

24.01.19

다변량 스팀 사용 이상 감지 및 영향 변수의 원인 분석

서울과학기술대학교 데이터사이언스학과

이성호 sean0310@seoultech.ac.kr

배소희 shbae2819@g.seoultech.ac.kr

심재웅 jaewoong@seoultech.ac.kr

Tg04에 대한 예측

데이터

이용 데이터

제품 2종에 대한 6달간 센서 데이터

df_ext(2023-03,04)(5123,0385)_2023-11-16 seoultech

df_ext(2023-05,06)(5123,0385)_2023-11-16 seoultech

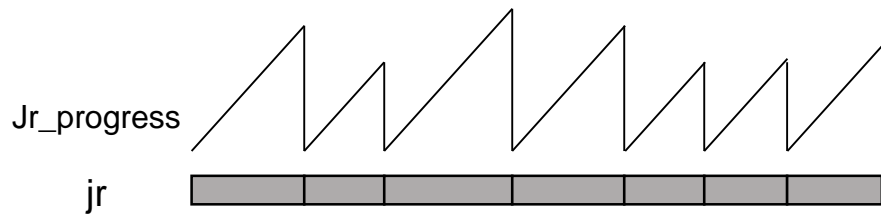
df_ext(2023-07,08)(5123,0385)_2023-11-16 seoultech

기간 : 2023-03-02 08:00:00 ~ 2023-08-27 03:00:00 (분)

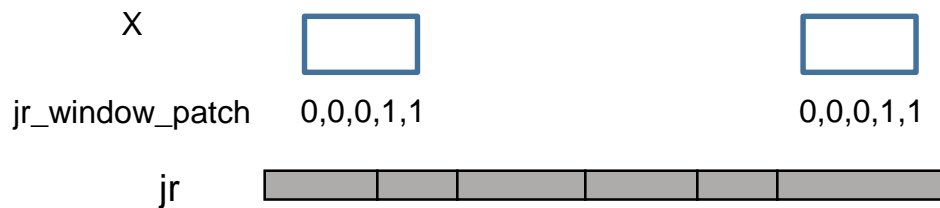
date	날짜
tg	Sensor (38개)
stop	공정 분석값 0 : 가동 1 : 중지 이벤트 발생 2 : 중지 복구
jr	단위 공정값 / 제품 생산 주기 (생산품 번호)

이용 데이터 전처리

- jr_progress : jr을 기준으로 시간에 따른 정수를 새로운 변수로 추가



- jr_window_patch : window 사이즈 만큼의 데이터 패치 후, 예측하고자 하는 시점의 jr과 다른지 같은지 표기하는 새로운 변수 추가



데이터

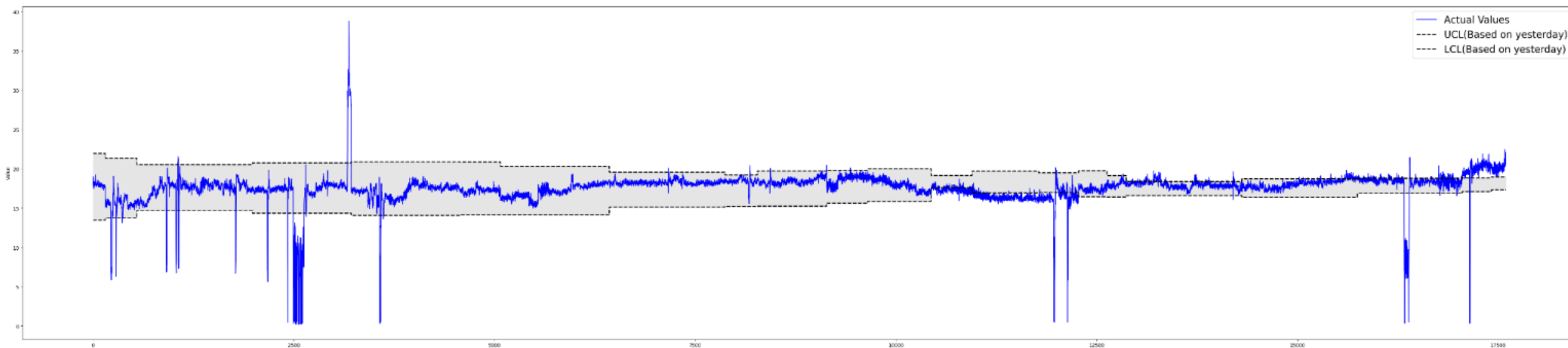
- 이용 데이터
 - 이용 데이터 정의
 - **input(X) : tg (38개 sensor 데이터), 공정진행도(jr_progress) , 공정변화도(jr_window_patch)**
 - Window size : 30
 - **output(y) : 5분 후의 tg 04(스팀 순간값)**
 - Data split
 - Train/Validation/Test = 60:20:20
 - Train : 2023-03-02 08:00:00 ~ 2023-06-16 18:37:00 (51840)
 - Validation : ~ 2023-07-16 12:47:00 (17340)
 - Test : ~ 2023-08-27 03:00:00 (17268)

UCL & LCL 설정(이상치 제외한 이전의 5일 누적)

Sigma = 3

누적 날짜 = 5

{0(LCL-UCL 구간 내): 14585, 1(LCL 이하): 1834, 2(UCL 이상): 1179}



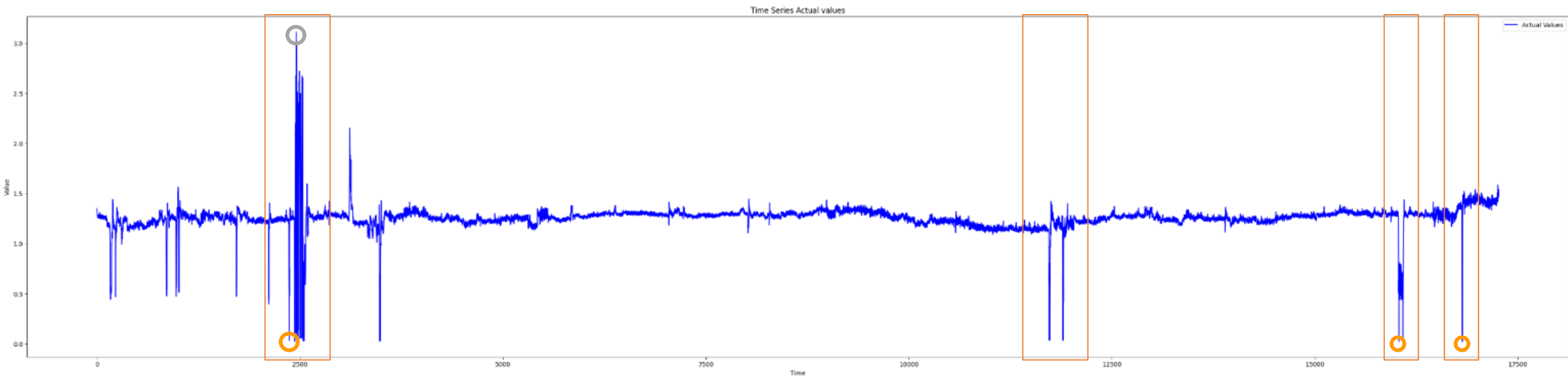
Instance 설정

● Value가 3이상인 index

index	date	ei
2452	2023-07-23 08:48:00	3.1125
2457	2023-07-23 08:53:00	3.0158

● Value가 0.025이하인 index

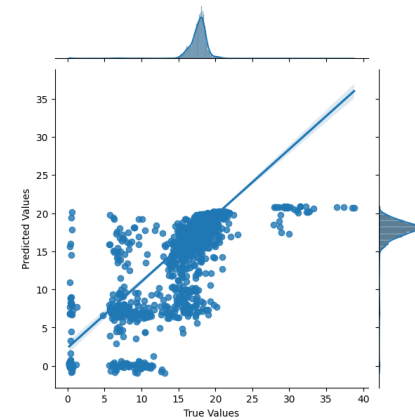
index	date	ei
2433	2023-07-23 08:29:00	0.0241
16036	2023-08-23 09:47:00	0.0241
16818	2023-08-26 19:31:00	0.0226



1D CNN

● Experiment setting

- Epoch : 100
- optimizer : Adam(lr=1e-4)
- val loss가 가장 낮은 모델로 test
- Layer : 1D Conv layer(16->32->64->128) / kernel=3),
Linear layer (128 -> 32 -> 1)

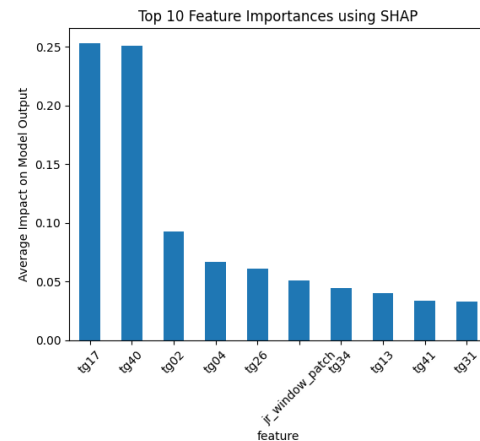
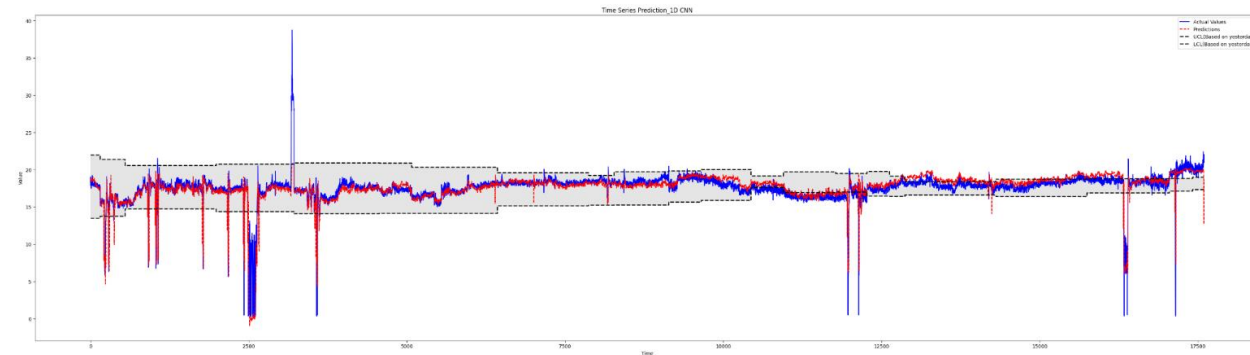
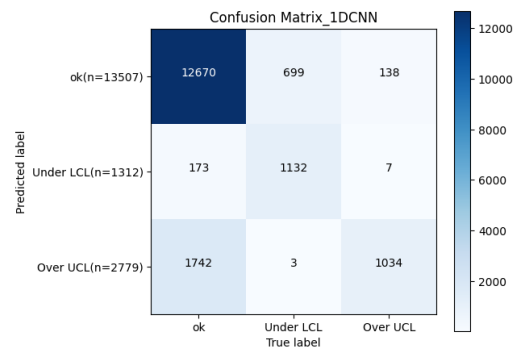


