RNN

38 visit_pop_cnt_sf

1666 non-null

float64

```
In [1]:
        import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.preprocessing import MinMaxScaler
        from sklearn.model_selection import train_test_split
In [3]:
        # 파일
        df = pd.read_csv('전체데이터_병합.csv',encoding='cp949',parse_dates=['y_m'])
        df.info()
        df.head()
       <class 'pandas.core.frame.DataFrame'>
       RangeIndex: 1666 entries, 0 to 1665
       Data columns (total 39 columns):
        #
            Column
                           Non-Null Count Dtype
                            _____
        0
                           1666 non-null datetime64[ns]
            y_m
                           1666 non-null object
            city
        1
                          1666 non-null object
1666 non-null float64
        2
            location
           area_cnt
        3
        4
           em_cnt
                           1666 non-null int64
        5
           pay_amt 1666 non-null int64
제주도민_여 1666 **
                           1666 non-null int64
        6
                            1666 non-null float64
1666 non-null float64
1666 non-null float64
        7
            외국인거주_여
        8
        9
            제주도민_남
                                1666 non-null float64
        10 외국인거주_남
        11 제주도민_60이상
                                1666 non-null float64
                                 1666 non-null float64
        12 제주도민_60미만
                           1666 non-null float64
        13 total_pop

      14 패스트푸드_결제건수
      1666 non-null float64

      15 패스트푸드_결제금액
      1666 non-null float64

        15 패스트푸드_결제금액
        16 간식_결제건수
                                 1666 non-null float64
        17
           간식_결제금액
                                  1666 non-null float64
        18 농축수산물_결제건수
                                  1666 non-null float64
        19 농축수산물_결제금액
                                   1666 non-null float64
        20 마트/슈퍼마켓_결제건수
                                   1666 non-null float64
        21 마트/슈퍼마켓_결제금액
                                    1666 non-null float64
        22 식품_결제건수
                                  1666 non-null float64
        23 식품_결제금액
                                  1666 non-null float64
                                  1666 non-null float64
        24 배달_결제건수
        25
                                  1666 non-null float64
           배달_결제금액
        26 식당_결제건수
                                  1666 non-null float64
        27 식당_결제금액
                                  1666 non-null float64
        28 풍속
                              1666 non-null float64
        29 기온
                              1666 non-null float64
                              1666 non-null float64
        30 습도
                              1666 non-null float64
        31 강수
        32 전국_누적확진자
                                  1666 non-null float64
        33 전국_월별확진자
                                  1666 non-null float64
        34 제주_누적확진자
                                  1666 non-null float64
        35 제주_월별확진자
                                  1666 non-null float64
        36 visit_pop_cnt
                            1666 non-null
                                         float64
                            1666 non-null
        37 visit_pop_cnt_lf
                                          float64
```

dtypes: datetime64[ns](1), float64(33), int64(3), object(2)

memory usage: 507.7+ KB

Out[3]:

	y_m	city	location	area_cnt	em_cnt	em_g	pay_amt	제주도 민_여	외국 인거 주_여	제주도 민_남	•••	기
0	2018- 01-01	서 귀 포 시	남원읍	52.0	9570	42437700	1270773	9306.0	200.0	9806.0		6.25658
1	2018- 01-01	서 귀 포 시	대륜동	38.0	21666	57612600	1676850	6637.0	95.0	6836.0		8.0043(
2	2018- 01-01	서 귀 포 시	대정읍	89.0	10185	38885550	1164122	10725.0	677.0	10360.0		5.41787
3	2018- 01-01	서 귀 포 시	대천동	37.0	20280	53858550	1593709	6475.0	137.0	6685.0		8.0043(
4	2018- 01-01	서 귀 포 시	동홍동	49.0	45936	118701000	3501286	11569.0	642.0	11124.0		5.7715(

