IoT Botnet Detection-LSTM

June 23, 2023

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2 1. DDoS Detection on IoT - Unique Model - L5 - 23 Features - LSTM

2.1 1.1 Imports

```
[1]: import os
    os.environ['PYTHONHASHSEED'] = '0'
    os.environ["CUDA_VISIBLE_DEVICES"]="-1"
    os.environ["TF_CUDNN_USE_AUTOTUNE"]="0"

import numpy as np
    np.random.seed(2023)

import random as rn
    rn.seed(2023)

import pandas as pd
```

```
import matplotlib.pyplot as plt
%matplotlib inline
from datetime import datetime
import time
import tensorflow as tf
tf.random.set_seed(2023)
from tensorflow import keras
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import classification_report, accuracy_score
from sklearn.metrics import confusion_matrix
from sklearn.metrics import ConfusionMatrixDisplay
import warnings
warnings.filterwarnings('ignore')
from IPython.display import set_matplotlib_formats
set_matplotlib_formats('pdf', 'svg')
import seaborn as sns
# sns.set(color_codes=True)
from keras.layers import Input, Dropout, Dense, LSTM, TimeDistributed, __
  →RepeatVector
from keras.models import Model
from keras import regularizers
import json, codecs
2023-06-22 09:42:11.696610: I tensorflow/tsl/cuda/cudart_stub.cc:28] Could not
```

find cuda drivers on your machine, GPU will not be used.

2023-06-22 09:42:11.742599: I tensorflow/tsl/cuda/cudart_stub.cc:28] Could not find cuda drivers on your machine, GPU will not be used.

2023-06-22 09:42:11.743768: I tensorflow/core/platform/cpu_feature_guard.cc:182] This TensorFlow binary is optimized to use available CPU instructions in performance-critical operations.

To enable the following instructions: AVX2 FMA, in other operations, rebuild TensorFlow with the appropriate compiler flags.

2023-06-22 09:42:12.784296: W

tensorflow/compiler/tf2tensorrt/utils/py_utils.cc:38] TF-TRT Warning: Could not find TensorRT

3 2. Import Datasets and Normalize

3.1 2.1 Danmini Doorbell

```
[2]: # Benign traffic
    dd benign = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
     Danmini_Doorbell/benign_traffic.csv', encoding = "utf-8", sep = ',')
    df_dd_benign = dd_benign.copy(deep=True)
    columns = list(df_dd_benign.columns)
    chosen_columns = []
    for column in columns:
        if column.find('L5') != -1: # selection and a genus intervalo L5|
     →(100 ms)
            chosen_columns.append(column)
    df_dd_benign = pd.DataFrame(df_dd_benign, columns = chosen_columns)
     # Mirai attacks
    dd mirai_ack = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
      →Danmini_Doorbell/mirai_attacks/ack.csv', encoding = "utf-8", sep = ',' )
    df_dd_mirai_ack = dd_mirai_ack.copy(deep=True)
    df dd mirai ack = pd.DataFrame(df dd mirai ack, columns = chosen columns)
    dd mirai_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
     Danmini_Doorbell/mirai_attacks/scan.csv', encoding = "utf-8", sep = ',' )
    df dd mirai scan = dd mirai scan.copy(deep=True)
    df_dd_mirai_scan = pd.DataFrame(df_dd_mirai_scan, columns = chosen_columns)
    dd_mirai_syn = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/

    Danmini_Doorbell/mirai_attacks/syn.csv', encoding = "utf-8", sep = ',')

    df_dd_mirai_syn = dd_mirai_syn.copy(deep=True)
    df dd mirai syn = pd.DataFrame(df dd mirai syn, columns = chosen columns)
    dd mirai_udp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
     Danmini_Doorbell/mirai_attacks/udp.csv', encoding = "utf-8", sep = ',' )
    df dd mirai udp = dd mirai udp.copy(deep=True)
    df_dd_mirai_udp = pd.DataFrame(df_dd_mirai_udp, columns = chosen_columns)
    dd_mirai_udpplain = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
      ⇔Danmini Doorbell/mirai_attacks/udpplain.csv', encoding = "utf-8", sep = ',' )
    df_dd_mirai_udpplain = dd_mirai_udpplain.copy(deep=True)
    df_dd_mirai_udpplain = pd.DataFrame(df_dd_mirai_udpplain, columns =_u
      ⇔chosen_columns)
```

```
# Bashlite attacks
    dd bashlite_combo = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
      Danmini_Doorbell/gafgyt_attacks/combo.csv', encoding = "utf-8", sep = ',')
    df dd bashlite combo = dd bashlite combo.copy(deep=True)
    df_dd_bashlite_combo = pd.DataFrame(df_dd_bashlite_combo, columns =_
      ⇔chosen columns)
    dd_bashlite_junk = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
      Danmini_Doorbell/gafgyt_attacks/junk.csv', encoding = "utf-8", sep = ',')
    df dd bashlite junk = dd bashlite junk.copy(deep=True)
    df_dd_bashlite_junk = pd.DataFrame(df_dd_bashlite_junk, columns = u
      ⇔chosen columns)
    dd bashlite scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
      →Danmini_Doorbell/gafgyt_attacks/scan.csv', encoding = "utf-8", sep = ',')
    df dd bashlite scan = dd bashlite scan.copy(deep=True)
    df_dd_bashlite_scan = pd.DataFrame(df_dd_bashlite_scan, columns =__
      ⇔chosen columns)
    dd bashlite_udp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
      →Danmini_Doorbell/gafgyt_attacks/udp.csv', encoding = "utf-8", sep = ',')
    df_dd_bashlite_udp = dd_bashlite_udp.copy(deep=True)
    df dd bashlite udp = pd.DataFrame(df dd bashlite udp, columns = chosen columns)
    dd bashlite_tcp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
      Goding = "utf-8", sep = ',' )
    df dd bashlite tcp = dd bashlite tcp.copy(deep=True)
    df_dd_bashlite_tcp = pd.DataFrame(df_dd_bashlite_tcp, columns = chosen_columns)
[3]: df_dd_benign.head()
[3]:
       MI dir L5 weight MI dir L5 mean MI dir L5 variance H L5 weight \
    0
               1.000000
                              60.000000
                                                  0.000000
                                                               1.000000
    1
               1.000000
                             354.000000
                                                  0.000000
                                                               1.000000
    2
               1.857879
                             360.458980
                                                 35.789338
                                                               1.857879
    3
               1.000000
                             337.000000
                                                  0.000000
                                                               1.000000
               1.680223
                             172.140917
                                              18487.448750
                                                               1.680223
        H_L5_mean H_L5_variance HH_L5_weight HH_L5_mean HH_L5_std \
    0 60.000000
                        0.000000
                                      1.000000
                                                60.000000
                                                            0.000000
    1 354.000000
                        0.000005
                                      1.000000
                                               354.000000
                                                            0.002143
    2 360.458979
                       35.789342
                                      1.857879
                                               360.458979
                                                            5.982419
    3 337.000000
                                      1.000000 337.000000
                                                            0.000000
                        0.000000
    4 172.140917
                    18487.448750
                                      1.000000
                                                60.000000
                                                            0.000000
       HH_L5_magnitude ... HH_jit_L5_weight HH_jit_L5_mean HH_jit_L5_variance \
```

```
0
         60.000000
                                1.000000
                                             1.505662e+09
                                                                  0.000000e+00
1
        354.000000
                                             4.980575e+00
                                                                  4.230000e-07
                                1.000000
2
        360.458979
                                1.857879
                                             2.323596e+00
                                                                  6.056226e+00
3
                                                                  0.000000e+00
        337.000000
                                1.000000
                                             1.505662e+09
4
        524.399648
                                1.000000
                                             1.505662e+09
                                                                  0.000000e+00
   HpHp_L5_weight HpHp_L5_mean HpHp_L5_std HpHp_L5_magnitude
         1.000000
                       60.000000
                                      0.000000
                                                         60.000000
0
1
         1.000000
                      354.000000
                                      0.002143
                                                        354.000000
2
         1.857879
                      360.458979
                                      5.982419
                                                        360.458979
3
                      337.000000
                                      0.000000
                                                        337.000000
         1.000000
4
         1.000000
                       60.000000
                                      0.000000
                                                         60.000000
   HpHp_L5_radius
                    HpHp_L5_covariance
                                         HpHp_L5_pcc
0
         0.000000
                                   0.0
                                                 0.0
         0.000005
                                   0.0
                                                 0.0
1
2
        35.789342
                                   0.0
                                                 0.0
3
         0.000000
                                   0.0
                                                 0.0
4
         0.000000
                                   0.0
                                                 0.0
```

