Regression_3112_1037_1037_Hanyang_Securities

February 8, 2021

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[1]: import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sns
    import warnings
    import os
    import sys
    import tensorflow as tf
    from sklearn.preprocessing import MinMaxScaler
    from tqdm import tqdm
[2]: # 예측할 종목은 한양증권(001750) 입니다
     # 기간은 00-01-04 ~ 21-02-05 입니다
     # Yahoo Finanace에서 다운로드 받아 null값을 제거한 데이터를 사용합니다
    stock = pd.read_csv('C:\Jupyter_Project\Hanyang_Securities_F.csv')
    df = stock.dropna()
    df.head()
[2]:
             Date Open High
                                    Close
                                             Adj Close
                                                       Volume
                               Low
    0 2000-01-04 6300 7100
                              6300
                                     7000 1619.266357
                                                        56800
    1 2000-01-05 6700 7100
                              6610
                                     6700 1549.868774
                                                        52100
                                     6300 1457.339844
    2 2000-01-06 7000 7000
                              6250
                                                        64900
    3 2000-01-07
                   6350 6600
                              6300
                                     6370 1473.532349
                                                        61800
    4 2000-01-10 6610 6700
                                     6500 1503.603882
                              6300
                                                        56100
[3]: df.tail()
[3]:
                Date
                       Open
                             High
                                     Low Close Adj Close Volume
    5181 2021-02-01
                      9200
                             9480
                                    9100
                                           9380
                                                   9380.0
                                                            81355
    5182 2021-02-02
                      9460
                             9810
                                    9460
                                           9700
                                                   9700.0 105755
    5183 2021-02-03
                      9850 10200
                                    9800
                                           9990
                                                   9990.0 170966
    5184 2021-02-04 10100
                            10200
                                    9940
                                         10150
                                                  10150.0 133504
    5185 2021-02-05 10200 10800 10150
                                         10650
                                                  10650.0 247224
```

```
[4]: # OHLC를 Adj OHLC로 바꾸기 위한 비율입니다
     # Adj OHLC는 과거의 절대가격을 현재 가격의 시점으로 보기위한 수정된 가격입니다
     # 과거 발생한 액면분할과 현금배당을 반영한 Adj Close를 기준으로 조정합니다
     ratio = df['Adj Close']/df['Close']
     ratio
[4]: 0
            0.231324
     1
            0.231324
     2
            0.231324
     3
            0.231324
            0.231324
               . . .
     5181
            1.000000
     5182
            1.000000
     5183
            1.000000
     5184
            1.000000
     5185
            1.000000
    Length: 5186, dtype: float64
[5]: df['Adj Open'] = df['Open']*ratio
     df['Adj High'] = df['High']*ratio
     df['Adj Low'] = df['Low']*ratio
[6]: df.drop(['Open', 'High', 'Low', 'Close'], axis=1, inplace=True)
[7]: df.rename(columns={'Adj Open':'Open', 'Adj High':'High', 'Adj Low':'Low', 'Adj
      →Close':'Close'}, inplace=True)
[8]: df = df[['Open', 'High', 'Low', 'Close', 'Volume']]
     df
[8]:
                                                           Close Volume
                   Open
                                High
                                               Low
                                                                   56800
     0
            1457.339721
                         1642.398734
                                       1457.339721
                                                     1619.266357
     1
            1549.868774
                         1642.398253
                                       1529.049641
                                                     1549.868774
                                                                   52100
     2
            1619.266493
                         1619.266493
                                       1445.773655
                                                     1457.339844
                                                                   64900
            1468.905874
                         1526.736814
                                       1457.339686
                                                     1473.532349
                                                                   61800
           1529.049486
                         1549.868617
                                       1457.339147
                                                     1503.603882
                                                                   56100
                                                                     . . .
     5181
           9200.000000
                         9480.000000
                                       9100.000000
                                                     9380.000000
                                                                   81355
           9460.000000
                         9810.000000
                                       9460.000000
                                                     9700.000000
     5182
                                                                  105755
     5183
           9850.000000
                        10200.000000
                                       9800.000000
                                                     9990.000000
                                                                  170966
     5184
          10100.000000
                        10200.000000
                                       9940.000000
                                                    10150.000000
                                                                  133504
          10200.000000
                        10800.000000
                                      10150.000000
     5185
                                                    10650.000000
                                                                  247224
     [5186 rows x 5 columns]
```

```
[9]: # MinMaxScaler 클래스를 사용하여 데이터를 스케일링 합니다.
     # MinMaxScaler는 데이터의 최대값이 1, 최소값이 0이 되도록 변환합니다
     scaler = MinMaxScaler()
     scale_cols = ['Open', 'High', 'Low', 'Close', 'Volume']
     df_scaled = scaler.fit_transform(df[scale_cols])
     # 정규화가 완료된 데이터들은 pandas dataframe으로 변환합니다
     # pandas는 시계열 자료에 대한 다양한 기능을 제공하여 LSTM에서 사용하는 window를 만들
     때 유용합니다
     df_scaled = pd.DataFrame(df_scaled)
     df scaled.columns = scale cols
     print(df_scaled)
             Open
                      High
                               Low
                                      Close
                                              Volume
         0.069093 0.078420 0.072692 0.082280 0.020301
    0
    1
         0.076891 0.078420 0.078924 0.076473 0.018620
         0.082740 0.076587 0.071686 0.068730 0.023197
    2
    3
         0.070068 0.069256 0.072692 0.070085 0.022088
    4
         0.075136 0.071089 0.072692 0.072601 0.020050
    5181 0.721622 0.699387 0.736878 0.731697 0.029080
    5182 0.743534 0.725532 0.768164 0.758474 0.037804
    5183 0.776402 0.756432 0.797711 0.782742 0.061119
    5184 0.797472 0.756432 0.809878 0.796130 0.047725
    [5186 rows x 5 columns]
[10]: # window는 LSTM을 훈련하기 위한 단위로 고정된 사이즈를 가집니다
     # window가 12개라면 과거 시간 데이터 12개를 사용해서 다음 시간 단위의 값의 예측하게 됩
     LICH
     # 테스트 기간은 21일, 따라서 5299-21 : train / 21 : test
     window size = 20
     TEST SIZE = 1037
[11]: train = df_scaled[:-TEST_SIZE]
     test = df_scaled[-TEST_SIZE:]
[12]: test.describe()
[12]:
                 Open
                            High
                                        Low
                                                  Close
                                                             Volume
     count 1037.000000 1037.000000 1037.000000 1037.000000 1037.000000
             0.555879
                         0.526694
                                    0.567086
                                               0.551764
                                                           0.012608
     mean
     std
             0.079533
                         0.076859
                                    0.080378
                                               0.079329
                                                           0.036624
```

```
0.319620
                       0.315917
                                    0.326686
                                                 0.314971
                                                              0.000169
min
25%
          0.502964
                       0.475662
                                    0.513952
                                                 0.498957
                                                              0.002769
50%
          0.539574
                       0.509858
                                    0.550341
                                                 0.535049
                                                              0.005645
75%
          0.568782
                       0.538644
                                    0.580720
                                                 0.565572
                                                              0.012352
          0.831182
                       0.803969
                                    0.828128
                                                 0.837970
                                                              1.000000
max
def make_dataset(data, label, window_size=20):
```

