UROP 1100 - Victoria A. Junaedi

Code from https://github.com/BenjiKCF/Neural-Network-with-Financial-Time-Series-Data (https://github.com/BenjiKCF/Neural-Network-with-Financial-Time-Series-Data)

In [2]:

```
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.pyplot as plt2
import pandas as pd
from pandas import datetime
import math, time
from sklearn import preprocessing
import datetime
from sklearn.metrics import mean_squared_error
from math import sqrt
from keras.models import Sequential
from keras.layers.core import Dense, Dropout, Activation
from keras.layers.recurrent import LSTM
from keras.models import load_model
import keras
import pandas_datareader.data as web
import h5py
from keras import backend as K
import quandl
```

Using TensorFlow backend.

Getting the data from Quandl

and determining other parameters

In [49]:

```
quandl.ApiConfig.api_key = 'ETPyWitxJn3wiSRvxUjk'
seq_len = 22
shape = [seq_len, 9, 1]
neurons = [256, 256, 32, 1]
dropout = 0.3
decay = 0.5
epochs = 300
stock_name = 'CathayPacific'
```

Pulling the data from Quandl

In [50]:

```
def get_stock_data(stock_name, normalize=True, ma=[]):
    Return a dataframe of that stock and normalize all the values.
    (Optional: create moving average)
      df = quandl.get_table('HKEX', ticker = stock_name)
\# Date, Nominal Price, Net Change, Change (%), Bid, Ask, P/E(x), High, Low, Previous Close, Share
 Volume ('000), Turnover ('000), Lot Size
    df = quandl.get("HKEX/00293", authtoken="ETPyWitxJn3wiSRvxUjk")
    df.drop(['Net Change','P/E(x)', 'Bid', 'Ask', 'Lot Size', 'Turnover (\'000)'], 1, i
nplace=True)
      df.set_index('Date', inplace=True)
    # Renaming all the columns so that we can use the old version code
    df.rename(columns={'Nominal Price': 'Open', 'Share Volume (\'000)': 'Volume', 'Previ
ous Close': 'Adj Close', 'Change (%)': 'Pct'}, inplace=True)
#
      # Percentage change
      df['Pct'] = df['Adj Close'].pct_change()
#
    df.dropna(inplace=True)
    # Moving Average
    if ma != []:
        for moving in ma:
            df['{}ma'.format(moving)] = df['Adj Close'].rolling(window=moving).mean()
    df.dropna(inplace=True)
    if normalize:
        min_max_scaler = preprocessing.MinMaxScaler()
        df['Open'] = min_max_scaler.fit_transform(df.Open.values.reshape(-1,1))
        df['High'] = min_max_scaler.fit_transform(df.High.values.reshape(-1,1))
        df['Low'] = min_max_scaler.fit_transform(df.Low.values.reshape(-1,1))
        df['Volume'] = min max scaler.fit transform(df.Volume.values.reshape(-1,1))
        df['Adj Close'] = min_max_scaler.fit_transform(df['Adj
Close'].values.reshape(-1,1))
        df['Pct'] = min_max_scaler.fit_transform(df['Pct'].values.reshape(-1,1))
        if ma != []:
            for moving in ma:
                df['{}ma'.format(moving)] = min max scaler.fit transform(df['{}ma'.form
at(moving)].values.reshape(-1,1))
    # Move Adj Close to the rightmost for the ease of training
    adj_close = df['Adj Close']
    df.drop(labels=['Adj Close'], axis=1, inplace=True)
    df = pd.concat([df, adj close], axis=1)
    return df
```

```
In [51]:
```

```
df = get_stock_data(stock_name, ma=[50, 100, 200]) # putting 50 days, 100 days, and 200
days moving average
```

Visualizing the data

In [52]:

```
def plot_stock(df):
    print(df.head())
    plt.subplot(211)
    plt.plot(df['Adj Close'], color='red', label='Adj Close')
    plt.legend(loc='best')
    plt.subplot(212)
    plt.plot(df['Pct'], color='blue', label='Percentage change')
    plt.legend(loc='best')
    plt.show()
```

In [53]:

0.5

```
plot_stock(df)
                0pen
                            Pct
                                     High
                                                 Low
                                                        Volume
                                                                     50ma
Date
2014-12-19
            0.628352
                      0.350632
                                 0.623352
                                           0.629412
                                                      0.062676
                                                                0.560608
2014-12-22
            0.649425
                                 0.629002
                                           0.649020
                                                                0.564995
                      0.375036
                                                      0.018500
2014-12-23
            0.632184
                      0.350439
                                 0.627119
                                           0.643137
                                                      0.020874
                                                                0.570337
2014-12-24
            0.634100
                      0.258898
                                 0.623352
                                           0.643137
                                                      0.009915
                                                                0.575722
2014-12-29
            0.636015
                      0.258898
                                 0.629002 0.654902
                                                      0.017850
                                                                0.581759
               100ma
                          200ma Adj Close
Date
2014-12-19
           0.525715
                      0.569391
                                  0.639469
2014-12-22
            0.527974
                      0.570507
                                  0.622391
2014-12-23
            0.530558
                      0.571352
                                  0.643264
2014-12-24
            0.533049
                      0.572354
                                  0.626186
2014-12-29 0.535634 0.573413
                                  0.628083
 1.0
                                          Adj Close
 0.5
  2014-122015-042015-082015-122016-042016-082016-122017-042017-08
 1.0
```