[5 rows x 23 columns]

```
[4]: scaler = MinMaxScaler()

df_dd_benign_norm = scaler.fit_transform(df_dd_benign)
df_dd_miraiack_norm = scaler.fit_transform(df_dd_mirai_ack)
df_dd_miraiscan_norm = scaler.fit_transform(df_dd_mirai_scan)
df_dd_miraisyn_norm = scaler.fit_transform(df_dd_mirai_syn)
df_dd_miraiudp_norm = scaler.fit_transform(df_dd_mirai_udp)
df_dd_miraiudpplain_norm = scaler.fit_transform(df_dd_mirai_udpplain)

df_dd_bashlitecombo_norm = scaler.fit_transform(df_dd_bashlite_combo)
df_dd_bashlitejunk_norm = scaler.fit_transform(df_dd_bashlite_junk)
df_dd_bashlitescan_norm = scaler.fit_transform(df_dd_bashlite_scan)
df_dd_bashliteudp_norm = scaler.fit_transform(df_dd_bashlite_udp)
df_dd_bashlitetcp_norm = scaler.fit_transform(df_dd_bashlite_tcp)
```

3.2 2.2 Ecobee Thermostat

```
et mirai ack = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
 ⇒Ecobee Thermostat/mirai_attacks/ack.csv', encoding = "utf-8", sep = ',')
df et mirai ack = et mirai ack.copy(deep=True)
df_et_mirai_ack = pd.DataFrame(df_et_mirai_ack, columns = chosen_columns)
et mirai scan = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
df_et_mirai_scan = et_mirai_scan.copy(deep=True)
df_et_mirai_scan = pd.DataFrame(df_et_mirai_scan, columns = chosen_columns)
et mirai syn = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
Geta to Ecobee_Thermostat/mirai_attacks/syn.csv', encoding = "utf-8", sep = ',' )
df_et_mirai_syn = et_mirai_syn.copy(deep=True)
df_et_mirai_syn = pd.DataFrame(df_et_mirai_syn, columns = chosen_columns)
et_mirai_udp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
GEcobee Thermostat/mirai_attacks/udp.csv', encoding = "utf-8", sep = ',' )
df_et_mirai_udp = et_mirai_udp.copy(deep=True)
df_et_mirai_udp = pd.DataFrame(df_et_mirai_udp, columns = chosen_columns)
et mirai udpplain = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
df_et_mirai_udpplain = et_mirai_udpplain.copy(deep=True)
df_et_mirai_udpplain = pd.DataFrame(df_et_mirai_udpplain, columns =__
 ⇔chosen_columns)
# Bashlite
et_bashlite_combo = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
Getable = "Looping = "utf-8", sep = ',' )
df_et_bashlite_combo = et_bashlite_combo.copy(deep=True)
df et bashlite combo = pd.DataFrame(df et bashlite combo, columns = 11
 ⇔chosen columns)
et_bashlite_junk = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 ⇒Ecobee_Thermostat/gafgyt_attacks/junk.csv', encoding = "utf-8", sep = ',')
df_et_bashlite_junk = et_bashlite_junk.copy(deep=True)
df_et_bashlite_junk = pd.DataFrame(df_et_bashlite_junk, columns =__
 ⇔chosen_columns)
et_bashlite_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 df_et_bashlite_scan = et_bashlite_scan.copy(deep=True)
```

```
[6]: df_et_benign_norm = scaler.fit_transform(df_et_benign)

df_et_miraiack_norm = scaler.fit_transform(df_et_mirai_ack)

df_et_miraiscan_norm = scaler.fit_transform(df_et_mirai_scan)

df_et_miraisyn_norm = scaler.fit_transform(df_et_mirai_syn)

df_et_miraiudp_norm = scaler.fit_transform(df_et_mirai_udp)

df_et_miraiudpplain_norm = scaler.fit_transform(df_et_mirai_udpplain)

df_et_bashlitecombo_norm = scaler.fit_transform(df_et_bashlite_combo)

df_et_bashlitejunk_norm = scaler.fit_transform(df_et_bashlite_junk)

df_et_bashlitescan_norm = scaler.fit_transform(df_et_bashlite_scan)

df_et_bashliteudp_norm = scaler.fit_transform(df_et_bashlite_udp)

df_et_bashlitetcp_norm = scaler.fit_transform(df_et_bashlite_tcp)
```

3.3 2.3 Ennio Doorbell

```
df_ed_bashlite_junk = ed_bashlite_junk.copy(deep=True)
df_ed_bashlite_junk = pd.DataFrame(df_ed_bashlite_junk, columns =__
 ⇔chosen_columns)
ed_bashlite_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 ⇔Ennio Doorbell/gafgyt attacks/scan.csv', encoding = "utf-8", sep = ',' )
df_ed_bashlite_scan = ed_bashlite_scan.copy(deep=True)
df_ed_bashlite_scan = pd.DataFrame(df_ed_bashlite_scan, columns =__
 ⇔chosen_columns)
ed_bashlite_udp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 ⇔Ennio_Doorbell/gafgyt_attacks/udp.csv', encoding = "utf-8", sep = ',' )
df_ed_bashlite_udp = ed_bashlite_udp.copy(deep=True)
df_ed_bashlite_udp = pd.DataFrame(df_ed_bashlite_udp, columns = chosen_columns)
ed_bashlite_tcp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
Genio_Doorbell/gafgyt_attacks/tcp.csv', encoding = "utf-8", sep = ',' )
df ed bashlite tcp = ed bashlite tcp.copy(deep=True)
df_ed_bashlite_tcp = pd.DataFrame(df_ed_bashlite_tcp, columns = chosen_columns)
```

```
[8]: df_ed_benign_norm = scaler.fit_transform(df_ed_benign)

df_ed_bashlitecombo_norm = scaler.fit_transform(df_ed_bashlite_combo)

df_ed_bashlitejunk_norm = scaler.fit_transform(df_ed_bashlite_junk)

df_ed_bashlitescan_norm = scaler.fit_transform(df_ed_bashlite_scan)

df_ed_bashliteudp_norm = scaler.fit_transform(df_ed_bashlite_udp)

df_ed_bashlitetcp_norm = scaler.fit_transform(df_ed_bashlite_tcp)
```

3.4 2.4 Philips Baby Monitor

```
pb_mirai_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →Philips B120N10 Baby Monitor/mirai attacks/scan.csv', encoding = "utf-8", □
 ⇔sep = ',' )
df_pb_mirai_scan = pb_mirai_scan.copy(deep=True)
df_pb_mirai_scan = pd.DataFrame(df_pb_mirai_scan, columns = chosen_columns)
pb_mirai_syn = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 ⇔Philips B120N10 Baby Monitor/mirai attacks/syn.csv', encoding = "utf-8", sep,
 df_pb_mirai_syn = pb_mirai_syn.copy(deep=True)
df pb_mirai_syn = pd.DataFrame(df pb_mirai_syn, columns = chosen_columns)
pb_mirai_udp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
→Philips_B120N10_Baby_Monitor/mirai_attacks/udp.csv', encoding = "utf-8", sep_
df_pb_mirai_udp = pb_mirai_udp.copy(deep=True)
df_pb_mirai_udp = pd.DataFrame(df_pb_mirai_udp, columns = chosen_columns)
pb_mirai_udpplain = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →Philips_B120N10_Baby_Monitor/mirai_attacks/udpplain.csv', encoding = U
df pb mirai udpplain = pb mirai udpplain.copy(deep=True)
df_pb_mirai_udpplain = pd.DataFrame(df_pb_mirai_udpplain, columns = u
 ⇔chosen_columns)
# Bashlite
pb_bashlite_combo = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
⇔Philips_B120N10_Baby_Monitor/gafgyt_attacks/combo.csv', encoding = "utf-8", ⊔
⇔sep = ',' )
df_pb_bashlite_combo = pb_bashlite_combo.copy(deep=True)
df_pb_bashlite_combo = pd.DataFrame(df_pb_bashlite_combo, columns = ___
 ⇔chosen_columns)
pb_bashlite_junk = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →Philips B120N10 Baby Monitor/gafgyt attacks/junk.csv', encoding = "utf-8",,,
⇔sep = ',' )
df_pb_bashlite_junk = pb_bashlite_junk.copy(deep=True)
df_pb_bashlite_junk = pd.DataFrame(df_pb_bashlite_junk, columns =__
 ⇔chosen columns)
pb_bashlite_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →Philips_B120N10_Baby_Monitor/gafgyt_attacks/scan.csv', encoding = "utf-8", __
⇔sep = ',' )
df_pb_bashlite_scan = pb_bashlite_scan.copy(deep=True)
```

```
df_pb_bashlite_scan = pd.DataFrame(df_pb_bashlite_scan, columns =__
       ⇔chosen_columns)
      pb bashlite udp = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
       →Philips_B120N10_Baby_Monitor/gafgyt_attacks/udp.csv', encoding = "utf-8", __
       ⇔sep = ',' )
      df_pb_bashlite_udp = pb_bashlite_udp.copy(deep=True)
      df_pb_bashlite_udp = pd.DataFrame(df_pb_bashlite_udp, columns = chosen_columns)
      pb bashlite_tcp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
       →Philips_B120N10_Baby_Monitor/gafgyt_attacks/tcp.csv', encoding = "utf-8", 
       ⇔sep = ',' )
      df_pb_bashlite_tcp = pb_bashlite_tcp.copy(deep=True)
      df_pb_bashlite_tcp = pd.DataFrame(df_pb_bashlite_tcp, columns = chosen_columns)
[10]: df pb benign norm = scaler.fit transform(df pb benign)
      df_pb_miraiack_norm = scaler.fit_transform(df_pb_mirai_ack)
      df_pb_miraiscan_norm = scaler.fit_transform(df_pb_mirai_scan)
      df_pb_miraisyn_norm = scaler.fit_transform(df_pb_mirai_syn)
      df_pb_miraiudp_norm = scaler.fit_transform(df_pb_mirai_udp)
      df_pb_miraiudpplain_norm = scaler.fit_transform(df_pb_mirai_udpplain)
      df_pb_bashlitecombo_norm = scaler.fit_transform(df_pb_bashlite_combo)
      df_pb_bashlitejunk_norm = scaler.fit_transform(df_pb_bashlite_junk)
      df_pb_bashlitescan_norm = scaler.fit_transform(df_pb_bashlite_scan)
      df_pb_bashliteudp_norm = scaler.fit_transform(df_pb_bashlite_udp)
      df_pb_bashlitetcp_norm = scaler.fit_transform(df_pb_bashlite_tcp)
```

