

# Power Consumption

**Get Started** 



## Members



**Dechatorn** 



Siravit



Chansinee

### About Dataset:

This dataset is related to the power consumption of three different distribution networks of Tetouan city which is located in north Morocco.

	DateTime	Temperature	Humidity	Wind Speed	general diffuse flows	diffuse flows	Zone 1 Power Consumption	Zone 2 Power Consumption	Zone 3 Power Consumption
0	2017-01-01 00:00:00	6.559	73.8	0.083	0.051	0.119	34055.69620	16128.87538	20240.96386
1	2017-01-01 00:10:00	6.414	74.5	0.083	0.070	0.085	29814.68354	19375.07599	20131.08434
2	2017-01-01 00:20:00	6.313	74.5	0.080	0.062	0.100	29128.10127	19006.68693	19668.43373
3	2017-01-01 00:30:00	6.121	75.0	0.083	0.091	0.096	28228.86076	18361.09422	18899.27711
4	2017-01-01 00:40:00	5.921	75.7	0.081	0.048	0.085	27335.69620	17872.34043	18442.40964
52411	2017-12-30 23:10:00	7.010	72.4	0.080	0.040	0.096	31160.45627	26857.31820	14780.31212
52412	2017-12-30 23:20:00	6.947	72.6	0.082	0.051	0.093	30430.41825	26124.57809	14428.81152
52413	2017-12-30 23:30:00	6.900	72.8	0.086	0.084	0.074	29590.87452	25277.69254	13806.48259
52414	2017-12-30 23:40:00	6.758	73.0	0.080	0.066	0.089	28958.17490	24692.23688	13512.60504
52415	2017-12-30 23:50:00	6.580	74.1	0.081	0.062	0.111	28349.80989	24055.23167	13345.49820

52416 rows × 9 columns

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50.446	0 1								

data is stored every 10 minutes

in 1 year (2017-01-01: 00:00:00

to 2017-12- 31: 23:50:00.)

