

Ensemble Learning and Reinforcement Learning for Energy Demand Prediction

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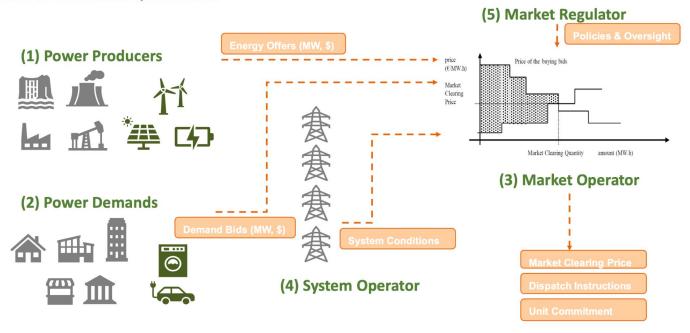




CAISO (California Independent System Operator)

Basic Concepts from Energy Markets

The role of resources optimization





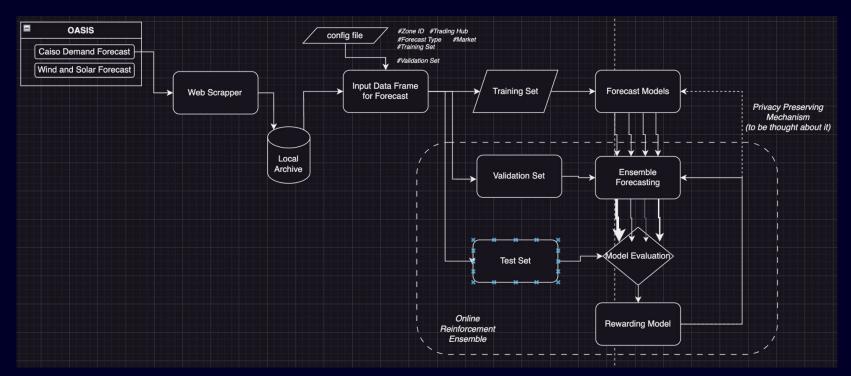
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Overall Plan for demand forecasting





CAISO Dataset

DAM: Day Ahead Market

ACTUAL: Ground Truth Data

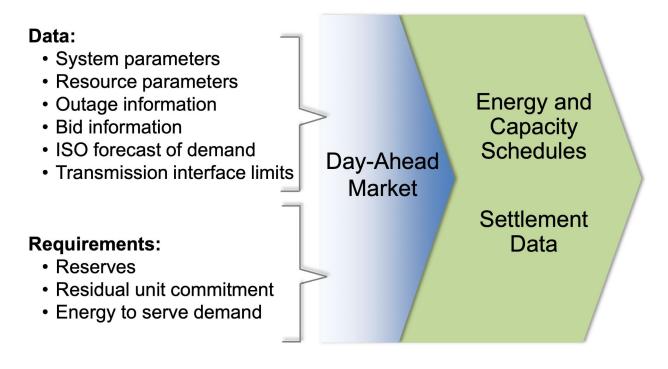
RTD: Real Time Dispatch

RTPD: Read Time Pre-Dispatch



DAM: Day Ahead Market

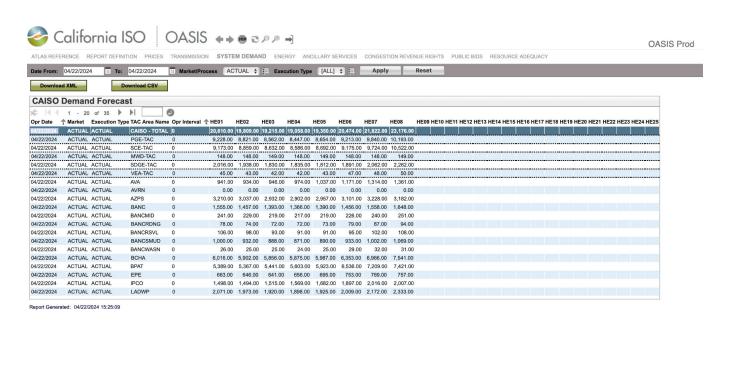
Inputs and outputs of the day-ahead market