3.5 2.5 Security Camera

```
p7_mirai_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 ⇔Provision PT_737E_Security_Camera/mirai_attacks/scan.csv', encoding = □
 df_p7_mirai_scan = p7_mirai_scan.copy(deep=True)
df p7 mirai scan = pd.DataFrame(df p7 mirai scan, columns = chosen columns)
p7_mirai_syn = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →Provision PT_737E Security Camera/mirai_attacks/syn.csv', encoding =
 df_p7_mirai_syn = p7_mirai_syn.copy(deep=True)
df p7 mirai syn = pd.DataFrame(df p7 mirai syn, columns = chosen columns)
p7_mirai_udp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 ⊶Provision_PT_737E_Security_Camera/mirai_attacks/udp.csv', encoding = U
 df_p7_mirai_udp = p7_mirai_udp.copy(deep=True)
df p7_mirai_udp = pd.DataFrame(df_p7_mirai_udp, columns = chosen_columns)
p7_mirai_udpplain = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →Provision_PT_737E_Security_Camera/mirai_attacks/udpplain.csv', encoding = U

y"utf-8", sep = ',' )

df_p7_mirai_udpplain = p7_mirai_udpplain.copy(deep=True)
df_p7_mirai_udpplain = pd.DataFrame(df_p7_mirai_udpplain, columns =_u
 ⇔chosen_columns)
# Bashlite
p7_bashlite_combo = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 ⊶Provision_PT_737E_Security_Camera/gafgyt_attacks/combo.csv', encoding =

"utf-8", sep = ',' )

df_p7_bashlite_combo = p7_bashlite_combo.copy(deep=True)
df_p7_bashlite_combo = pd.DataFrame(df_p7_bashlite_combo, columns =__
 ⇔chosen_columns)
p7 bashlite junk = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
 →Provision_PT_737E_Security_Camera/gafgyt_attacks/junk.csv', encoding = 
 df_p7_bashlite_junk = p7_bashlite_junk.copy(deep=True)
df_p7_bashlite_junk = pd.DataFrame(df_p7_bashlite_junk, columns =_
 ⇔chosen_columns)
p7_bashlite_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →Provision_PT_737E_Security_Camera/gafgyt_attacks/scan.csv', encoding =

''utf-8", sep = ',' )

df_p7_bashlite_scan = p7_bashlite_scan.copy(deep=True)
```

```
df_p7_bashlite scan = pd.DataFrame(df_p7_bashlite scan, columns =__
       ⇔chosen_columns)
      p7 bashlite udp = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
       →Provision_PT_737E_Security_Camera/gafgyt_attacks/udp.csv', encoding = U

"utf-8", sep = ',' )

      df_p7_bashlite_udp = p7_bashlite_udp.copy(deep=True)
      df_p7_bashlite_udp = pd.DataFrame(df_p7_bashlite_udp, columns = chosen_columns)
      p7_bashlite_tcp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
       →Provision_PT_737E_Security_Camera/gafgyt_attacks/tcp.csv', encoding = U

''utf-8", sep = ',' )

      df_p7_bashlite_tcp = p7_bashlite_tcp.copy(deep=True)
      df_p7_bashlite_tcp = pd.DataFrame(df_p7_bashlite_tcp, columns = chosen_columns)
[12]: df p7 benign norm = scaler.fit transform(df p7 benign)
      df_p7_miraiack_norm = scaler.fit_transform(df_p7_mirai_ack)
      df_p7_miraiscan_norm = scaler.fit_transform(df_p7_mirai_scan)
      df_p7_miraisyn_norm = scaler.fit_transform(df_p7_mirai_syn)
      df_p7_miraiudp_norm = scaler.fit_transform(df_p7_mirai_udp)
      df_p7_miraiudpplain_norm = scaler.fit_transform(df_p7_mirai_udpplain)
      df_p7_bashlitecombo_norm = scaler.fit_transform(df_p7_bashlite_combo)
      df_p7_bashlitejunk_norm = scaler.fit_transform(df_p7_bashlite_junk)
      df_p7_bashlitescan_norm = scaler.fit_transform(df_p7_bashlite_scan)
      df_p7_bashliteudp_norm = scaler.fit_transform(df_p7_bashlite_udp)
      df_p7_bashlitetcp_norm = scaler.fit_transform(df_p7_bashlite_tcp)
```

3.6 2.6 Security Camera

```
p8_mirai_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 ⇔Provision PT_838_Security_Camera/mirai attacks/scan.csv', encoding = □
 df_p8_mirai_scan = p8_mirai_scan.copy(deep=True)
df p8 mirai scan = pd.DataFrame(df p8 mirai scan, columns = chosen columns)
p8_mirai_syn = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →Provision PT_838_Security_Camera/mirai_attacks/syn.csv', encoding = "utf-8", __
 ⇔sep = ',' )
df_p8_mirai_syn = p8_mirai_syn.copy(deep=True)
df p8 mirai syn = pd.DataFrame(df p8 mirai syn, columns = chosen columns)
p8_mirai_udp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 ⇔Provision_PT_838_Security_Camera/mirai_attacks/udp.csv', encoding = "utf-8", __
 ⇔sep = ',' )
df_p8_mirai_udp = p8_mirai_udp.copy(deep=True)
df p8_mirai_udp = pd.DataFrame(df p8 mirai_udp, columns = chosen_columns)
p8_mirai_udpplain = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →Provision_PT_838_Security_Camera/mirai_attacks/udpplain.csv', encoding = U

''utf-8", sep = ',' )

df_p8_mirai_udpplain = p8_mirai_udpplain.copy(deep=True)
df_p8_mirai_udpplain = pd.DataFrame(df_p8_mirai_udpplain, columns =_u
 ⇔chosen_columns)
# Bashlite
p8_bashlite_combo = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 ⊶Provision_PT_838_Security_Camera/gafgyt_attacks/combo.csv', encoding = U

"utf-8", sep = ',' )

df_p8_bashlite_combo = p8_bashlite_combo.copy(deep=True)
df_p8_bashlite_combo = pd.DataFrame(df_p8_bashlite_combo, columns =__
 ⇔chosen_columns)
p8 bashlite junk = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
 →Provision_PT_838_Security_Camera/gafgyt_attacks/junk.csv', encoding = 
df_p8_bashlite_junk = p8_bashlite_junk.copy(deep=True)
df_p8_bashlite_junk = pd.DataFrame(df_p8_bashlite_junk, columns =__
 ⇔chosen_columns)
p8_bashlite_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →Provision_PT_838_Security_Camera/gafgyt_attacks/scan.csv', encoding = U

''utf-8", sep = ',' )

df_p8_bashlite_scan = p8_bashlite_scan.copy(deep=True)
```

```
df_p8_bashlite_scan = pd.DataFrame(df_p8_bashlite_scan, columns =__
       ⇔chosen_columns)
      p8 bashlite udp = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
       ⊶Provision_PT_838_Security_Camera/gafgyt_attacks/udp.csv', encoding = U

"utf-8", sep = ',' )

      df_p8_bashlite_udp = p8_bashlite_udp.copy(deep=True)
      df_p8_bashlite_udp = pd.DataFrame(df_p8_bashlite_udp, columns = chosen_columns)
      p8 bashlite_tcp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
       →Provision PT_838 Security_Camera/gafgyt_attacks/tcp.csv', encoding = U

y"utf-8", sep = ',' )

      df_p8_bashlite_tcp = p8_bashlite_tcp.copy(deep=True)
      df_p8_bashlite_tcp = pd.DataFrame(df_p8_bashlite_tcp, columns = chosen_columns)
[14]: df p8 benign norm = scaler.fit transform(df p8 benign)
      df_p8_miraiack_norm = scaler.fit_transform(df_p8_mirai_ack)
      df_p8_miraiscan_norm = scaler.fit_transform(df_p8_mirai_scan)
      df_p8_miraisyn_norm = scaler.fit_transform(df_p8_mirai_syn)
      df_p8_miraiudp_norm = scaler.fit_transform(df_p8_mirai_udp)
      df_p8_miraiudpplain_norm = scaler.fit_transform(df_p8_mirai_udpplain)
      df_p8_bashlitecombo_norm = scaler.fit_transform(df_p8_bashlite_combo)
      df_p8_bashlitejunk_norm = scaler.fit_transform(df_p8_bashlite_junk)
      df_p8_bashlitescan_norm = scaler.fit_transform(df_p8_bashlite_scan)
      df_p8_bashliteudp_norm = scaler.fit_transform(df_p8_bashlite_udp)
      df_p8_bashlitetcp_norm = scaler.fit_transform(df_p8_bashlite_tcp)
```

