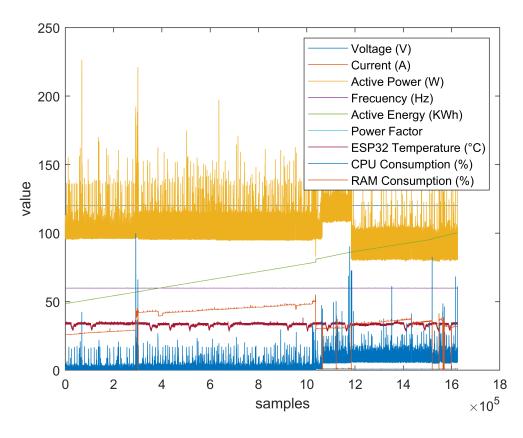
Load Data, Sampling Rate=1sample/1seg

Date = '2021-05-27 00:00:00' AND '2021-06-17 22:04:21'

```
clear;clc;%clear all
%Prepare the raw dataset
%load('./data/dataset_raw.mat');%Load table data
%data(:,19)=[];%delete empty data
%data(:,13:17)=[];%delete empty data
%data(:,1)=[];%delete the topics
load('./data/dataset.mat');%Load table data
dataNew=table2array(data(:,4:12));% Array Double
%csvwrite('data.csv',dataNew);%Save as .csv file
```

Plot Data (Dataset)

```
figure
plot(dataNew);xlabel('samples');ylabel('value');
legend('Voltage (V)','Current (A)','Active Power (W)','Frecuency (Hz)','Active Energy (KWh)','I
```



Statital Information (Dataset)

```
xds_v = datastats(dataNew(:,1))%Voltage (V)
```

 $xds_v = struct with fields:$

num: 1625175 max: 120.2000 min: 119.4000

```
mean: 120.0236
median: 120
range: 0.8000
std: 0.0648

xds_C = datastats
```

xds_C = datastats(dataNew(:,2))%Current (A)

```
xds_C = struct with fields:
    num: 1625175
    max: 1.9800
    min: 0.7000
    mean: 0.9240
    median: 0.9300
    range: 1.2800
    std: 0.0934
```

xds_AP = datastats(dataNew(:,3))%Active Power (W)

```
xds_AP = struct with fields:
    num: 1625175
    max: 226.6000
    min: 79.1000
    mean: 98.9189
    median: 96.6000
    range: 147.5000
    std: 9.3631
```

xds_F = datastats(dataNew(:,4))%Frecuency (Hz)

```
xds_F = struct with fields:
    num: 1625175
    max: 60.1000
    min: 59.7000
    mean: 59.9668
    median: 60
    range: 0.4000
    std: 0.0488
```

xds_AE = datastats(dataNew(:,5))%Active Energy (KWh)

```
xds_AE = struct with fields:
    num: 1625175
    max: 100.1200
    min: 48.5600
    mean: 73.3169
    median: 72.2300
    range: 51.5600
    std: 14.9472
```

xds_AE = datastats(dataNew(:,6))%Power Factor

```
xds_AE = struct with fields:
    num: 1625175
    max: 1
    min: 0.7300
    mean: 0.8946
    median: 0.8600
    range: 0.2700
    std: 0.0443
```

xds_T = datastats(dataNew(:,7))%ESP32 Temperature (°C)

```
xds_T = struct with fields:
      num: 1625175
      max: 38.3300
      min: 28.3300
     mean: 33.2923
   median: 33.8900
    range: 10
      std: 1.0580
xds_CPU = datastats(dataNew(:,8))%CPU Consumption (%)
xds CPU = struct with fields:
      num: 1625175
      max: 99.8000
      min: 0
     mean: 3.8781
   median: 1.3000
    range: 99.8000
      std: 4.5249
xds RAM = datastats(dataNew(:,9))%RAM Consumption (%)
xds RAM = struct with fields:
      num: 1625175
      max: 55.2000
      min: 0
     mean: 37.6688
   median: 36.3500
    range: 55.2000
      std: 7.0824
```

1 - Preprocessing (Normalization)

```
DataNorm=[];
for i = 1:size(dataNew,2)
    DataNorm=[DataNorm (dataNew(:,i)-min(dataNew(:,i)))/(max(dataNew(:,i))-min(dataNew(:,i)))]
end
```