CAISO (California Independent System Operator)





45 | SMEC=\$14.37012



QUICK LINKS

OASIS Home

∅ CONNECTED

For RT Interval 08:30 - LMP for ZP26=\$-32.42717 NP15=\$22.2

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Support System Technical Information

CAISO (California Independent System Operator)

INTERVALST/ INTERVALENI L	OAD_TYPE	OPR_DT T	OPR_HR T	OPR_INTERV	MARKET_RUI	TAC_AREA_N	LABEL T	XML_DATA_I	POS T	MW T
2023-01-01T2 2023-01-01T2	1	2023-01-01	15	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1426.98
2023-01-01T1 2023-01-01T1	1	2023-01-01	10	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1461.02
2023-01-01T1 2023-01-01T1	1	2023-01-01	9	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1423.28
2023-01-01TC 2023-01-01T1	1	2023-01-01	2	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1164.2
2023-01-02T(2023-01-02T(1	2023-01-01	22	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1424.59
2023-01-02T(2023-01-02T(1	2023-01-01	18	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1597.37
2023-01-01T2 2023-01-02T0		2023-01-01	16	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1446.45
2023-01-01T1 2023-01-01T2	1	2023-01-01	12	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1462.8
2023-01-01T1 2023-01-01T1	1	2023-01-01	11	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1466.69
2023-01-01T1 2023-01-01T1	1	2023-01-01	8	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1362.17
2023-01-01T2 2023-01-01T2	1	2023-01-01	13	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1452.11
2023-01-01T1 2023-01-01T1	1	2023-01-01	5	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1187.78
2023-01-02T(2023-01-02T(2023-01-01	24	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1270.49
2023-01-01T1 2023-01-01T1	1	2023-01-01	4	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1164.94
2023-01-01T1 2023-01-01T1	1	2023-01-01	3	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1157.8
2023-01-01TC 2023-01-01TC	1	2023-01-01	1	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1183.83
2023-01-02T(2023-01-02T(1	2023-01-01	19	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1580.71
2023-01-02T(2023-01-02T(1	2023-01-01	17	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1526.89
2023-01-01T1 2023-01-01T1		2023-01-01	6	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1230.45
2023-01-01T1 2023-01-01T1	1	2023-01-01	7	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1290.14
2023-01-02T(2023-01-02T(1	2023-01-01	23	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1336.41
2023-01-02T(2023-01-02T(1	2023-01-01	21	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1501.16
2023-01-02T(2023-01-02T(2023-01-01	20	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1549.48
2023-01-01T2 2023-01-01T2	1	2023-01-01	14	0	DAM	AVA	Demand Fored	SYS_FCST_D/	2.8	1435.78
2023-01-02T(2023-01-02T(2023-01-01	21	0	DAM	AZPS	Demand Fored	SYS_FCST_D/	2.8	3218.27
2023-01-01T2 2023-01-01T2	1	2023-01-01	13	0	DAM	AZPS	Demand Fored	SYS_FCST_D/	2.8	2319.39
2023-01-01TC 2023-01-01TC	1	2023-01-01	1	0	DAM	AZPS	Demand Fored	SYS_FCST_DA	2.8	2708.88



Dataset split

Training set: Jan 1st, 2023 – July 1st, 2023

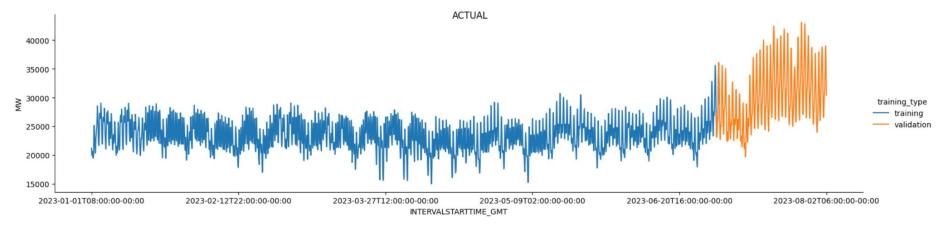
Validation set: July 1st, 2023 – August 1st, 2023