#### power consumption per 10 every 10 minutes

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52416 rd	ws × 9 columns								

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E0/16 **	ave v 0 columns								

other factors affecting the power consumption



Fig1. power consumption for the year of 2017 at each hour for 3 zones

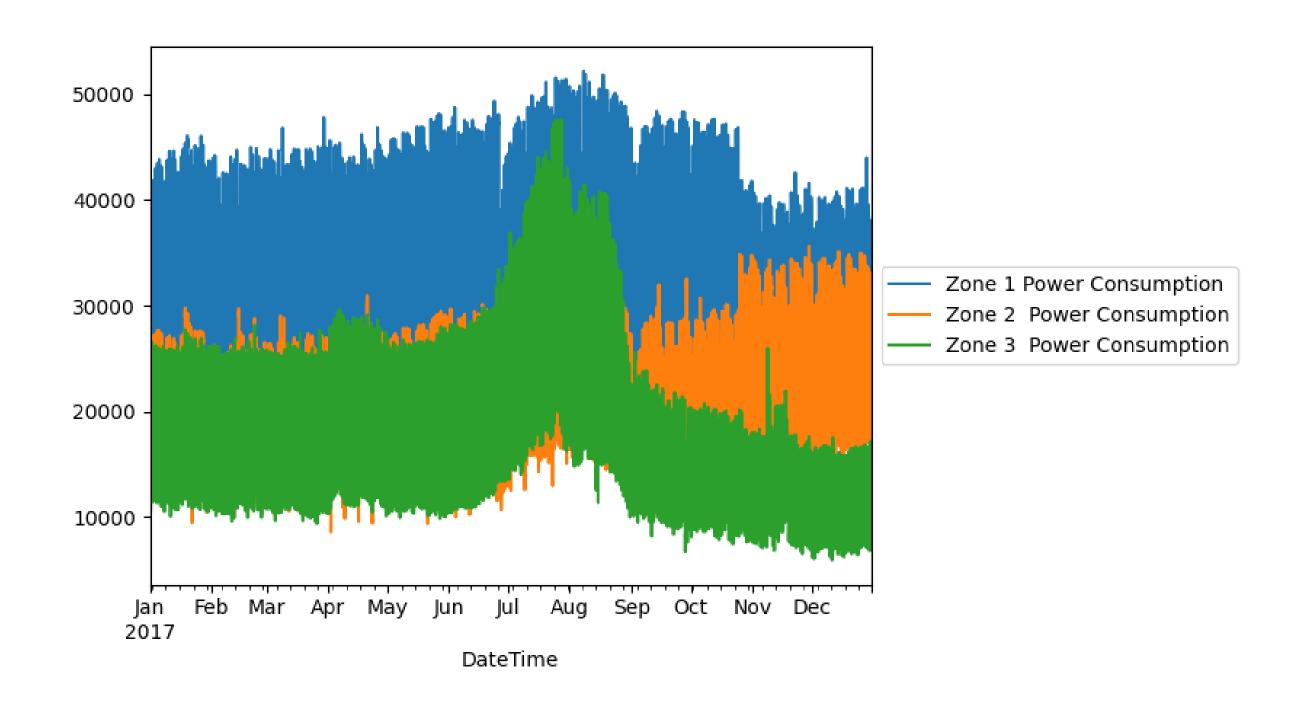


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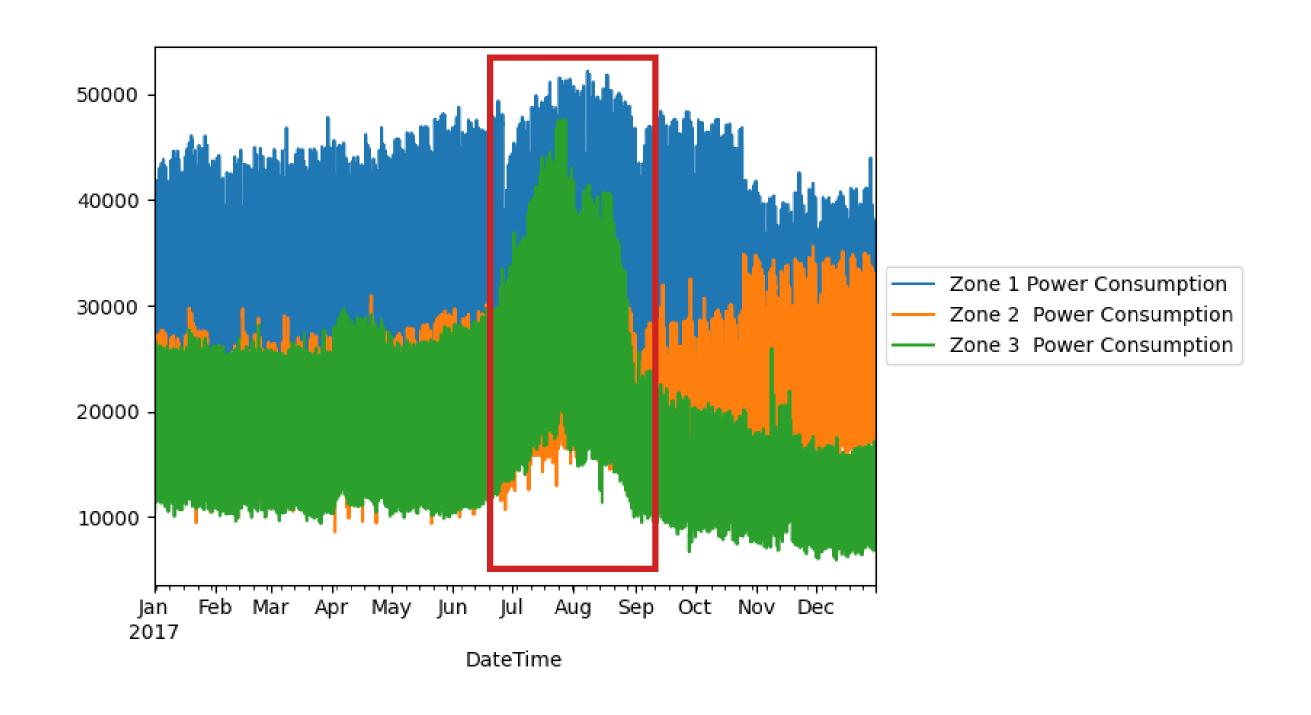