2 - Feature Extraction (rms by temporary window)

```
% Feature 1: rms(Active Power (W))
 xds_F1 = datastats(DataFeatures(:,1))
 xds_F1 = struct with fields:
        num: 27086
        max: 0.8693
        min: 0.4401
       mean: 0.7819
     median: 0.7953
      range: 0.4292
        std: 0.0542
 % Feature 2: rms(Active Energy (KWh))
 xds_F2 = datastats(DataFeatures(:,2))
 xds F2 = struct with fields:
        num: 27086
        max: 0.5796
        min: 0.0797
       mean: 0.1780
     median: 0.1887
      range: 0.5000
        std: 0.0654
 % Feature 3: rms(CPU Consumption (%))
 xds_F3 = datastats(DataFeatures(:,3))
 xds_F3 = struct with fields:
        num: 27086
        max: 0.5551
        min: 0.0922
       mean: 0.1386
     median: 0.1253
      range: 0.4630
        std: 0.0536
 % Feature 4: rms(RAM Consumption (%))
 xds F4 = datastats(DataFeatures(:,4))
 xds F4 = struct with fields:
        num: 27086
        max: 0.7828
        min: 0.2772
       mean: 0.6762
     median: 0.6820
      range: 0.5055
        std: 0.0494
3 - Feature Selection
 %corrplot(DataNorm)
```

```
R = corrcoef(DataFeatures)
R = 9 \times 9
   1.0000
             -0.7712
                       -0.9131
                                  0.2255
                                           -0.0330
                                                     -0.0177
                                                                 0.2061
                                                                          -0.3039 ...
   -0.7712
             1.0000
                        0.8662
                                  0.0108
                                           -0.4388
                                                     -0.5201
                                                                -0.0306
                                                                          -0.2088
   -0.9131
             0.8662
                        1.0000
                                  0.0064
                                           -0.0161
                                                     -0.0243
                                                                -0.0489
                                                                           0.2964
   0.2255
             0.0108
                       0.0064
                                  1.0000
                                           -0.0134
                                                     -0.0117
                                                                0.0063
                                                                          -0.0098
   -0.0330
             -0.4388
                       -0.0161
                                 -0.0134
                                            1.0000
                                                      0.8466
                                                                -0.0452
                                                                           0.7864
   -0.0177
             -0.5201
                       -0.0243
                                 -0.0117
                                            0.8466
                                                      1.0000
                                                                -0.0195
                                                                           0.9216
   0.2061
             -0.0306
                       -0.0489
                                  0.0063
                                           -0.0452
                                                      -0.0195
                                                                 1.0000
                                                                          -0.0266
```

```
    -0.3039
    -0.2088
    0.2964
    -0.0098
    0.7864
    0.9216
    -0.0266
    1.0000

    0.0392
    0.0872
    -0.0541
    -0.0107
    0.0403
    -0.2858
    -0.0923
    -0.2628
```

```
%'Voltage (V)','Current (A)','Active Power (W)','Frecuency (Hz)','Active Energy (KWh)','Power (WHeat map)
% Experimento: ESP32 Temperature
DataFeatures(:,7)=[];
% Based on corrcoef To delete:
% PowerFactor -> alta correlacion con Active Energy (85%), entonces nos
% quedamos con Active Energy
% Voltage, current -> Alta correlacion con Active Power (91% y 86%),
% entonces nos quedfamos Active Power
DataFeatures(:,6)=[];DataFeatures(:,2)=[];DataFeatures(:,1)=[];
% Parametros electricos: Frecuency
DataFeatures(:,2)=[];
%Nos quedamos con estas variables: 'Active Power (W)','Active Energy (KWh)','CPU Consumption (S)
```

4 - Regression Learner (Input output Data)

```
training=0.85; %70 Training, 15% Validation, 15% Testing
testing=0.15;
%New Testing adicional...
% NOTA: hacer una prueba sin CPU Consumption (%),RAM Consumption (%) y ver
% si ayudan a mejorar la prediccion...

input=[];output=[];
% inputs: Active Power (W),Active Energy (KWh),CPU Consumption (%),RAM Consumption (%)
input=DataFeatures(1:round(length(DataFeatures)*training)-1,:);
% output: Energía in the following time step
output=DataFeatures(1+1:round(length(DataFeatures)*training),2);
Samples=length(DataFeatures)*0.85-1
```

Samples = 2.3022e+04

RMSE (Testing): week, day, hour, minute, second

% RMSE