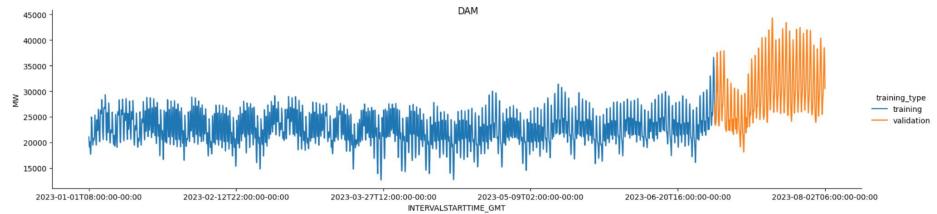
Test set: August 1st, 2023 – August 20th, 2023



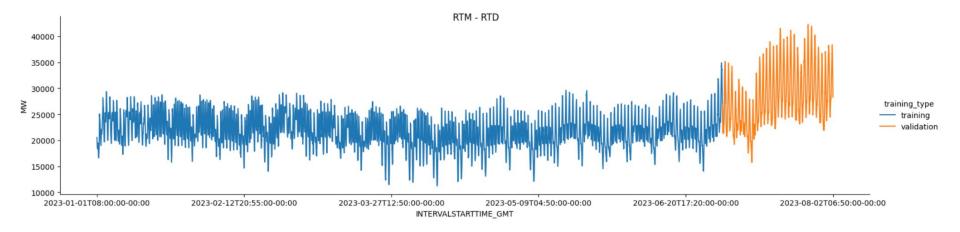


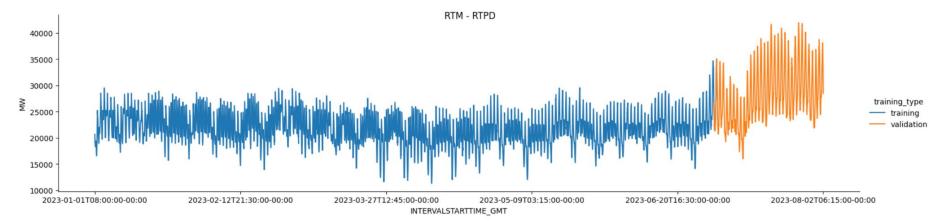
Day Ahead Market



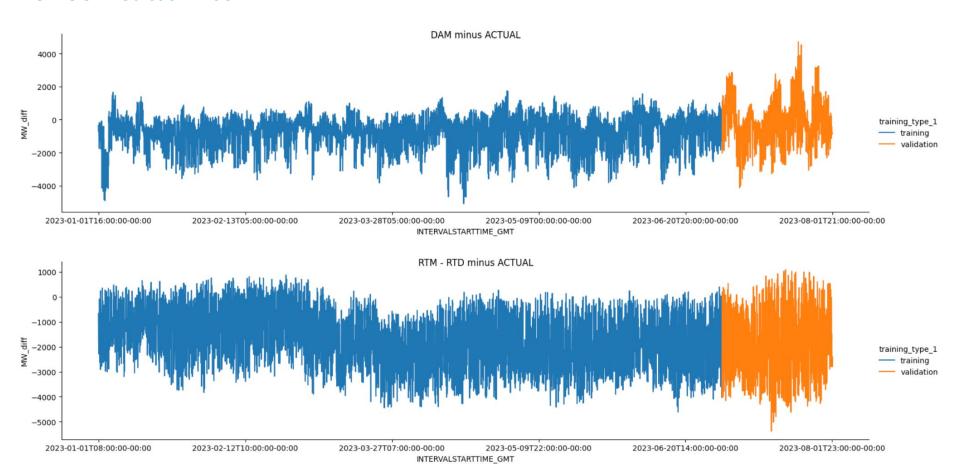


Real Time Market



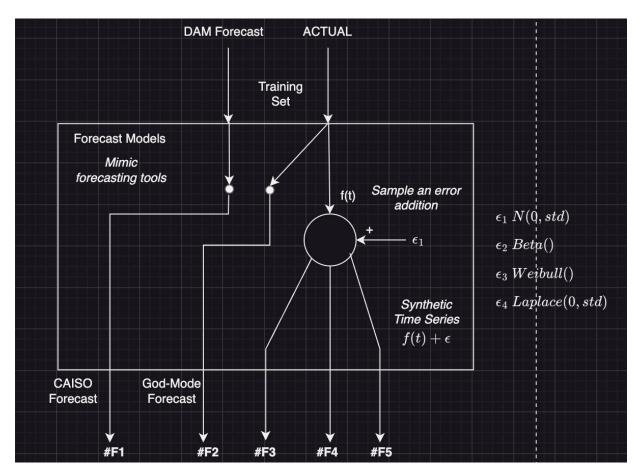


CAISO Prediction Bias



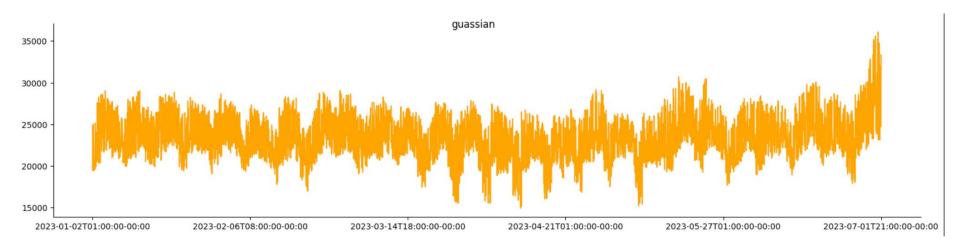