```
[13]: # 정해진 window_size에 기반하여 20일 기간의 데이터 셋을 묶어준다

def make_dataset(data, label, window_size=20):
    feature_list = []
    label_list = []
    for i in range(len(data) - window_size):
        feature_list.append(np.array(data.iloc[i:i+window_size]))
        label_list.append(np.array(label.iloc[i+window_size]))
    return np.array(feature_list), np.array(label_list)
```

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[14]: from sklearn.model_selection import train_test_split
     feature_cols = ['Open', 'High', 'Low', 'Volume']
     label_cols = ['Close']
     train_feature = train[feature_cols]
     train_label = train[label_cols]
     # train dataset
     train_feature, train_label = make_dataset(train_feature, train_label, 20)
     # train set : 모델을 학습하는 유일한 dataset
     # validation set : 학습이 이미 완료된 모델을 검증하기 위한 dataset(비율 0.2)
     # validation_split : X_train과 y_train에서 일정 비율을 분리하여 검증데이터로 사용
     # 훈련 자체에는 반영되지 않고 훈련 과정을 지켜보기 위한 용도로 사용된다
     # train_test_split 함수는 전체 데이터 셋 배열을 받아서 랜덤하게 훈련/테스트 데이터 셋
     으로 분리해주는 함수입니다
     x_train, x_valid, y_train, y_valid = train_test_split(train_feature,_

→train_label, test_size=0.25)
     x_train.shape, x_valid.shape
     # ((4206, 20, 4), (1052, 20, 4))
     # test dataset : 학습과 검증이 완료된 모델의 성능을 평가하기 위한 dataset
     x_test = test[feature_cols]
     y_test = test[label_cols]
     x_test.shape, y_test.shape
                           (21, 1)
          (21, 4),
```

```
[14]: ((1037, 4), (1037, 1))
[15]: x_test, y_test = make_dataset(x_test, y_test, 20)
     x_test.shape, y_test.shape
     # (21-20, 20, 4),
                        (21-20, 1)
[15]: ((1017, 20, 4), (1017, 1))
[16]: x_train.shape, x_valid.shape, y_train.shape, y_valid.shape, x_test.shape, y_test.
      ⇔shape
[16]: ((3096, 20, 4), (1033, 20, 4), (3096, 1), (1033, 1), (1017, 20, 4), (1017, 1))
[17]: # print proportions
     print('train: {}% | validation: {}% | test {}%'.format(round(len(y_train)/
      \rightarrowlen(df_scaled),2),
                                                        round(len(y_valid)/
      \rightarrowlen(df_scaled),2),
                                                        round(len(y_test)/
      \rightarrowlen(df_scaled),2)))
     train: 0.6% | validation: 0.2% | test 0.2%
             데이터 비율을 지정하는 방법에 대한 규칙은 없다. 다만 모델에 제공하는 데이터
             를 제한하면 학습할 수 있는 내용이 제한된다. 그러나 테스트 세트가 너무 작으면
             모델 성능에 대한 정확한 추정치를 제공하지 않는다. 교차 검증을 통해 이 상황을
             쉽게 처리할 수 있다
[18]: from keras.models import Sequential
     from keras.layers import Dense
     from keras.callbacks import EarlyStopping, ModelCheckpoint
     from keras.layers import LSTM
     # LSTM은 RNN 알고리즘의 특별한 한 종류입니다.
     # LSTM은 긴 의존기간을 필요로 하는 데이터를 학습하는데 효과적인 모델입니다
     # 이 모델은 add함수를 사용하여 레이어들을 선형으로 쌓는 Sequential Model 입니다
     # 16 메모리 셀을 가진 LSTM 레이어 하나와 Dense 레이어 하나(output)을 사용합니다
     # input_shape는 input이 어떤 모양으로 들어올지에 대한 정보입니다. 데이터 개수는 중요하
     지 않기에 window_size와 feature만 알려주면 된다
     \# train_feature.shape[1] = window_size / train_feature.shape[2] = ['Open', _ \dots ]
      \rightarrow 'High', 'Low', 'Volume']
     # 예측하고자 하는 target 개수가 하나이므로 Dense(1)dl 출력으로 사용됩니다
     model = Sequential()
     model.add(LSTM(20,
                   input_shape=(x_test.shape[1], x_test.shape[2]),
                   activation='relu',
```

```
return_sequences=False)
            )
    model.add(Dense(1))
[19]: model.summary()
    Model: "sequential"
    _____
                          Output Shape
    Layer (type)
    ______
    1stm (LSTM)
                          (None, 20)
                                              2000
    dense (Dense)
                          (None, 1)
                                              21
    ______
    Total params: 2,021
    Trainable params: 2,021
    Non-trainable params: 0
[20]: x_test.shape
[20]: (1017, 20, 4)
[21]: # val_loss가 10회 같을 시 early_stop, batch_size(=K)는 K문제 풀고 답보고 하는 식
    # 위에서 모델을 구성한 후 compile 메서드를 호출하여 학습과정을 설정합니다
    # optimizer : 훈련 과정을 설정하다
    # loss : 최적화 과정에서 최소화될 손실 함수(loss function)을 설정합니다
    # metrics : 훈련을 모니터링하기 위해 사용됩니다
    # validation_data = 검증 데이터를 사용합니다. 각 에포크마다 정확도도 함께 출력됩니다
    # 이 정확도는 훈련이 잘 되고 있는지를 보여줄 뿐이며 실제로 모델이 검증데이터를 학습하지
    는 않습니다
    # 검증 데이터의 loss가 낮아지다가 높아지기 시작하면 overfitting의 신호입니다
    # verbose / 0 : 출력 없음 / 1 : 훈련 진행도 보여주는 진행 막대 보여줌 / 2 : 미니 배치
    마다 손실 정보 출력
    from numpy import array
    from keras.models import Sequential
    from keras.layers import Dense
    from keras import backend as K
    def RMSE(y_true, y_pred):
       return K.sqrt(K.mean(K.square(y_pred - y_true)))
    def soft_acc(y_true, y_pred):
       return K.mean(K.equal(K.round(y_true), K.round(y_pred)))
    def MPE(y_true, y_pred):
       return K.mean((y_true - y_pred) / y_true) * 100
```

```
def MSLE(y_true, y_pred):
    return K.mean(K.square(K.log(y_true+1) - K.log(y_pred+1)), axis=-1)
def RMSLE(y_true, y_pred):
    return K.sqrt(K.mean(K.square(K.log(y_true+1) - K.log(y_pred+1)), axis=-1))
def R2(y_true, y_pred):
    SS_res = K.sum(K.square(y_true - y_pred))
    SS_tot = K.sum(K.square(y_true - K.mean(y_true)))
    return ( 1 - SS_res/(SS_tot + K.epsilon()))
model.compile(loss = RMSE, optimizer='adam', metrics=[soft_acc, 'mse', 'mae', _
 →RMSE, 'mape', MPE, MSLE, RMSLE, R2])
early_stop = EarlyStopping(monitor='val_loss', patience=10)
filename = os.path.join('tmp', 'ckeckpointer.ckpt')
checkpoint = ModelCheckpoint(filename, monitor='val_loss', verbose=1,_
 ⇒save_best_only=True, mode='auto')
history = model.fit(x_train, y_train,
                  epochs=200,
                  batch_size=128,
                  validation_data=(x_valid, y_valid),
                  callbacks=[early_stop, checkpoint])
# score_test를 만들면 테스트가 더이상 테스트가 아니고, 처음부터 모든 데이터에 대해 학습
한 것과 같기 때문에 일반화 할 수 없는 모델을 만드는 것과 같다.
score_train = model.evaluate(x_train, y_train, batch_size=128)
score_validation = model.evaluate(x_valid, y_valid, batch_size=128)
Epoch 1/200
0.8765 - mse: 0.1513 - mae: 0.3176 - RMSE: 0.3878 - mape: 1869.4305 - MPE: -inf
- MSLE: 0.1038 - RMSLE: 0.2679 - R2: -2.7005 - val_loss: 0.3190 - val_soft_acc:
0.8687 - val_mse: 0.1020 - val_mae: 0.2512 - val_RMSE: 0.3133 - val_mape:
85.8997 - val_MPE: 38.0505 - val_MSLE: 0.0626 - val_RMSLE: 0.2014 - val_R2:
-1.2675
Epoch 00001: val_loss improved from inf to 0.31901, saving model to
tmp\ckeckpointer.ckpt
{\tt INFO: tensorflow: Assets \ written \ to: \ tmp\ckeckpointer.ckpt\assets}
Epoch 2/200
0.8759 - mse: 0.0778 - mae: 0.2219 - RMSE: 0.2761 - mape: 26597.1178 - MPE: -inf
```

```
- MSLE: 0.0469 - RMSLE: 0.1762 - R2: -0.8739 - val_loss: 0.0954 - val_soft_acc:
0.9269 - val_mse: 0.0091 - val_mae: 0.0882 - val_RMSE: 0.0959 - val_mape:
143.6386 - val_MPE: -110.3536 - val_MSLE: 0.0055 - val_RMSLE: 0.0691 - val_R2:
0.7838
Epoch 00002: val_loss improved from 0.31901 to 0.09544, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 3/200
0.9552 - mse: 0.0070 - mae: 0.0652 - RMSE: 0.0829 - mape: 16373.9791 - MPE: -inf
- MSLE: 0.0042 - RMSLE: 0.0518 - R2: 0.8344 - val_loss: 0.0638 - val_soft_acc:
0.9651 - val_mse: 0.0041 - val_mae: 0.0460 - val_RMSE: 0.0639 - val_mape:
117.1206 - val_MPE: -107.0139 - val_MSLE: 0.0025 - val_RMSLE: 0.0370 - val_R2:
0.9043
Epoch 00003: val_loss improved from 0.09544 to 0.06384, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 4/200
0.9657 - mse: 0.0028 - mae: 0.0407 - RMSE: 0.0522 - mape: 7257.6892 - MPE: -inf
- MSLE: 0.0017 - RMSLE: 0.0323 - R2: 0.9312 - val_loss: 0.0358 - val_soft_acc:
0.9668 - val_mse: 0.0013 - val_mae: 0.0208 - val_RMSE: 0.0363 - val_mape:
33.0372 - val_MPE: -26.7043 - val_MSLE: 5.3926e-04 - val_RMSLE: 0.0152 - val_R2:
0.9682
Epoch 00004: val_loss improved from 0.06384 to 0.03581, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 5/200
0.9705 - mse: 7.8456e-04 - mae: 0.0165 - RMSE: 0.0275 - mape: 2546.9461 - MPE:
-inf - MSLE: 3.4091e-04 - RMSLE: 0.0120 - R2: 0.9813 - val_loss: 0.0278 -
val_soft_acc: 0.9555 - val_mse: 7.9019e-04 - val_mae: 0.0176 - val_RMSE: 0.0292
- val_mape: 10.9165 - val_MPE: 3.8695 - val_MSLE: 3.4885e-04 - val_RMSLE: 0.0127
- val_R2: 0.9786
Epoch 00005: val_loss improved from 0.03581 to 0.02781, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 6/200
0.9732 - mse: 6.1030e-04 - mae: 0.0152 - RMSE: 0.0243 - mape: 6330.5447 - MPE:
-inf - MSLE: 2.7210e-04 - RMSLE: 0.0110 - R2: 0.9858 - val_loss: 0.0234 -
val_soft_acc: 0.9694 - val_mse: 5.5810e-04 - val_mae: 0.0132 - val_RMSE: 0.0248
- val_mape: 10.3574 - val_MPE: -3.9734 - val_MSLE: 2.3525e-04 - val_RMSLE:
```