● Result

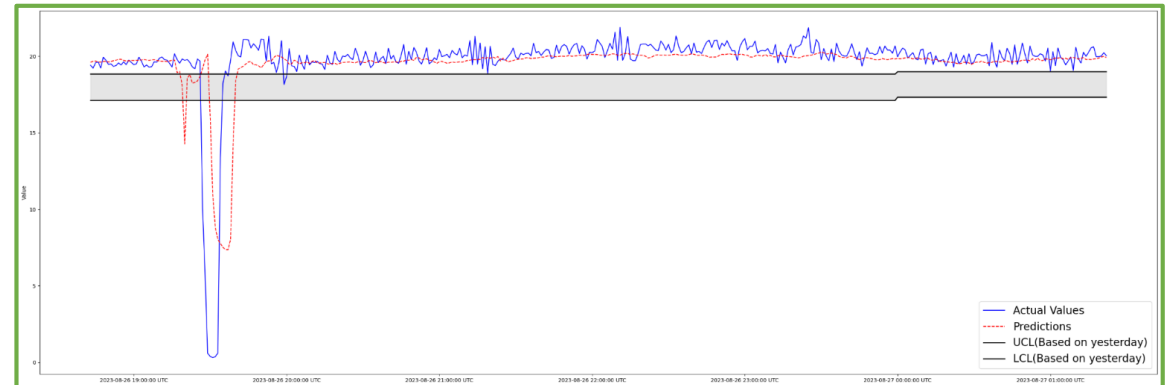
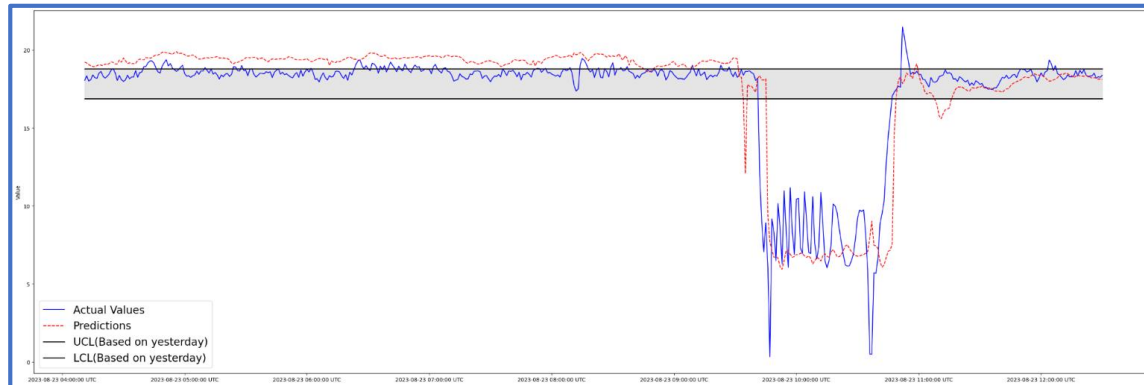
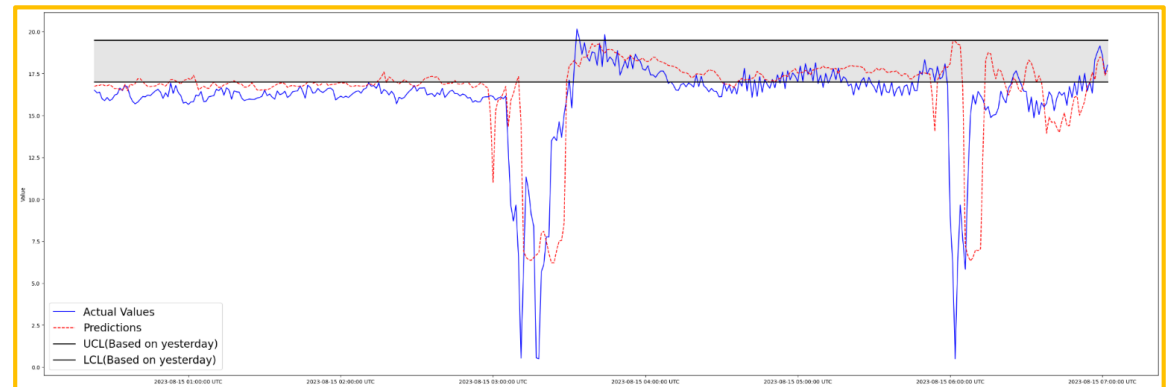
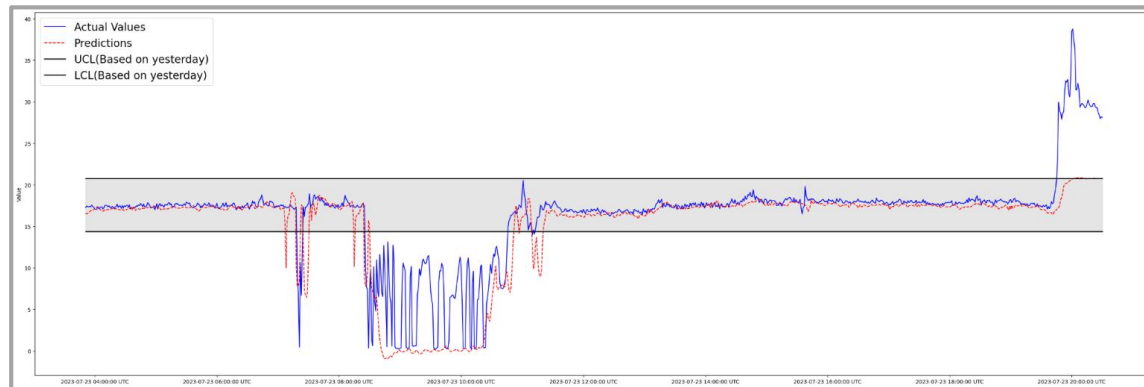
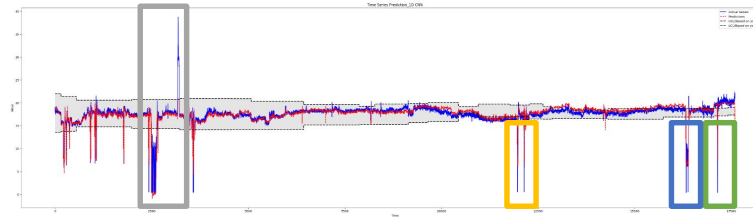
- R^2 : 0.5151
- MSE : 1.8069

● 변수 별 중요도 해석

- tg17, tg40



1D CNN



LSTM

● Experiment setting

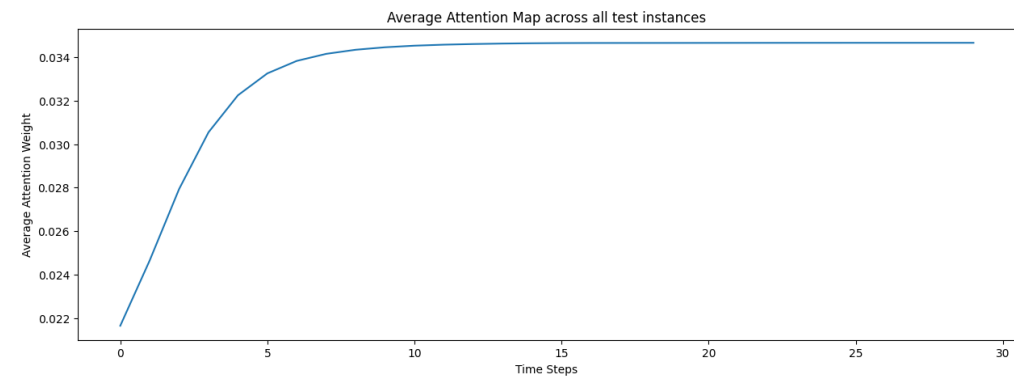
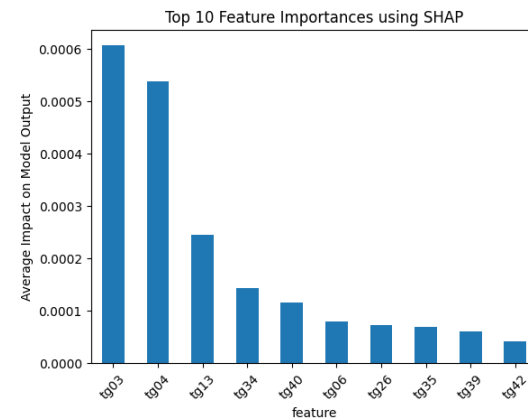
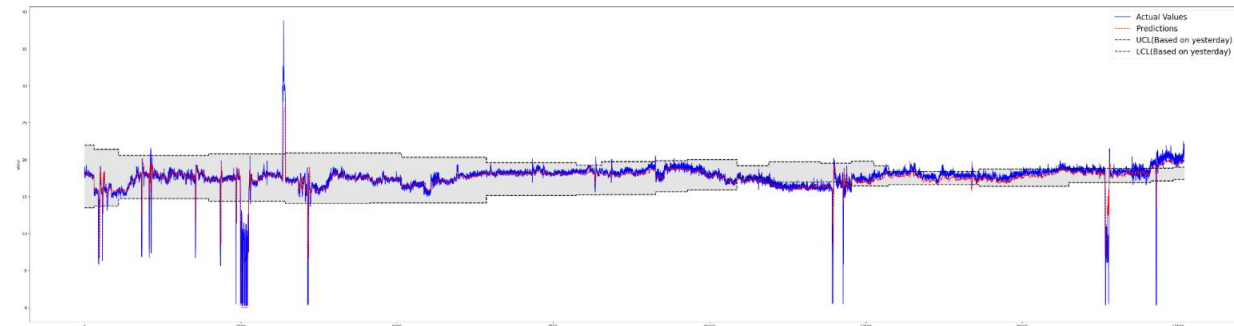
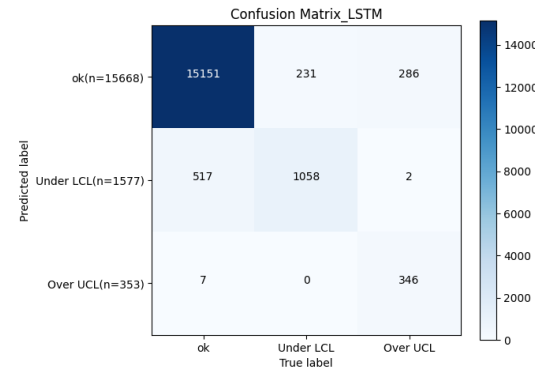
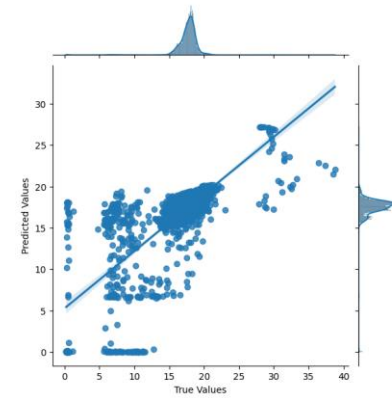
- Epoch : 100
- optimizer : Adam(lr=1e-4)
- val loss가 가장 낮은 모델로 test
- Layer : lstm layer(hidden=512, layer=6) + attention layer