Data Augmentation



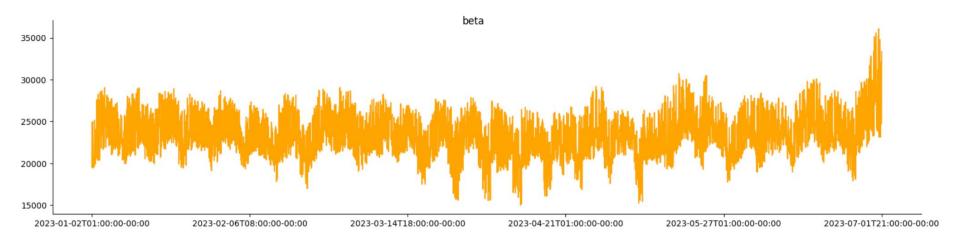


Data Augmentation: Gaussian Noise



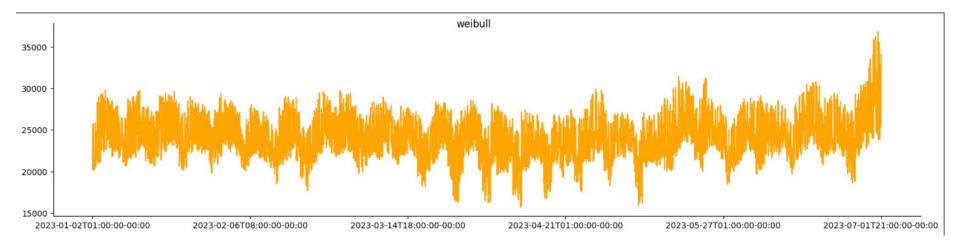


Data Augmentation: Beta Noise



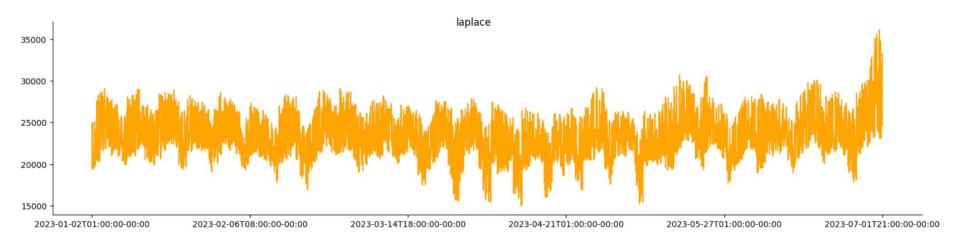


Data Augmentation: Weibull Noise





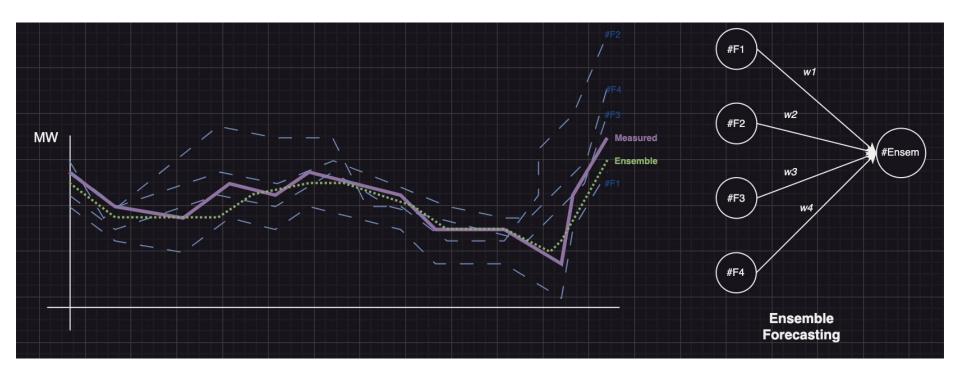
Data Augmentation: Laplace Noise