3.7 2.7 Security Camera

```
s2 mirai_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 SimpleHome XCS7_1002 WHT_Security_Camera/mirai_attacks/scan.csv', encoding = 1
df_s2_mirai_scan = s2_mirai_scan.copy(deep=True)
df s2 mirai scan = pd.DataFrame(df s2 mirai scan, columns = chosen columns)
s2 mirai syn = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
 →SimpleHome XCS7 1002 WHT Security Camera/mirai attacks/syn.csv', encoding =
df_s2_mirai_syn = s2_mirai_syn.copy(deep=True)
df s2 mirai syn = pd.DataFrame(df s2 mirai syn, columns = chosen columns)
s2_mirai_udp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →SimpleHome_XCS7_1002_WHT_Security_Camera/mirai_attacks/udp.csv', encoding =
 df_s2_mirai_udp = s2_mirai_udp.copy(deep=True)
df s2_mirai_udp = pd.DataFrame(df s2_mirai_udp, columns = chosen_columns)
s2_mirai_udpplain = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →SimpleHome_XCS7_1002_WHT_Security_Camera/mirai_attacks/udpplain.csv', _
 ⊖encoding = "utf-8", sep = ',')
df_s2_mirai_udpplain = s2_mirai_udpplain.copy(deep=True)
df_s2_mirai_udpplain = pd.DataFrame(df_s2_mirai_udpplain, columns =_u
 ⇔chosen_columns)
# Bashlite
s2_bashlite_combo = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 SimpleHome_XCS7_1002_WHT_Security_Camera/gafgyt_attacks/combo.csv', encoding_
⇔= "utf-8", sep = ',' )
df_s2_bashlite_combo = s2_bashlite_combo.copy(deep=True)
df_s2_bashlite_combo = pd.DataFrame(df_s2_bashlite_combo, columns = ___
 ⇔chosen_columns)
s2 bashlite junk = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
 SimpleHome_XCS7_1002_WHT_Security_Camera/gafgyt_attacks/junk.csv', encoding_
⇔= "utf-8", sep = ',' )
df_s2_bashlite_junk = s2_bashlite_junk.copy(deep=True)
df_s2_bashlite_junk = pd.DataFrame(df_s2_bashlite_junk, columns =_
 ⇔chosen_columns)
s2_bashlite_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →SimpleHome_XCS7_1002_WHT_Security_Camera/gafgyt_attacks/scan.csv', encoding_
 ⇔= "utf-8", sep = ',' )
df_s2_bashlite_scan = s2_bashlite_scan.copy(deep=True)
```

```
df_s2_bashlite scan = pd.DataFrame(df_s2_bashlite scan, columns =_
       ⇔chosen_columns)
     s2 bashlite udp = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
       SimpleHome_XCS7_1002_WHT_Security_Camera/gafgyt_attacks/udp.csv', encoding = 
      df_s2_bashlite_udp = s2_bashlite_udp.copy(deep=True)
     df_s2_bashlite_udp = pd.DataFrame(df_s2_bashlite_udp, columns = chosen_columns)
     s2 bashlite_tcp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
      SimpleHome_XCS7_1002_WHT_Security_Camera/gafgyt_attacks/tcp.csv', encoding = ___
       df_s2_bashlite_tcp = s2_bashlite_tcp.copy(deep=True)
     df_s2_bashlite_tcp = pd.DataFrame(df_s2_bashlite_tcp, columns = chosen_columns)
[16]: df s2 benign norm = scaler.fit transform(df s2 benign)
     df_s2_miraiack_norm = scaler.fit_transform(df_s2_mirai_ack)
     df_s2_miraiscan_norm = scaler.fit_transform(df_s2_mirai_scan)
     df_s2_miraisyn_norm = scaler.fit_transform(df_s2_mirai_syn)
     df_s2_miraiudp_norm = scaler.fit_transform(df_s2_mirai_udp)
     df_s2_miraiudpplain_norm = scaler.fit_transform(df_s2_mirai_udpplain)
     df_s2_bashlitecombo_norm = scaler.fit_transform(df_s2_bashlite_combo)
     df_s2_bashlitejunk_norm = scaler.fit_transform(df_s2_bashlite_junk)
     df_s2_bashlitescan_norm = scaler.fit_transform(df_s2_bashlite_scan)
     df_s2_bashliteudp_norm = scaler.fit_transform(df_s2_bashlite_udp)
     df_s2_bashlitetcp_norm = scaler.fit_transform(df_s2_bashlite_tcp)
```

3.8 2.8 Security Camera

```
s3 mirai_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 SimpleHome XCS7 1003 WHT Security Camera/mirai_attacks/scan.csv', encoding = 1
df_s3_mirai_scan = s3_mirai_scan.copy(deep=True)
df s3 mirai scan = pd.DataFrame(df s3 mirai scan, columns = chosen columns)
s3_mirai_syn = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →SimpleHome XCS7 1003 WHT Security Camera/mirai attacks/syn.csv', encoding =
df_s3_mirai_syn = s3_mirai_syn.copy(deep=True)
df s3 mirai syn = pd.DataFrame(df s3 mirai syn, columns = chosen columns)
s3_mirai_udp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →SimpleHome_XCS7_1003_WHT_Security_Camera/mirai_attacks/udp.csv', encoding =

¬"utf-8", sep = ',' )

df_s3_mirai_udp = s3_mirai_udp.copy(deep=True)
df s3_mirai_udp = pd.DataFrame(df s3_mirai_udp, columns = chosen_columns)
s3_mirai_udpplain = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
SimpleHome_XCS7_1003_WHT_Security_Camera/mirai_attacks/udpplain.csv',
 ⊖encoding = "utf-8", sep = ',')
df_s3_mirai_udpplain = s3_mirai_udpplain.copy(deep=True)
df_s3_mirai_udpplain = pd.DataFrame(df_s3_mirai_udpplain, columns =_u
 ⇔chosen_columns)
# Bashlite
s3_bashlite_combo = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 SimpleHome_XCS7_1003_WHT_Security_Camera/gafgyt_attacks/combo.csv', encoding_
df_s3_bashlite_combo = s3_bashlite_combo.copy(deep=True)
df_s3_bashlite_combo = pd.DataFrame(df_s3_bashlite_combo, columns = ___
 ⇔chosen_columns)
s3 bashlite junk = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
 SimpleHome_XCS7_1003_WHT_Security_Camera/gafgyt_attacks/junk.csv', encoding_
⇔= "utf-8", sep = ',' )
df_s3_bashlite_junk = s3_bashlite_junk.copy(deep=True)
df_s3_bashlite_junk = pd.DataFrame(df_s3_bashlite_junk, columns =_
 ⇔chosen_columns)
s3_bashlite_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
 →SimpleHome_XCS7_1003_WHT_Security_Camera/gafgyt_attacks/scan.csv', encoding_
 ⇔= "utf-8", sep = ',' )
df_s3_bashlite_scan = s3_bashlite_scan.copy(deep=True)
```

```
df_s3_bashlite scan = pd.DataFrame(df_s3_bashlite scan, columns =_
       ⇔chosen_columns)
     s3 bashlite udp = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
       SimpleHome_XCS7_1003_WHT_Security_Camera/gafgyt_attacks/udp.csv', encoding = 
      df_s3_bashlite_udp = s3_bashlite_udp.copy(deep=True)
     df_s3_bashlite_udp = pd.DataFrame(df_s3_bashlite_udp, columns = chosen_columns)
     s3 bashlite_tcp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
       SimpleHome_XCS7_1003_WHT_Security_Camera/gafgyt_attacks/tcp.csv', encoding = ___

y"utf-8", sep = ',' )

     df_s3_bashlite_tcp = s3_bashlite_tcp.copy(deep=True)
     df_s3_bashlite_tcp = pd.DataFrame(df_s3_bashlite_tcp, columns = chosen_columns)
[18]: df s3 benign norm = scaler.fit transform(df s3 benign)
     df_s3_miraiack_norm = scaler.fit_transform(df_s3_mirai_ack)
     df_s3_miraiscan_norm = scaler.fit_transform(df_s3_mirai_scan)
     df_s3_miraisyn_norm = scaler.fit_transform(df_s3_mirai_syn)
     df_s3_miraiudp_norm = scaler.fit_transform(df_s3_mirai_udp)
     df_s3_miraiudpplain_norm = scaler.fit_transform(df_s3_mirai_udpplain)
     df_s3_bashlitecombo_norm = scaler.fit_transform(df_s3_bashlite_combo)
     df_s3_bashlitejunk_norm = scaler.fit_transform(df_s3_bashlite_junk)
     df_s3_bashlitescan_norm = scaler.fit_transform(df_s3_bashlite_scan)
     df_s3_bashliteudp_norm = scaler.fit_transform(df_s3_bashlite_udp)
     df_s3_bashlitetcp_norm = scaler.fit_transform(df_s3_bashlite_tcp)
```

