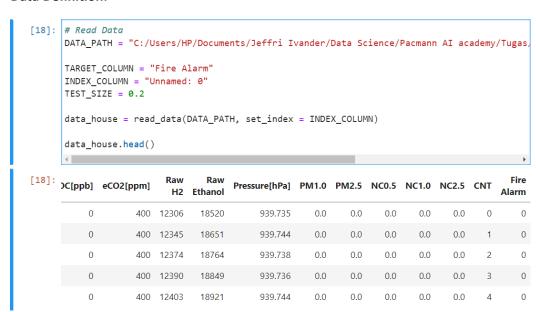
F1 Score of AI based Smoke Detector

2. Try Data pipelines, Especially definition; Validation; Defense; Splitting; and Serialization

Data Collection:



Data Definition:



```
datadescription.txt - Notepad
<u>File Edit Format View Help</u>
Timestamp UTC seconds
Temperature[C]
Air Temperature
Humidity[%]
Air Humidity
TVOC[ppb]
otal Volatile Organic Compounds; measured in parts per billion
co2 equivalent concentration; calculated from different values like TVCO
raw molecular hydrogen; not compensated (Bias, temperature, etc.)
Raw Ethanol
raw ethanol gas
Pressure[hPa]
Air Pressure
PM1.0
particulate matter size < 1.0 \mum (PM1.0). 1.0 \mum < 2.5 \mum (PM2.5)
```

Data Validation

All data is valid and No Null identified

```
[23]: #Data Validation
      print(data_house.info())
      print(f'data shape:{data_house.shape}')
      <class 'pandas.core.frame.DataFrame'>
      Int64Index: 62630 entries, 0 to 62629
      Data columns (total 15 columns):
                        Non-Null Count Dtype
       # Column
      ---
                         -----
                        62630 non-null int64
       0 UTC
          Temperature[C] 62630 non-null float64
       1
          Humidity[%]
                         62630 non-null float64
       2
                         62630 non-null int64
       3
          TVOC[ppb]
       4
          eCO2[ppm]
                         62630 non-null int64
       5
          Raw H2
                         62630 non-null int64
                         62630 non-null int64
          Raw Ethanol
       6
          Pressure[hPa] 62630 non-null float64
       8
          PM1.0
                       62630 non-null float64
          PM2.5
       9
                         62630 non-null float64
       10 NC0.5
                        62630 non-null float64
       11 NC1.0
                        62630 non-null float64
       12 NC2.5
                        62630 non-null float64
       13 CNT
                        62630 non-null int64
       14 Fire Alarm
                         62630 non-null int64
      dtypes: float64(8), int64(7)
      memory usage: 7.6 MB
      None
      data shape: (62630, 15)
```

----, ---,

```
[27]: data_house.describe()
[27]:
                      UTC Temperature[C] Humidity[%]
                                                           TVOC[ppb]
                                                                         eCO2[ppm]
                                                                                          Raw H2
                                                                                                   Raw Fth
       count 6.263000e+04
                               62630.000000 62630.000000
                                                          62630.000000
                                                                       62630.000000 62630.000000 62630.000
       mean 1.654792e+09
                                  15.970424
                                               48.539499
                                                           1942.057528
                                                                         670.021044 12942.453936 19754.25
         std 1.100025e+05
                                  14.359576
                                                8.865367
                                                           7811.589055
                                                                        1905.885439
                                                                                       272.464305
                                                                                                     609.513
         min 1.654712e+09
                                 -22.010000
                                               10.740000
                                                              0.000000
                                                                         400.000000 10668.000000 15317.000
             1.654743e+09
                                  10.994250
                                               47.530000
                                                            130.000000
                                                                         400.000000 12830.000000 19435.000
                                  20.130000
                                               50.150000
                                                            981.000000
                                                                         400.000000 12924.000000 19501.000
        50%
             1.654762e+09
         75% 1.654778e+09
                                  25.409500
                                               53.240000
                                                           1189.000000
                                                                         438.000000 13109.000000 20078.000
        max 1.655130e+09
                                  59.930000
                                               75.200000 60000.000000 60000.000000 13803.000000 21410.000
```

Data Defense

```
params = {
    'float_columns': [ 'Temperature[C]', 'Humidity[%]', 'Pressure[hPa]', 'PM1.0', 'PM2.5', 'N
    'int32_columns': ['UTC','TVOC[ppb]', 'eCO2[ppm]', 'Raw H2', 'Raw Ethanol', 'CNT', 'Fire
    }

print(params)

{'float_columns': ['Temperature[C]', 'Humidity[%]', 'Pressure[hPa]', 'PM1.0', 'PM2.5',
    'NC0.5', 'NC1.0', 'NC2.5'], 'int32_columns': ['UTC', 'TVOC[ppb]', 'eCO2[ppm]', 'Raw H
    2', 'Raw Ethanol', 'CNT', 'Fire Alarm']}

[34]: def check_data(input_data, params):
    # check data types
    assert input_data.select_dtypes("float").columns.to_list() == params["float_columns' assert input_data.select_dtypes("int").columns.to_list() == params["int32_columns"].

[35]: # jika tidak ada error berarti data sudah sesuai dengan desain kita
    check_data(data_house, params)
```

Data Splitting and Data Serialization

```
# pisahkan data x dan y (x adalah fitur, y adalah LabeL)

x = data_house[params["predictors"]].copy()

y = data_house['Fire Alarm'].copy()

[60]: x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.3, random_state = 42, stratify = y)

x_valid, x_test, y_valid, y_test = train_test_split(x_test, y_test, test_size = 0.5, random_state = 42, stratify = y)

x_valid, x_test, y_valid, y_test = train_test_split(x_test, y_test, test_size = 0.5, random_state = 42, stratify = y)

[61]: #Data SeriaLization
joblib.dump(x_train, "C:/Users/HP/Documents/Jeffri Ivander/Data Science/Pacmann AI academy/Tugas/ML Process/e
joblib.dump(x_train, "C:/Users/HP/Documents/Jeffri Ivander/Data Science/Pacmann AI academy/Tugas/ML Process/e
joblib.dump(x_valid, "C:/Users/HP/Documents/Jeffri Ivander/Data Science/Pacmann AI academy/Tugas/ML Process/e
joblib.dump(x_test, "C:/Users/HP/Documents/Jeffri Ivander/Data Science/Pacmann AI academy/Tugas/ML Process/ex
joblib.dump(x_test, "C:/Users/HP/Documents/Jeffri Ivander/Data Science/Pacmann AI academy/Tugas/M
```