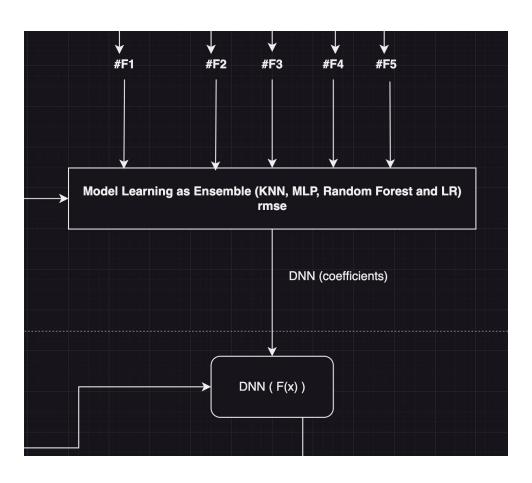


Model Ensembling





Model Ensembling



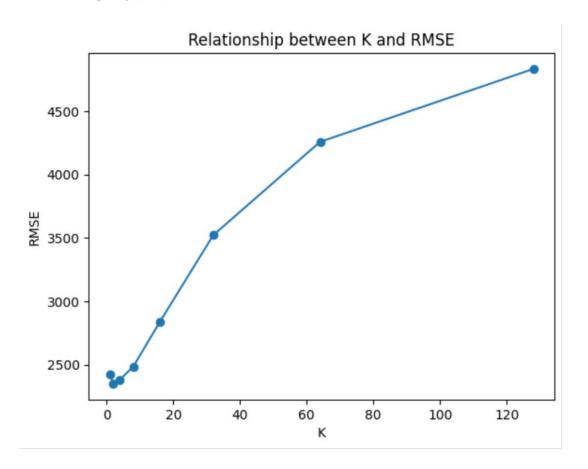


Linear Regression with Actual Values as a feature: Parameters

```
Intercept: 3.637978807091713e-12
Coefficients: [-1.56102205e-16 -5.55111512e-17 -5.07042306e-17 -1.15643556e-16
-6.97027300e-17 1.00000000e+00]
LR: 2.7630033906398518e-12
```