3.9 2.9 Samsung Webcam

```
df_sw_bashlite_combo = pd.DataFrame(df_sw_bashlite_combo, columns =__
       ⇔chosen columns)
      sw bashlite junk = pd.read csv('/mnt/extra/2023-1 10Periodo/Poc II/nbaiot/
       Samsung_SNH_1011_N_Webcam/gafgyt_attacks/junk.csv', encoding = "utf-8", sep_
       df_sw_bashlite_junk = sw_bashlite_junk.copy(deep=True)
      df_sw_bashlite_junk = pd.DataFrame(df_sw_bashlite_junk, columns =_
       ⇔chosen_columns)
      sw_bashlite_scan = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
       →Samsung_SNH_1011_N_Webcam/gafgyt_attacks/scan.csv', encoding = "utf-8", sep_
       df_sw_bashlite_scan = sw_bashlite_scan.copy(deep=True)
      df_sw_bashlite_scan = pd.DataFrame(df_sw_bashlite_scan, columns =__
       ⇔chosen_columns)
      sw_bashlite_udp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
       Samsung_SNH_1011_N_Webcam/gafgyt_attacks/udp.csv', encoding = "utf-8", sep = ∪
       \hookrightarrow 1, 1)
      df_sw_bashlite_udp = sw_bashlite_udp.copy(deep=True)
      df_sw_bashlite_udp = pd.DataFrame(df_sw_bashlite_udp, columns = chosen_columns)
      sw_bashlite_tcp = pd.read_csv('/mnt/extra/2023-1_10Periodo/Poc_II/nbaiot/
       Samsung_SNH_1011_N_Webcam/gafgyt_attacks/tcp.csv', encoding = "utf-8", sep =
      \hookrightarrow<sup>1</sup>, <sup>1</sup>)
      df_sw_bashlite_tcp = sw_bashlite_tcp.copy(deep=True)
      df_sw_bashlite_tcp = pd.DataFrame(df_sw_bashlite_tcp, columns = chosen_columns)
[20]: df_sw_benign_norm = scaler.fit_transform(df_sw_benign)
      df_sw_bashlitecombo_norm = scaler.fit_transform(df_sw_bashlite_combo)
      df_sw_bashlitejunk_norm = scaler.fit_transform(df_sw_bashlite_junk)
      df_sw_bashlitescan_norm = scaler.fit_transform(df_sw_bashlite scan)
      df_sw_bashliteudp_norm = scaler.fit_transform(df_sw_bashlite_udp)
      df_sw_bashlitetcp_norm = scaler.fit_transform(df_sw_bashlite_tcp)
```

4 3. LSTM Autoencoder - Attack Detection

4.1 3.1 Model

```
[21]: # Train set
len_dd_benign_train = int(0.7 * len(df_dd_benign_norm))
X_train_dd_benign = df_dd_benign_norm[:len_dd_benign_train]
len_et_benign_train = int(0.7 * len(df_et_benign_norm))
```

```
X_train_et_benign = df_et_benign_norm[:len_et_benign_train]
len_ed_benign_train = int(0.7 * len(df_ed_benign_norm))
X_train_ed_benign = df_ed_benign_norm[:len_ed_benign_train]
len_pb_benign_train = int(0.7 * len(df_pb_benign_norm))
X_train_pb_benign = df_pb_benign_norm[:len_pb_benign_train]
len p7 benign train = int(0.7 * len(df p7 benign norm))
X_train_p7_benign = df_p7_benign_norm[:len_p7_benign_train]
len_p8_benign_train = int(0.7 * len(df_p8_benign_norm))
X_train_p8_benign = df_p8_benign_norm[:len_p8_benign_train]
len_s2_benign_train = int(0.7 * len(df_s2_benign_norm))
X_train_s2_benign = df_s2_benign_norm[:len_s2_benign_train]
len_s3_benign_train = int(0.7 * len(df_s3_benign_norm))
X_train_s3_benign = df_s3_benign_norm[:len_s3_benign_train]
len_sw_benign_train = int(0.7 * len(df_sw_benign_norm))
X_train_sw_benign = df_sw_benign_norm[:len_sw_benign_train]
X train AE = np.concatenate([X train dd benign, X train et benign, |
 →X_train_ed_benign, X_train_pb_benign, X_train_p7_benign,
                              X_train_p8_benign, X_train_s2_benign,__

→X_train_s3_benign])
# Test set - 30% benign and the rest is attack
X_test_dd_benign = df_dd_benign_norm[len_dd_benign_train:]
X_test_et_benign = df_et_benign_norm[len_et_benign_train:]
X_test_ed_benign = df_ed_benign_norm[len_ed_benign_train:]
X_test_pb_benign = df_pb_benign_norm[len_pb_benign_train:]
X test p7 benign = df p7 benign norm[len p7 benign train:]
X_test_p8_benign = df_p8_benign_norm[len_p8_benign_train:]
X_test_s2_benign = df_s2_benign_norm[len_s2_benign_train:]
X_test_s3_benign = df_s3_benign_norm[len_s3_benign_train:]
X_test_sw_benign = df_sw_benign_norm[len_sw_benign_train:]
X test_benign = np.concatenate([X_test_dd_benign, X_test_et_benign, u
 →X_test_ed_benign, X_test_pb_benign, X_test_p7_benign,
                                X_test_p8_benign, X_test_s2_benign,__
→X_test_s3_benign, X_test_sw_benign])
# 30% benign + attacks
X_test_AE = np.concatenate([X_test_benign,
```

```
df_dd_miraiack_norm, df_dd_miraiscan_norm,__
 odf_dd miraisyn norm, df_dd miraiudp norm, df_dd miraiudpplain norm,
                             df_dd_bashlitecombo_norm, df_dd_bashlitejunk_norm,
 df_dd_bashlitescan_norm, df_dd_bashliteudp_norm, df_dd_bashlitetcp_norm,
                             df_et_miraiack_norm, df_et_miraiscan_norm,__
 df_et_miraisyn_norm, df_et_miraiudp_norm, df_et_miraiudpplain_norm,
                             df_et_bashlitecombo_norm, df_et_bashlitejunk_norm,__
 df_et_bashlitescan_norm, df_et_bashliteudp_norm, df_et_bashlitetcp_norm,
                             df_ed_bashlitecombo_norm, df_ed_bashlitejunk_norm,
 df_ed_bashlitescan_norm, df_ed_bashliteudp_norm, df_ed_bashlitetcp_norm,
                             df_pb_miraiack_norm, df_pb_miraiscan_norm,__
 df_pb_miraisyn_norm, df_pb_miraiudp_norm, df_pb_miraiudpplain_norm,
                             df_pb_bashlitecombo_norm, df_pb_bashlitejunk_norm,_
 →df_pb_bashlitescan_norm, df_pb_bashliteudp_norm, df_pb_bashlitetcp_norm,
                             df_p7_miraiack_norm, df_p7_miraiscan_norm,
 →df_p7_miraisyn_norm, df_p7_miraiudp_norm, df_p7_miraiudpplain_norm,
                             df p7 bashlitecombo norm, df p7 bashlitejunk norm,
 df_p7_bashlitescan_norm, df_p7_bashliteudp_norm, df_p7_bashlitetcp_norm,
                             df_p8_miraiack_norm, df_p8_miraiscan_norm,_
 →df_p8_miraisyn_norm, df_p8_miraiudp_norm, df_p8_miraiudpplain_norm,
                             df_p8_bashlitecombo_norm, df_p8_bashlitejunk_norm,
 df_p8_bashlitescan_norm, df_p8_bashliteudp_norm, df_p8_bashlitetcp_norm,
                             df_s2_miraiack_norm, df_s2_miraiscan_norm,
 df_s2_miraisyn_norm, df_s2_miraiudp_norm, df_s2_miraiudpplain_norm,
                             df_s2_bashlitecombo_norm, df_s2_bashlitejunk_norm,
 df_s2_bashlitescan_norm, df_s2_bashliteudp_norm, df_s2_bashlitetcp_norm,
                             df_s3_miraiack_norm, df_s3_miraiscan_norm,
 df_s3_miraisyn_norm, df_s3_miraiudp_norm, df_s3_miraiudpplain_norm,
                             df s3 bashlitecombo norm, df s3 bashlitejunk norm,
 df_s3_bashlitescan_norm, df_s3_bashliteudp_norm, df_s3_bashlitetcp_norm,
                             df_sw_bashlitecombo_norm, df_sw_bashlitejunk_norm,
 df_sw_bashlitescan_norm, df_sw_bashliteudp_norm, df_sw_bashlitetcp_norm])
X_train_label = np.zeros(len(X_train_AE))
Y_test_AE = np.ones(len(X_test_AE))
Y_test_AE[:len(X_test_benign)] = 0
```