0.0094 - val_R2: 0.9845

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Epoch 00006: val_loss improved from 0.02781 to 0.02339, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 7/200
0.9788 - mse: 4.5956e-04 - mae: 0.0123 - RMSE: 0.0210 - mape: 1740.6762 - MPE:
-inf - MSLE: 2.1007e-04 - RMSLE: 0.0089 - R2: 0.9891 - val_loss: 0.0231 -
val_soft_acc: 0.9712 - val_mse: 5.4654e-04 - val_mae: 0.0131 - val_RMSE: 0.0244
- val_mape: 10.3307 - val_MPE: -5.1015 - val_MSLE: 2.2895e-04 - val_RMSLE:
0.0093 - val_R2: 0.9851
Epoch 00007: val_loss improved from 0.02339 to 0.02314, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 8/200
0.9790 - mse: 4.7408e-04 - mae: 0.0121 - RMSE: 0.0214 - mape: 333.6890 - MPE:
-inf - MSLE: 2.0479e-04 - RMSLE: 0.0087 - R2: 0.9889 - val_loss: 0.0221 -
val_soft_acc: 0.9720 - val_mse: 4.9916e-04 - val_mae: 0.0123 - val_RMSE: 0.0233
- val_mape: 10.0658 - val_MPE: -5.0937 - val_MSLE: 2.0983e-04 - val_RMSLE:
0.0087 - val_R2: 0.9862
Epoch 00008: val_loss improved from 0.02314 to 0.02208, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 9/200
0.9738 - mse: 4.1408e-04 - mae: 0.0122 - RMSE: 0.0200 - mape: 12253.4055 - MPE:
-inf - MSLE: 1.8608e-04 - RMSLE: 0.0088 - R2: 0.9907 - val_loss: 0.0214 -
val_soft_acc: 0.9729 - val_mse: 4.6988e-04 - val_mae: 0.0116 - val_RMSE: 0.0226
- val_mape: 7.4660 - val_MPE: -2.4444 - val_MSLE: 1.9435e-04 - val_RMSLE: 0.0082
- val R2: 0.9871
Epoch 00009: val_loss improved from 0.02208 to 0.02138, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 10/200
0.9830 - mse: 4.1242e-04 - mae: 0.0112 - RMSE: 0.0199 - mape: 7.1733 - MPE:
-1.1896 - MSLE: 1.8140e-04 - RMSLE: 0.0081 - R2: 0.989 - Os 8ms/step - loss:
0.0197 - soft_acc: 0.9823 - mse: 4.0163e-04 - mae: 0.0111 - RMSE: 0.0196 - mape:
293.6921 - MPE: -inf - MSLE: 1.7680e-04 - RMSLE: 0.0080 - R2: 0.9903 - val_loss:
0.0217 - val_soft_acc: 0.9703 - val_mse: 4.8178e-04 - val_mae: 0.0124 -
val_RMSE: 0.0229 - val_mape: 7.9044 - val_MPE: -1.5238 - val_MSLE: 2.0242e-04 -
val_RMSLE: 0.0088 - val_R2: 0.9866
```

Epoch 00010: val_loss did not improve from 0.02138

```
Epoch 11/200
0.9789 - mse: 3.6545e-04 - mae: 0.0110 - RMSE: 0.0188 - mape: 1170.1042 - MPE:
-inf - MSLE: 1.6173e-04 - RMSLE: 0.0079 - R2: 0.9915 - val_loss: 0.0205 -
val_soft_acc: 0.9729 - val_mse: 4.3492e-04 - val_mae: 0.0110 - val_RMSE: 0.0217
- val_mape: 6.9849 - val_MPE: -1.8560 - val_MSLE: 1.7883e-04 - val_RMSLE: 0.0077
- val_R2: 0.9880
Epoch 00011: val_loss improved from 0.02138 to 0.02054, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 12/200
0.9783 - mse: 3.6398e-04 - mae: 0.0103 - RMSE: 0.0187 - mape: 780.7187 - MPE:
-inf - MSLE: 1.5757e-04 - RMSLE: 0.0073 - R2: 0.9914 - val_loss: 0.0203 -
val_soft_acc: 0.9720 - val_mse: 4.2666e-04 - val_mae: 0.0108 - val_RMSE: 0.0215
- val_mape: 6.6287 - val_MPE: -1.2335 - val_MSLE: 1.7467e-04 - val_RMSLE: 0.0076
- val_R2: 0.9883
Epoch 00012: val_loss improved from 0.02054 to 0.02032, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 13/200
0.9835 - mse: 3.5768e-04 - mae: 0.0101 - RMSE: 0.0187 - mape: 48.1493 - MPE:
-inf - MSLE: 1.5274e-04 - RMSLE: 0.0072 - R2: 0.9914 - val_loss: 0.0226 -
val_soft_acc: 0.9616 - val_mse: 5.2460e-04 - val_mae: 0.0138 - val_RMSE: 0.0234
- val_mape: 10.9509 - val_MPE: -8.8762 - val_MSLE: 2.1896e-04 - val_RMSLE:
0.0098 - val_R2: 0.9865
Epoch 00013: val_loss did not improve from 0.02032
Epoch 14/200
0.9802 - mse: 4.3468e-04 - mae: 0.0120 - RMSE: 0.0203 - mape: 974.8507 - MPE:
-inf - MSLE: 1.8545e-04 - RMSLE: 0.0086 - R2: 0.9901 - val_loss: 0.0200 -
val_soft_acc: 0.9712 - val_mse: 4.1229e-04 - val_mae: 0.0109 - val_RMSE: 0.0212
- val_mape: 6.3493 - val_MPE: 0.1461 - val_MSLE: 1.7014e-04 - val_RMSLE: 0.0077
- val_R2: 0.9885
Epoch 00014: val_loss improved from 0.02032 to 0.01998, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
0.9808 - mse: 3.8097e-04 - mae: 0.0105 - RMSE: 0.0192 - mape: 375.2020 - MPE:
-inf - MSLE: 1.6094e-04 - RMSLE: 0.0075 - R2: 0.9913 - val_loss: 0.0198 -
val_soft_acc: 0.9720 - val_mse: 4.0708e-04 - val_mae: 0.0113 - val_RMSE: 0.0209
- val_mape: 6.3737 - val_MPE: 1.6391 - val_MSLE: 1.6979e-04 - val_RMSLE: 0.0080
```