● Result

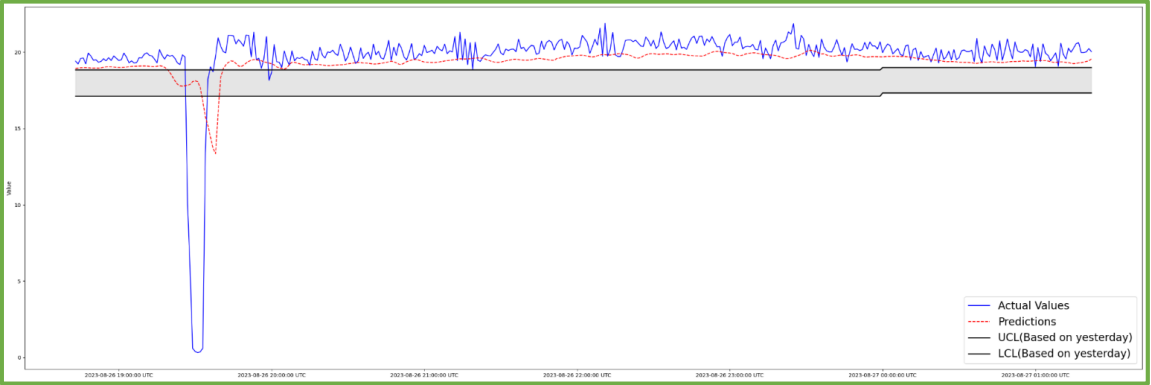
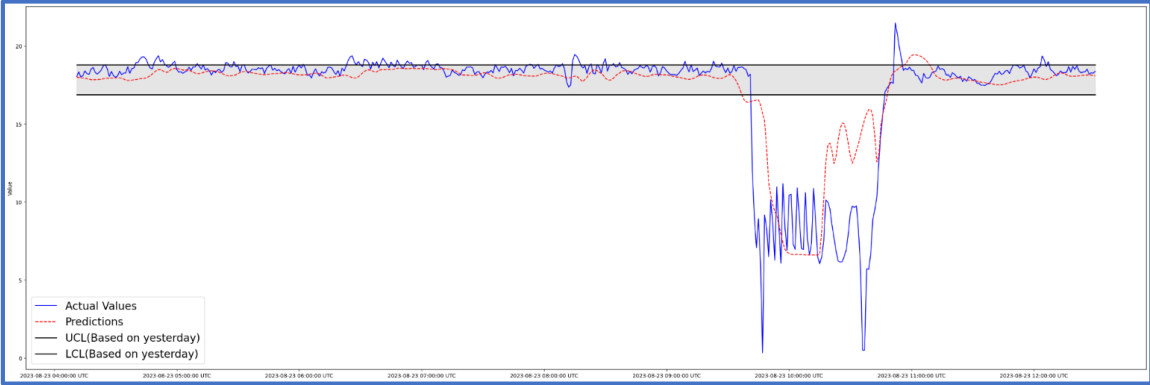
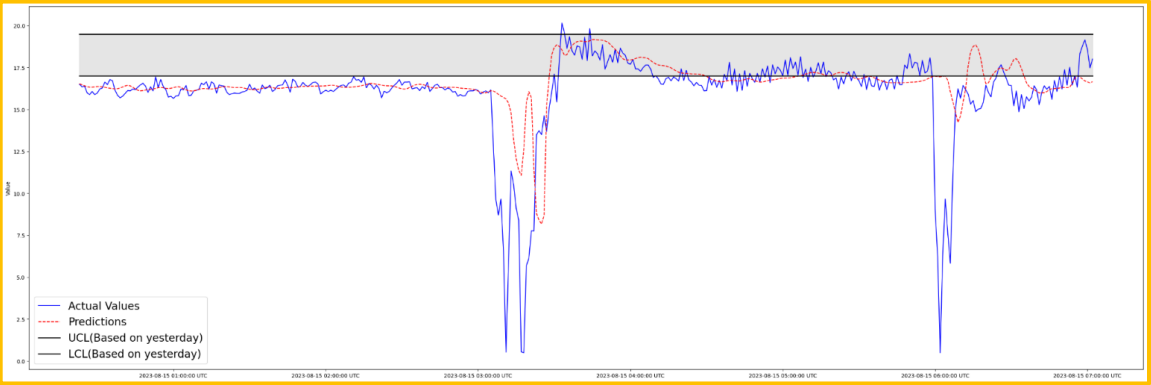
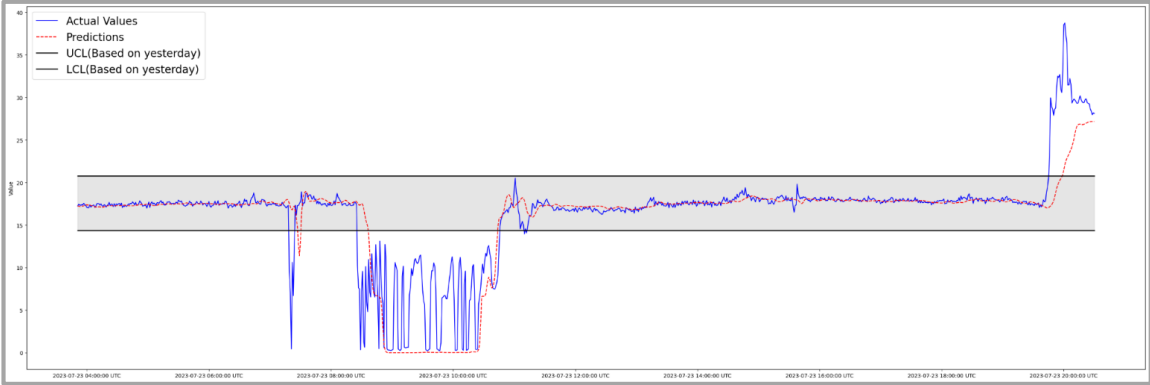
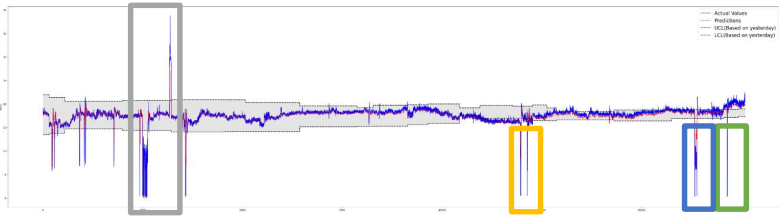
- R^2 : 0.5682
- MSE : 1.6091

● 변수 별 중요도 해석

- tg03, tg04



LSTM



IMV-LSTM

● Experiment setting

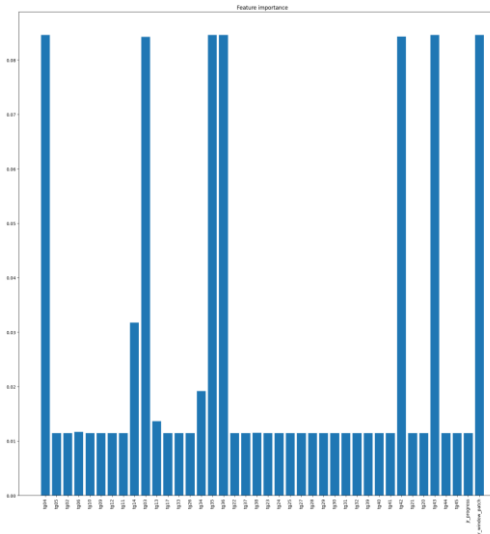
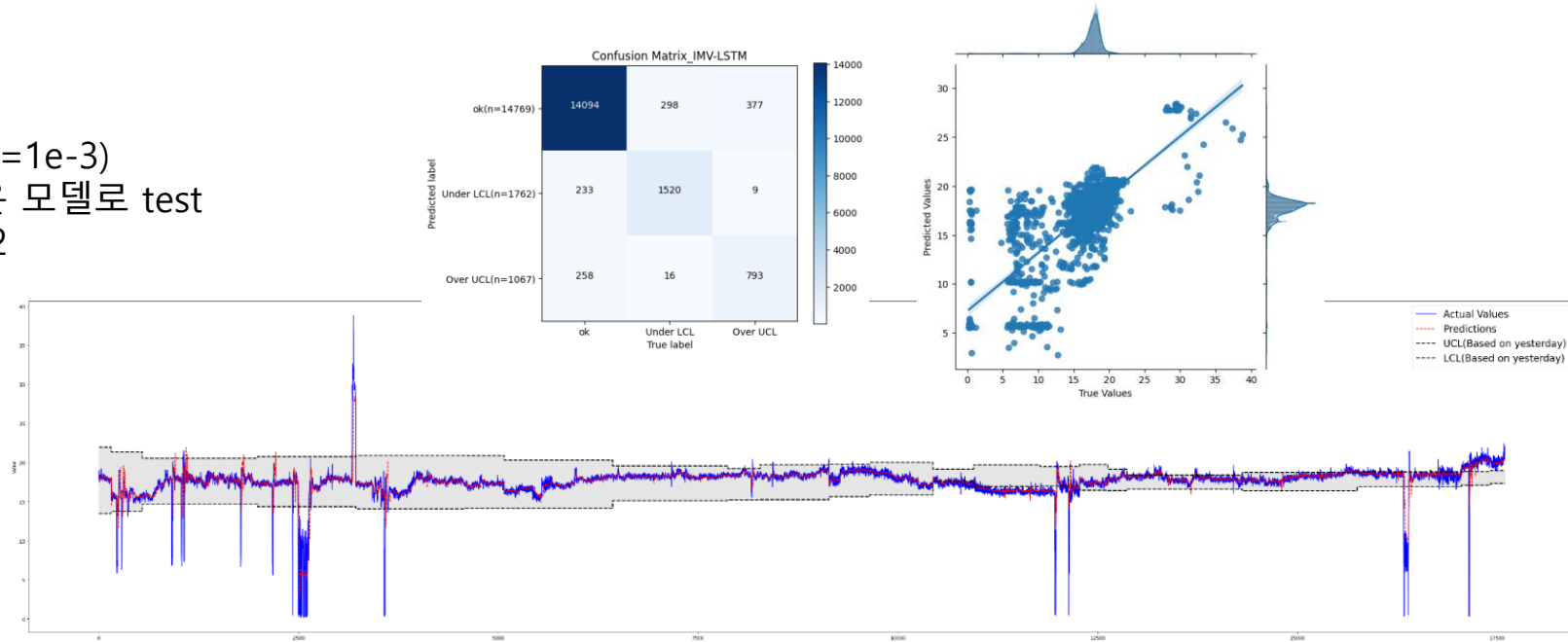
- Epoch : 100
- optimizer : Adam(lr=1e-3)
- val loss가 가장 낮은 모델로 test
- Lstm layer 노드 : 32