Fig2. average daily power consumption at each hour for 3 zones

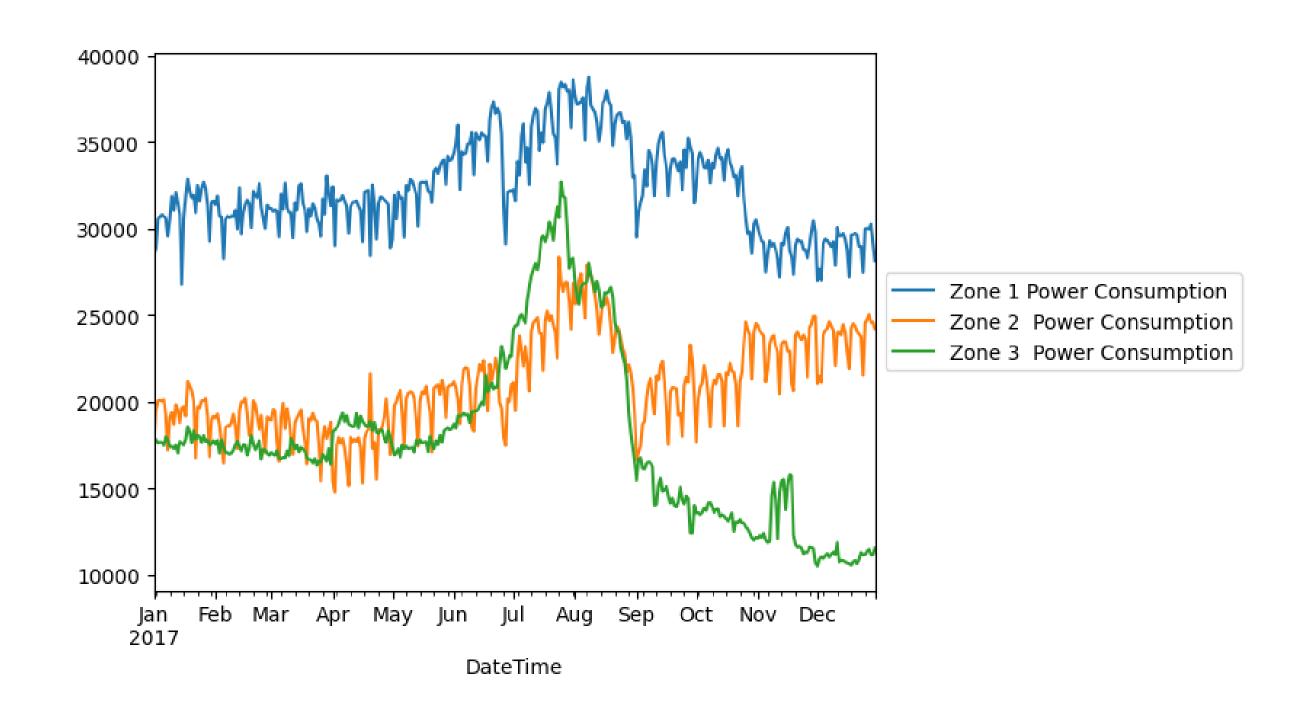


Fig2. average daily power consumption at each hour for 3 zones

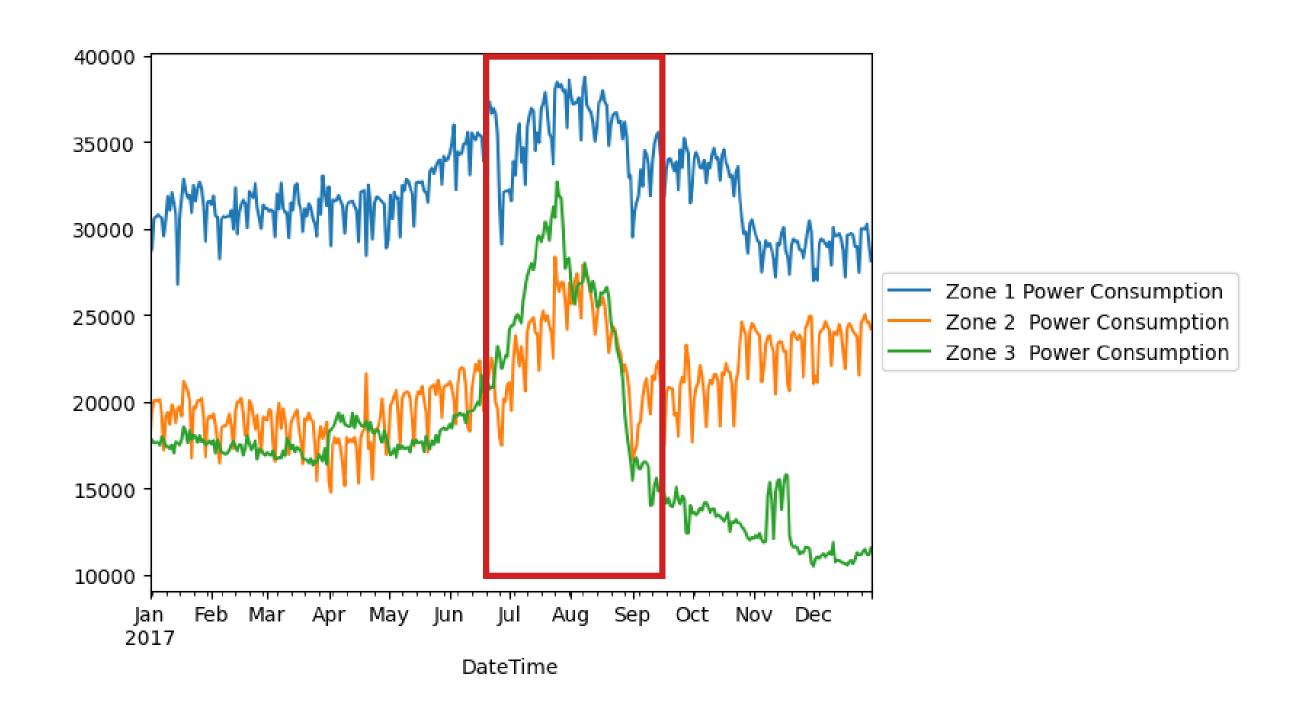