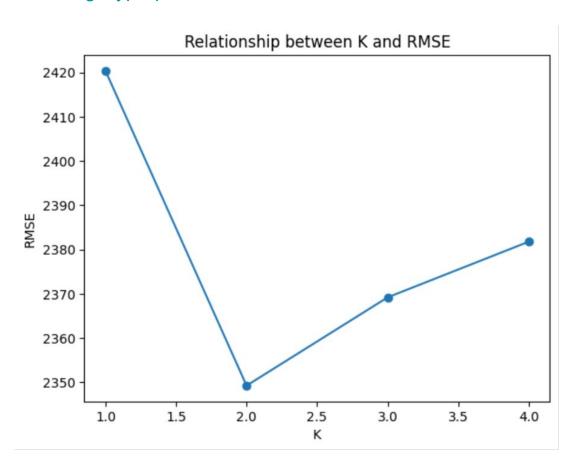


K Nearest Neighbors: Tuning Hyperparameters



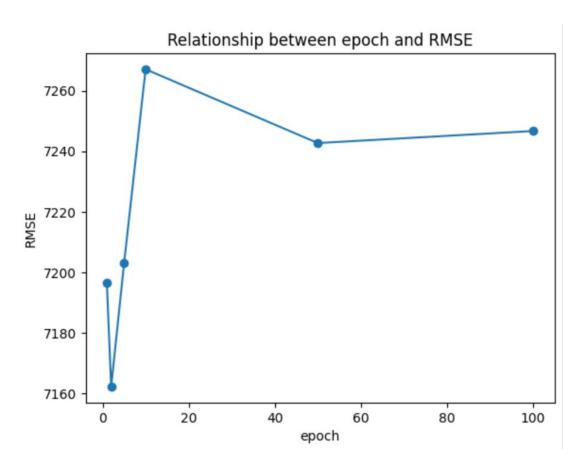


K Nearest Neighbors: Tuning Hyperparameters



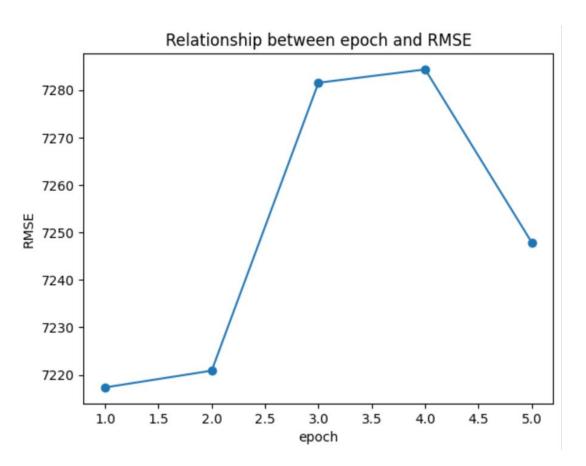


Multilayer Perceptron: Tuning Hyperparameters



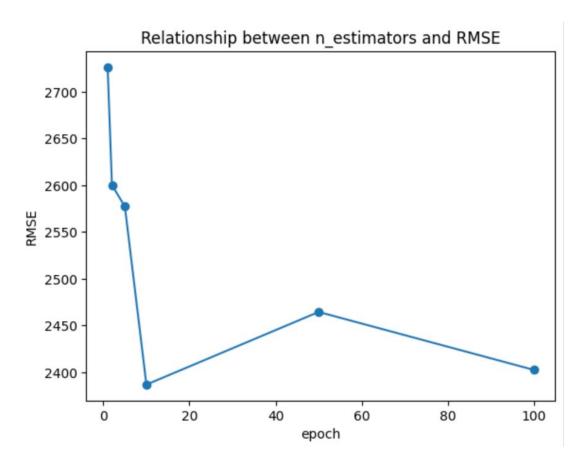


Multilayer Perceptron: Tuning Hyperparameters



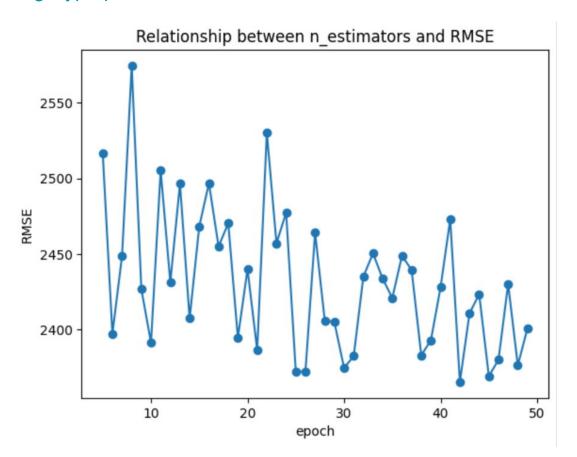


Random Forest: Tuning Hyperparameters





Random Forest: Tuning Hyperparameters





Linear Regression: Parameters

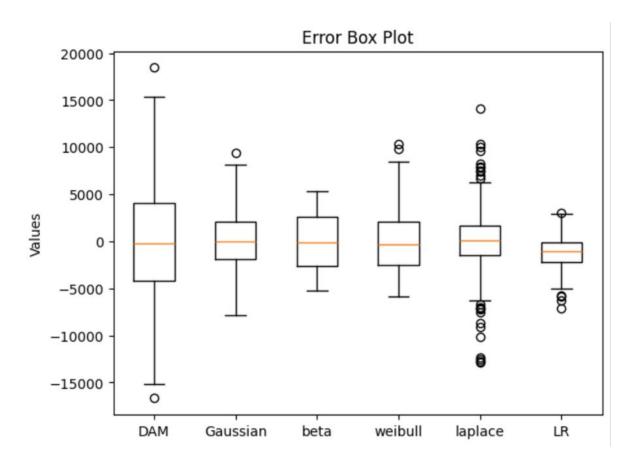
Intercept: 3809.3155696293616