● Result

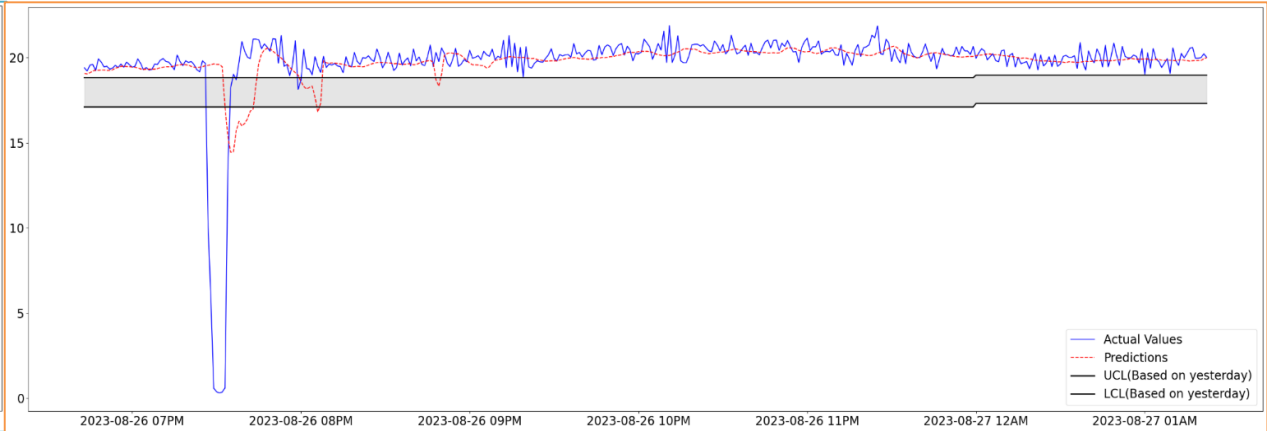
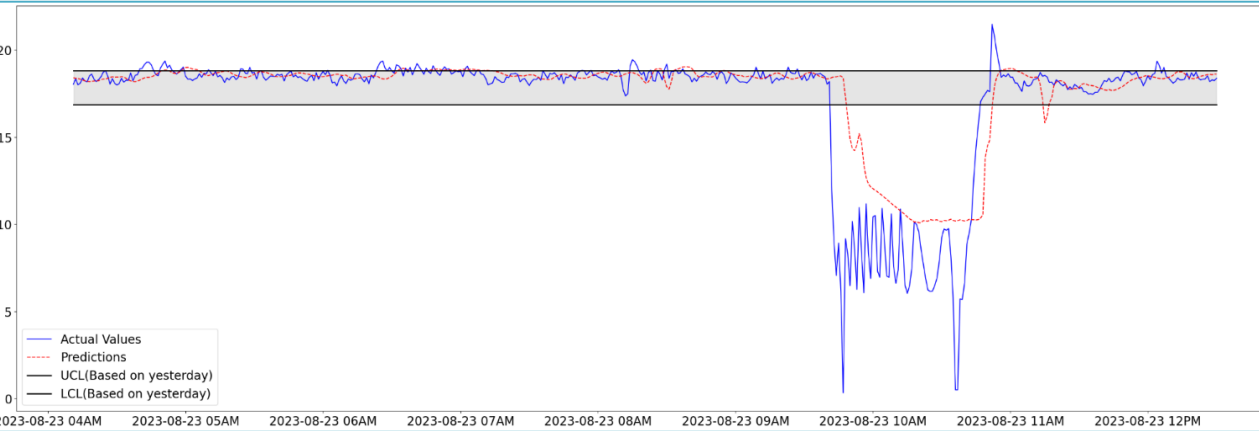
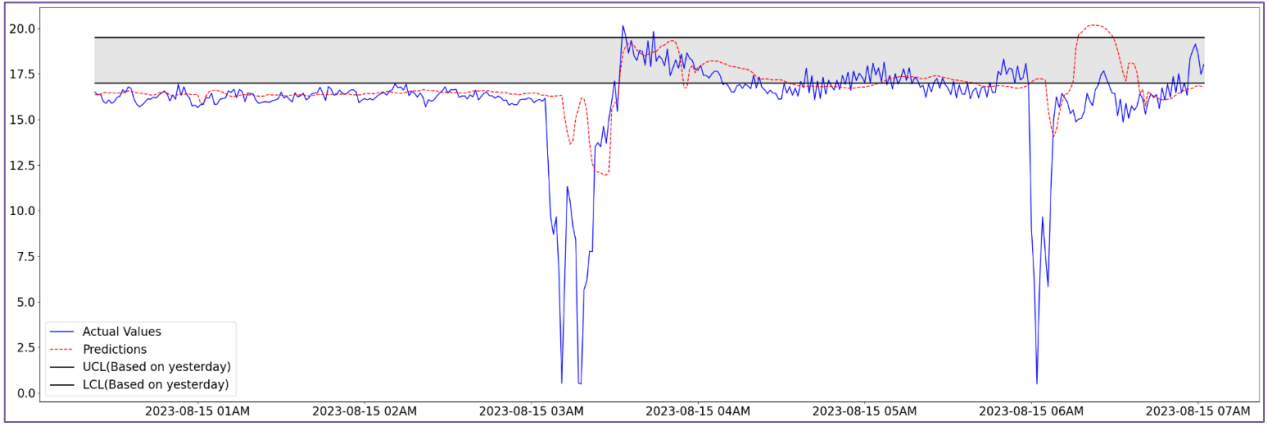
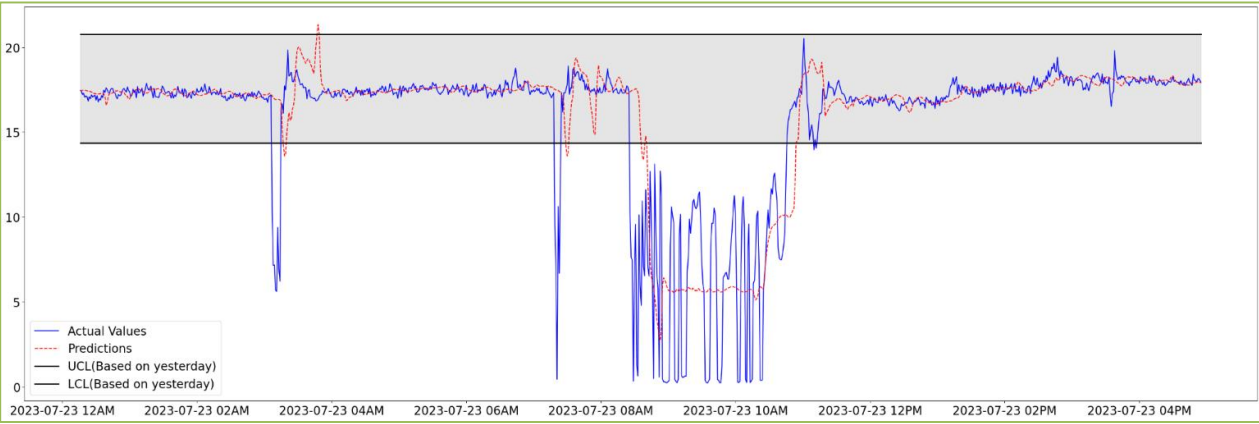
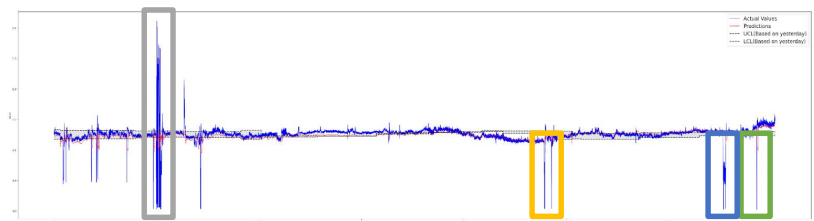
- R^2 : 0.6028
- MSE : 1.48
- Acc : 0.9323
- Macro F1 : 0.8373

● 모델 해석(Attention Map)

- tg04, tg03, tg35, tg36, tg42, tg43, jr_window_patch



IMV-LSTM



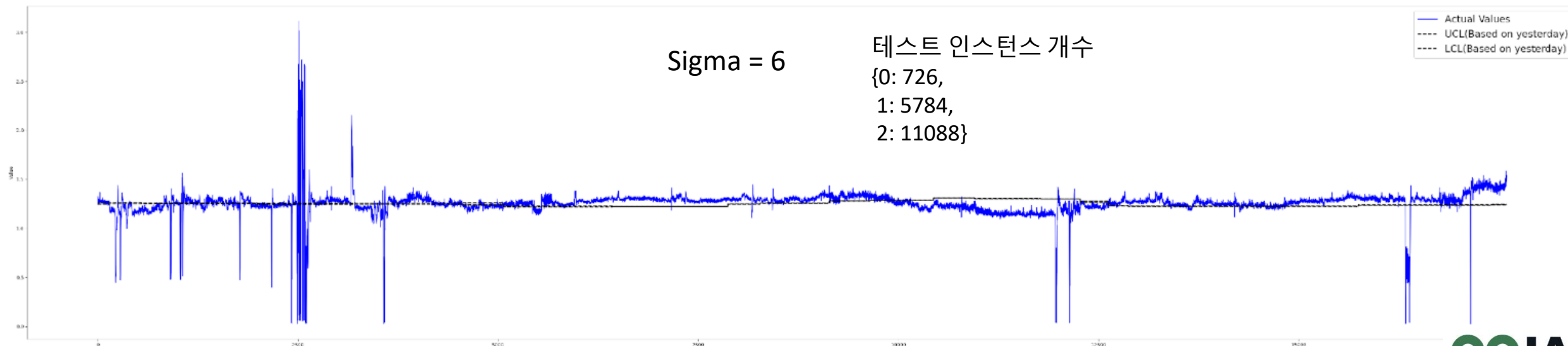
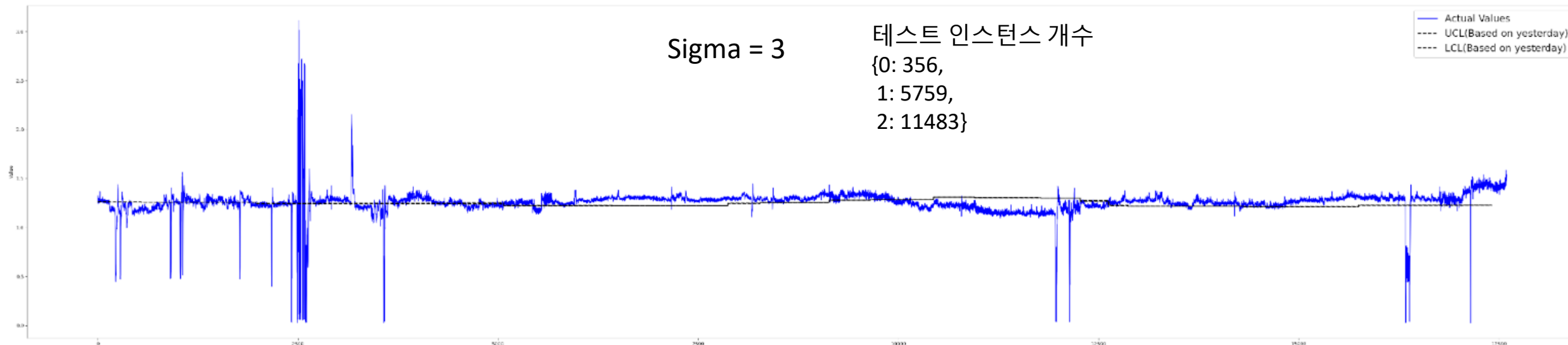
ei값에 대한 예측

데이터

- 이용 데이터
 - 이용 데이터 정의
 - **input(X) : tg (38개 sensor 데이터), 공정진행도(jr_progress) , 공정변화도(jr_window_patch)**
 - Window size : 30
 - **output(y) : 5분 후의 ei계산값 $tg04/(tg02*tg03*0.0003)$**
 - Data split
 - Train/Validation/Test = 60:20:20
 - Train : 2023-03-02 08:00:00 ~ 2023-06-16 18:37:00 (51840)
 - Validation : ~ 2023-07-16 12:47:00 (17340)
 - Test : ~ 2023-08-27 03:00:00 (17268)

UCL & LCL 설정(이상치 제외한 이전의 7일 누적)

