

Second Programming Exam. Deep Learning, 2022

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```
import pandas as pd
import numpy as np

from random import randrange
from matplotlib import pyplot as plt
import pandas as pd
import numpy as np
import xgboost as xgb
import pickle, os, pymysql
from sklearn.preprocessing import MinMaxScaler
from tensorflow.keras.layers import LSTM, Dropout, Dense, Activation
from tensorflow.keras.models import Sequential, load_model
from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping
from sys import exit

def plot_loss(x, y):
    plt.figure(figsize=(12,6))
    plt.plot(x)
    plt.plot(y)
    plt.title('LSTM Model loss')
    plt.ylabel('MSE')
    plt.xlabel('epoch')
    plt.legend(['Train_loss', 'Val_loss'], loc='upper right')

def Get_data():
    df_train = pd.read_csv('predict1_trn.csv')
    df_test = pd.read_csv('predict1_tst.csv')
    X_train, y_train = df_train.iloc[:, :24], df_train.iloc[:, -6:]
    X_test = df_test.iloc[:, :24]
    return X_train, y_train, X_test

def Normalize():
    df_train = pd.read_csv('predict1_trn.csv')
    scaler = MinMaxScaler(feature_range=(0,1))
    scaled = scaler.fit_transform(df_train)
```

```
X_scaled, y_scaled = scaled[:, :24], scaled[:, -6:]
return scaler, X_scaled, y_scaled
```

```
def Build_Model(X, y):
    model = Sequential()
    model.add(LSTM(50, return_sequences = True, input_shape = (X.shape[1], 1)))
    model.add(Dropout(0.2))
    model.add(LSTM(50, return_sequences = True))
    model.add(Dropout(0.2))
    model.add(LSTM(50))
    model.add(Dropout(0.2))
    model.add(Dense(y.shape[1]))
    model.summary()
    model.compile(loss = 'mse', optimizer='adam', metrics=['mse']) #rmsprop
    return model
```

```
def Train_Model(X, y):
    re_X = X.reshape(X.shape[0], X.shape[1], 1)
    file1 = 'Q2/predict1.h5'
    checkpoint = ModelCheckpoint(file1, monitor='val_loss', verbose=2,
save_best_only = True, mode='min')
    earlyStopping = EarlyStopping(monitor='val_loss', patience=50, verbose=2,
mode='auto')
    callbacks_list= [checkpoint,earlyStopping]
    lstm_model = Build_Model(re_X, y)
    history = lstm_model.fit(re_X, y, epochs=90, batch_size=30,
callbacks=callbacks_list, validation_split=0.1)
    # savemodel
    lstm_model.save(file1)
    return history.history ['loss'], history.history ['val_loss']
```

```
def Predict(model, X_test):
    temp_x = np.concatenate((X_test.iloc[:, :].values, X_test.iloc[:, -6:].values), axis=1)
    temp_x = scaler.fit_transform(temp_x)
    y_hat = model.predict(temp_x[:, :24])
    # inverse
    all_test = np.concatenate((temp_x[:, :24], y_hat), axis=1)
    actual_test = scaler.inverse_transform(all_test)
```

```
return y_hat, actual_test
```

```
def Export_csv(data):
```

```
    cols = []
```

```
    for i in range(-23,7,1):
```

```
        if i < 0 :
```

```
            cols.append(f't-{{i*(-1)}}')
```

```
        elif i==0:
```

```
            cols.append('t')
```

```
        else:
```

```
            cols.append(f't+{{i}}')
```

```
df_fn = pd.DataFrame(data)
```

```
df_fn.to_csv('Q2/predict1_answer.csv', index=None, header=cols)
```

```
# read csv data
```

```
X_train, y_train, X_test = Get_data()
```

```
# normalize
```

```
scaler, X_scaled, y_scaled = Normalize()
```

```
# train model
```

```
loss, val_loss = Train_Model(X_scaled, y_scaled)
```

```
# plot loss
```

```
plot_loss(loss, val_loss)
```

```
# load model
```

```
pre_model= load_model('Q2/predict1.h5')
```

```
# predict value
```

```
y_hat, actual = Predict(pre_model, X_test)
```

```
# export csv
```

```
Export_csv = Export_csv(actual)
```