Split training and testing

2014-122015-042015-082015-122016-042016-082016-122017-042017-08

In [54]:

```
def load data(stock, seq len):
    amount_of_features = len(stock.columns)
    print ("Amount of features = {}".format(amount_of_features))
    data = stock.as matrix()
    sequence_length = seq_len + 1 # index starting from 0
    result = []
    for index in range(len(data) - sequence_length): # maxmimum date = Lastest date - s
equence Length
        result.append(data[index: index + sequence length]) # index : index + 22days
    result = np.array(result)
    row = round(0.8 * result.shape[0]) # 80% split
    print ("Amount of training data = {}".format(round(0.9 * result.shape[0])))
    print ("Amount of testing data = {}".format(round(0.1 * result.shape[0])))
    train = result[:int(row), :] # 90% date
    X_train = train[:, :-1] # all data until day m
    y_train = train[:, -1][:,-1] # day m + 1 adjusted close price
   X_test = result[int(row):, :-1]
    y_test = result[int(row):, -1][:,-1]
   X_train = np.reshape(X_train, (X_train.shape[0], X_train.shape[1], amount_of_featur
es))
   X_test = np.reshape(X_test, (X_test.shape[0], X_test.shape[1], amount_of_features))
    return [X_train, y_train, X_test, y_test]
```

In [55]:

```
X_train, y_train, X_test, y_test = load_data(df, seq_len)
```

```
Amount of features = 9
Amount of training data = 558
Amount of testing data = 62
```

In [56]:

```
def build model(shape, neurons, dropout, decay):
    model = Sequential()
    model.add(LSTM(neurons[0], input_shape=(shape[0], shape[1]),
return sequences=True))
    model.add(Dropout(dropout))
    model.add(LSTM(neurons[1], input_shape=(shape[0], shape[1]),
return sequences=False))
    model.add(Dropout(dropout))
    model.add(Dense(neurons[2],kernel_initializer="uniform",activation='relu'))
    model.add(Dense(neurons[3],kernel_initializer="uniform",activation='linear'))
    # model = load_model('my_LSTM_stock_model1000.h5')
    adam = keras.optimizers.Adam(decay=decay)
    model.compile(loss='mse',optimizer='adam', metrics=['accuracy'])
    model.summary()
    return model
```

In [59]:

model = build_model(shape, neurons, dropout, decay)

Layer (type)	Output Shape	Param #
lstm_7 (LSTM)	(None, 22, 256)	272384
dropout_7 (Dropout)	(None, 22, 256)	0
lstm_8 (LSTM)	(None, 256)	525312
dropout_8 (Dropout)	(None, 256)	0
dense_7 (Dense)	(None, 32)	8224
dense_8 (Dense)	(None, 1)	33