0 : LCL-UCL 구간 내
1 : LCL 이하
2 : UCL 이상



분산이 너무 작아 UCL - LCL 사이 구간이 너무 작아짐

1D CNN

● Experiment setting

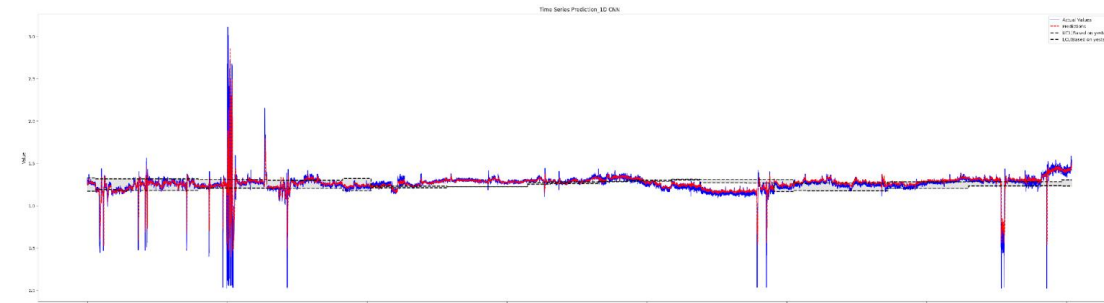
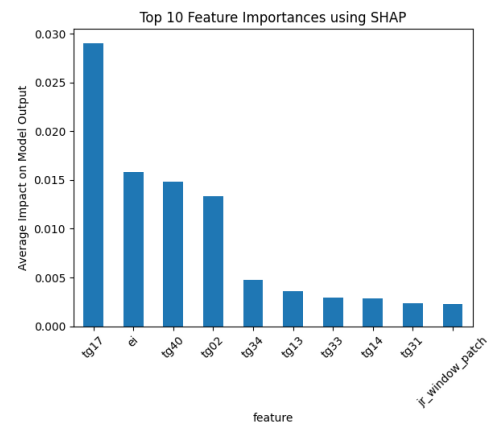
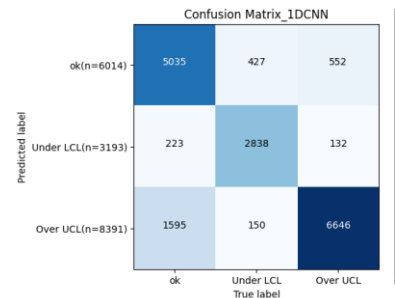
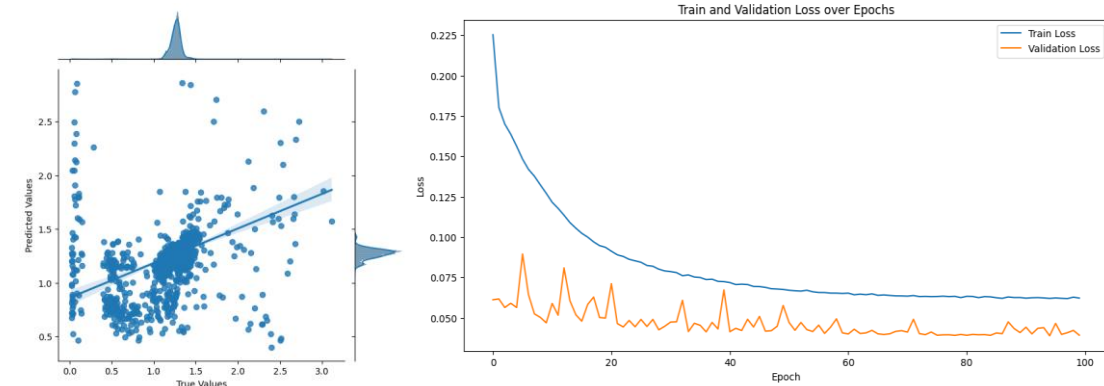
- Epoch : 100
- optimizer : Adam(lr=1e-4)
- val loss가 가장 낮은 모델로 test
- Layer : 1D Conv layer(16->32->64->128->256) / kernel=3),
Linear layer (128 -> 32 -> 1)

● Result

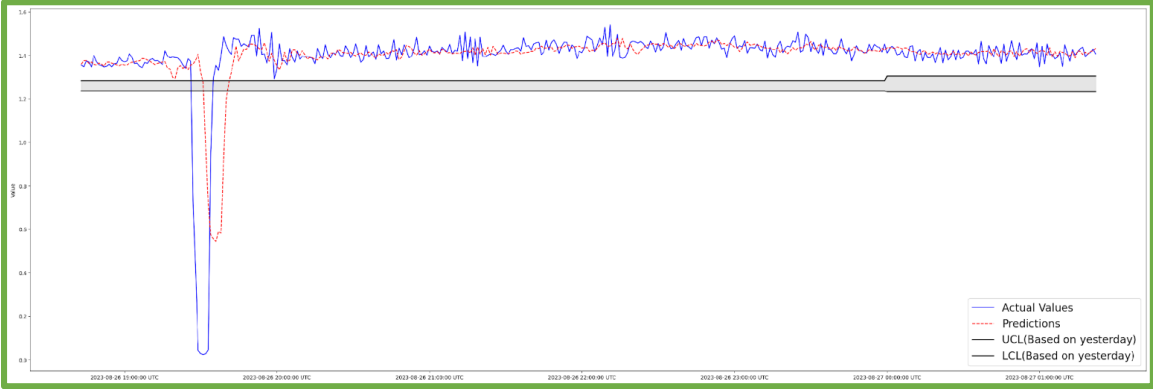
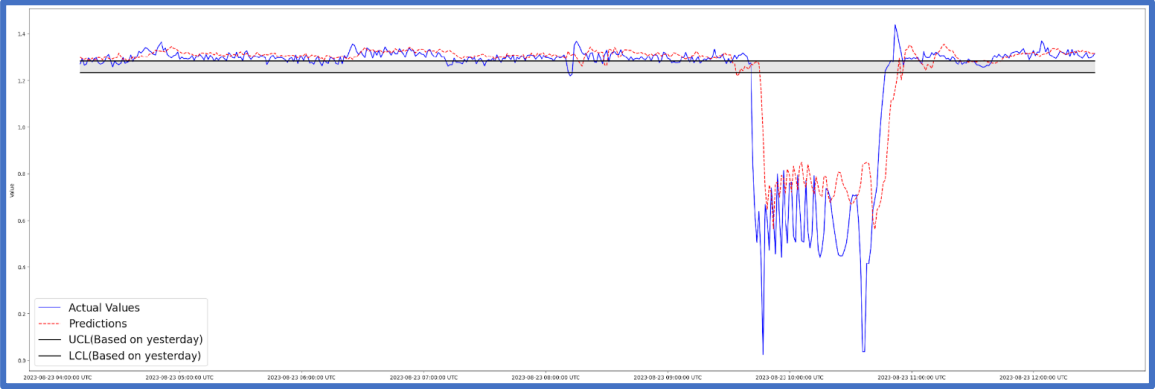
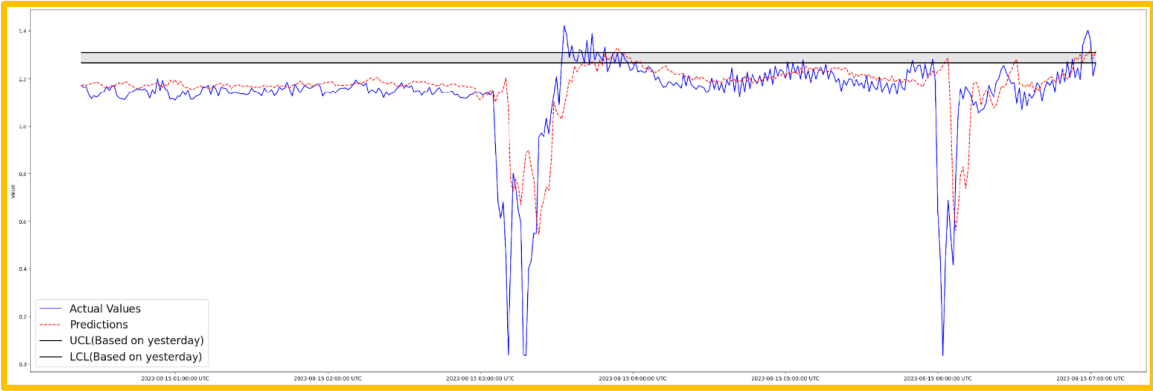
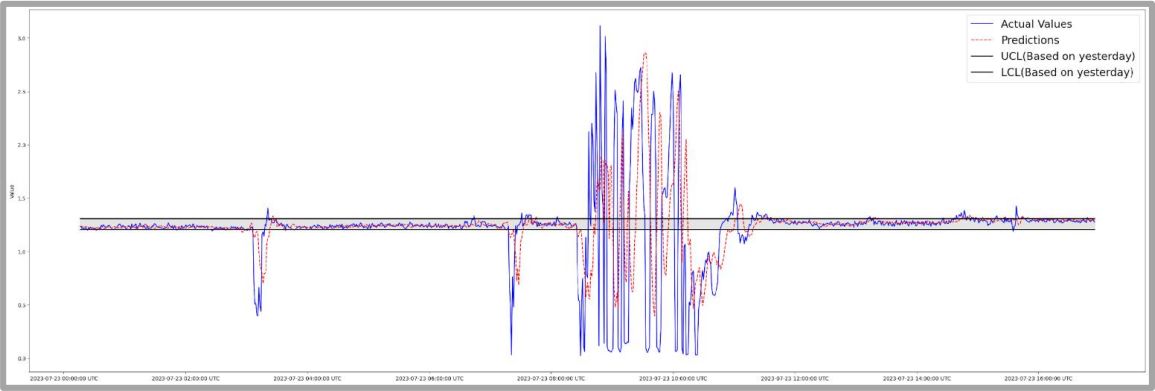
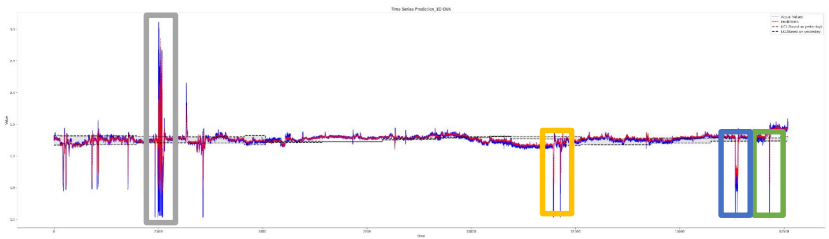
- R^2 : 0.1317
- MSE : 0.0146