```
Epoch 00015: val_loss improved from 0.01998 to 0.01985, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 16/200
0.9828 - mse: 3.1733e-04 - mae: 0.0099 - RMSE: 0.0175 - mape: 110.8316 - MPE:
-inf - MSLE: 1.4014e-04 - RMSLE: 0.0071 - R2: 0.9923 - val_loss: 0.0194 -
val_soft_acc: 0.9703 - val_mse: 3.9178e-04 - val_mae: 0.0106 - val_RMSE: 0.0204
- val_mape: 6.5014 - val_MPE: 2.6541 - val_MSLE: 1.6205e-04 - val_RMSLE: 0.0076
- val_R2: 0.9895
Epoch 00016: val_loss improved from 0.01985 to 0.01945, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 17/200
0.9826 - mse: 3.5554e-04 - mae: 0.0104 - RMSE: 0.0186 - mape: 690.9058 - MPE:
-inf - MSLE: 1.5286e-04 - RMSLE: 0.0075 - R2: 0.9918 - val_loss: 0.0190 -
val_soft_acc: 0.9703 - val_mse: 3.7433e-04 - val_mae: 0.0102 - val_RMSE: 0.0200
- val_mape: 5.9632 - val_MPE: 0.6051 - val_MSLE: 1.5380e-04 - val_RMSLE: 0.0072
- val_R2: 0.9900
Epoch 00017: val_loss improved from 0.01945 to 0.01901, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 18/200
0.9855 - mse: 3.2335e-04 - mae: 0.0097 - RMSE: 0.0174 - mape: 79.2738 - MPE: inf
- MSLE: 1.4420e-04 - RMSLE: 0.0070 - R2: 0.9926 - val_loss: 0.0188 -
val_soft_acc: 0.9729 - val_mse: 3.6781e-04 - val_mae: 0.0101 - val_RMSE: 0.0198
- val_mape: 6.3844 - val_MPE: -1.9029 - val_MSLE: 1.5069e-04 - val_RMSLE: 0.0071
- val_R2: 0.9901
Epoch 00018: val_loss improved from 0.01901 to 0.01884, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 19/200
0.9841 - mse: 3.2168e-04 - mae: 0.0099 - RMSE: 0.0173 - mape: 6298.2121 - MPE:
-inf - MSLE: 1.3933e-04 - RMSLE: 0.0072 - R2: 0.9928 - val_loss: 0.0184 -
val_soft_acc: 0.9720 - val_mse: 3.5183e-04 - val_mae: 0.0102 - val_RMSE: 0.0194
- val_mape: 8.1452 - val_MPE: -4.7442 - val_MSLE: 1.4613e-04 - val_RMSLE: 0.0073
```

- val R2: 0.9889

- val_R2: 0.9904

Epoch 00019: val_loss improved from 0.01884 to 0.01842, saving model to tmp\ckeckpointer.ckpt

```
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 20/200
0.9810 - mse: 3.0739e-04 - mae: 0.0096 - RMSE: 0.0172 - mape: 2382.2863 - MPE:
-inf - MSLE: 1.3669e-04 - RMSLE: 0.0069 - R2: 0.9925 - val_loss: 0.0189 -
val_soft_acc: 0.9712 - val_mse: 3.6875e-04 - val_mae: 0.0105 - val_RMSE: 0.0197
- val_mape: 7.5009 - val_MPE: -4.6798 - val_MSLE: 1.5221e-04 - val_RMSLE: 0.0074
- val_R2: 0.9903
Epoch 00020: val_loss did not improve from 0.01842
Epoch 21/200
0.9805 - mse: 2.9841e-04 - mae: 0.0100 - RMSE: 0.0170 - mape: 444.5656 - MPE:
-inf - MSLE: 1.3373e-04 - RMSLE: 0.0072 - R2: 0.9928 - val_loss: 0.0192 -
val_soft_acc: 0.9712 - val_mse: 3.8082e-04 - val_mae: 0.0107 - val_RMSE: 0.0200
- val_mape: 7.1463 - val_MPE: -4.2206 - val_MSLE: 1.5638e-04 - val_RMSLE: 0.0075
- val_R2: 0.9901
Epoch 00021: val_loss did not improve from 0.01842
Epoch 22/200
0.9823 - mse: 3.1395e-04 - mae: 0.0100 - RMSE: 0.0173 - mape: 7.5945 - MPE: -inf
- MSLE: 1.3748e-04 - RMSLE: 0.0072 - R2: 0.9925 - val_loss: 0.0180 -
val_soft_acc: 0.9720 - val_mse: 3.3642e-04 - val_mae: 0.0100 - val_RMSE: 0.0190
- val_mape: 5.9931 - val_MPE: 2.0080 - val_MSLE: 1.4036e-04 - val_RMSLE: 0.0071
- val_R2: 0.9908
Epoch 00022: val_loss improved from 0.01842 to 0.01803, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 23/200
0.9843 - mse: 2.8995e-04 - mae: 0.0095 - RMSE: 0.0167 - mape: 3838.1649 - MPE:
-inf - MSLE: 1.2549e-04 - RMSLE: 0.0068 - R2: 0.9931 - val_loss: 0.0179 -
val_soft_acc: 0.9720 - val_mse: 3.3162e-04 - val_mae: 0.0100 - val_RMSE: 0.0189
- val_mape: 5.8429 - val_MPE: 0.8401 - val_MSLE: 1.3806e-04 - val_RMSLE: 0.0071
- val_R2: 0.9909
Epoch 00023: val_loss improved from 0.01803 to 0.01790, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 24/200
0.9820 - mse: 2.8415e-04 - mae: 0.0094 - RMSE: 0.0165 - mape: 1147.4901 - MPE:
-inf - MSLE: 1.2470e-04 - RMSLE: 0.0067 - R2: 0.9933 - val_loss: 0.0179 -
val_soft_acc: 0.9703 - val_mse: 3.3238e-04 - val_mae: 0.0097 - val_RMSE: 0.0188
- val_mape: 5.8527 - val_MPE: 1.1157 - val_MSLE: 1.3738e-04 - val_RMSLE: 0.0069
- val_R2: 0.9911
```