Total params: 805,953 Trainable params: 805,953 Non-trainable params: 0

In [67]:

```
model.fit(
   X_train,
   y_train,
   batch_size=512,
   epochs=epochs,
   validation_split=0.33,
   verbose=1)
```

```
Train on 332 samples, validate on 164 samples
Epoch 1/300
332/332 [=============== ] - 6s - loss: 0.0079 - acc: 0.0030
- val_loss: 0.0348 - val_acc: 0.0061
Epoch 2/300
332/332 [================ ] - 3s - loss: 0.0105 - acc: 0.0030
- val_loss: 0.0340 - val_acc: 0.0061
Epoch 3/300
- val_loss: 0.0327 - val_acc: 0.0061
Epoch 4/300
332/332 [================ ] - 2s - loss: 0.0079 - acc: 0.0030
- val loss: 0.0326 - val acc: 0.0061
Epoch 5/300
332/332 [============= ] - 2s - loss: 0.0086 - acc: 0.0030
- val_loss: 0.0342 - val_acc: 0.0061
Epoch 6/300
332/332 [================ ] - 2s - loss: 0.0058 - acc: 0.0030
- val_loss: 0.0362 - val_acc: 0.0061
Epoch 7/300
332/332 [================ ] - 2s - loss: 0.0072 - acc: 0.0030
- val_loss: 0.0371 - val_acc: 0.0061
Epoch 8/300
332/332 [================ ] - 2s - loss: 0.0072 - acc: 0.0030
- val_loss: 0.0366 - val_acc: 0.0061
Epoch 9/300
332/332 [================ ] - 2s - loss: 0.0058 - acc: 0.0030
- val_loss: 0.0367 - val_acc: 0.0061
Epoch 10/300
- val_loss: 0.0379 - val_acc: 0.0061
Epoch 11/300
332/332 [================ ] - 3s - loss: 0.0056 - acc: 0.0030
- val_loss: 0.0381 - val_acc: 0.0061
Epoch 12/300
- val_loss: 0.0367 - val_acc: 0.0061
Epoch 13/300
- val_loss: 0.0367 - val_acc: 0.0061
Epoch 14/300
- val_loss: 0.0377 - val_acc: 0.0061
Epoch 15/300
- val_loss: 0.0373 - val_acc: 0.0061
Epoch 16/300
- val_loss: 0.0362 - val_acc: 0.0061
Epoch 17/300
- val_loss: 0.0371 - val_acc: 0.0061
Epoch 18/300
332/332 [============== ] - 2s - loss: 0.0054 - acc: 0.0030
- val loss: 0.0365 - val acc: 0.0061
Epoch 19/300
332/332 [=============== ] - 2s - loss: 0.0053 - acc: 0.0030
- val_loss: 0.0347 - val_acc: 0.0061
Epoch 20/300
- val loss: 0.0351 - val acc: 0.0061
```

```
Epoch 21/300
- val loss: 0.0346 - val acc: 0.0061
Epoch 22/300
332/332 [=============== ] - 2s - loss: 0.0048 - acc: 0.0030
- val_loss: 0.0330 - val_acc: 0.0061
Epoch 23/300
332/332 [================ ] - 2s - loss: 0.0046 - acc: 0.0030
- val_loss: 0.0330 - val_acc: 0.0061
Epoch 24/300
- val_loss: 0.0335 - val_acc: 0.0061
Epoch 25/300
- val_loss: 0.0329 - val_acc: 0.0061
Epoch 26/300
332/332 [================ ] - 2s - loss: 0.0039 - acc: 0.0030
- val_loss: 0.0324 - val_acc: 0.0061
Epoch 27/300
332/332 [================ ] - 2s - loss: 0.0049 - acc: 0.0030
- val_loss: 0.0333 - val_acc: 0.0061
Epoch 28/300
332/332 [================ ] - 2s - loss: 0.0042 - acc: 0.0030
- val_loss: 0.0332 - val_acc: 0.0061
Epoch 29/300
- val loss: 0.0323 - val acc: 0.0061
Epoch 30/300
- val_loss: 0.0325 - val_acc: 0.0061
Epoch 31/300
- val_loss: 0.0330 - val_acc: 0.0061
Epoch 32/300
332/332 [================ ] - 2s - loss: 0.0051 - acc: 0.0030
- val_loss: 0.0323 - val_acc: 0.0061
Epoch 33/300
332/332 [================ ] - 2s - loss: 0.0041 - acc: 0.0030
- val_loss: 0.0324 - val_acc: 0.0061
Epoch 34/300
332/332 [================ ] - 2s - loss: 0.0036 - acc: 0.0030
- val_loss: 0.0328 - val_acc: 0.0061
Epoch 35/300
- val loss: 0.0325 - val acc: 0.0061
Epoch 36/300
- val_loss: 0.0321 - val_acc: 0.0061
Epoch 37/300
- val_loss: 0.0330 - val_acc: 0.0061
Epoch 38/300
- val_loss: 0.0328 - val_acc: 0.0061
Epoch 39/300
- val_loss: 0.0318 - val_acc: 0.0061
Epoch 40/300
- val_loss: 0.0334 - val_acc: 0.0061
Epoch 41/300
```

```
- val_loss: 0.0332 - val_acc: 0.0061
Epoch 42/300
- val loss: 0.0318 - val acc: 0.0061
Epoch 43/300
- val_loss: 0.0322 - val_acc: 0.0061
Epoch 44/300
332/332 [================ ] - 2s - loss: 0.0035 - acc: 0.0030
- val_loss: 0.0331 - val_acc: 0.0061
Epoch 45/300
- val_loss: 0.0328 - val_acc: 0.0061
Epoch 46/300
332/332 [============= ] - 2s - loss: 0.0040 - acc: 0.0030
- val_loss: 0.0318 - val_acc: 0.0061
Epoch 47/300
- val_loss: 0.0319 - val_acc: 0.0061
Epoch 48/300
332/332 [================ ] - 2s - loss: 0.0034 - acc: 0.0030
- val_loss: 0.0323 - val_acc: 0.0061
Epoch 49/300
- val_loss: 0.0316 - val_acc: 0.0061
Epoch 50/300
332/332 [================ ] - 2s - loss: 0.0037 - acc: 0.0030
- val_loss: 0.0304 - val_acc: 0.0061
Epoch 51/300
- val_loss: 0.0305 - val_acc: 0.0061
Epoch 52/300
332/332 [================ ] - 2s - loss: 0.0034 - acc: 0.0030
- val_loss: 0.0309 - val_acc: 0.0061
Epoch 53/300
- val_loss: 0.0308 - val_acc: 0.0061
Epoch 54/300
332/332 [================ ] - 2s - loss: 0.0035 - acc: 0.0030
- val_loss: 0.0305 - val_acc: 0.0061
Epoch 55/300
- val_loss: 0.0316 - val_acc: 0.0061
Epoch 56/300
- val_loss: 0.0309 - val_acc: 0.0061
Epoch 57/300
- val_loss: 0.0297 - val_acc: 0.0061
Epoch 58/300
332/332 [=============== ] - 2s - loss: 0.0042 - acc: 0.0030
- val_loss: 0.0306 - val_acc: 0.0061
Epoch 59/300
- val_loss: 0.0311 - val_acc: 0.0061
Epoch 60/300