● 변수 별 중요도 해석

- tg17, ei



1D CNN



LSTM

● Experiment setting

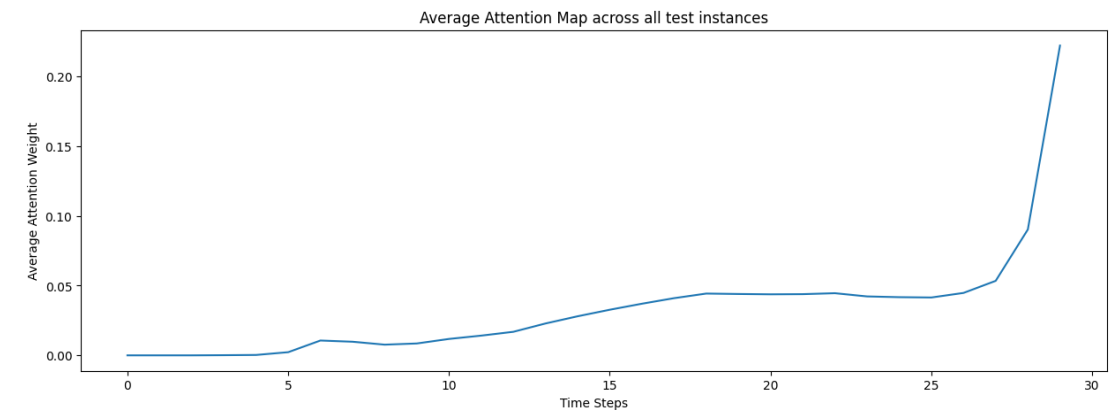
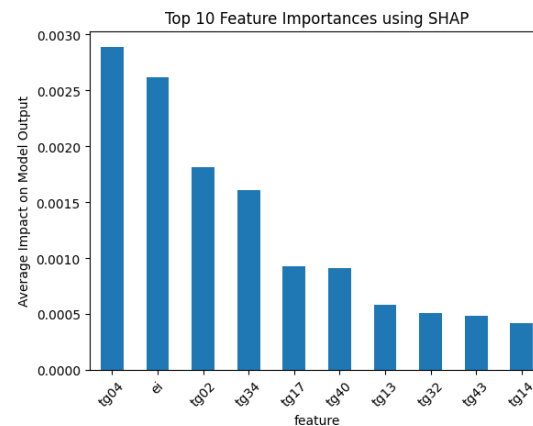
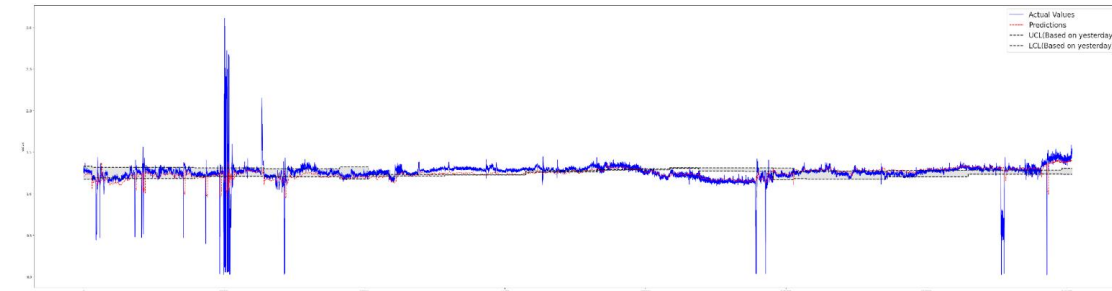
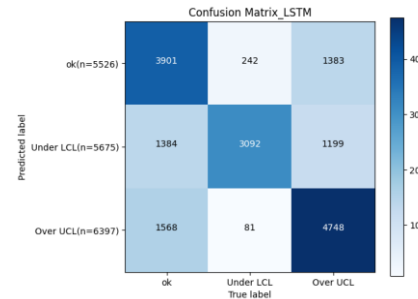
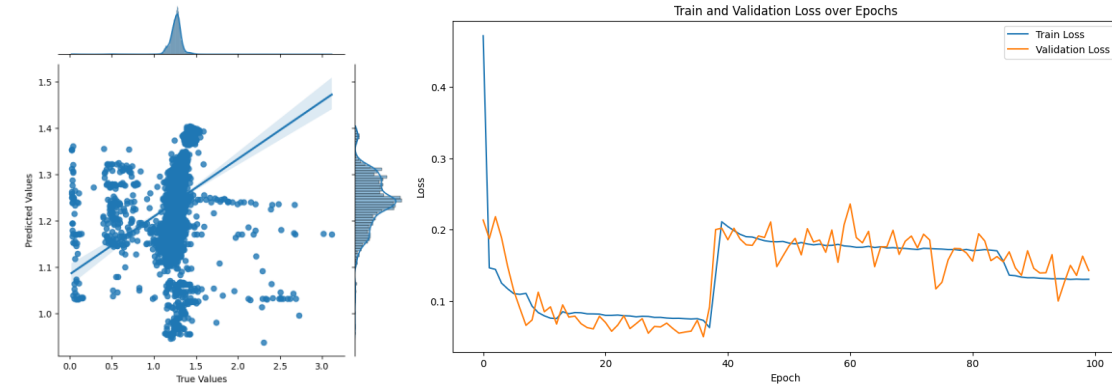
- Epoch : 100
- optimizer : Adam(lr=1e-4)
- val loss가 가장 낮은 모델로 test
- Layer : lstm layer(hidden=512, layer=6) + attention layer

● Result

- R^2 : 0.0238
- MSE : 0.0164

● 변수 별 중요도 해석

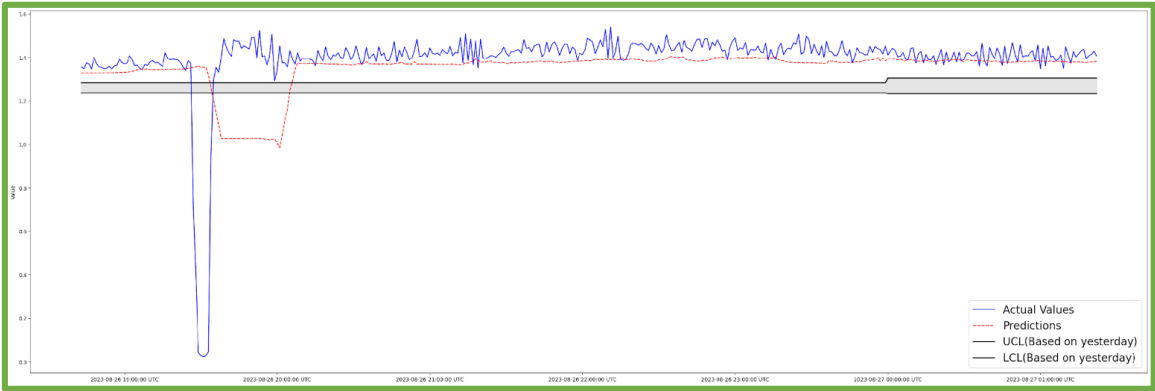
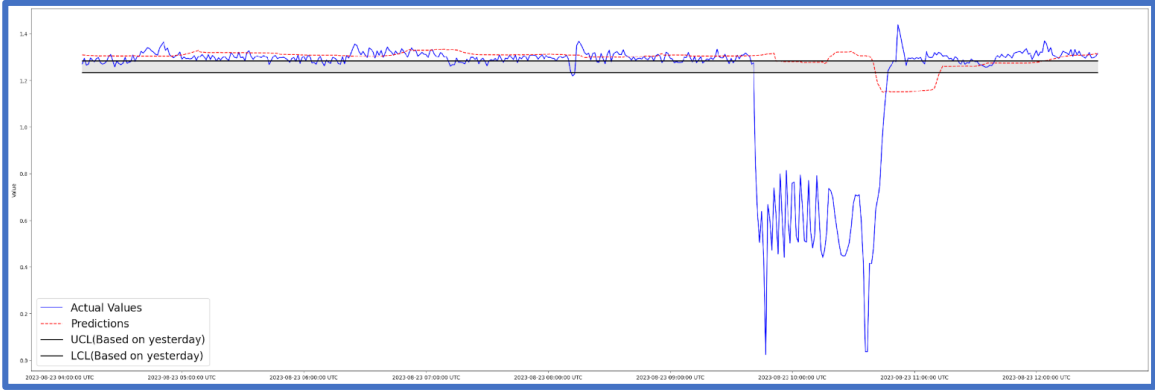
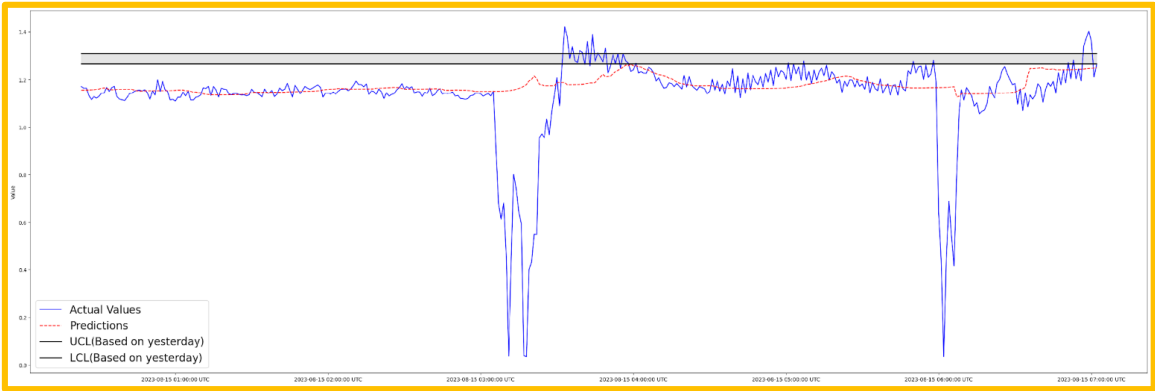
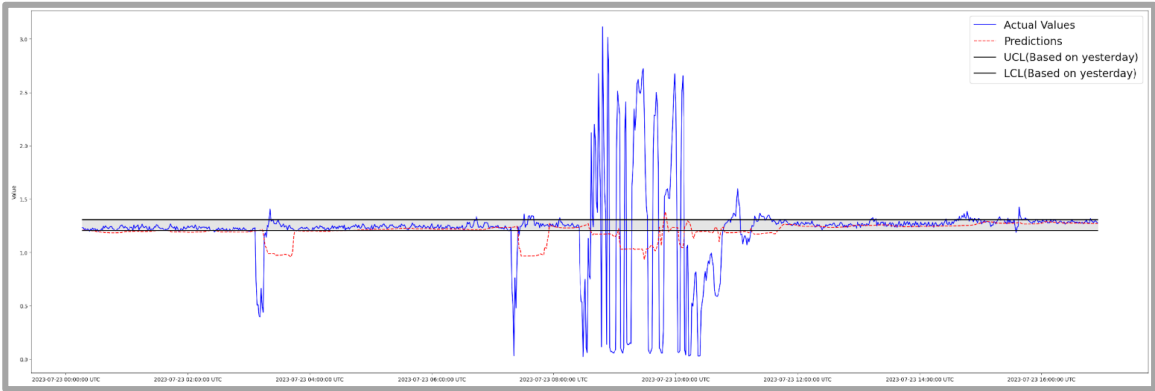
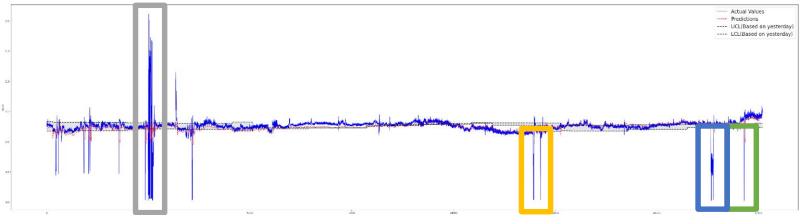
- tg04, ei