Fig3. average power consumption per week in 1 year at each hour for 3 zones

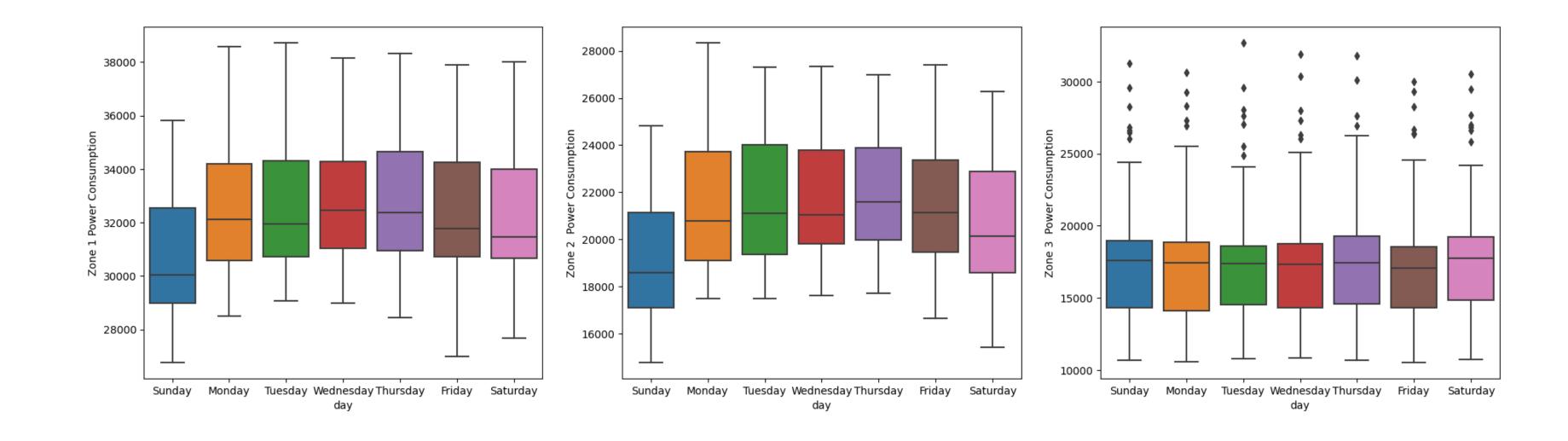




Fig3. average power consumption per week in 1 year at each hour for 3 zones