[22]: 6506674 + 166784

[22]: 6673458

[23]: 166784/6673458

[23]: 0.024992140506466063

```
[24]: X_train_AE.shape, X_test_AE.shape, Y_test_AE.shape
[24]: ((352643, 23), (6673458, 23), (6673458,))
[25]: # reshape inputs for LSTM [samples, timesteps, features]
      X_train = X_train_AE.reshape(X_train_AE.shape[0], 1, X_train_AE.shape[1])
      print(f'Training data shape: {X train.shape}')
      X test = X test_AE.reshape(X_test_AE.shape[0], 1, X_test_AE.shape[1])
      print(f'Test data shape: {X_test.shape}')
     Training data shape: (352643, 1, 23)
     Test data shape: (6673458, 1, 23)
[26]: # define the autoencoder network model
      def autoencoder_model(X):
          inputs = Input(shape=(X.shape[1], X.shape[2]))
          L1 = LSTM(16, activation='relu', return_sequences=True,_
       ⇒kernel regularizer=regularizers.12(0.00))(inputs)
          L2 = LSTM(4, activation='relu', return_sequences=False)(L1)
          L3 = RepeatVector(X.shape[1])(L2)
          L4 = LSTM(4, activation='relu', return_sequences=True)(L3)
          L5 = LSTM(16, activation='relu', return_sequences=True)(L4)
          output = TimeDistributed(Dense(X.shape[2]))(L5)
          model = Model(inputs=inputs, outputs=output)
          return model
      # KL Loss function
      def vae loss(x, x decoded mean):
          # Compute the average MSE error, then scale it up (sum on all axes)
          reconstruction_loss = K.sum(K.square(x - x_decoded_mean))
          # Compute the KL loss
          kl_loss = -0.5 * K.sum(1 + z_var - K.square(z_mean) - K.square(K.
       →exp(z_var)), axis=-1)
          # Return the average loss over all
          total_loss = K.mean(reconstruction_loss + kl_loss) # Total_loss = __
       \neg reconstruction\_loss + kl\_loss
          return total_loss
      # (1) Reconstruction Loss - Forces the encoder to generate latent features that \Box
       ⇔minimize the reconstruction error, or else is penalized
```

```
# (2) KL Loss - Forces the distribution generated by the encoder to be similar,
       →to the prior probability of the input vector, pushing latent feature space
       ⇔to normality
      def saveHist(path, history):
          with codecs.open(path, 'w', encoding='utf-8') as f:
              json.dump(history, f, separators=(',', ':'), sort_keys=True, indent=4)
      def loadHist(path):
          n = {} # set history to empty
          if os.path.exists(path): # reload history if it exists
              with codecs.open(path, 'r', encoding='utf-8') as f:
                  n = json.loads(f.read())
          return n
      def appendHist(h1, h2):
          if h1 == {}:
              return h2
          else:
              dest = \{\}
              for key, value in h1.items():
                  dest[key] = value + h2[key]
              return dest
      model_loaded = False
[27]: learning_rate = 0.001
                                     # learning rate for optimizer
      # maybe the random results of metrics is because Adam opt - Use SGD instead
      # opt = tf.keras.optimizers.experimental.SGD(lr = learning rate, momentum=0.9)
      opt = tf.keras.optimizers.Adam(learning rate = learning rate)
      # create the autoencoder model
      model = autoencoder_model(X_train)
      model.compile(optimizer=opt, loss='mae', metrics=['accuracy'])
      model.summary()
     2023-06-22 09:45:34.562228: E
     tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:266] failed call to
     cuInit: CUDA_ERROR_NO_DEVICE: no CUDA-capable device is detected
     2023-06-22 09:45:34.562401: I
     tensorflow/compiler/xla/stream_executor/cuda/cuda_diagnostics.cc:168] retrieving
     CUDA diagnostic information for host: pop-os
```

tensorflow/compiler/xla/stream_executor/cuda/cuda_diagnostics.cc:175] hostname:

2023-06-22 09:45:34.562410: I

pop-os

2023-06-22 09:45:34.562556: I

tensorflow/compiler/xla/stream_executor/cuda/cuda_diagnostics.cc:199] libcuda

reported version is: 525.116.4 2023-06-22 09:45:34.562593: I

tensorflow/compiler/xla/stream_executor/cuda/cuda_diagnostics.cc:203] kernel

reported version is: 525.116.4 2023-06-22 09:45:34.562599: I

tensorflow/compiler/xla/stream_executor/cuda/cuda_diagnostics.cc:309] kernel version seems to match DSO: 525.116.4

Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 1, 23)]	0
lstm (LSTM)	(None, 1, 16)	2560
lstm_1 (LSTM)	(None, 4)	336
<pre>repeat_vector (RepeatVector)</pre>	(None, 1, 4)	0
lstm_2 (LSTM)	(None, 1, 4)	144
lstm_3 (LSTM)	(None, 1, 16)	1344
<pre>time_distributed (TimeDistr ibuted)</pre>	(None, 1, 23)	391

Total params: 4,775 Trainable params: 4,775 Non-trainable params: 0

4.2 3.2 Train

```
[28]: # fit the model to the data
# nb_epochs = 100
# batch_size = 32

# train_start = time.time()
# history = model.fit(X_train, X_train,
# epochs = nb_epochs,
# batch_size = batch_size,
# validation_split = 0.10,
```

```
# shuffle = False).history
# train_end = time.time()
# train_time = train_end - train_start

# print(f"Fit time: {train_time:.2f}s")

[29]: # history_filename = 'history_26-05.json'

# model.save("lstm_model_26-05")
# saveHist(history_filename, history)

[30]: # It can be used to reconstruct the model identically.
lstm_model = keras.models.load_model("lstm_model_26-05")
history = loadHist('history_26-05.json')

model_loaded = True

lstm_model.summary()
# result = lstm_model.score(X_test, Y_test)
# print(result)
```

Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 1, 23)]	0
lstm (LSTM)	(None, 1, 16)	2560
lstm_1 (LSTM)	(None, 4)	336
<pre>repeat_vector (RepeatVector)</pre>	(None, 1, 4)	0
lstm_2 (LSTM)	(None, 1, 4)	144
lstm_3 (LSTM)	(None, 1, 16)	1344
<pre>time_distributed (TimeDistr ibuted)</pre>	(None, 1, 23)	391