* Validation loss가 매우 불안정, epoch 조절해도 동일

* 예측값의 변동 폭이 매우 작음

LSTM



IMV-LSTM

● Experiment setting

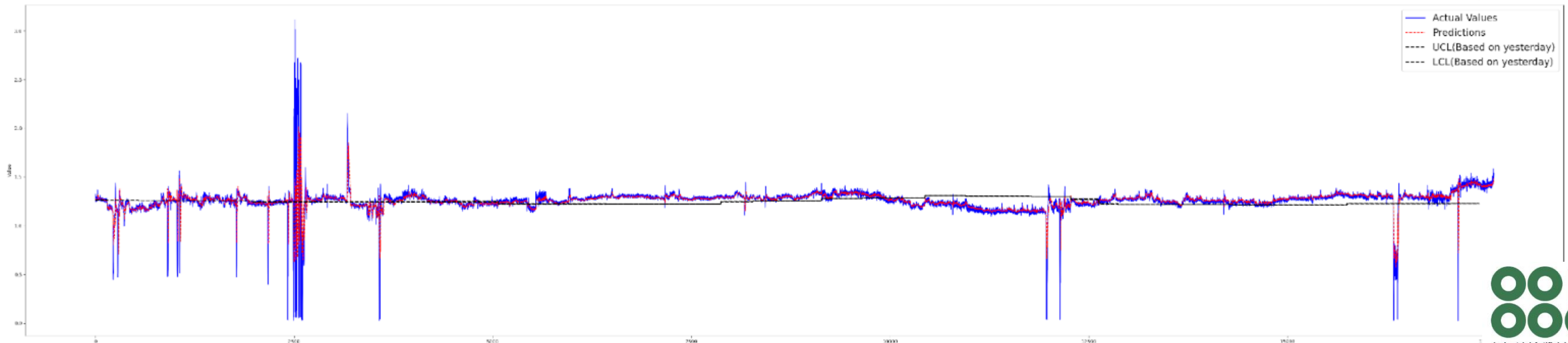
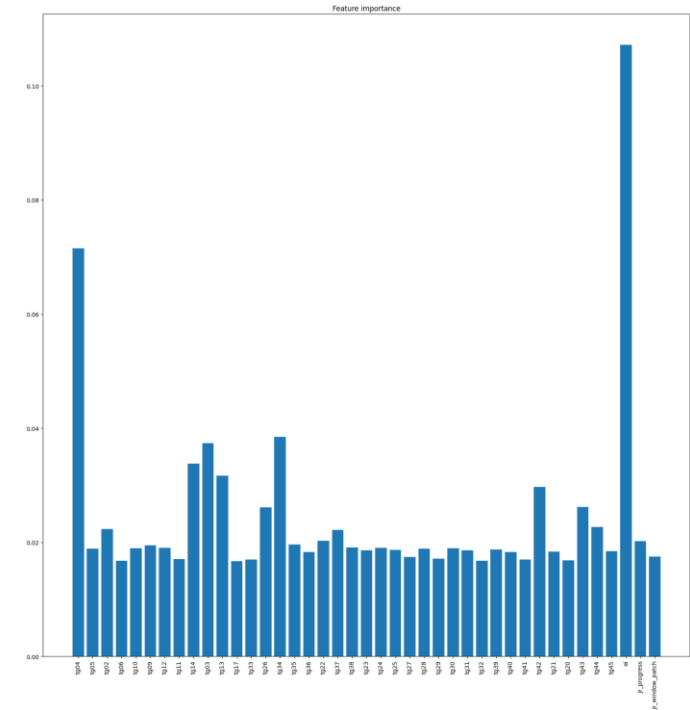
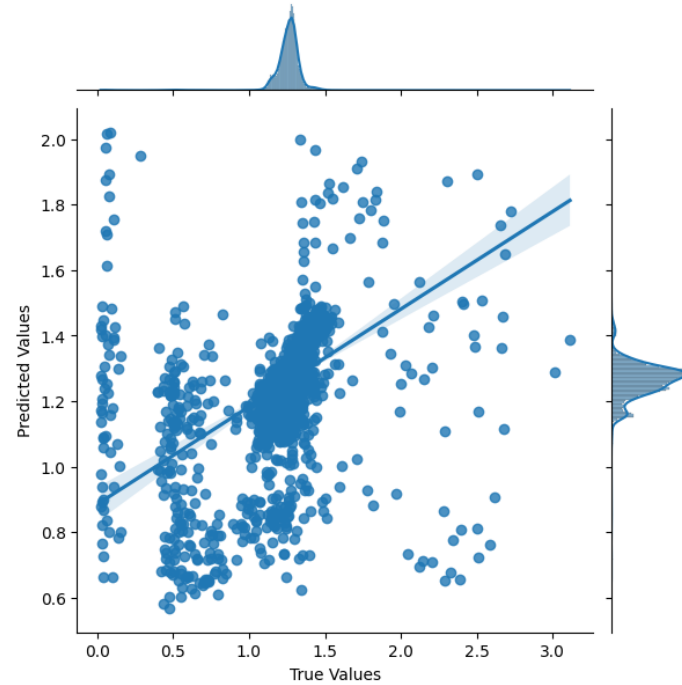
- Epoch : 100
- optimizer : Adam(lr=1e-3)
- val loss가 가장 낮은 모델로 test
- Lstm layer 노드 : 32

● Result

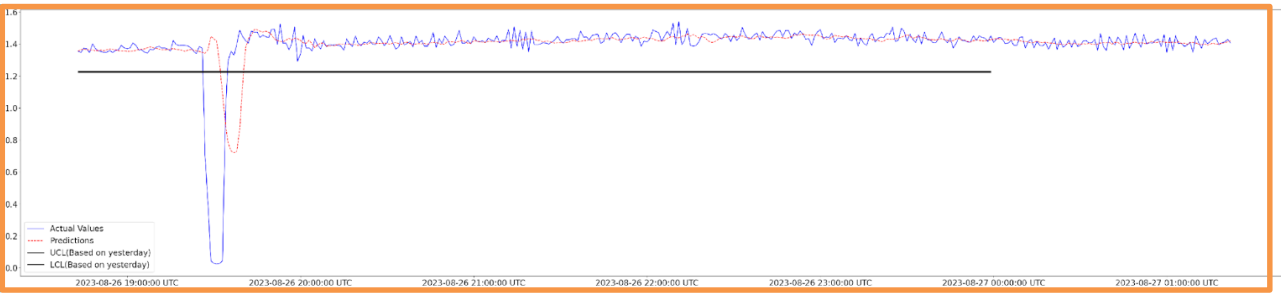
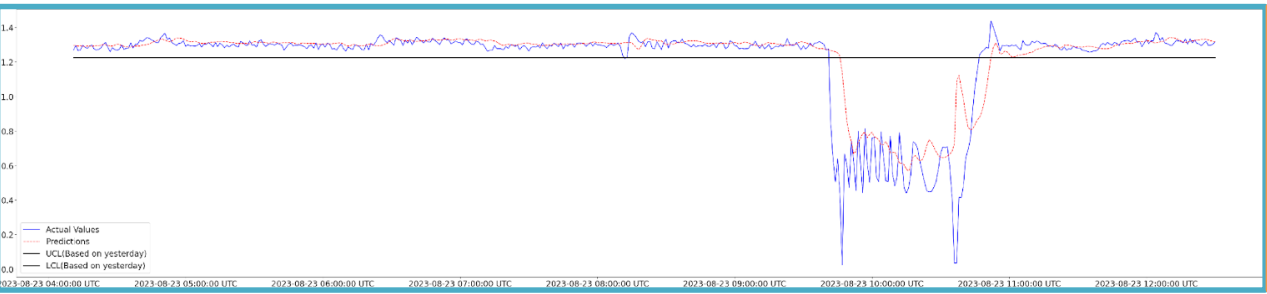
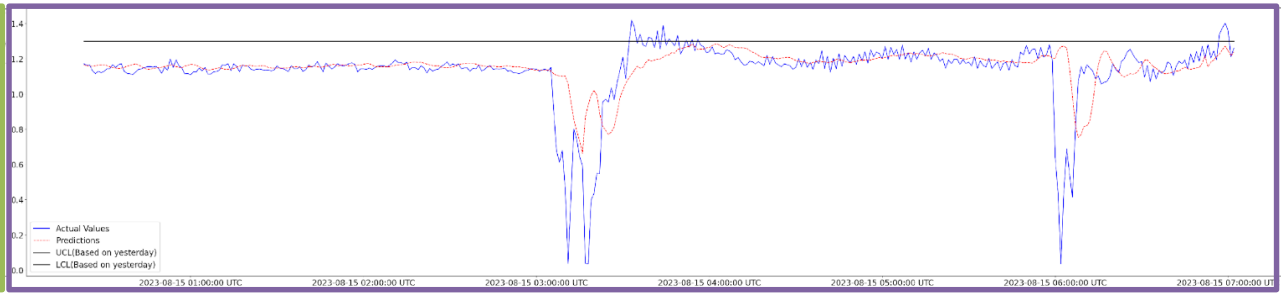
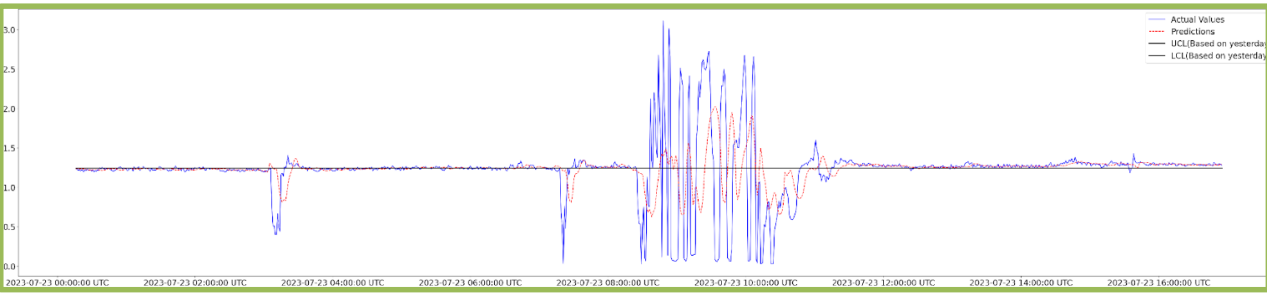
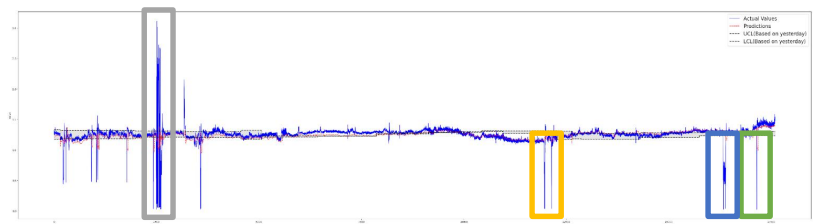
- R^2 : 0.2052
- MSE : 0.0134

● 모델 해석(Attention Map)

- ei, tg04



IMV-LSTM



예측 설명 예시

IMV-LSTM (Tg04에 대한 예측)