5 rows × 39 columns

```
In [4]:
       df.isnull().sum()
                     0
Out[4]:
                     0
      city
      location
                     0
                     0
      area_cnt
                     0
      em_cnt
      em_g
                     0
      pay_amt
      제주도민_여
                         0
      외국인거주_여
                          0
      제주도민_남
                         0
      외국인거주_남
                          0
      제주도민_60이상
                          0
      제주도민_60미만
                          0
      total_pop
      패스트푸드_결제건수
                            0
      패스트푸드_결제금액
                            0
      간식_결제건수
                          0
      간식_결제금액
      농축수산물_결제건수
                            0
      농축수산물_결제금액
                            0
      마트/슈퍼마켓_결제건수
                             0
      마트/슈퍼마켓_결제금액
                             0
```

```
식품_결제건수
                   0
식품_결제금액
                   0
배달_결제건수
                   0
배달_결제금액
                   0
식당_결제건수
                   0
식당_결제금액
                   0
풍속
                 0
기온
                0
습도
                 0
강수
                 0
전국_누적확진자
                    0
전국_월별확진자
                    0
제주_누적확진자
                    0
제주_월별확진자
                    0
visit_pop_cnt
               0
visit_pop_cnt_lf
               0
               0
visit_pop_cnt_sf
dtype: int64
```

```
In [5]: df = df.fillna(0)
```

```
In [6]: df['y_m'] = pd.to_datetime(df['y_m'],format='%Y%m')
df['year'] = df['y_m'].dt.year
df
```

Out[6]:

	y_m	city	location	area_cnt	em_cnt	em_g	pay_amt	제주도 민_여	외국 인거 주_여	제주도 민_남	•••	
0	2018- 01-01	서 귀 포 시	남원읍	52.0	9570	42437700	1270773	9306.0	200.0	9806.0		62
1	2018- 01-01	서 귀 포 시	대륜동	38.0	21666	57612600	1676850	6637.0	95.0	6836.0		66
2	2018- 01-01	서 귀 포 시	대정읍	89.0	10185	38885550	1164122	10725.0	677.0	10360.0		70
3	2018- 01-01	서 귀 포 시	대천동	37.0	20280	53858550	1593709	6475.0	137.0	6685.0		66
4	2018- 01-01	서 귀 포 시	동홍동	49.0	45936	118701000	3501286	11569.0	642.0	11124.0		69
•••												
1661	2021- 06-01	제 주 시	일도2동	87.0	84360	147438200	4402149	16569.0	200.0	16077.0		75
1662	2021- 06-01	제 주 시	조천읍	141.0	27732	63927750	1911187	12422.0	242.0	13017.0		82

	y_m	city	location	area_cnt	em_cnt	em_g	pay_amt	제주도 민_여	외국 인거 주_여	제주도 민_남	•••	
1663	2021- 06-01	제 주 시	한경면	71.0	8031	27060150	809898	4531.0	100.0	4627.0		93
1664	2021- 06-01	제 주 시	한림읍	112.0	25653	82746990	2476292	10341.0	1140.0	10891.0		92
1665	2021- 06-01	제 주 시	화북동	84.0	66088	110750050	3306029	12238.0	161.0	12062.0		75

1666 rows × 40 columns

```
In [7]:
          # df_l = df['location'] == '노형동'
          \# dfI = df[df_I]
          # dfl
 In [8]:
          # df_l = df['location'] == '건입동'
          \# dfI = df[df_I]
          # dfl
In [9]:
          # df_| = df['location'] == '연동'
          \# dfI = df[df_I]
          # dfl
In [10]:
          df_l = df['location'] == '이도2동'
          dfI = df[df_I]
          dfl.head()
```

Out[10]:

	y_m	city	location	area_cnt	em_cnt	em_g	pay_amt	제주도 민_여	외국 인거 주_여	제주도 민_남	•••	
33	2018- 01-01	제 주 시	이도2동	132.0	99670	246651600	7371540	25593.0	344.0	24510.0		66.3
71	2018- 02-01	제 주 시	이도2동	132.0	92603	230440750	6886558	25488.0	332.0	24417.0		60.74
109	2018- 03-01	제 주 시	이도2동	132.0	103323	249276500	7448453	25421.0	341.0	24400.0		73.1

y_m city location area_cnt em_cnt em_g pay_amt 제주도 외국 제주도 인거 민_남 ...

```
제
    2018-
147
            주
                이도2동
                           132.0 102728 227249550 6788163 25420.0 342.0 24393.0 ... 68.83
     04-01
            시
            제
    2018-
            주
185
                이도2동
                           132.0 113606 234533700 7004233 25435.0 354.0 24448.0 ... 75.80
     05-01
            시
```