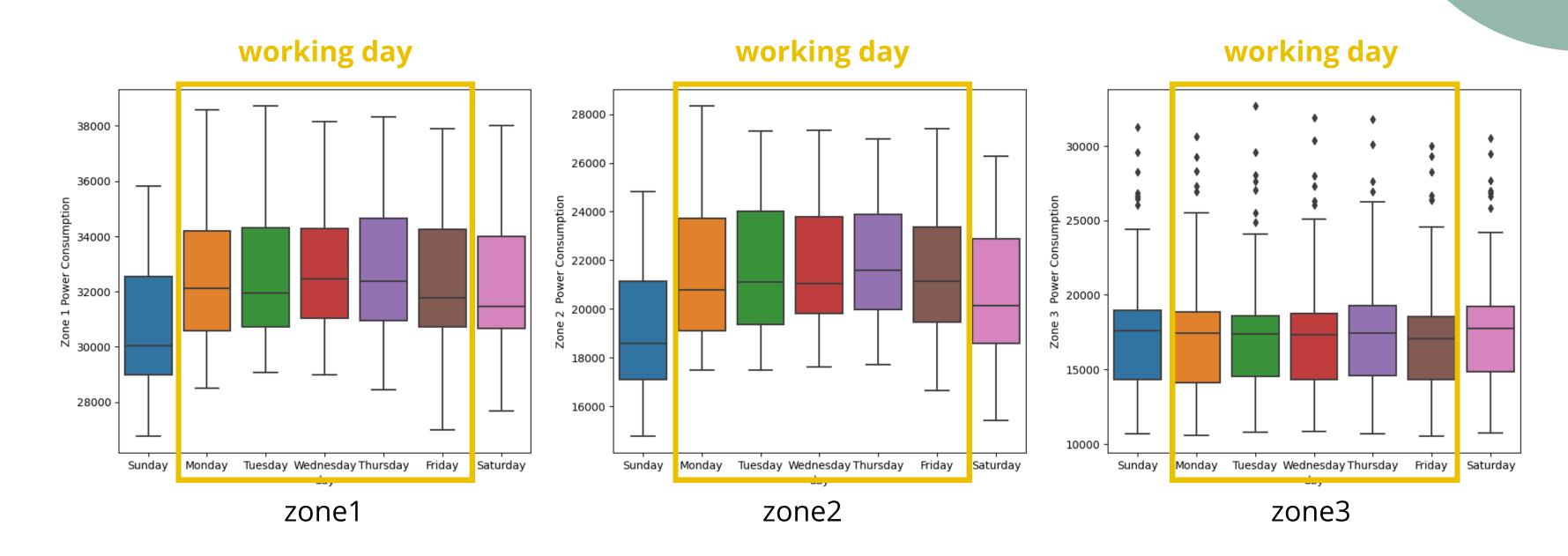
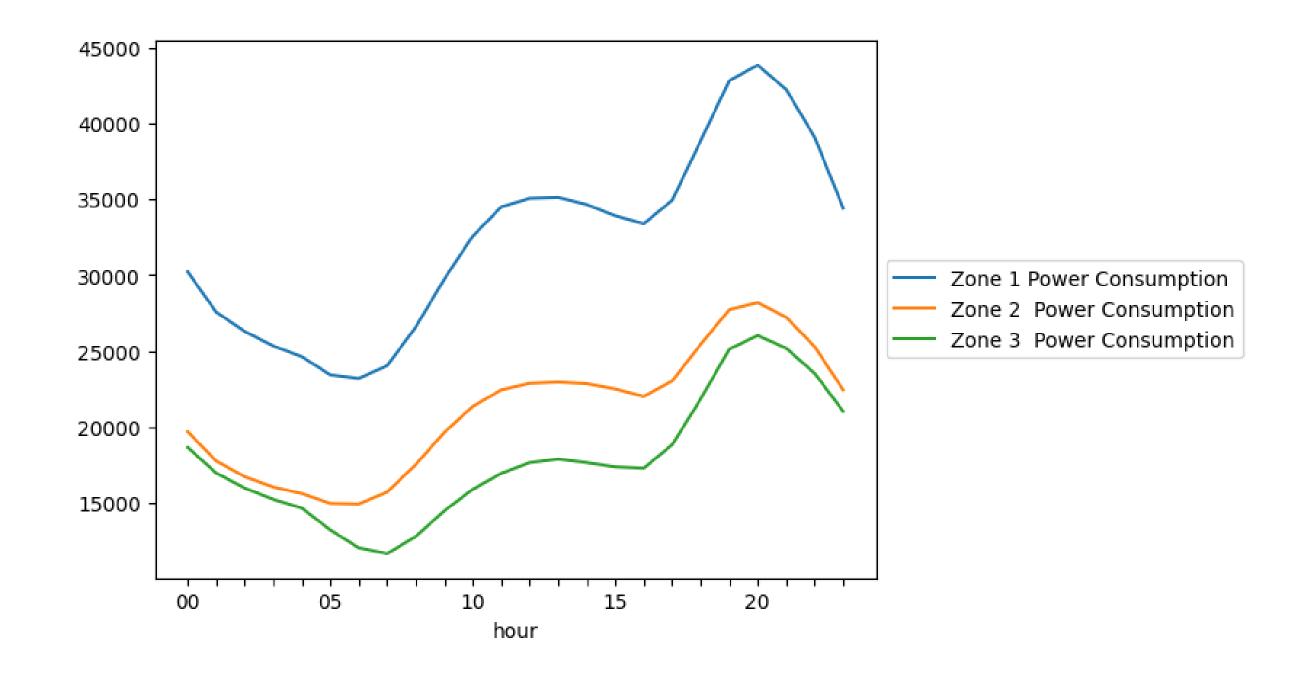


Fig4. average hourly power consumption per day in 1 year at each hour for 3 zones