332/332 [=============== ] - 2s - loss: 0.0039 - acc: 0.0030
- val_loss: 0.0307 - val_acc: 0.0061
Epoch 61/300
```

```
- val_loss: 0.0308 - val_acc: 0.0061
Epoch 62/300
332/332 [============== ] - 2s - loss: 0.0038 - acc: 0.0030
- val_loss: 0.0312 - val_acc: 0.0061
Epoch 63/300
332/332 [================ ] - 2s - loss: 0.0037 - acc: 0.0030
- val_loss: 0.0306 - val_acc: 0.0061
Epoch 64/300
332/332 [================ ] - 2s - loss: 0.0034 - acc: 0.0030
- val_loss: 0.0296 - val_acc: 0.0061
Epoch 65/300
- val_loss: 0.0296 - val_acc: 0.0061
Epoch 66/300
332/332 [================ ] - 2s - loss: 0.0033 - acc: 0.0030
- val_loss: 0.0305 - val_acc: 0.0061
Epoch 67/300
- val_loss: 0.0305 - val_acc: 0.0061
Epoch 68/300
332/332 [================ ] - 2s - loss: 0.0035 - acc: 0.0030
- val_loss: 0.0291 - val_acc: 0.0061
Epoch 69/300
332/332 [================ ] - 2s - loss: 0.0030 - acc: 0.0030
- val_loss: 0.0285 - val_acc: 0.0061
Epoch 70/300
- val_loss: 0.0295 - val_acc: 0.0061
Epoch 71/300
332/332 [================ ] - 2s - loss: 0.0033 - acc: 0.0030
- val_loss: 0.0296 - val_acc: 0.0061
Epoch 72/300
332/332 [================ ] - 2s - loss: 0.0035 - acc: 0.0030
- val_loss: 0.0289 - val_acc: 0.0061
Epoch 73/300
- val_loss: 0.0286 - val_acc: 0.0061
Epoch 74/300
332/332 [================ ] - 2s - loss: 0.0031 - acc: 0.0030
- val_loss: 0.0284 - val_acc: 0.0061
Epoch 75/300
- val_loss: 0.0281 - val_acc: 0.0061
Epoch 76/300
332/332 [=============== ] - 2s - loss: 0.0032 - acc: 0.0030
- val_loss: 0.0281 - val_acc: 0.0061
Epoch 77/300
- val_loss: 0.0292 - val_acc: 0.0061
Epoch 78/300
332/332 [=============== ] - 2s - loss: 0.0034 - acc: 0.0030
- val loss: 0.0289 - val acc: 0.0061
Epoch 79/300
- val_loss: 0.0276 - val_acc: 0.0061
Epoch 80/300
332/332 [=============== ] - 2s - loss: 0.0032 - acc: 0.0030
- val_loss: 0.0273 - val_acc: 0.0061
Epoch 81/300
332/332 [============== ] - 2s - loss: 0.0029 - acc: 0.0030
- val_loss: 0.0279 - val_acc: 0.0061
```

```
Epoch 82/300
332/332 [================= ] - 3s - loss: 0.0028 - acc: 0.0030
- val loss: 0.0286 - val acc: 0.0061
Epoch 83/300
332/332 [=============== ] - 3s - loss: 0.0034 - acc: 0.0030
- val_loss: 0.0283 - val_acc: 0.0061
Epoch 84/300
332/332 [================ ] - 2s - loss: 0.0027 - acc: 0.0030
- val_loss: 0.0284 - val_acc: 0.0061
Epoch 85/300
- val_loss: 0.0289 - val_acc: 0.0061
Epoch 86/300
- val_loss: 0.0277 - val_acc: 0.0061
Epoch 87/300
332/332 [================ ] - 2s - loss: 0.0029 - acc: 0.0030
- val_loss: 0.0262 - val_acc: 0.0061
Epoch 88/300
332/332 [================ ] - 2s - loss: 0.0028 - acc: 0.0030
- val_loss: 0.0263 - val_acc: 0.0061
Epoch 89/300
332/332 [================ ] - 2s - loss: 0.0031 - acc: 0.0030
- val_loss: 0.0266 - val_acc: 0.0061
Epoch 90/300
- val loss: 0.0264 - val acc: 0.0061
Epoch 91/300
332/332 [================= ] - 2s - loss: 0.0034 - acc: 0.0030
- val_loss: 0.0274 - val_acc: 0.0061
Epoch 92/300
- val_loss: 0.0275 - val_acc: 0.0061
Epoch 93/300
332/332 [================ ] - 2s - loss: 0.0030 - acc: 0.0030
- val_loss: 0.0261 - val_acc: 0.0061
Epoch 94/300
332/332 [================ ] - 2s - loss: 0.0034 - acc: 0.0030
- val_loss: 0.0271 - val_acc: 0.0061
Epoch 95/300
332/332 [================ ] - 2s - loss: 0.0030 - acc: 0.0030
- val_loss: 0.0293 - val_acc: 0.0061
Epoch 96/300
332/332 [================ ] - 2s - loss: 0.0030 - acc: 0.0030
- val loss: 0.0291 - val acc: 0.0061
Epoch 97/300
- val_loss: 0.0276 - val_acc: 0.0061
Epoch 98/300
- val_loss: 0.0279 - val_acc: 0.0061
Epoch 99/300
- val_loss: 0.0277 - val_acc: 0.0061
Epoch 100/300
- val_loss: 0.0262 - val_acc: 0.0061
Epoch 101/300
- val_loss: 0.0260 - val_acc: 0.0061
Epoch 102/300
```

```
- val_loss: 0.0262 - val_acc: 0.0061
Epoch 103/300
- val loss: 0.0250 - val acc: 0.0061
Epoch 104/300
- val_loss: 0.0243 - val_acc: 0.0061
Epoch 105/300
332/332 [================ ] - 2s - loss: 0.0027 - acc: 0.0030
- val_loss: 0.0253 - val_acc: 0.0061
Epoch 106/300
- val_loss: 0.0271 - val_acc: 0.0061
Epoch 107/300
332/332 [================= ] - 2s - loss: 0.0029 - acc: 0.0030
- val_loss: 0.0268 - val_acc: 0.0061
Epoch 108/300
- val_loss: 0.0259 - val_acc: 0.0061
Epoch 109/300
332/332 [================ ] - 2s - loss: 0.0028 - acc: 0.0030
- val_loss: 0.0267 - val_acc: 0.0061
Epoch 110/300
- val_loss: 0.0265 - val_acc: 0.0061
Epoch 111/300
332/332 [================ ] - 2s - loss: 0.0029 - acc: 0.0030
- val_loss: 0.0258 - val_acc: 0.0061
Epoch 112/300
- val_loss: 0.0256 - val_acc: 0.0061
Epoch 113/300
332/332 [================= ] - 2s - loss: 0.0027 - acc: 0.0030
- val_loss: 0.0255 - val_acc: 0.0061
Epoch 114/300
- val_loss: 0.0255 - val_acc: 0.0061
Epoch 115/300
332/332 [================ ] - 2s - loss: 0.0030 - acc: 0.0030
- val_loss: 0.0278 - val_acc: 0.0061
Epoch 116/300
- val_loss: 0.0275 - val_acc: 0.0061
Epoch 117/300
- val_loss: 0.0240 - val_acc: 0.0061
Epoch 118/300
- val_loss: 0.0229 - val_acc: 0.0061
Epoch 119/300
332/332 [=============== ] - 2s - loss: 0.0025 - acc: 0.0030
- val_loss: 0.0237 - val_acc: 0.0061
Epoch 120/300
- val_loss: 0.0252 - val_acc: 0.0061
Epoch 121/300
332/332 [=============== ] - 2s - loss: 0.0028 - acc: 0.0030
- val_loss: 0.0255 - val_acc: 0.0061
Epoch 122/300
```

```
- val_loss: 0.0266 - val_acc: 0.0061
Epoch 123/300
332/332 [============== ] - 2s - loss: 0.0025 - acc: 0.0030
- val_loss: 0.0257 - val_acc: 0.0061
Epoch 124/300
332/332 [================ ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0253 - val_acc: 0.0061
Epoch 125/300
332/332 [================ ] - 2s - loss: 0.0025 - acc: 0.0030
- val_loss: 0.0255 - val_acc: 0.0061
Epoch 126/300
- val_loss: 0.0248 - val_acc: 0.0061
Epoch 127/300
332/332 [================ ] - 2s - loss: 0.0022 - acc: 0.0030
- val_loss: 0.0234 - val_acc: 0.0061
Epoch 128/300
- val_loss: 0.0227 - val_acc: 0.0061
Epoch 129/300
332/332 [================ ] - 2s - loss: 0.0025 - acc: 0.0030
- val_loss: 0.0251 - val_acc: 0.0061
Epoch 130/300
332/332 [================ ] - 2s - loss: 0.0027 - acc: 0.0030
- val_loss: 0.0262 - val_acc: 0.0061
Epoch 131/300
- val_loss: 0.0257 - val_acc: 0.0061
Epoch 132/300
332/332 [================ ] - 2s - loss: 0.0031 - acc: 0.0030
- val_loss: 0.0271 - val_acc: 0.0061
Epoch 133/300
332/332 [================ ] - 2s - loss: 0.0028 - acc: 0.0030
- val_loss: 0.0254 - val_acc: 0.0061
Epoch 134/300
332/332 [================= ] - 2s - loss: 0.0029 - acc: 0.0030
- val_loss: 0.0220 - val_acc: 0.0061
Epoch 135/300
332/332 [================ ] - 2s - loss: 0.0029 - acc: 0.0030
- val_loss: 0.0231 - val_acc: 0.0061
Epoch 136/300
- val_loss: 0.0267 - val_acc: 0.0061
Epoch 137/300
332/332 [=============== ] - 2s - loss: 0.0032 - acc: 0.0030
- val_loss: 0.0250 - val_acc: 0.0061
Epoch 138/300
- val_loss: 0.0218 - val_acc: 0.0061
Epoch 139/300
332/332 [=============== ] - 2s - loss: 0.0028 - acc: 0.0030
- val_loss: 0.0230 - val_acc: 0.0061
Epoch 140/300
- val_loss: 0.0262 - val_acc: 0.0061
Epoch 141/300
- val_loss: 0.0254 - val_acc: 0.0061
Epoch 142/300
- val_loss: 0.0218 - val_acc: 0.0061
```

```
Epoch 143/300
- val loss: 0.0216 - val acc: 0.0061
Epoch 144/300
332/332 [=============== ] - 2s - loss: 0.0028 - acc: 0.0030
- val_loss: 0.0243 - val_acc: 0.0061
Epoch 145/300
332/332 [================ ] - 2s - loss: 0.0025 - acc: 0.0030
- val_loss: 0.0244 - val_acc: 0.0061
Epoch 146/300
- val_loss: 0.0225 - val_acc: 0.0061
Epoch 147/300
- val_loss: 0.0231 - val_acc: 0.0061
Epoch 148/300
332/332 [================ ] - 2s - loss: 0.0034 - acc: 0.0030
- val_loss: 0.0228 - val_acc: 0.0061
Epoch 149/300
332/332 [================ ] - 2s - loss: 0.0027 - acc: 0.0030
- val_loss: 0.0222 - val_acc: 0.0061
Epoch 150/300
332/332 [================ ] - 2s - loss: 0.0030 - acc: 0.0030
- val_loss: 0.0239 - val_acc: 0.0061
Epoch 151/300
- val loss: 0.0250 - val acc: 0.0061
Epoch 152/300
332/332 [================= ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0235 - val_acc: 0.0061
Epoch 153/300
- val_loss: 0.0217 - val_acc: 0.0061
Epoch 154/300
332/332 [================ ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0216 - val_acc: 0.0061
Epoch 155/300
332/332 [================ ] - 2s - loss: 0.0025 - acc: 0.0030
- val_loss: 0.0228 - val_acc: 0.0061
Epoch 156/300
332/332 [================ ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0238 - val_acc: 0.0061
Epoch 157/300
332/332 [================ ] - 2s - loss: 0.0025 - acc: 0.0030
- val loss: 0.0241 - val acc: 0.0061
Epoch 158/300
- val_loss: 0.0227 - val_acc: 0.0061
Epoch 159/300
- val_loss: 0.0212 - val_acc: 0.0061
Epoch 160/300
- val_loss: 0.0216 - val_acc: 0.0061
Epoch 161/300
- val_loss: 0.0233 - val_acc: 0.0061
Epoch 162/300
- val_loss: 0.0228 - val_acc: 0.0061
Epoch 163/300
```

```
- val_loss: 0.0219 - val_acc: 0.0061
Epoch 164/300
- val loss: 0.0221 - val acc: 0.0061
Epoch 165/300
- val_loss: 0.0223 - val_acc: 0.0061
Epoch 166/300
332/332 [================ ] - 2s - loss: 0.0025 - acc: 0.0030
- val_loss: 0.0227 - val_acc: 0.0061
Epoch 167/300
- val_loss: 0.0230 - val_acc: 0.0061
Epoch 168/300
- val_loss: 0.0226 - val_acc: 0.0061
Epoch 169/300
- val_loss: 0.0205 - val_acc: 0.0061
Epoch 170/300
332/332 [================ ] - 2s - loss: 0.0026 - acc: 0.0030
- val_loss: 0.0201 - val_acc: 0.0061
Epoch 171/300
- val_loss: 0.0226 - val_acc: 0.0061
Epoch 172/300
332/332 [================ ] - 2s - loss: 0.0026 - acc: 0.0030
- val_loss: 0.0232 - val_acc: 0.0061
Epoch 173/300
- val_loss: 0.0227 - val_acc: 0.0061
Epoch 174/300
- val_loss: 0.0242 - val_acc: 0.0061
Epoch 175/300
- val_loss: 0.0254 - val_acc: 0.0061
Epoch 176/300
332/332 [================ ] - 2s - loss: 0.0023 - acc: 0.0030
- val_loss: 0.0234 - val_acc: 0.0061
Epoch 177/300
- val_loss: 0.0206 - val_acc: 0.0061
Epoch 178/300
- val_loss: 0.0194 - val_acc: 0.0061
Epoch 179/300
- val_loss: 0.0195 - val_acc: 0.0061
Epoch 180/300
332/332 [=============== ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0213 - val_acc: 0.0061
Epoch 181/300
- val_loss: 0.0218 - val_acc: 0.0061
Epoch 182/300
332/332 [=============== ] - 2s - loss: 0.0020 - acc: 0.0030
- val_loss: 0.0213 - val_acc: 0.0061
Epoch 183/300
```

```
- val_loss: 0.0208 - val_acc: 0.0061
Epoch 184/300
332/332 [============== ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0222 - val_acc: 0.0061
Epoch 185/300
332/332 [================ ] - 2s - loss: 0.0025 - acc: 0.0030
- val_loss: 0.0232 - val_acc: 0.0061
Epoch 186/300
332/332 [================ ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0220 - val_acc: 0.0061
Epoch 187/300
- val_loss: 0.0223 - val_acc: 0.0061
Epoch 188/300
332/332 [================ ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0225 - val_acc: 0.0061
Epoch 189/300
- val_loss: 0.0220 - val_acc: 0.0061
Epoch 190/300
332/332 [================ ] - 2s - loss: 0.0022 - acc: 0.0030
- val_loss: 0.0215 - val_acc: 0.0061
Epoch 191/300
332/332 [================ ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0221 - val_acc: 0.0061
Epoch 192/300
- val_loss: 0.0216 - val_acc: 0.0061
Epoch 193/300
332/332 [================ ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0197 - val_acc: 0.0061
Epoch 194/300
332/332 [================ ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0205 - val_acc: 0.0061
Epoch 195/300
332/332 [================= ] - 2s - loss: 0.0022 - acc: 0.0030
- val_loss: 0.0210 - val_acc: 0.0061
Epoch 196/300
332/332 [================ ] - 2s - loss: 0.0026 - acc: 0.0030
- val_loss: 0.0195 - val_acc: 0.0061
Epoch 197/300
- val_loss: 0.0203 - val_acc: 0.0061
Epoch 198/300
332/332 [=============== ] - 2s - loss: 0.0023 - acc: 0.0030
- val_loss: 0.0221 - val_acc: 0.0061
Epoch 199/300
- val_loss: 0.0209 - val_acc: 0.0061
Epoch 200/300
332/332 [=============== ] - 2s - loss: 0.0022 - acc: 0.0030
- val_loss: 0.0206 - val_acc: 0.0061
Epoch 201/300
- val_loss: 0.0229 - val_acc: 0.0061
Epoch 202/300
- val_loss: 0.0224 - val_acc: 0.0061
Epoch 203/300
- val_loss: 0.0202 - val_acc: 0.0061
```

```
Epoch 204/300
- val loss: 0.0204 - val acc: 0.0061
Epoch 205/300
332/332 [=============== ] - 2s - loss: 0.0023 - acc: 0.0030
- val_loss: 0.0210 - val_acc: 0.0061
Epoch 206/300
332/332 [================ ] - 2s - loss: 0.0026 - acc: 0.0030
- val_loss: 0.0191 - val_acc: 0.0061
Epoch 207/300
- val_loss: 0.0182 - val_acc: 0.0061
Epoch 208/300
- val_loss: 0.0209 - val_acc: 0.0061
Epoch 209/300
332/332 [================ ] - 2s - loss: 0.0026 - acc: 0.0030
- val_loss: 0.0231 - val_acc: 0.0061
Epoch 210/300
332/332 [================ ] - 2s - loss: 0.0025 - acc: 0.0030
- val_loss: 0.0210 - val_acc: 0.0061
Epoch 211/300
332/332 [================ ] - 2s - loss: 0.0023 - acc: 0.0030
- val_loss: 0.0203 - val_acc: 0.0061
Epoch 212/300
- val loss: 0.0223 - val acc: 0.0061
Epoch 213/300
- val_loss: 0.0225 - val_acc: 0.0061
Epoch 214/300
- val_loss: 0.0210 - val_acc: 0.0061
Epoch 215/300
332/332 [================ ] - 2s - loss: 0.0022 - acc: 0.0030
- val_loss: 0.0199 - val_acc: 0.0061
Epoch 216/300
332/332 [================ ] - 2s - loss: 0.0029 - acc: 0.0030
- val_loss: 0.0210 - val_acc: 0.0061
Epoch 217/300
332/332 [================ ] - 2s - loss: 0.0027 - acc: 0.0030
- val_loss: 0.0201 - val_acc: 0.0061
Epoch 218/300
- val loss: 0.0187 - val acc: 0.0061
Epoch 219/300
- val_loss: 0.0215 - val_acc: 0.0061
Epoch 220/300
- val_loss: 0.0235 - val_acc: 0.0061
Epoch 221/300
- val_loss: 0.0210 - val_acc: 0.0061
Epoch 222/300
- val_loss: 0.0190 - val_acc: 0.0061
Epoch 223/300
- val_loss: 0.0200 - val_acc: 0.0061
Epoch 224/300
```

```
- val_loss: 0.0220 - val_acc: 0.0061
Epoch 225/300
- val loss: 0.0219 - val acc: 0.0061
Epoch 226/300
- val_loss: 0.0202 - val_acc: 0.0061
Epoch 227/300
332/332 [================ ] - 2s - loss: 0.0023 - acc: 0.0030
- val_loss: 0.0208 - val_acc: 0.0061
Epoch 228/300
- val_loss: 0.0217 - val_acc: 0.0061
Epoch 229/300
- val_loss: 0.0214 - val_acc: 0.0061
Epoch 230/300
- val_loss: 0.0202 - val_acc: 0.0061
Epoch 231/300
332/332 [================ ] - 2s - loss: 0.0023 - acc: 0.0030
- val_loss: 0.0190 - val_acc: 0.0061
Epoch 232/300
- val_loss: 0.0174 - val_acc: 0.0061
Epoch 233/300
332/332 [================ ] - 2s - loss: 0.0022 - acc: 0.0030
- val_loss: 0.0174 - val_acc: 0.0061
Epoch 234/300
- val_loss: 0.0198 - val_acc: 0.0061
Epoch 235/300
- val_loss: 0.0232 - val_acc: 0.0061
Epoch 236/300
- val_loss: 0.0222 - val_acc: 0.0061
Epoch 237/300
332/332 [================ ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0197 - val_acc: 0.0061
Epoch 238/300
- val_loss: 0.0202 - val_acc: 0.0061
Epoch 239/300
- val_loss: 0.0222 - val_acc: 0.0061
Epoch 240/300
- val_loss: 0.0222 - val_acc: 0.0061
Epoch 241/300
332/332 [=============== ] - 2s - loss: 0.0023 - acc: 0.0030
- val_loss: 0.0198 - val_acc: 0.0061
Epoch 242/300
- val_loss: 0.0187 - val_acc: 0.0061
Epoch 243/300
332/332 [=============== ] - 2s - loss: 0.0021 - acc: 0.0030
- val_loss: 0.0190 - val_acc: 0.0061
Epoch 244/300
```

```
- val_loss: 0.0200 - val_acc: 0.0061
Epoch 245/300
332/332 [============== ] - 2s - loss: 0.0021 - acc: 0.0030
- val_loss: 0.0209 - val_acc: 0.0061
Epoch 246/300
332/332 [================ ] - 2s - loss: 0.0021 - acc: 0.0030
- val_loss: 0.0208 - val_acc: 0.0061
Epoch 247/300
332/332 [================ ] - 2s - loss: 0.0023 - acc: 0.0030
- val_loss: 0.0188 - val_acc: 0.0061
Epoch 248/300
- val_loss: 0.0184 - val_acc: 0.0061
Epoch 249/300
332/332 [================ ] - 2s - loss: 0.0023 - acc: 0.0030
- val_loss: 0.0202 - val_acc: 0.0061
Epoch 250/300
- val_loss: 0.0210 - val_acc: 0.0061
Epoch 251/300
332/332 [================ ] - 2s - loss: 0.0019 - acc: 0.0030
- val_loss: 0.0206 - val_acc: 0.0061
Epoch 252/300
332/332 [================ ] - 2s - loss: 0.0021 - acc: 0.0030
- val_loss: 0.0206 - val_acc: 0.0061
Epoch 253/300
- val_loss: 0.0211 - val_acc: 0.0061
Epoch 254/300
332/332 [================ ] - 2s - loss: 0.0019 - acc: 0.0030
- val_loss: 0.0211 - val_acc: 0.0061
Epoch 255/300
332/332 [================ ] - 3s - loss: 0.0021 - acc: 0.0030
- val_loss: 0.0214 - val_acc: 0.0061
Epoch 256/300
- val_loss: 0.0208 - val_acc: 0.0061
Epoch 257/300
332/332 [================ ] - 2s - loss: 0.0018 - acc: 0.0030
- val_loss: 0.0203 - val_acc: 0.0061
Epoch 258/300
- val_loss: 0.0200 - val_acc: 0.0061
Epoch 259/300
332/332 [=============== ] - 2s - loss: 0.0022 - acc: 0.0030
- val_loss: 0.0209 - val_acc: 0.0061
Epoch 260/300
- val_loss: 0.0218 - val_acc: 0.0061
Epoch 261/300
332/332 [=============== ] - 2s - loss: 0.0022 - acc: 0.0030
- val_loss: 0.0208 - val_acc: 0.0061
Epoch 262/300
- val_loss: 0.0198 - val_acc: 0.0061
Epoch 263/300
- val_loss: 0.0207 - val_acc: 0.0061
Epoch 264/300
- val_loss: 0.0212 - val_acc: 0.0061
```

```
Epoch 265/300
332/332 [============== ] - 2s - loss: 0.0019 - acc: 0.0030
- val loss: 0.0207 - val acc: 0.0061
Epoch 266/300
332/332 [=============== ] - 2s - loss: 0.0023 - acc: 0.0030
- val_loss: 0.0185 - val_acc: 0.0061
Epoch 267/300
332/332 [================ ] - 2s - loss: 0.0019 - acc: 0.0030
- val_loss: 0.0185 - val_acc: 0.0061
Epoch 268/300
- val_loss: 0.0200 - val_acc: 0.0061
Epoch 269/300
- val_loss: 0.0212 - val_acc: 0.0061
Epoch 270/300
332/332 [================ ] - 2s - loss: 0.0021 - acc: 0.0030
- val_loss: 0.0211 - val_acc: 0.0061
Epoch 271/300
332/332 [================ ] - 2s - loss: 0.0020 - acc: 0.0030
- val_loss: 0.0203 - val_acc: 0.0061
Epoch 272/300
332/332 [================ ] - 2s - loss: 0.0023 - acc: 0.0030
- val_loss: 0.0195 - val_acc: 0.0061
Epoch 273/300
- val loss: 0.0199 - val acc: 0.0061
Epoch 274/300
332/332 [================= ] - 2s - loss: 0.0022 - acc: 0.0030
- val_loss: 0.0204 - val_acc: 0.0061
Epoch 275/300
- val_loss: 0.0206 - val_acc: 0.0061
Epoch 276/300
332/332 [================ ] - 2s - loss: 0.0020 - acc: 0.0030
- val_loss: 0.0199 - val_acc: 0.0061
Epoch 277/300
332/332 [================ ] - 2s - loss: 0.0022 - acc: 0.0030
- val_loss: 0.0192 - val_acc: 0.0061
Epoch 278/300
332/332 [================ ] - 2s - loss: 0.0023 - acc: 0.0030
- val_loss: 0.0216 - val_acc: 0.0061
Epoch 279/300
- val loss: 0.0229 - val acc: 0.0061
Epoch 280/300
- val_loss: 0.0211 - val_acc: 0.0061
Epoch 281/300
- val_loss: 0.0185 - val_acc: 0.0061
Epoch 282/300
- val_loss: 0.0191 - val_acc: 0.0061
Epoch 283/300
- val_loss: 0.0181 - val_acc: 0.0061
Epoch 284/300
- val_loss: 0.0180 - val_acc: 0.0061
Epoch 285/300
```

```
- val_loss: 0.0222 - val_acc: 0.0061
Epoch 286/300
- val_loss: 0.0274 - val_acc: 0.0061
Epoch 287/300
- val_loss: 0.0242 - val_acc: 0.0061
Epoch 288/300
332/332 [================ ] - 2s - loss: 0.0027 - acc: 0.0030
- val_loss: 0.0192 - val_acc: 0.0061
Epoch 289/300
- val_loss: 0.0191 - val_acc: 0.0061
Epoch 290/300
332/332 [============= ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0227 - val_acc: 0.0061
Epoch 291/300
- val_loss: 0.0236 - val_acc: 0.0061
Epoch 292/300
332/332 [================ ] - 2s - loss: 0.0026 - acc: 0.0030
- val_loss: 0.0197 - val_acc: 0.0061
Epoch 293/300
- val_loss: 0.0181 - val_acc: 0.0061
Epoch 294/300
332/332 [================ ] - 2s - loss: 0.0021 - acc: 0.0030
- val_loss: 0.0196 - val_acc: 0.0061
Epoch 295/300
- val_loss: 0.0202 - val_acc: 0.0061
Epoch 296/300
- val_loss: 0.0188 - val_acc: 0.0061
Epoch 297/300
- val_loss: 0.0176 - val_acc: 0.0061
Epoch 298/300
332/332 [================ ] - 2s - loss: 0.0024 - acc: 0.0030
- val_loss: 0.0183 - val_acc: 0.0061
Epoch 299/300
332/332 [================= ] - 2s - loss: 0.0019 - acc: 0.0030
- val_loss: 0.0194 - val_acc: 0.0061
Epoch 300/300
- val_loss: 0.0201 - val_acc: 0.0061
Out[67]:
```

<keras.callbacks.History at 0xe99f4fa518>

Result

In [68]:

```
def model_score(model, X_train, y_train, X_test, y_test):
    trainScore = model.evaluate(X_train, y_train, verbose=0)
    print('Train Score: %.5f MSE (%.2f RMSE)' % (trainScore[0],
math.sqrt(trainScore[0])))

    testScore = model.evaluate(X_test, y_test, verbose=0)
    print('Test Score: %.5f MSE (%.2f RMSE)' % (testScore[0], math.sqrt(testScore[0])))
    return trainScore[0], testScore[0]

model_score(model, X_train, y_train, X_test, y_test)

Train Score: 0.00735 MSE (0.09 RMSE)
Test Score: 0.01370 MSE (0.12 RMSE)

Out[68]:
(0.0073529773301655244, 0.013702700447831903)
```

Prediction vs Real Results

In [69]:

```
def percentage_difference(model, X_test, y_test):
    percentage_diff=[]

p = model.predict(X_test)
    for u in range(len(y_test)): # for each data index in test data
        pr = p[u][0] # pr = prediction on day u

        percentage_diff.append((pr-y_test[u]/pr)*100)
    return p
```

```
In [70]:
```

```
p = percentage_difference(model, X_test, y_test)
```

In [71]:

```
def denormalize(normalized value):
    Return a dataframe of that stock and normalize all the values.
    (Optional: create moving average)
    df = quandl.get("HKEX/00293", authtoken="ETPyWitxJn3wiSRvxUjk")
    df.drop(['Net Change','P/E(x)', 'Bid', 'Ask', 'Lot Size', 'Turnover (\'000)'], 1, i
nplace=True)
     df.set_index('Date', inplace=True)
    # Renaming all the columns so that we can use the old version code
    df.rename(columns={'Nominal Price': 'Open','Share Volume (\'000)': 'Volume', 'Previ
ous Close': 'Adj Close', 'Change (%)': 'Pct'}, inplace=True)
    df.dropna(inplace=True)
    df = df['Adj Close'].values.reshape(-1,1)
    normalized_value = normalized_value.reshape(-1,1)
    #return df.shape, p.shape
    min_max_scaler = preprocessing.MinMaxScaler()
    a = min_max_scaler.fit_transform(df)
    new = min max scaler.inverse transform(normalized value)
    return new
```

In [73]:

```
def plot_result(stock_name, normalized_value_p, normalized_value_y_test):
    newp = denormalize(normalized_value_p)
    newy_test = denormalize(normalized_value_y_test)
    plt2.plot(newp, color='red', label='Prediction')
    plt2.plot(newy_test,color='blue', label='Actual')
    plt2.legend(loc='best')
    plt2.title('The test result for {}'.format(stock_name))
    plt2.xlabel('Days')
    plt2.ylabel('Adjusted Close')
    plt2.show()
```