```
Epoch 00024: val_loss did not improve from 0.01790
Epoch 25/200
0.9788 - mse: 3.3881e-04 - mae: 0.0110 - RMSE: 0.0182 - mape: 1339.5308 - MPE:
-inf - MSLE: 1.5040e-04 - RMSLE: 0.0080 - R2: 0.9918 - val_loss: 0.0178 -
val_soft_acc: 0.9703 - val_mse: 3.2603e-04 - val_mae: 0.0097 - val_RMSE: 0.0186
- val_mape: 5.8893 - val_MPE: 1.4070 - val_MSLE: 1.3537e-04 - val_RMSLE: 0.0069
- val_R2: 0.9913
Epoch 00025: val_loss improved from 0.01790 to 0.01777, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 26/200
0.9786 - mse: 3.0087e-04 - mae: 0.0098 - RMSE: 0.0169 - mape: 2374.9665 - MPE:
-inf - MSLE: 1.3281e-04 - RMSLE: 0.0071 - R2: 0.9932 - val_loss: 0.0177 -
val_soft_acc: 0.9703 - val_mse: 3.2450e-04 - val_mae: 0.0096 - val_RMSE: 0.0185
- val_mape: 6.2016 - val_MPE: -2.5596 - val_MSLE: 1.3371e-04 - val_RMSLE: 0.0067
- val_R2: 0.9914
Epoch 00026: val_loss improved from 0.01777 to 0.01770, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 27/200
0.9835 - mse: 2.6730e-04 - mae: 0.0086 - RMSE: 0.0159 - mape: 1857.4715 - MPE:
-inf - MSLE: 1.1299e-04 - RMSLE: 0.0062 - R2: 0.9933 - val_loss: 0.0172 -
val_soft_acc: 0.9738 - val_mse: 3.0634e-04 - val_mae: 0.0093 - val_RMSE: 0.0181
- val_mape: 5.9751 - val_MPE: -1.3736 - val_MSLE: 1.2709e-04 - val_RMSLE: 0.0066
- val_R2: 0.9917
Epoch 00027: val_loss improved from 0.01770 to 0.01721, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 28/200
0.9868 - mse: 2.8310e-04 - mae: 0.0089 - RMSE: 0.0165 - mape: 2446.6873 - MPE:
-inf - MSLE: 1.2080e-04 - RMSLE: 0.0064 - R2: 0.9933 - val_loss: 0.0175 -
val_soft_acc: 0.9712 - val_mse: 3.1682e-04 - val_mae: 0.0099 - val_RMSE: 0.0183
- val_mape: 8.6990 - val_MPE: -6.4365 - val_MSLE: 1.3339e-04 - val_RMSLE: 0.0071
- val_R2: 0.9916
Epoch 00028: val_loss did not improve from 0.01721
Epoch 29/200
0.9846 - mse: 2.9124e-04 - mae: 0.0091 - RMSE: 0.0169 - mape: 1831.8570 - MPE:
-inf - MSLE: 1.2642e-04 - RMSLE: 0.0066 - R2: 0.9932 - val_loss: 0.0171 -
```

```
val_soft_acc: 0.9729 - val_mse: 3.0252e-04 - val_mae: 0.0098 - val_RMSE: 0.0180
- val_mape: 5.7162 - val_MPE: 1.2424 - val_MSLE: 1.2721e-04 - val_RMSLE: 0.0069
- val_R2: 0.9918
Epoch 00029: val_loss improved from 0.01721 to 0.01712, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 30/200
0.9803 - mse: 2.6481e-04 - mae: 0.0090 - RMSE: 0.0159 - mape: 916.9306 - MPE:
-inf - MSLE: 1.1269e-04 - RMSLE: 0.0064 - R2: 0.9940 - val_loss: 0.0171 -
val_soft_acc: 0.9729 - val_mse: 3.0074e-04 - val_mae: 0.0098 - val_RMSE: 0.0180
- val_mape: 5.7832 - val_MPE: 1.7475 - val_MSLE: 1.2723e-04 - val_RMSLE: 0.0070
- val R2: 0.9918
Epoch 00030: val_loss improved from 0.01712 to 0.01708, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 31/200
0.9802 - mse: 2.9786e-04 - mae: 0.0098 - RMSE: 0.0169 - mape: 357.1914 - MPE:
-inf - MSLE: 1.2675e-04 - RMSLE: 0.0070 - R2: 0.9931 - val_loss: 0.0179 -
val_soft_acc: 0.9712 - val_mse: 3.2770e-04 - val_mae: 0.0114 - val_RMSE: 0.0188
- val_mape: 7.3403 - val_MPE: 5.4104 - val_MSLE: 1.4370e-04 - val_RMSLE: 0.0082
- val_R2: 0.9911
Epoch 00031: val_loss did not improve from 0.01708
Epoch 32/200
0.9835 - mse: 2.9948e-04 - mae: 0.0098 - RMSE: 0.0169 - mape: 2251.9771 - MPE:
-inf - MSLE: 1.3007e-04 - RMSLE: 0.0071 - R2: 0.9931 - val_loss: 0.0165 -
val_soft_acc: 0.9746 - val_mse: 2.8149e-04 - val_mae: 0.0091 - val_RMSE: 0.0173
- val_mape: 7.3964 - val_MPE: -4.6022 - val_MSLE: 1.1812e-04 - val_RMSLE: 0.0065
- val_R2: 0.9925
Epoch 00032: val_loss improved from 0.01708 to 0.01651, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 33/200
0.9821 - mse: 2.4631e-04 - mae: 0.0090 - RMSE: 0.0153 - mape: 4213.2886 - MPE:
-inf - MSLE: 1.0888e-04 - RMSLE: 0.0065 - R2: 0.9943 - val_loss: 0.0164 -
val_soft_acc: 0.9729 - val_mse: 2.7815e-04 - val_mae: 0.0089 - val_RMSE: 0.0172
- val_mape: 5.5174 - val_MPE: -1.1442 - val_MSLE: 1.1596e-04 - val_RMSLE: 0.0063
- val_R2: 0.9926
Epoch 00033: val_loss improved from 0.01651 to 0.01643, saving model to
```

tmp\ckeckpointer.ckpt

```
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 34/200
0.9847 - mse: 2.7626e-04 - mae: 0.0088 - RMSE: 0.0161 - mape: 372.5754 - MPE:
-inf - MSLE: 1.1942e-04 - RMSLE: 0.0062 - R2: 0.9937 - val_loss: 0.0163 -
val_soft_acc: 0.9746 - val_mse: 2.7321e-04 - val_mae: 0.0091 - val_RMSE: 0.0171
- val_mape: 6.1135 - val_MPE: -1.9166 - val_MSLE: 1.1473e-04 - val_RMSLE: 0.0064
- val_R2: 0.9926
Epoch 00034: val_loss improved from 0.01643 to 0.01628, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 35/200
0.9839 - mse: 2.3452e-04 - mae: 0.0086 - RMSE: 0.0152 - mape: 6208.0151 - MPE:
-inf - MSLE: 1.0278e-04 - RMSLE: 0.0062 - R2: 0.9945 - val_loss: 0.0161 -
val_soft_acc: 0.9746 - val_mse: 2.6850e-04 - val_mae: 0.0088 - val_RMSE: 0.0170
- val_mape: 6.0768 - val_MPE: -2.6731 - val_MSLE: 1.1247e-04 - val_RMSLE: 0.0062
- val_R2: 0.9928
Epoch 00035: val_loss improved from 0.01628 to 0.01615, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 36/200
0.9842 - mse: 2.3840e-04 - mae: 0.0089 - RMSE: 0.0152 - mape: 437.0283 - MPE:
-inf - MSLE: 1.0652e-04 - RMSLE: 0.0064 - R2: 0.9941 - val_loss: 0.0162 -
val_soft_acc: 0.9712 - val_mse: 2.6981e-04 - val_mae: 0.0088 - val_RMSE: 0.0169
- val_mape: 5.4708 - val_MPE: -1.2429 - val_MSLE: 1.1285e-04 - val_RMSLE: 0.0062
- val_R2: 0.9928
Epoch 00036: val_loss did not improve from 0.01615
Epoch 37/200
0.9784 - mse: 2.5590e-04 - mae: 0.0089 - RMSE: 0.0157 - mape: 1690.8989 - MPE:
-inf - MSLE: 1.1299e-04 - RMSLE: 0.0064 - R2: 0.9939 - val_loss: 0.0163 -
val_soft_acc: 0.9738 - val_mse: 2.7187e-04 - val_mae: 0.0094 - val_RMSE: 0.0171
- val_mape: 5.4822 - val_MPE: 1.2683 - val_MSLE: 1.1566e-04 - val_RMSLE: 0.0067
- val_R2: 0.9926
Epoch 00037: val_loss did not improve from 0.01615
Epoch 38/200
0.9812 - mse: 2.5378e-04 - mae: 0.0089 - RMSE: 0.0157 - mape: 1160.1034 - MPE:
-inf - MSLE: 1.0897e-04 - RMSLE: 0.0064 - R2: 0.9939 - val_loss: 0.0160 -
val_soft_acc: 0.9764 - val_mse: 2.6231e-04 - val_mae: 0.0086 - val_RMSE: 0.0167
- val_mape: 5.2619 - val_MPE: 0.4669 - val_MSLE: 1.0971e-04 - val_RMSLE: 0.0061
- val_R2: 0.9930
```