**EDA** 

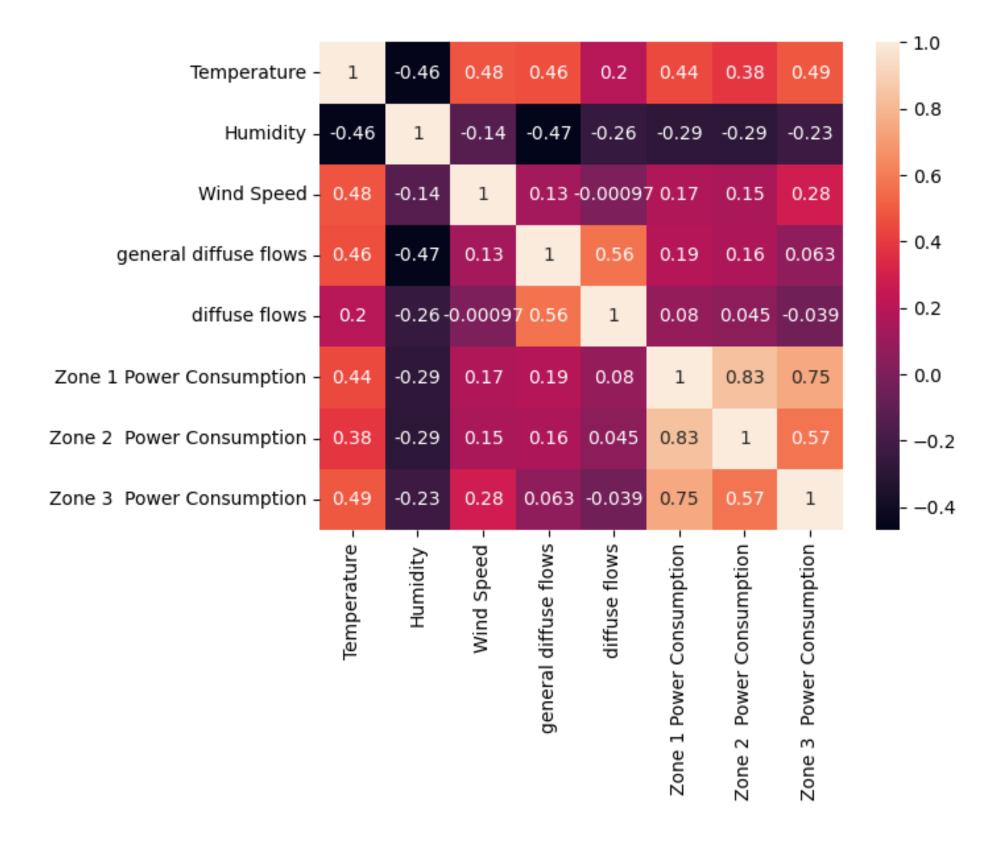


Fig5. the correlation of attributes in dataset

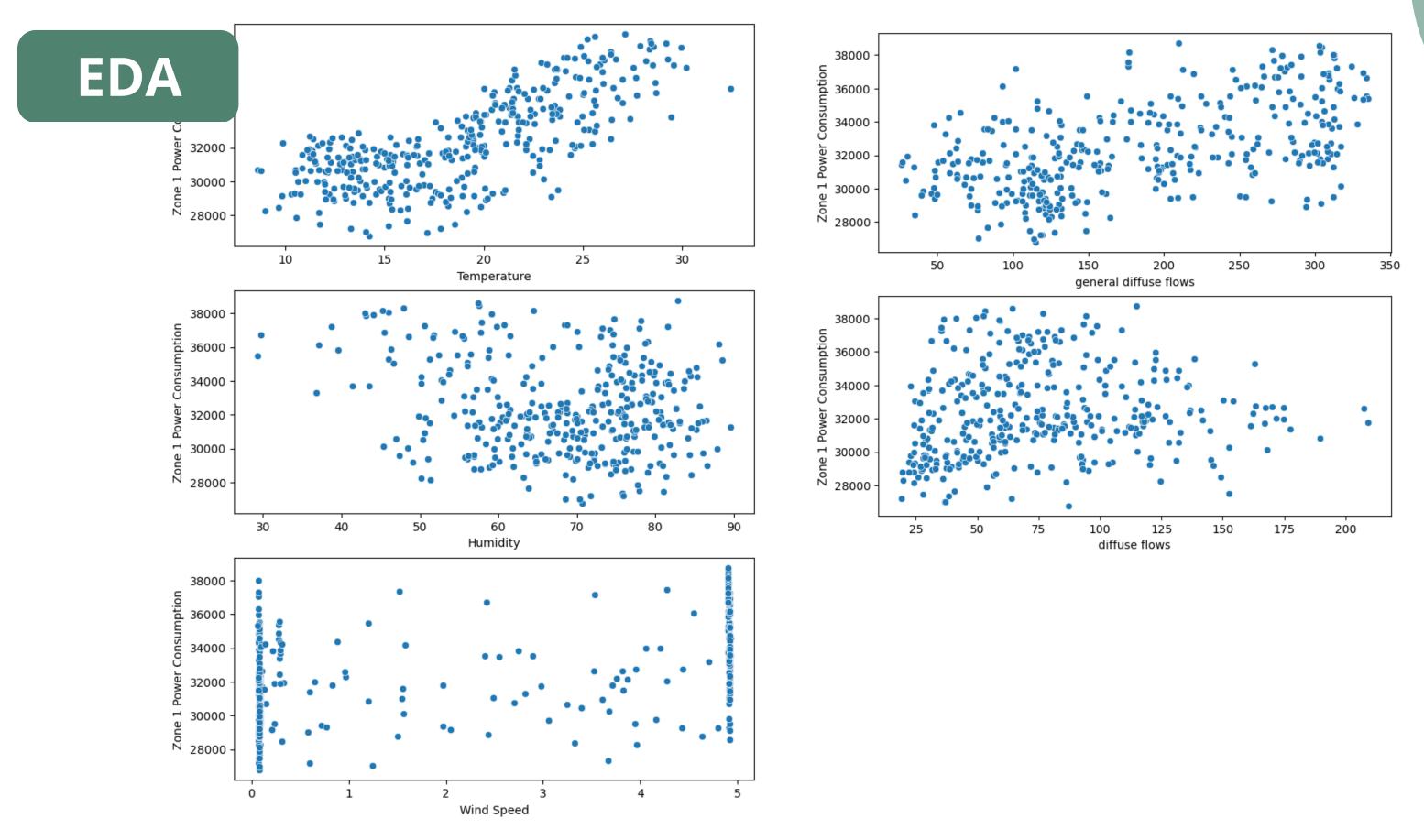


Fig6. the correlation of weather attributes