Coefficients: [0.05976227 0.19431258 0.19846912 0.19496223 0.19161918]

LR: 1940.2586104000395

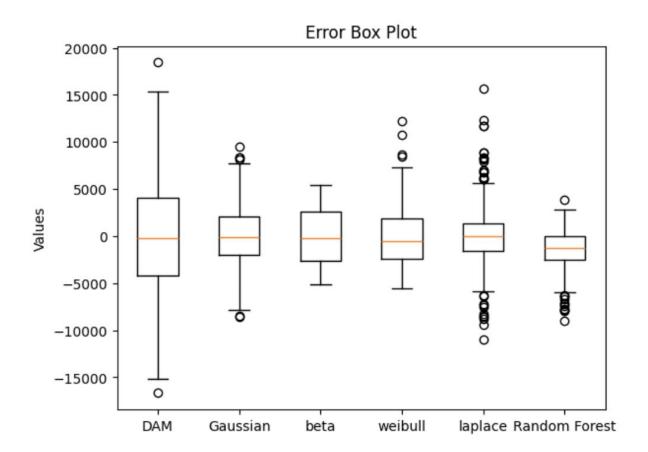


Error compared with single models





Error compared with single models





Conclusion

In this research project, we explore the whole procedure of deploying and optimizing machine learning models for increasing prediction accuracy, by experimenting with data processing and machine learning methods, we improved the accuracy of the power demand forecasting.





- Reinforcement learning methods
- Combining with data markets
- Real world CAISO model improvement

SIEMENS