```
Epoch 00038: val_loss improved from 0.01615 to 0.01596, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 39/200
0.9843 - mse: 2.4178e-04 - mae: 0.0084 - RMSE: 0.0152 - mape: 267.0457 - MPE:
-inf - MSLE: 1.0312e-04 - RMSLE: 0.0060 - R2: 0.9945 - val_loss: 0.0164 -
val_soft_acc: 0.9738 - val_mse: 2.7692e-04 - val_mae: 0.0090 - val_RMSE: 0.0171
- val_mape: 6.8158 - val_MPE: -4.2522 - val_MSLE: 1.1516e-04 - val_RMSLE: 0.0064
- val_R2: 0.9928
Epoch 00039: val_loss did not improve from 0.01596
Epoch 40/200
0.9825 - mse: 2.5872e-04 - mae: 0.0089 - RMSE: 0.0158 - mape: 211.0300 - MPE:
-inf - MSLE: 1.1244e-04 - RMSLE: 0.0064 - R2: 0.9939 - val_loss: 0.0171 -
val_soft_acc: 0.9703 - val_mse: 2.9813e-04 - val_mae: 0.0102 - val_RMSE: 0.0176
- val_mape: 8.0166 - val_MPE: -6.2196 - val_MSLE: 1.2719e-04 - val_RMSLE: 0.0073
- val_R2: 0.9924
Epoch 00040: val_loss did not improve from 0.01596
Epoch 41/200
0.9863 - mse: 2.5294e-04 - mae: 0.0088 - RMSE: 0.0156 - mape: 456.3876 - MPE:
-inf - MSLE: 1.0975e-04 - RMSLE: 0.0063 - R2: 0.9940 - val_loss: 0.0155 -
val_soft_acc: 0.9755 - val_mse: 2.4728e-04 - val_mae: 0.0085 - val_RMSE: 0.0163
- val_mape: 5.1752 - val_MPE: 0.4380 - val_MSLE: 1.0442e-04 - val_RMSLE: 0.0061
- val_R2: 0.9933
Epoch 00041: val_loss improved from 0.01596 to 0.01550, saving model to
tmp\ckeckpointer.ckpt
{\tt INFO: tensorflow: Assets \ written \ to: \ tmp\ckeckpointer.ckpt\assets}
Epoch 42/200
0.9886 - mse: 2.2062e-04 - mae: 0.0080 - RMSE: 0.0144 - mape: 3459.1965 - MPE:
-inf - MSLE: 9.8270e-05 - RMSLE: 0.0058 - R2: 0.9948 - val_loss: 0.0162 -
val_soft_acc: 0.9720 - val_mse: 2.6871e-04 - val_mae: 0.0090 - val_RMSE: 0.0168
- val_mape: 5.9219 - val_MPE: -3.0058 - val_MSLE: 1.1255e-04 - val_RMSLE: 0.0064
- val_R2: 0.9930
Epoch 00042: val_loss did not improve from 0.01550
0.9824 - mse: 2.7160e-04 - mae: 0.0089 - RMSE: 0.0159 - mape: 603.6849 - MPE:
-inf - MSLE: 1.1694e-04 - RMSLE: 0.0064 - R2: 0.9937 - val_loss: 0.0164 -
val_soft_acc: 0.9720 - val_mse: 2.7514e-04 - val_mae: 0.0088 - val_RMSE: 0.0169
- val_mape: 5.1839 - val_MPE: -0.6919 - val_MSLE: 1.1364e-04 - val_RMSLE: 0.0062
```