● Experiment setting

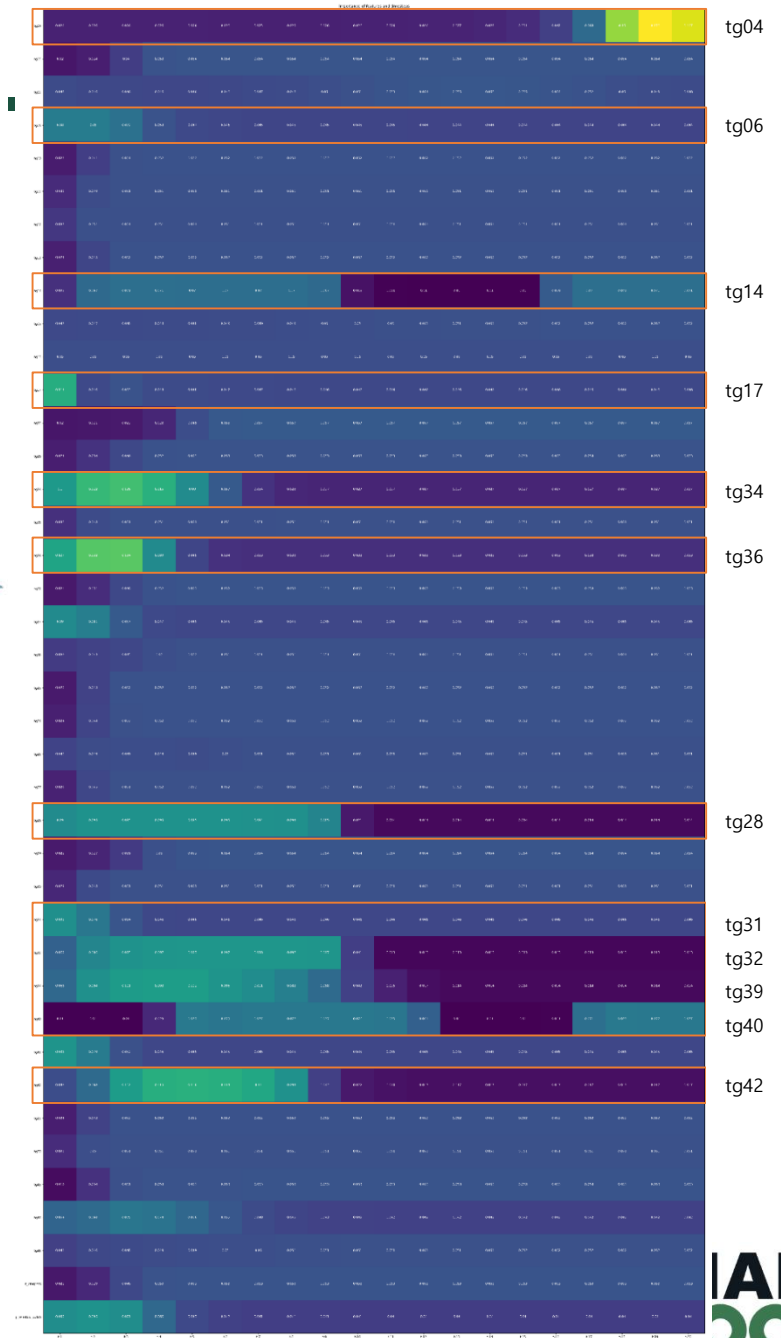
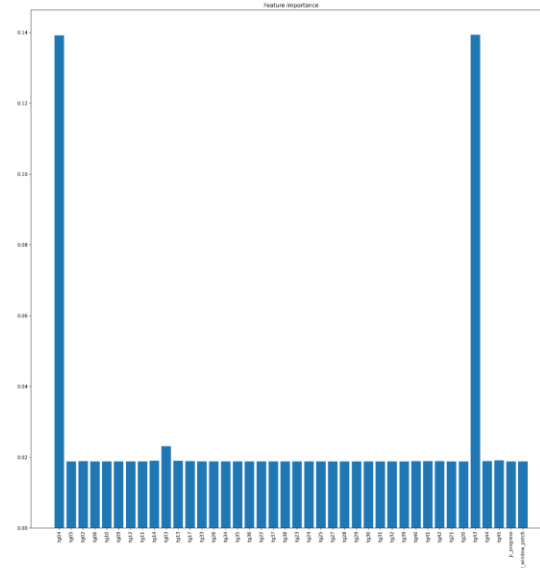
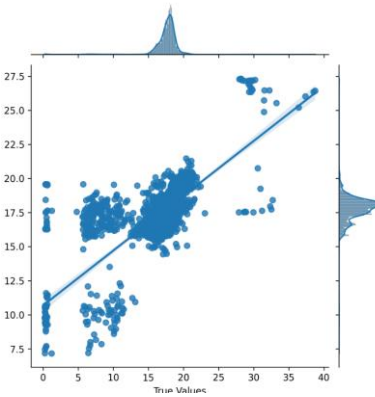
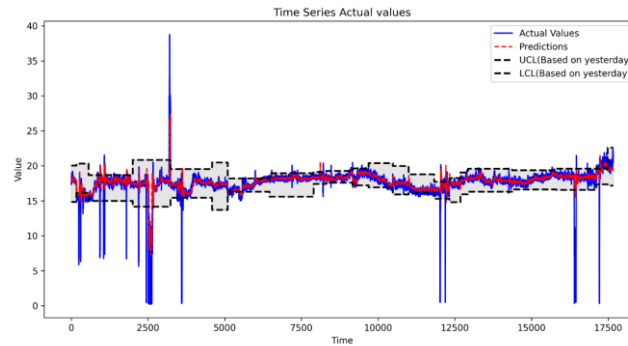
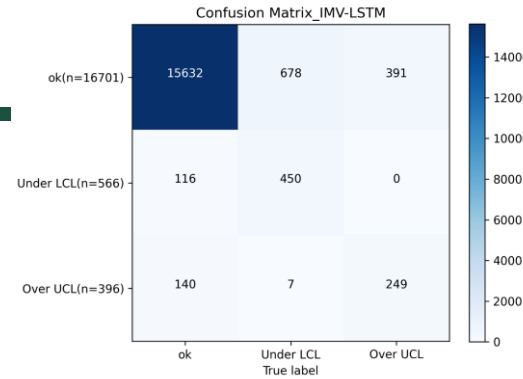
- Epoch : 100
- optimizer : Adam(lr=1e-3)
- val loss가 가장 낮은 모델로 test
- Lstm layer 노드 : 32

● Result

- R^2 : 0.4881
- MSE : 1.9021

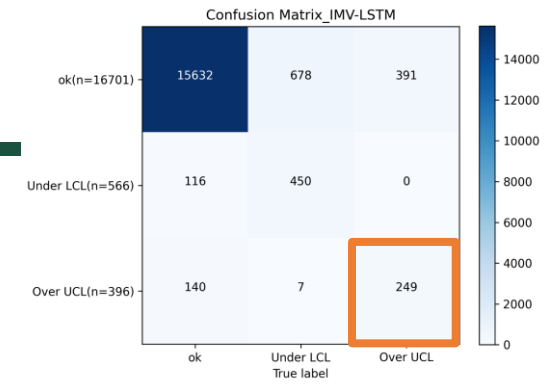
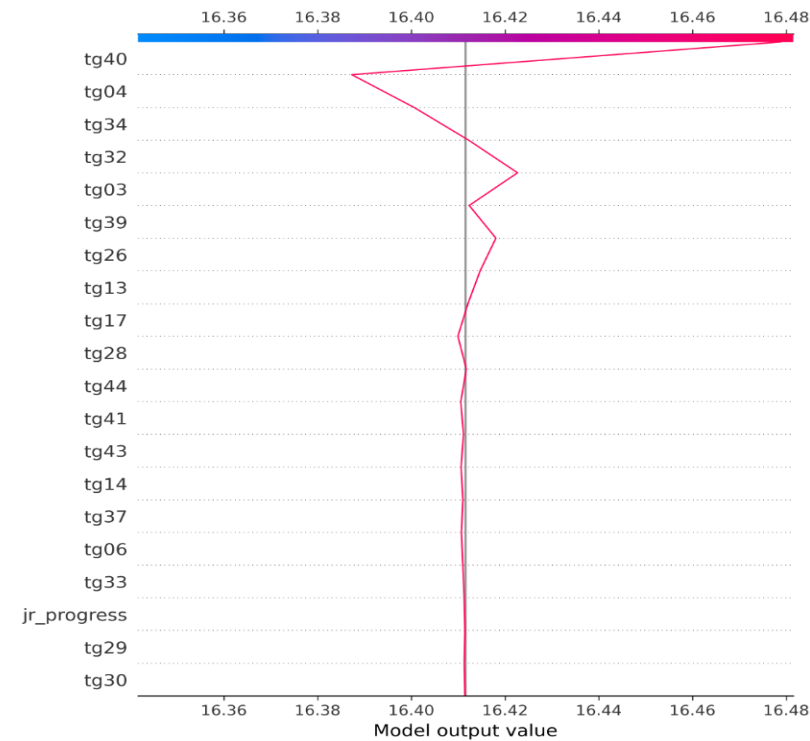
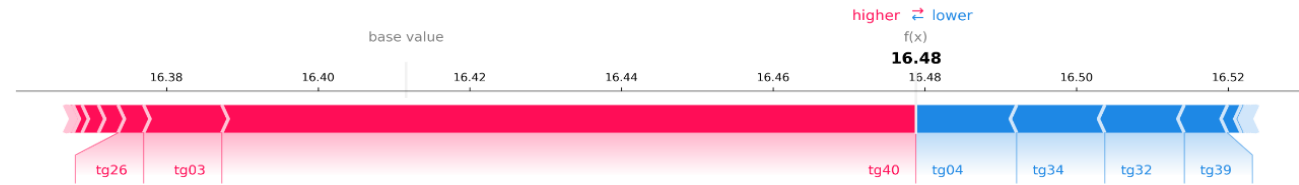
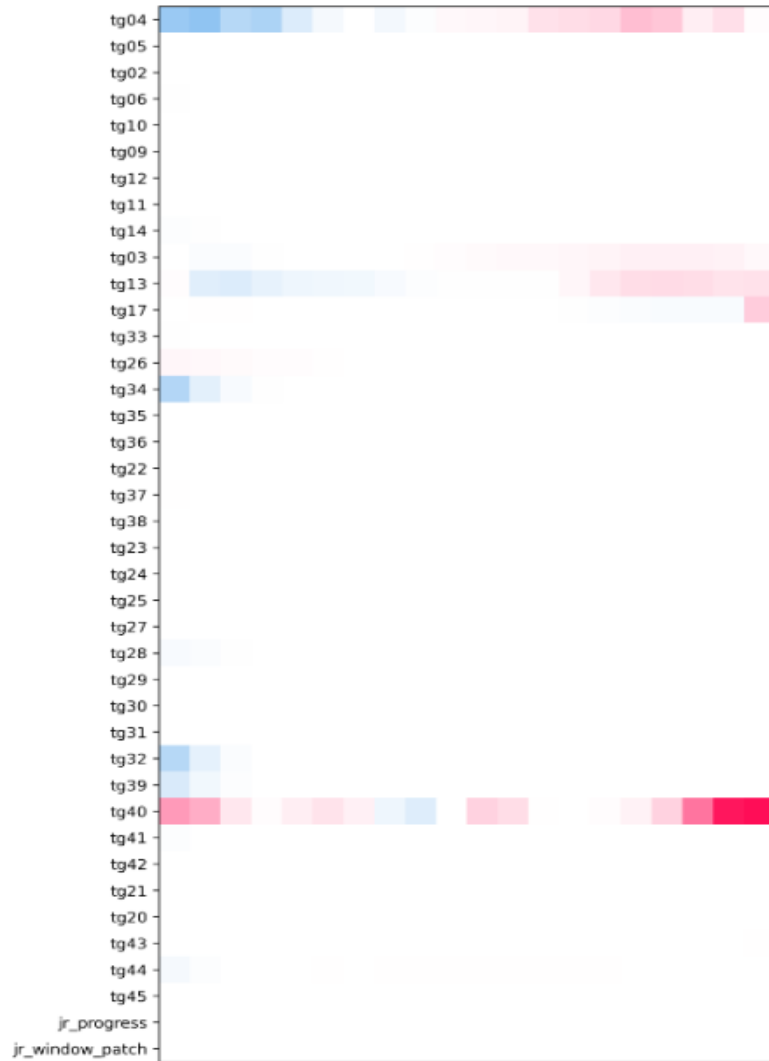
● Global Attention Map

- tg04, tg43, tg03



IMV-LSTM

- 개별 예측에 대한 해석 (local interpretation)
SHAP - Correct_over (249)



추후 계획

추후 계획

- Classification 문제로 변환하여 모델링 시도
- 코드 정리 / 모듈화
- 자료 작성

감사합니다