and power consumption



Fig6. Temperature correlation with Power consumption in each zone

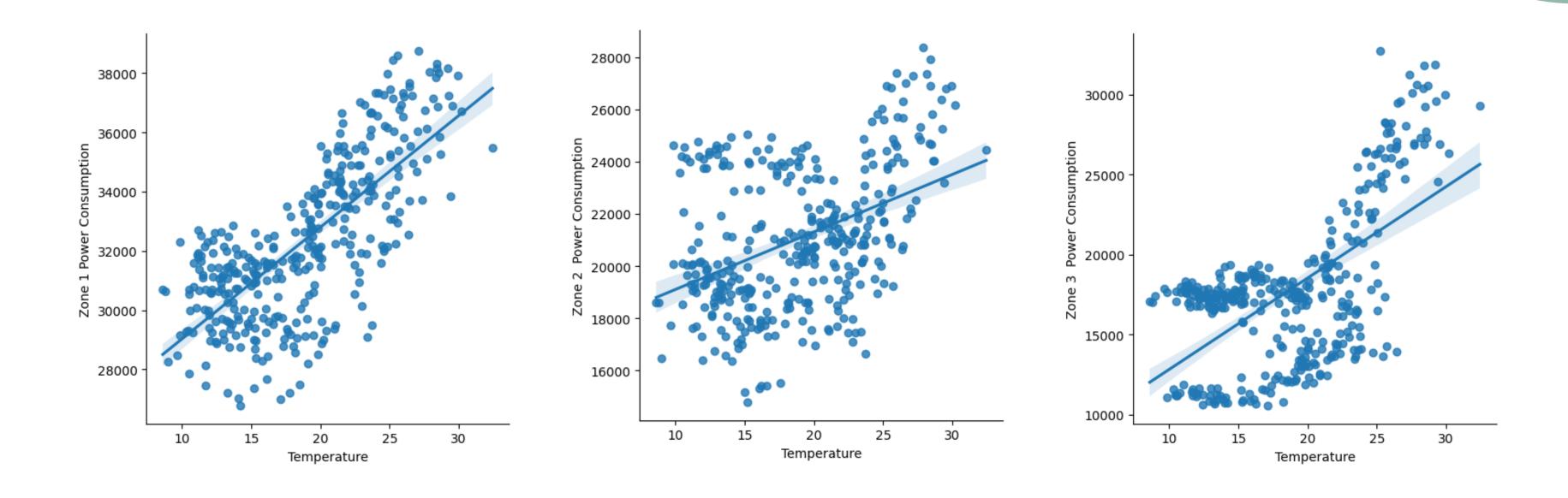
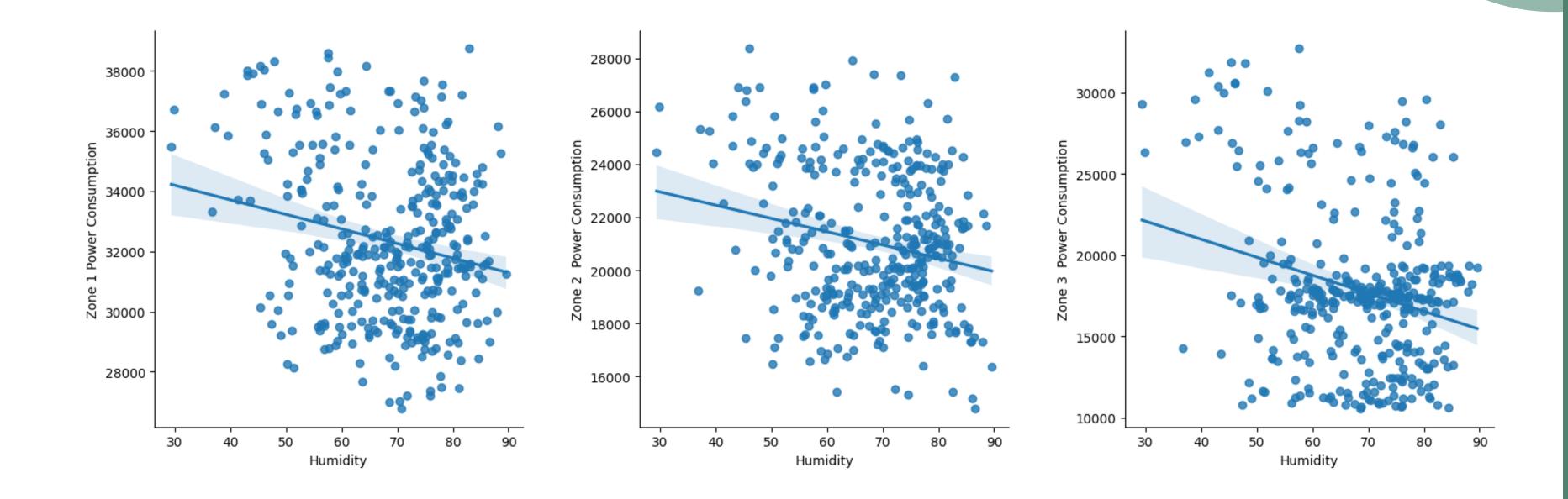