```
- val R2: 0.9929
Epoch 00043: val_loss did not improve from 0.01550
0.9809 - mse: 2.6917e-04 - mae: 0.0091 - RMSE: 0.0160 - mape: 1159.5668 - MPE:
-inf - MSLE: 1.1488e-04 - RMSLE: 0.0065 - R2: 0.9937 - val_loss: 0.0155 -
val_soft_acc: 0.9764 - val_mse: 2.4720e-04 - val_mae: 0.0083 - val_RMSE: 0.0163
- val_mape: 5.2653 - val_MPE: -1.2176 - val_MSLE: 1.0375e-04 - val_RMSLE: 0.0059
- val_R2: 0.9934
Epoch 00044: val_loss did not improve from 0.01550
Epoch 45/200
0.9845 - mse: 2.3497e-04 - mae: 0.0083 - RMSE: 0.0151 - mape: 294.7668 - MPE:
-inf - MSLE: 9.9587e-05 - RMSLE: 0.0059 - R2: 0.9946 - val_loss: 0.0155 -
val_soft_acc: 0.9755 - val_mse: 2.4735e-04 - val_mae: 0.0089 - val_RMSE: 0.0163
- val_mape: 8.3404 - val_MPE: -6.1452 - val_MSLE: 1.0654e-04 - val_RMSLE: 0.0064
- val_R2: 0.9934
Epoch 00045: val_loss did not improve from 0.01550
Epoch 46/200
0.9860 - mse: 2.1565e-04 - mae: 0.0082 - RMSE: 0.0144 - mape: 2779.3721 - MPE:
-inf - MSLE: 9.5685e-05 - RMSLE: 0.0060 - R2: 0.9948 - val_loss: 0.0154 -
val_soft_acc: 0.9764 - val_mse: 2.4420e-04 - val_mae: 0.0084 - val_RMSE: 0.0161
- val_mape: 5.2409 - val_MPE: -0.8130 - val_MSLE: 1.0221e-04 - val_RMSLE: 0.0059
- val_R2: 0.9935
Epoch 00046: val_loss improved from 0.01550 to 0.01542, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 47/200
0.9878 - mse: 2.4544e-04 - mae: 0.0083 - RMSE: 0.0152 - mape: 651.2839 - MPE:
-inf - MSLE: 1.0323e-04 - RMSLE: 0.0059 - R2: 0.9943 - val_loss: 0.0152 -
val_soft_acc: 0.9764 - val_mse: 2.3618e-04 - val_mae: 0.0084 - val_RMSE: 0.0159
- val_mape: 5.6613 - val_MPE: 2.6650 - val_MSLE: 1.0111e-04 - val_RMSLE: 0.0061
- val_R2: 0.9937
Epoch 00047: val_loss improved from 0.01542 to 0.01518, saving model to
tmp\ckeckpointer.ckpt
INFO:tensorflow:Assets written to: tmp\ckeckpointer.ckpt\assets
Epoch 48/200
0.9857 - mse: 2.4011e-04 - mae: 0.0083 - RMSE: 0.0152 - mape: 1562.8400 - MPE:
-inf - MSLE: 1.0407e-04 - RMSLE: 0.0060 - R2: 0.9943 - val_loss: 0.0152 -
val_soft_acc: 0.9729 - val_mse: 2.3743e-04 - val_mae: 0.0083 - val_RMSE: 0.0159
```

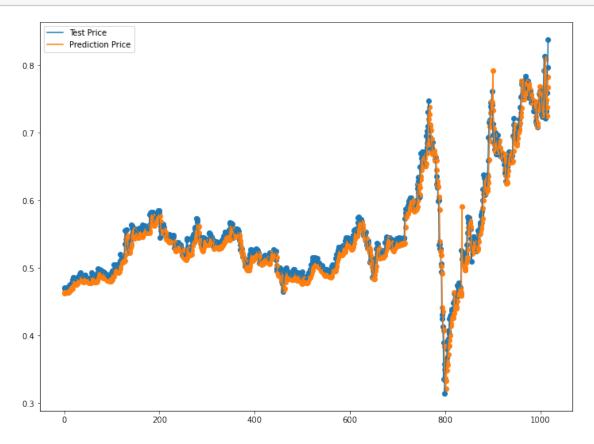
```
- val_mape: 5.2055 - val_MPE: 1.0778 - val_MSLE: 1.0079e-04 - val_RMSLE: 0.0059
- val_R2: 0.9937
Epoch 00048: val_loss did not improve from 0.01518
Epoch 49/200
0.9847 - mse: 2.6783e-04 - mae: 0.0090 - RMSE: 0.0162 - mape: 349.3843 - MPE:
-inf - MSLE: 1.1463e-04 - RMSLE: 0.0065 - R2: 0.9937 - val_loss: 0.0160 -
val_soft_acc: 0.9738 - val_mse: 2.6133e-04 - val_mae: 0.0101 - val_RMSE: 0.0168
- val_mape: 6.5535 - val_MPE: 4.6092 - val_MSLE: 1.1626e-04 - val_RMSLE: 0.0073
- val_R2: 0.9929
Epoch 00049: val_loss did not improve from 0.01518
Epoch 50/200
0.9814 - mse: 2.2277e-04 - mae: 0.0083 - RMSE: 0.0147 - mape: 2574.2157 - MPE:
-inf - MSLE: 9.6752e-05 - RMSLE: 0.0060 - R2: 0.9948 - val_loss: 0.0155 -
val_soft_acc: 0.9720 - val_mse: 2.4425e-04 - val_mae: 0.0086 - val_RMSE: 0.0161
- val_mape: 6.2979 - val_MPE: -3.6401 - val_MSLE: 1.0329e-04 - val_RMSLE: 0.0062
- val_R2: 0.9936
Epoch 00050: val_loss did not improve from 0.01518
Epoch 51/200
0.9823 - mse: 2.3845e-04 - mae: 0.0091 - RMSE: 0.0154 - mape: 127.7951 - MPE:
-inf - MSLE: 1.0409e-04 - RMSLE: 0.0065 - R2: 0.9945 - val_loss: 0.0152 -
val_soft_acc: 0.9764 - val_mse: 2.3685e-04 - val_mae: 0.0084 - val_RMSE: 0.0159
- val_mape: 5.4101 - val_MPE: 2.1620 - val_MSLE: 1.0086e-04 - val_RMSLE: 0.0060
- val_R2: 0.9937
Epoch 00051: val_loss did not improve from 0.01518
0.9864 - mse: 2.3084e-04 - mae: 0.0079 - RMSE: 0.0150 - mape: 1809.9790 - MPE:
-inf - MSLE: 9.6272e-05 - RMSLE: 0.0057 - R2: 0.9946 - val_loss: 0.0168 -
val_soft_acc: 0.9712 - val_mse: 2.8834e-04 - val_mae: 0.0114 - val_RMSE: 0.0176
- val_mape: 7.4638 - val_MPE: 5.9537 - val_MSLE: 1.3111e-04 - val_RMSLE: 0.0083
- val_R2: 0.9923
Epoch 00052: val_loss did not improve from 0.01518
Epoch 53/200
0.9819 - mse: 2.3990e-04 - mae: 0.0092 - RMSE: 0.0153 - mape: 2224.9102 - MPE:
-inf - MSLE: 1.0726e-04 - RMSLE: 0.0067 - R2: 0.9941 - val_loss: 0.0153 -
val_soft_acc: 0.9746 - val_mse: 2.4133e-04 - val_mae: 0.0090 - val_RMSE: 0.0161
- val_mape: 5.2285 - val_MPE: 1.6214 - val_MSLE: 1.0401e-04 - val_RMSLE: 0.0064
- val_R2: 0.9934
```

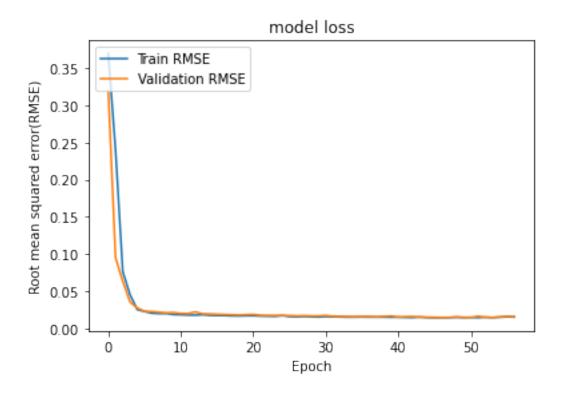
```
Epoch 54/200
    0.9833 - mse: 2.2280e-04 - mae: 0.0085 - RMSE: 0.0147 - mape: 42.4492 - MPE: inf
    - MSLE: 9.7055e-05 - RMSLE: 0.0061 - R2: 0.9947 - val_loss: 0.0152 -
    val_soft_acc: 0.9746 - val_mse: 2.3687e-04 - val_mae: 0.0088 - val_RMSE: 0.0159
    - val_mape: 8.8350 - val_MPE: -6.6440 - val_MSLE: 1.0291e-04 - val_RMSLE: 0.0064
    - val_R2: 0.9936
    Epoch 00054: val_loss did not improve from 0.01518
    Epoch 55/200
    0.9818 - mse: 2.4979e-04 - mae: 0.0093 - RMSE: 0.0157 - mape: 2354.6518 - MPE:
    -inf - MSLE: 1.1090e-04 - RMSLE: 0.0068 - R2: 0.9943 - val_loss: 0.0159 -
    val_soft_acc: 0.9720 - val_mse: 2.5746e-04 - val_mae: 0.0091 - val_RMSE: 0.0164
    - val_mape: 7.3763 - val_MPE: -5.4096 - val_MSLE: 1.0940e-04 - val_RMSLE: 0.0066
    - val_R2: 0.9934
    Epoch 00055: val_loss did not improve from 0.01518
    Epoch 56/200
    0.9796 - mse: 2.4692e-04 - mae: 0.0096 - RMSE: 0.0155 - mape: 404.0689 - MPE:
    -inf - MSLE: 1.1181e-04 - RMSLE: 0.0070 - R2: 0.9942 - val_loss: 0.0165 -
    val_soft_acc: 0.9703 - val_mse: 2.7844e-04 - val_mae: 0.0101 - val_RMSE: 0.0170
    - val_mape: 8.1351 - val_MPE: -6.4739 - val_MSLE: 1.1985e-04 - val_RMSLE: 0.0073
    - val_R2: 0.9930
    Epoch 00056: val_loss did not improve from 0.01518
    Epoch 57/200
    0.9778 - mse: 2.3164e-04 - mae: 0.0097 - RMSE: 0.0150 - mape: 190.1999 - MPE:
    -inf - MSLE: 1.0646e-04 - RMSLE: 0.0071 - R2: 0.9946 - val_loss: 0.0161 -
    val_soft_acc: 0.9746 - val_mse: 2.6542e-04 - val_mae: 0.0105 - val_RMSE: 0.0169
    - val_mape: 9.0060 - val_MPE: 7.3535 - val_MSLE: 1.2074e-04 - val_RMSLE: 0.0077
    - val_R2: 0.9929
    Epoch 00057: val_loss did not improve from 0.01518
    0.9859 - mse: 2.4505e-04 - mae: 0.0098 - RMSE: 0.0154 - mape: 802.3535 - MPE:
    inf - MSLE: 1.1239e-04 - RMSLE: 0.0073 - R2: 0.9943
    0.9746 - mse: 2.6542e-04 - mae: 0.0105 - RMSE: 0.0169 - mape: 9.0060 - MPE:
    7.3535 - MSLE: 1.2074e-04 - RMSLE: 0.0077 - R2: 0.9929
[22]: pred = model.predict(x_test)
    pred.shape
```

Epoch 00053: val_loss did not improve from 0.01518