5 rows × 40 columns

```
In [13]:
# 정규화
scaler = MinMaxScaler()
scale_cols = dfl.drop(columns=['y_m', 'city', 'location'], axis=1)
scale_cols[:] = scaler.fit_transform(scale_cols[:])
# 유의변수 추출
scale_cols = scale_cols.drop(columns=['제주도민_60미만', 'total_pop', '식당_결제금액', 'scale_cols.head()

Out[13]:

WASER NOTER NOTER AND SCALE TO SCAL
```

```
제주도민
                                                 외국인거
                                                          제주도민
                                                                                      식품
                                                                             퍼마켓
    area cnt
              em cnt
                         em g
                               pay amt
                                                                     물 결제
                                             여
                                                    주_남
                                                           60이상
                                                                                       제금
                                                                       건수
                                                                            결제금액
 33
         0.0 0.245119 0.530298
                               0.531984
                                        1.000000
                                                 0.000000
                                                          0.000000
                                                                   0.186319 0.599766
                                                                                     0.0033
71
         0.0 0.140992 0.403869
                               0.405325 0.732824
                                                 0.015873
                                                          0.018366 0.190725 0.558672
                                                                                    0.1304
109
         0.0 0.298944 0.550769
                               0.552070 0.562341
                                                 0.095238
                                                          0.035465
                                                                  0.000000
                                                                            0.528745
                                                                                     0.0819
         0.0 0.290177 0.378981 0.379629 0.559796 0.126984
147
                                                          0.050032 0.097623
                                                                           0.610611 0.0000
185
         0.0 0.450456 0.435790 0.436058 0.597964 1.000000 0.089297 0.273507 0.852501 0.1526
```

•

```
In [14]: # 데이터셋 분리(시계열)

TEST_SIZE = 36# 3년 데이터
WINDOW_SIZE = 6 # 6개월 데이터

test = scale_cols[:-TEST_SIZE]
train = scale_cols[-TEST_SIZE:]
```

```
In [15]: # 훈련데이터와 테스트데이터 분리에 사용

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)
```

```
# 훈련데이터와 테스트데이터 분리
In [16]:
          feature_cols = scale_cols.columns
          label_cols = ['em_g']
          train_feature = train[feature_cols]
          train_label = train[label_cols]
          train_feature, train_label = make_dataset(train_feature, train_label, 20)
          x_train, x_valid, y_train, y_valid = train_test_split(train_feature, train_label, test
          x_train.shape, x_valid.shape
         ((12, 20, 17), (4, 20, 17))
Out[16]:
In [17]:
          test_feature = test[feature_cols]
          test_label = test[label_cols]
          test_feature.shape, test_label.shape
         ((6, 17), (6, 1))
Out[17]:
In [18]:
          test_feature, test_label = make_dataset(test_feature, test_label, 4)
          test_feature.shape, test_label.shape
         ((2, 4, 17), (2, 1))
Out[18]:
In [19]:
          # tensorflow, keras
          import tensorflow as tf
          from tensorflow import keras
In [20]:
          # keras 모형
          from keras.models import Sequential
          from keras. Layers import Dense
          from keras.callbacks import EarlyStopping, ModelCheckpoint
          from keras.layers import LSTM
          model = Sequential()
          model.add(LSTM(16,
                         input_shape=(train_feature.shape[1], train_feature.shape[2]),
                         activation='relu',
                         return_sequences=False)
                    )
          model.add(Dense(1))
In [21]:
          import os
          model.compile(loss='mean_squared_error', optimizer='adam',metrics=["acc"]) # acc 안나
          early_stop = EarlyStopping(monitor='val_loss', patience=5)
          model_path = 'model'
          filename = os.path.join(model_path, 'tmp_checkpoint.h5')
          checkpoint = ModelCheckpoint(filename, monitor='val_loss', verbose=1, save_best_only=
          history = model.fit(x_train, y_train,
                                              epochs=200,
                                              batch_size=16.
```

```
Epoch 1/200
1/1 [============== ] - 2s 2s/step - loss: 0.1544 - acc: 0.0833 - val_l
oss: 0.0746 - val_acc: 0.0000e+00
Epoch 00001: val_loss improved from inf to 0.07465, saving model to model\tmp_checkpoi
nt.h5
Epoch 2/200
1/1 [============== ] - Os 44ms/step - loss: 0.1339 - acc: 0.0833 - val
_loss: 0.0593 - val_acc: 0.0000e+00
Epoch 00002: val_loss improved from 0.07465 to 0.05925, saving model to model\tmp_chec
kpoint.h5
Epoch 3/200
1/1 [============== ] - Os 48ms/step - loss: 0.1152 - acc: 0.0833 - val
_loss: 0.0462 - val_acc: 0.0000e+00
Epoch 00003: val_loss improved from 0.05925 to 0.04617, saving model to model\tmp_chec
kpoint.h5
Epoch 4/200
1/1 [============= ] - Os 48ms/step - loss: 0.0985 - acc: 0.0833 - val
_loss: 0.0352 - val_acc: 0.0000e+00
Epoch 00004: val_loss improved from 0.04617 to 0.03521, saving model to model\tmp_chec
kpoint.h5
Epoch 5/200
_loss: 0.0265 - val_acc: 0.0000e+00
Epoch 00005: val_loss improved from 0.03521 to 0.02648, saving model to model\tmp_chec
kpoint.h5
Epoch 6/200
1/1 [========================== ] - Os 48ms/step - loss: 0.0704 - acc: 0.0833 - val
_loss: 0.0198 - val_acc: 0.0000e+00
Epoch 00006: val_loss improved from 0.02648 to 0.01981, saving model to model\tmp_chec
kpoint.h5
Epoch 7/200
1/1 [========================== ] - Os 52ms/step - loss: 0.0591 - acc: 0.0833 - val
_loss: 0.0151 - val_acc: 0.0000e+00
Epoch 00007: val_loss improved from 0.01981 to 0.01508, saving model to model\tmp_chec
kpoint.h5
Epoch 8/200
1/1 [=========] - Os 45ms/step - Ioss: 0.0493 - acc: 0.0833 - val
_loss: 0.0123 - val_acc: 0.0000e+00
Epoch 00008: val_loss improved from 0.01508 to 0.01229, saving model to model\tmp_chec
kpoint.h5
Epoch 9/200
1/1 [=============== ] - Os 53ms/step - loss: 0.0411 - acc: 0.0833 - val
_loss: 0.0114 - val_acc: 0.0000e+00
Epoch 00009: val_loss improved from 0.01229 to 0.01137, saving model to model\tmp_chec
kpoint.h5
Epoch 10/200
1/1 [============== ] - Os 50ms/step - loss: 0.0345 - acc: 0.0833 - val
_loss: 0.0123 - val_acc: 0.0000e+00
Epoch 00010: val_loss did not improve from 0.01137
Epoch 11/200
1/1 [=========================] - Os 52ms/step - loss: 0.0295 - acc: 0.0833 - val
```