Fig7. Humidity correlation with Power consumption in each zone



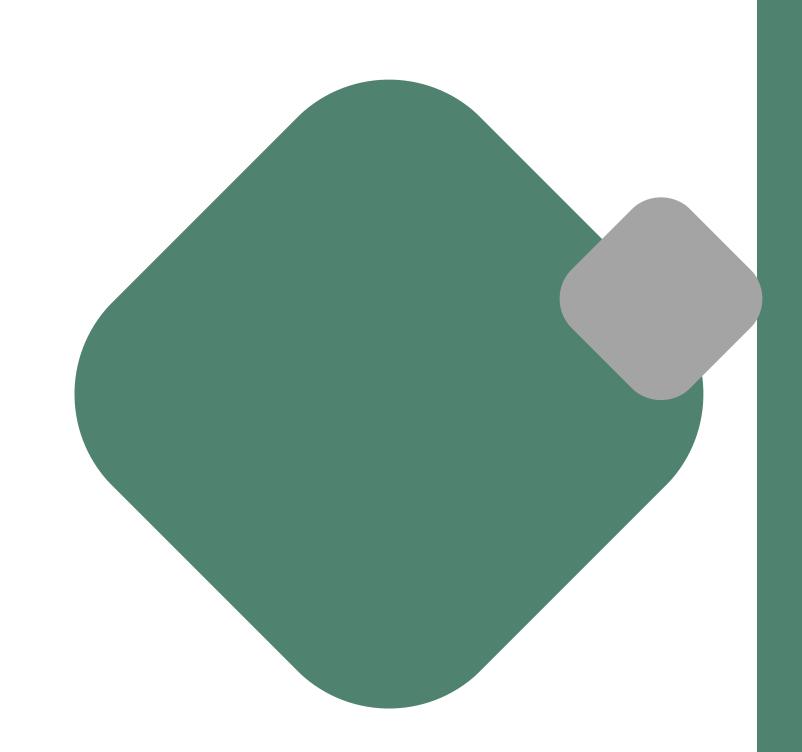
#### IMPORTANCE PREDICTION VARIABLE

#### **Calender & Time**

- Hour
- day of weeks

#### **Weather Attributes**

Temperature



- Linear / Non-linear Regression
- Support Vector Machine for Regression
- Decision Tree
- Random Forest

- Able to handle complex data.
- Able to predict in case of multiple parameters.
- Have high performance than the decision tree model.
- Reduce the risk of overfitting by using a subset of features to build each decision tree.

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## Random Forest

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