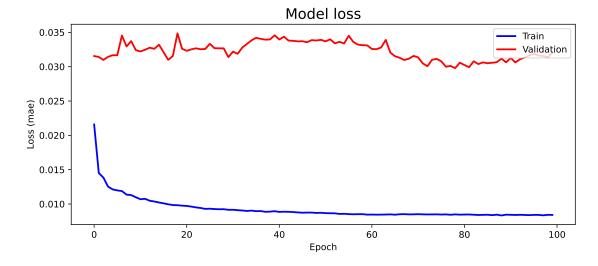
Total params: 4,775 Trainable params: 4,775 Non-trainable params: 0

```
# sns.set(color_codes=True)

# plot the training and validation losses
fig, ax = plt.subplots(figsize=(10, 4), dpi=80)

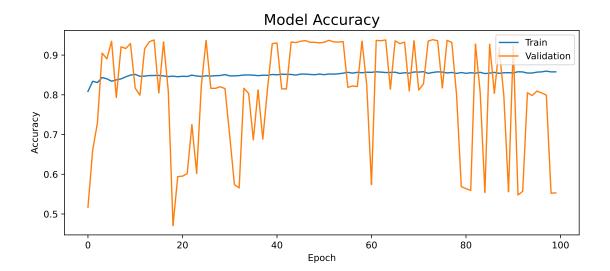
ax.plot(history['loss'], 'b', label='Train', linewidth=2)
ax.plot(history['val_loss'], 'r', label='Validation', linewidth=2)
ax.set_title('Model loss', fontsize=16)
#ax.set_ylim([0, 0.04])
ax.set_ylabel('Loss (mae)')
ax.set_xlabel('Epoch')
ax.legend(loc='upper right')

plt.show()
```



```
[32]: # plot the training and validation accuracys
fig, ax = plt.subplots(figsize=(10, 4), dpi=80)
ax.plot(history['accuracy'], label='Train')
ax.plot(history['val_accuracy'], label='Validation')
ax.set_title('Model Accuracy', fontsize=16)
ax.set_ylabel('Accuracy')
#ax.set_ylabel('Accuracy')
ax.set_ylim([0.45, 1.05])
ax.set_xlabel('Epoch')
ax.legend(loc='upper right')

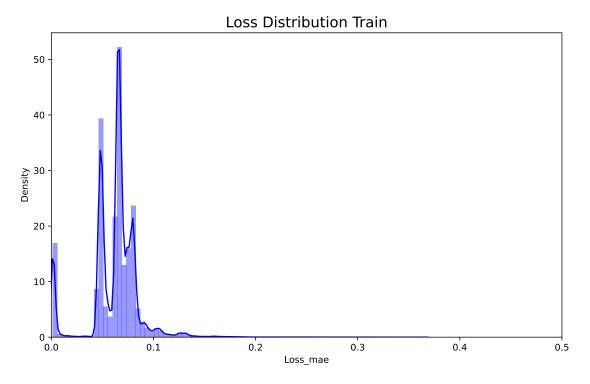
plt.show()
```



Predict on X_train [33]: train start = time.time() if model loaded: X_pred_train = lstm_model.predict(X_train) else: X_pred_train = model.predict(X_train) train_end = time.time() train_time = train_end - train_start print(f"Training time: {train_time:.2f}s") 11021/11021 [===========] - 16s 1ms/step Training time: 20.28s [34]: X_pred_train = X_pred_train.reshape(X_pred_train.shape[0], X_pred_train. ⇔shape[2]) X_pred_train = pd.DataFrame(X_pred_train, columns=df_sw_benign.columns) X_pred_train.index = pd.RangeIndex.from_range(range(0, len(X_train_AE))) scored_train = pd.DataFrame(index=pd.RangeIndex.from_range(range(0, __ ⇔len(X_train_AE)))) Xtrain = X_train.reshape(X_train.shape[0], X_train.shape[2]) [35]: scored_train['Loss_mae'] = np.mean(np.abs(X_pred_train - Xtrain), axis = 1) plt.figure(figsize=(10, 6), dpi=80) plt.title("Loss Distribution Train", fontsize=16)

sns.distplot(scored_train['Loss_mae'], bins = 80, kde=True, color='blue');

```
plt.xlim([0.0, .5]);
```



```
[36]: # threshold = 0.09 # threshold from loss_mae plot above
      threshold = np.quantile(scored_train['Loss_mae'], 0.95)
      print(threshold)
      scored_train['Threshold'] = threshold
      scored_train['Anomaly'] = scored_train['Loss_mae'] > scored_train['Threshold']
     0.09107424507651884
[37]: scored_train['Anomaly'].value_counts()
[37]: Anomaly
     False
               335010
      True
                17633
     Name: count, dtype: int64
[38]: target_names = ['Benign', 'Attack']
      print(f"Training time: {train_time / 60:.2f} min")
      print(classification_report(X_train_label, scored_train['Anomaly'],__

¬target_names=target_names, digits = 5))
```

```
Training time: 0.34 min
                   precision recall f1-score
                                                   support
                     1.00000 0.95000
                                         0.97436
                                                    352643
           Benign
                     0.00000 0.00000
           Attack
                                         0.00000
                                                         0
         accuracy
                                         0.95000
                                                    352643
        macro avg
                     0.50000
                               0.47500
                                         0.48718
                                                    352643
     weighted avg
                     1.00000
                               0.95000
                                         0.97436
                                                    352643
[39]: # Percentage of attacks predicted in training data
      print(f"Attacks Predicted: {np.count_nonzero(scored_train['Anomaly']==1) /__
       →len(scored_train['Anomaly']):.6f}")
     Attacks Predicted: 0.050002
     Predicao somente com dados Benignos (30%) == Cenário Normal
[40]: Xtest_benign = X_test_benign.reshape(X_test_benign.shape[0], 1, X_test_benign.
       \hookrightarrowshape [1])
      test_start = time.time()
      if model loaded:
          X_pred_LSTM_ = lstm_model.predict(Xtest_benign)
      else:
          X_pred_LSTM_ = model.predict(Xtest_benign)
      test_end = time.time()
      test_time = test_end - test_start
      print(f"Testing time: {test_time:.2f}s\n")
     5212/5212 [========== ] - 8s 2ms/step
     Testing time: 10.64s
[41]: X_pred_test_benign = X_pred_LSTM_.reshape(X_pred_LSTM_.shape[0], X_pred_LSTM_.
       \hookrightarrowshape [2])
      X_pred_test_benign = pd.DataFrame(X_pred_test_benign, columns=df_sw_benign.
      ⇔columns)
      X_pred_test_benign.index = pd.RangeIndex.from_range(range(0,__
       →len(X_test_benign))) # cria um objeto RangeIndex do tam de X_test_AE
      scored_test_benign = pd.DataFrame(index=pd.RangeIndex.from_range(range(0,_
       →len(X_test_benign))))
```

```
Xtest_benign = Xtest_benign.reshape(Xtest_benign.shape[0], Xtest_benign.
       \hookrightarrowshape [2])
[42]: # calculate the loss on the test set
     scored_test_benign['Loss_mae'] = np.mean(np.abs(X_pred_test_benign -_
       # error vector in testing data
     scored_test_benign['Threshold'] = threshold
     scored_test_benign['Anomaly'] = scored_test_benign['Loss_mae'] >__
       scored_test_benign['Threshold'] # if > then == attack
[43]: scored_test_benign.head(10)
[43]:
       Loss mae Threshold Anomaly
     0 0.069983
                  0.091074
                               False
     1 0.113978 0.091074
                                True
     2 0.182516 0.091074
                                True
     3 0.078173 0.091074
                              False
     4 0.081123 0.091074
                             False
     5 0.079617 0.091074
                             False
     6 0.079120 0.091074
                             False
     7 0.080085 0.091074
                              False
     8 0.082644 0.091074
                              False
     9 0.079993
                              False
                  0.091074
[44]: # Percentage of attacks predicted
     print(f"Attacks Predicted (only benign): {np.

count_nonzero(scored_test_benign['Anomaly']==1) /

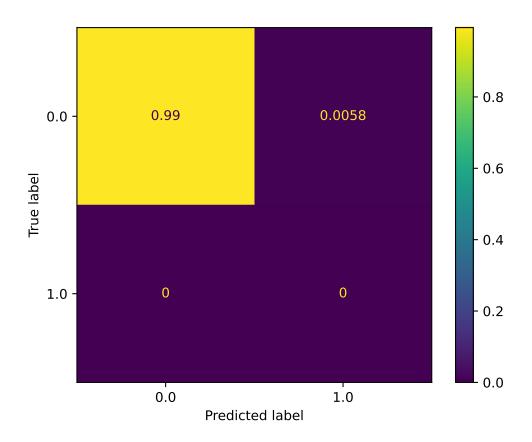
□
       ⇔len(scored_test_benign['Anomaly']):.6f}")
     Attacks Predicted (only benign): 0.040346
     Majority Vote
[45]: # Assuming you have the reconstruction errors stored in a
      # list or numpy array called 'reconstruction errors' and 'threshold'
      # represents the threshold value for determining anomalies.
     window_size = 96  # Size of the Sliding Window
     def detect_anomalies_with_sliding_window(reconstruction_errors, threshold):
         num_samples = len(reconstruction_errors)
         anomaly_labels = np.zeros(num_samples) # Initialize the anomaly labels
         for i in range(0, num_samples, window_size):
             window_errors = reconstruction_errors[i:i+window_size]
```