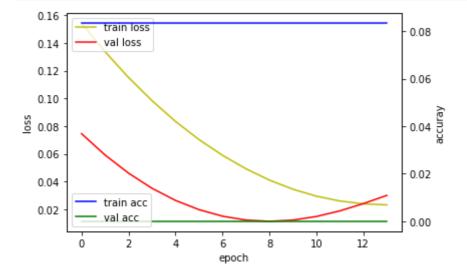
_loss: 0.0149 - val_acc: 0.0000e+00

Epoch 00014: val_loss did not improve from 0.01137

```
In [23]: fig, loss_ax = plt.subplots()
    acc_ax = loss_ax.twinx()
    loss_ax.plot(history.history['loss'], 'y', label='train loss')
    loss_ax.plot(history.history['val_loss'], 'r', label='val loss')
    acc_ax.plot(history.history['acc'], 'b', label='train acc')
    acc_ax.plot(history.history['val_acc'], 'g', label='val acc')

    loss_ax.set_xlabel('epoch')
    loss_ax.set_ylabel('loss')
    acc_ax.set_ylabel('loss')
    acc_ax.set_ylabel('accuray')
    loss_ax.legend(loc='upper left')
    acc_ax.legend(loc='lower left')

# plt.savefig('rnn_l.png')
```



```
In [24]: model.load_weights(filename)
    pred = model.predict(test_feature)
    pred.shape
```

WARNING:tensorflow:Model was constructed with shape (None, 20, 17) for input KerasTens or(type_spec=TensorSpec(shape=(None, 20, 17), dtype=tf.float32, name='lstm_input'), name='lstm_input', description="created by layer 'lstm_input'"), but it was called on an input with incompatible shape (None, 4, 17). (2, 1)

```
Out[24]:
```

```
In [25]: plt.figure(figsize=(12, 9))
    plt.plot(test_label, label = 'actual')
    plt.plot(pred, label = 'prediction')
    plt.legend()
    # plt.savefig('rnn_l_p.png')
```