```
[22]: (1017, 1)
```

```
[23]: pred = model.predict(x_test)
      pred.shape
      plt.figure(figsize=(12,9))
      plt.plot(np.asarray(y_test), label='Test Price')
      plt.plot(pred, label='Prediction Price')
      x_values = list(range(1017))
      plt.scatter(x_values, np.asarray(y_test))
      plt.scatter(x_values, pred)
      plt.legend()
      plt.show()
      plt.plot(history.history['loss'])
      plt.plot(history.history['val_loss'])
      plt.title('model loss')
      plt.ylabel('Root mean squared error(RMSE)')
      plt.xlabel('Epoch')
      plt.legend(['Train RMSE', 'Validation RMSE'], loc='upper left')
      plt.show()
```



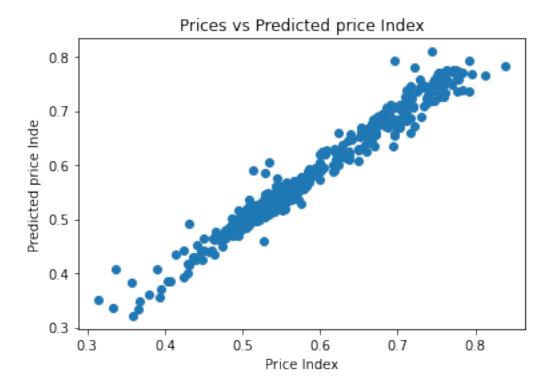


```
[24]: # 원래값과 예측 값이 일치하면 직선에 가깝게 분포가 된다

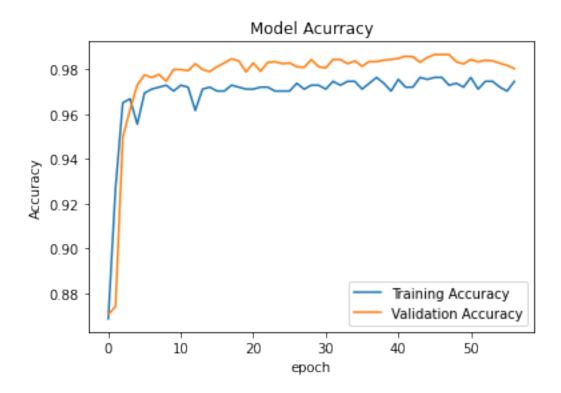
%matplotlib inline
import matplotlib.pyplot as plt

plt.scatter(np.asarray(y_test), pred)
plt.xlabel("Price Index")
plt.ylabel("Predicted price Inde")
plt.title("Prices vs Predicted price Index")
```

[24]: Text(0.5, 1.0, 'Prices vs Predicted price Index')



```
[25]: plt.plot(history.history['val_soft_acc'])
    plt.plot(history.history['soft_acc'])
    plt.title('Model Acurracy')
    plt.ylabel('Accuracy')
    plt.xlabel('epoch')
    plt.legend(['Training Accuracy', 'Validation Accuracy'], loc='lower right')
    plt.show()
```



```
[26]: accuracy_train = 100*score_train[1]
    accuracy_validation = 100*score_validation[1]

print('train accuracy: %.4f%%' % accuracy_train)
    print('validation accuracy: %.4f%%' % accuracy_validation)
```

train accuracy: 98.5937% validation accuracy: 97.4634%

R2_Score가 train data에서 계산되면 모델이 샘플 내 분산을 얼마나 설명하는지에 대해 알려주고 test data에서 계산되면 모델의 예측품질에 대해 알려준다. 기계학습 세계에서는 검증과 테스트 정확도를 모두 제시하는 것이 매우 일반적이지만가장 중요한 것은 테스트 정확도입니다. 그러나 한 쪽에서 낮은 R2 점수를 받고 그렇지 않다면 뭔가 꺼진 것입니다. R2 test « R2training이면 모델이 잘 일반화되지않았음을 나타낸다. 예를들어 테스트 세트에 future 데이타 만 포함되어 있으면모델이 잘 외삽되지 않는 것처럼 보입니다. 결론적으로 당신은 그것을 비교해야합니다. 그러나 대부분의 경우 가장 관심이 있는 테스트 세트 결과입니다

```
[27]: import numpy as np
  from sklearn.metrics import r2_score
  from sklearn.metrics import mean_squared_log_error

Y = np.asarray(y_test)
Y_hat = pred
```

```
def MSE(y_true, y_pred):
    return np.mean(np.square((y_true - y_pred)))
def MAE(y_true, y_pred):
    return np.mean(np.abs((y_true - y_pred)))
def RMSE(y_true, y_pred):
    return np.sqrt(np.mean((y_pred-y_true)**2))
def MAPE(y_true, y_pred):
    return np.mean(np.abs((y_true - y_pred) / y_true)) * 100
def MPE(y_true, y_pred):
    return np.mean((y_true - y_pred) / y_true) * 100
def root_mean_squared_log_error(y_true, y_pred):
    return np.sqrt(mean_squared_log_error(y_true, y_pred))
print('R2_Score')
print('-' * 40)
print('train error: {} |\nvalid error: {} |\ntest error : {}\n'.
 →format(score_train[9], score_validation[9], r2_score(Y, Y_hat)))
print('Mean Squared Error')
print('-' * 40)
print('train error: {} |\nvalid error: {} |\ntest error : {}\n'.
 →format(score_train[2], score_validation[2], MSE(Y, Y_hat)))
print('Mean Absolute Error')
print('-' * 40)
print('train error: {} |\nvalid error: {} |\ntest error : {}\n'.
 →format(score_train[3], score_validation[3], MAE(Y, Y_hat)))
print('Root Mean Squared Error')
print('-' * 40)
print('train error: {} |\nvalid error: {} |\ntest error : {}\n'.
→format(score_train[4], score_validation[3], RMSE(Y, Y_hat)))
print('Mean Squared Logarithmic Error')
print('-' * 40)
print('train error: {} |\nvalid error: {} |\ntest error : {}\n'.
 →format(score_train[7], score_validation[7], mean_squared_log_error(Y, Y_hat)))
print('Root Mean Squared Logarithmic Error')
print('-' * 40)
```

```
print('train error: {} |\nvalid error: {} |\ntest error : {}\n'.
 →format(score_train[8], score_validation[8], root_mean_squared_log_error(Y,_
 \rightarrowY hat)))
print('Mean Absolute Percentage Error')
print('-' * 40)
print('train error: {} |\nvalid error: {} |\ntest error : {}\n'.
 →format(score_train[5], score_validation[3], MAPE(Y, Y_hat)))
print('Mean Percentage Error')
print('-' * 40)
print('train error: {} |\nvalid error: {} |\ntest error : {}\n'.
 →format(score_train[6], score_validation[3], MPE(Y, Y_hat)))
R2_Score
_____
train error: 0.9942859411239624 |
valid error: 0.992901086807251 |
test error : 0.969569823591676
Mean Squared Error
train error: 0.00024504572502337396 |
valid error: 0.0002654152922332287 |
test error: 0.00019094254030111796
Mean Absolute Error
train error: 0.009836779907345772 |
valid error: 0.01048602070659399 |
test error: 0.009594846344983357
Root Mean Squared Error
______
train error: 0.015370653010904789
valid error: 0.01048602070659399
test error: 0.013818195985768835
Mean Squared Logarithmic Error
______
train error: 0.00011239355808356777
valid error: 0.0001207444875035435
test error: 7.643460276778266e-05
Root Mean Squared Logarithmic Error
_____
train error: 0.007310476154088974 |
```

```
valid error: 0.007747378665953875 |
     test error : 0.008742688532012488
     Mean Absolute Percentage Error
     train error: 802.353515625
     valid error: 0.01048602070659399 |
     test error : 1.7292413517444112
     Mean Percentage Error
     train error: inf |
     valid error: 0.01048602070659399 |
     test error: 1.037522508518689
[28]: # count : 개수, std : 표준편차
      test[label_cols].describe()
[28]:
                  Close
     count 1037.000000
     mean
               0.551764
     std
               0.079329
     min
               0.314971
     25%
               0.498957
     50%
               0.535049
     75%
               0.565572
     max
               0.837970
 []:
```