print(window_errors)

```
# If the majority of errors in the window exceed the threshold, label_{f \sqcup}
       ⇔the window as anomalous
              if sum(i > threshold for i in window_errors) > (len(window_errors) / 2):
                  anomaly labels[i:i+window size] = 1
          return anomaly labels
      anomaly labels =
       ⊸detect_anomalies_with_sliding_window(scored_test_benign['Anomaly'], ___
       →threshold)
      # print(f'\n\nAnomaly: {anomaly_labels}')
      print(np.unique(scored_test_benign['Anomaly'], return_counts=True))
      np.unique(anomaly_labels, return_counts=True)
     (array([False, True]), array([160055,
[45]: (array([0., 1.]), array([165824,
                                           960]))
     Metrics with Majority Vote
[47]: print(f"Testing time: {test_time / 60:.2f} min")
      print(classification report(np.zeros(len(anomaly_labels)), anomaly_labels,__

starget_names=target_names, digits = 5))

     Testing time: 0.18 min
                   precision
                                recall f1-score
                                                    support
           Benign
                     1.00000
                               0.99424
                                          0.99711
                                                     166784
           Attack
                     0.00000
                                0.00000
                                          0.00000
                                                          0
                                          0.99424
                                                     166784
         accuracy
        macro avg
                     0.50000
                                          0.49856
                                                     166784
                                0.49712
     weighted avg
                     1.00000
                                0.99424
                                          0.99711
                                                     166784
[48]: print(confusion_matrix(np.zeros(len(anomaly_labels)), anomaly_labels))
      ConfusionMatrixDisplay.from_predictions(np.zeros(len(anomaly_labels)),_
       →anomaly_labels, normalize='all')
      plt.show()
     [[165824
                 960]
      Γ
            0
                   0]]
```



```
[]: print(f"Attacks Predicted using Sliding Window: {np. 
count_nonzero(anomaly_labels==1) / len(anomaly_labels):.6f}")
```

Attacks Predicted using Sliding Window: 0.005756

4.3 3.3 Test

Predicao com dados Benignos (30%) + Ataques (70%) == Cenário sob Ataque

```
[49]: # Predict on test data
    test_start = time.time()

if model_loaded:
        X_pred_test = lstm_model.predict(X_test)

else:
        X_pred_test = model.predict(X_test)

test_end = time.time()
    test_time = test_end - test_start
    print(f"Testing Time: {test_time:.2f}s")
```

27/208546 [...] - ETA: 6:57

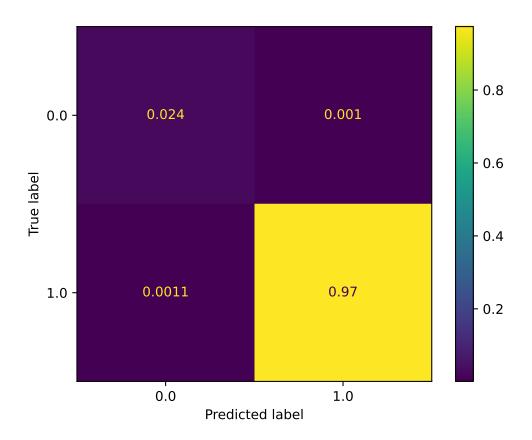
```
2023-06-22 09:58:56.337525: W tensorflow/tsl/framework/cpu_allocator_impl.cc:83]
     Allocation of 613958136 exceeds 10% of free system memory.
     2023-06-22 10:05:35.284799: W tensorflow/tsl/framework/cpu allocator impl.cc:83]
     Allocation of 613958136 exceeds 10% of free system memory.
     Testing Time: 400.14s
[50]: | X_pred_test = X_pred_test.reshape(X_pred_test.shape[0], X_pred_test.shape[2])
     X pred_test = pd.DataFrame(X_pred_test, columns=df_sw_benign.columns)
     X_pred_test.index = pd.RangeIndex.from_range(range(0, len(X_test_AE))) # cria__
      →um objeto RangeIndex do tam de X_test_AE
     scored_test = pd.DataFrame(index=pd.RangeIndex.from_range(range(0, __
      →len(X_test_AE))))
     Xtest = X_test.reshape(X_test.shape[0], X_test.shape[2])
[51]: # calculate the loss on the test set
     scored_test['Loss_mae'] = np.mean(np.abs(X_pred_test - Xtest), axis = 1)
                                                                              #
      ⇔error vector in testing data
     scored_test['Threshold'] = threshold
     scored_test['Anomaly'] = scored_test['Loss_mae'] > scored_test['Threshold'] #__
      \hookrightarrow if > then == attack
     scored_test.head(10)
[51]:
        Loss_mae Threshold Anomaly
     0 0.069983
                 0.091074
                              False
     1 0.113978 0.091074
                               True
     2 0.182516 0.091074
                               True
     3 0.078173 0.091074
                             False
     4 0.081123 0.091074
                            False
     5 0.079617 0.091074 False
     6 0.079120 0.091074
                            False
     7 0.080085 0.091074
                             False
     8 0.082644 0.091074
                              False
     9 0.079993
                  0.091074
                              False
     4.3.1 3.4 Metrics without Sliding Window
[52]: # Quantidade real de dados benignos e de ataque no teste
     print(np.unique(Y_test_AE, return_counts=True))
```

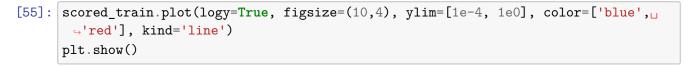
```
(array([0., 1.]), array([ 166784, 6506674]))
```

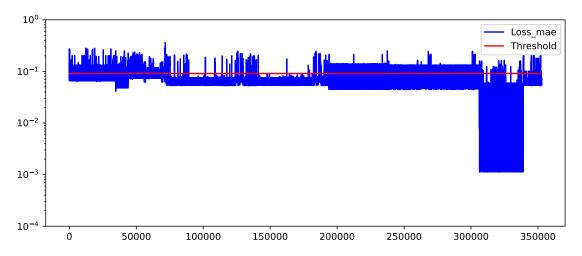
scored_test['Anomaly'].value_counts()

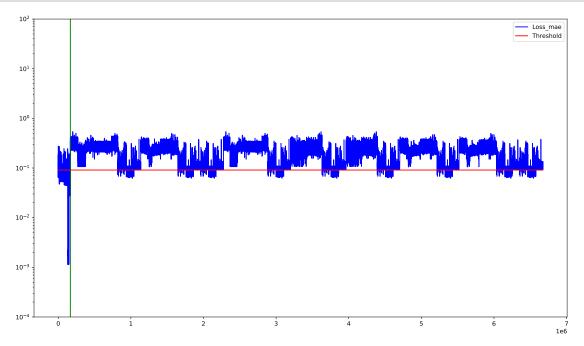
```
[52]: Anomaly
      True
               6506112
      False
                167346
      Name: count, dtype: int64
[53]: print(f"Testing time: {test_time / 60:.2f} min")
      print(classification_report(Y_test_AE, scored_test['Anomaly'],__
       →target_names=target_names, digits = 5))
     Testing time: 6.67 min
                   precision
                                                    support
                                recall f1-score
                                          0.95804
           Benign
                     0.95643
                              0.95965
                                                     166784
           Attack
                     0.99897
                               0.99888
                                          0.99892
                                                    6506674
         accuracy
                                          0.99790
                                                    6673458
        macro avg
                                          0.97848
                                                    6673458
                     0.97770
                               0.97927
     weighted avg
                     0.99790
                               0.99790
                                          0.99790
                                                    6673458
[54]: print(confusion_matrix(Y_test_AE, scored_test['Anomaly']))
      ConfusionMatrixDisplay.from_predictions(Y_test_AE, scored_test['Anomaly'],__

¬normalize='all')
      plt.show()
     [[ 160055
                  6729]
          7291 6499383]]
```







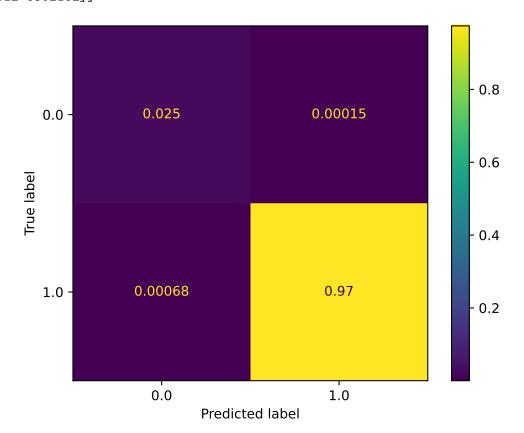


4.3.2 Majority Vote

	precision	recall	f1-score	support
Benign Attack	0.97351 0.99985	0.99405 0.99931	0.98367 0.99958	166784 6506674
accuracy macro avg	0.98668	0.99668	0.99918 0.99162	6673458 6673458
weighted avg	0.99919	0.99918	0.99918	6673458

```
[62]: print(confusion_matrix(Y_test_AE, anomaly_labels_))
ConfusionMatrixDisplay.from_predictions(Y_test_AE, anomaly_labels_,_
normalize='all')
plt.show()
```

[[165792 992] [4512 6502162]]



[]: