Thesis 20Dec

December 20, 2024

```
[]: # Install necessary libraries
     !pip install yfinance ccxt pandas numpy matplotlib pandas_ta tensorflow
    Requirement already satisfied: yfinance in /usr/local/lib/python3.10/dist-
    packages (0.2.50)
    Collecting ccxt
      Downloading ccxt-4.4.41-py2.py3-none-any.whl.metadata (116 kB)
                                116.4/116.4
    kB 3.3 MB/s eta 0:00:00
    Requirement already satisfied: pandas in /usr/local/lib/python3.10/dist-
    packages (2.2.2)
    Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages
    (1.26.4)
    Requirement already satisfied: matplotlib in /usr/local/lib/python3.10/dist-
    packages (3.8.0)
    Collecting pandas_ta
      Downloading pandas_ta-0.3.14b.tar.gz (115 kB)
                                115.1/115.1
    kB 8.3 MB/s eta 0:00:00
      Preparing metadata (setup.py) ... done
    Requirement already satisfied: tensorflow in /usr/local/lib/python3.10/dist-
    packages (2.17.1)
    Requirement already satisfied: requests>=2.31 in /usr/local/lib/python3.10/dist-
    packages (from yfinance) (2.32.3)
    Requirement already satisfied: multitasking>=0.0.7 in
    /usr/local/lib/python3.10/dist-packages (from yfinance) (0.0.11)
    Requirement already satisfied: lxml>=4.9.1 in /usr/local/lib/python3.10/dist-
    packages (from yfinance) (5.3.0)
    Requirement already satisfied: platformdirs>=2.0.0 in
    /usr/local/lib/python3.10/dist-packages (from yfinance) (4.3.6)
    Requirement already satisfied: pytz>=2022.5 in /usr/local/lib/python3.10/dist-
    packages (from yfinance) (2024.2)
    Requirement already satisfied: frozendict>=2.3.4 in
    /usr/local/lib/python3.10/dist-packages (from yfinance) (2.4.6)
    Requirement already satisfied: peewee>=3.16.2 in /usr/local/lib/python3.10/dist-
    packages (from yfinance) (3.17.8)
    Requirement already satisfied: beautifulsoup4>=4.11.1 in
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/usr/local/lib/python3.10/dist-packages (from yfinance) (4.12.3)
Requirement already satisfied: html5lib>=1.1 in /usr/local/lib/python3.10/dist-
packages (from yfinance) (1.1)
Requirement already satisfied: setuptools>=60.9.0 in
/usr/local/lib/python3.10/dist-packages (from ccxt) (75.1.0)
Requirement already satisfied: certifi>=2018.1.18 in
/usr/local/lib/python3.10/dist-packages (from ccxt) (2024.12.14)
Requirement already satisfied: cryptography>=2.6.1 in
/usr/local/lib/python3.10/dist-packages (from ccxt) (43.0.3)
Requirement already satisfied: typing-extensions>=4.4.0 in
/usr/local/lib/python3.10/dist-packages (from ccxt) (4.12.2)
Collecting aiohttp<=3.10.11 (from ccxt)</pre>
  Downloading aiohttp-3.10.11-cp310-cp310-
manylinux 2 17 x86 64.manylinux 2014 x86 64.whl.metadata (7.7 kB)
Collecting aiodns>=1.1.1 (from ccxt)
 Downloading aiodns-3.2.0-py3-none-any.whl.metadata (4.0 kB)
Requirement already satisfied: yarl>=1.7.2 in /usr/local/lib/python3.10/dist-
packages (from ccxt) (1.18.3)
Requirement already satisfied: python-dateutil>=2.8.2 in
/usr/local/lib/python3.10/dist-packages (from pandas) (2.8.2)
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.10/dist-
packages (from pandas) (2024.2)
Requirement already satisfied: contourpy>=1.0.1 in
/usr/local/lib/python3.10/dist-packages (from matplotlib) (1.3.1)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/dist-
packages (from matplotlib) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in
/usr/local/lib/python3.10/dist-packages (from matplotlib) (4.55.3)
Requirement already satisfied: kiwisolver>=1.0.1 in
/usr/local/lib/python3.10/dist-packages (from matplotlib) (1.4.7)
Requirement already satisfied: packaging>=20.0 in
/usr/local/lib/python3.10/dist-packages (from matplotlib) (24.2)
Requirement already satisfied: pillow>=6.2.0 in /usr/local/lib/python3.10/dist-
packages (from matplotlib) (11.0.0)
Requirement already satisfied: pyparsing>=2.3.1 in
/usr/local/lib/python3.10/dist-packages (from matplotlib) (3.2.0)
Requirement already satisfied: absl-py>=1.0.0 in /usr/local/lib/python3.10/dist-
packages (from tensorflow) (1.4.0)
Requirement already satisfied: astunparse>=1.6.0 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (1.6.3)
Requirement already satisfied: flatbuffers>=24.3.25 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (24.3.25)
Requirement already satisfied: gast!=0.5.0,!=0.5.1,!=0.5.2,>=0.2.1 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (0.6.0)
Requirement already satisfied: google-pasta>=0.1.1 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (0.2.0)
Requirement already satisfied: h5py>=3.10.0 in /usr/local/lib/python3.10/dist-
packages (from tensorflow) (3.12.1)
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Requirement already satisfied: libclang>=13.0.0 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (18.1.1)
Requirement already satisfied: ml-dtypes<0.5.0,>=0.3.1 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (0.4.1)
Requirement already satisfied: opt-einsum>=2.3.2 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (3.4.0)
Requirement already satisfied:
protobuf!=4.21.0,!=4.21.1,!=4.21.2,!=4.21.3,!=4.21.4,!=4.21.5,<5.0.0dev,>=3.20.3
in /usr/local/lib/python3.10/dist-packages (from tensorflow) (4.25.5)
Requirement already satisfied: six>=1.12.0 in /usr/local/lib/python3.10/dist-
packages (from tensorflow) (1.17.0)
Requirement already satisfied: termcolor>=1.1.0 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (2.5.0)
Requirement already satisfied: wrapt>=1.11.0 in /usr/local/lib/python3.10/dist-
packages (from tensorflow) (1.17.0)
Requirement already satisfied: grpcio<2.0,>=1.24.3 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (1.68.1)
Requirement already satisfied: tensorboard<2.18,>=2.17 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (2.17.1)
Requirement already satisfied: keras>=3.2.0 in /usr/local/lib/python3.10/dist-
packages (from tensorflow) (3.5.0)
Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in
/usr/local/lib/python3.10/dist-packages (from tensorflow) (0.37.1)
Collecting pycares>=4.0.0 (from aiodns>=1.1.1->ccxt)
 Downloading pycares-4.5.0-cp310-cp310-
manylinux_2_17_x86_64.manylinux2014_x86_64.whl.metadata (4.1 kB)
Requirement already satisfied: aiohappyeyeballs>=2.3.0 in
/usr/local/lib/python3.10/dist-packages (from aiohttp<=3.10.11->ccxt) (2.4.4)
Requirement already satisfied: aiosignal>=1.1.2 in
/usr/local/lib/python3.10/dist-packages (from aiohttp<=3.10.11->ccxt) (1.3.2)
Requirement already satisfied: attrs>=17.3.0 in /usr/local/lib/python3.10/dist-
packages (from aiohttp<=3.10.11->ccxt) (24.3.0)
Requirement already satisfied: frozenlist>=1.1.1 in
/usr/local/lib/python3.10/dist-packages (from aiohttp<=3.10.11->ccxt) (1.5.0)
Requirement already satisfied: multidict<7.0,>=4.5 in
/usr/local/lib/python3.10/dist-packages (from aiohttp<=3.10.11->ccxt) (6.1.0)
Requirement already satisfied: async-timeout<6.0,>=4.0 in
/usr/local/lib/python3.10/dist-packages (from aiohttp<=3.10.11->ccxt) (4.0.3)
Requirement already satisfied: wheel<1.0,>=0.23.0 in
/usr/local/lib/python3.10/dist-packages (from astunparse>=1.6.0->tensorflow)
(0.45.1)
Requirement already satisfied: soupsieve>1.2 in /usr/local/lib/python3.10/dist-
packages (from beautifulsoup4>=4.11.1->yfinance) (2.6)
Requirement already satisfied: cffi>=1.12 in /usr/local/lib/python3.10/dist-
packages (from cryptography>=2.6.1->ccxt) (1.17.1)
Requirement already satisfied: webencodings in /usr/local/lib/python3.10/dist-
packages (from html5lib>=1.1->yfinance) (0.5.1)
Requirement already satisfied: rich in /usr/local/lib/python3.10/dist-packages
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(from keras>=3.2.0->tensorflow) (13.9.4)
Requirement already satisfied: namex in /usr/local/lib/python3.10/dist-packages
(from keras>=3.2.0->tensorflow) (0.0.8)
Requirement already satisfied: optree in /usr/local/lib/python3.10/dist-packages
(from keras>=3.2.0->tensorflow) (0.13.1)
Requirement already satisfied: charset-normalizer<4,>=2 in
/usr/local/lib/python3.10/dist-packages (from requests>=2.31->yfinance) (3.4.0)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-
packages (from requests>=2.31->yfinance) (3.10)
Requirement already satisfied: urllib3<3,>=1.21.1 in
/usr/local/lib/python3.10/dist-packages (from requests>=2.31->yfinance) (2.2.3)
Requirement already satisfied: markdown>=2.6.8 in
/usr/local/lib/python3.10/dist-packages (from
tensorboard<2.18,>=2.17->tensorflow) (3.7)
Requirement already satisfied: tensorboard-data-server<0.8.0,>=0.7.0 in
/usr/local/lib/python3.10/dist-packages (from
tensorboard<2.18,>=2.17->tensorflow) (0.7.2)
Requirement already satisfied: werkzeug>=1.0.1 in
/usr/local/lib/python3.10/dist-packages (from
tensorboard<2.18,>=2.17->tensorflow) (3.1.3)
Requirement already satisfied: propcache>=0.2.0 in
/usr/local/lib/python3.10/dist-packages (from yarl>=1.7.2->ccxt) (0.2.1)
Requirement already satisfied: pycparser in /usr/local/lib/python3.10/dist-
packages (from cffi>=1.12->cryptography>=2.6.1->ccxt) (2.22)
Requirement already satisfied: MarkupSafe>=2.1.1 in
/usr/local/lib/python3.10/dist-packages (from
werkzeug>=1.0.1->tensorboard<2.18,>=2.17->tensorflow) (3.0.2)
Requirement already satisfied: markdown-it-py>=2.2.0 in
/usr/local/lib/python3.10/dist-packages (from rich->keras>=3.2.0->tensorflow)
(3.0.0)
Requirement already satisfied: pygments<3.0.0,>=2.13.0 in
/usr/local/lib/python3.10/dist-packages (from rich->keras>=3.2.0->tensorflow)
(2.18.0)
Requirement already satisfied: mdurl~=0.1 in /usr/local/lib/python3.10/dist-
packages (from markdown-it-py>=2.2.0->rich->keras>=3.2.0->tensorflow) (0.1.2)
Downloading ccxt-4.4.41-py2.py3-none-any.whl (5.6 MB)
                         5.6/5.6 MB
49.7 MB/s eta 0:00:00
Downloading aiodns-3.2.0-py3-none-any.whl (5.7 kB)
Downloading
aiohttp-3.10.11-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (1.2
MB)
                         1.2/1.2 MB
41.2 MB/s eta 0:00:00
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pycares-4.5.0-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (288
kB)
                         288.6/288.6 kB
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18.9 MB/s eta 0:00:00
    Building wheels for collected packages: pandas_ta
      Building wheel for pandas_ta (setup.py) ... done
      Created wheel for pandas_ta: filename=pandas_ta-0.3.14b0-py3-none-any.whl
    size=218909
    \verb|sha| 256 = \verb|a83a| 489d38e241b04fe4fdc7a| 72117554c3fdcad5b892e87ef61461abd66f911|
      Stored in directory: /root/.cache/pip/wheels/69/00/ac/f7fa862c34b0e2ef32017510
    0c233377b4c558944f12474cf0
    Successfully built pandas ta
    Installing collected packages: pycares, pandas_ta, aiohttp, aiodns, ccxt
      Attempting uninstall: aiohttp
        Found existing installation: aiohttp 3.11.10
        Uninstalling aiohttp-3.11.10:
          Successfully uninstalled aiohttp-3.11.10
    Successfully installed aiodns-3.2.0 aiohttp-3.10.11 ccxt-4.4.41
    pandas_ta-0.3.14b0 pycares-4.5.0
[]: from google.colab import drive
     drive.mount('/content/drive')
[]: # Import necessary libraries
     import yfinance as yf
     import ccxt
     import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     from sklearn.preprocessing import MinMaxScaler
     from tensorflow.keras.models import Sequential
     from tensorflow.keras.layers import LSTM, Dense, Dropout
     from tensorflow.keras.callbacks import EarlyStopping
     import pandas_ta as ta
[]: # Function to fetch stock data using yfinance
     def fetch stock_data(ticker, start_date='2010-01-01', end_date=None):
         stock = yf.Ticker(ticker)
         df = stock.history(start=start_date, end=end_date)
         df = df[['Open', 'High', 'Low', 'Close', 'Volume']].reset_index()
         df.columns = ['timestamp', 'open', 'high', 'low', 'close', 'volume']
         return df
     # Fetch data for Apple and Tesla
     apple_data = fetch_stock_data('AAPL', '2020-01-01')
     tesla_data = fetch_stock_data('TSLA', '2020-01-01')
     # Display first few rows
     print("Apple Stock Data:")
     print(apple_data.head())
```

```
print("\nTesla Stock Data:")
    print(tesla_data.head())
    Apple Stock Data:
                     timestamp
                                     open
                                               high
                                                           low
                                                                    close \
    0 2020-01-02 00:00:00-05:00 71.799858
                                          72.856598 71.545372
                                                                72.796005
    1 2020-01-03 00:00:00-05:00 72.020432 72.851761 71.862892
                                                               72.088295
    2 2020-01-06 00:00:00-05:00 71.206077 72.701500 70.954010
                                                                72,662720
    3 2020-01-07 00:00:00-05:00 72.672402 72.929314 72.100410 72.320969
    4 2020-01-08 00:00:00-05:00 72.022843 73.787300 72.022843 73.484337
          volume
    0 135480400
      146322800
    2 118387200
    3 108872000
    4 132079200
    Tesla Stock Data:
                     timestamp
                                     open
                                               high
                                                           low
                                                                    close
    0 2020-01-02 00:00:00-05:00 28.299999
                                          28.713333 28.114000
                                                                28.684000
    1 2020-01-03 00:00:00-05:00 29.366667
                                          30.266666 29.128000
                                                                29.534000
    2 2020-01-06 00:00:00-05:00 29.364668 30.104000 29.333332
                                                                30.102667
    3 2020-01-07 00:00:00-05:00 30.760000
                                          31.441999 30.224001
                                                                31.270666
    4 2020-01-08 00:00:00-05:00 31.580000 33.232666 31.215334 32.809334
          volume
    0 142981500
    1 266677500
    2 151995000
    3 268231500
    4 467164500
[]: # Function to fetch cryptocurrency data using ccxt
    def fetch_crypto_data(symbol='BTC/USD', exchange_name='kraken', timeframe='1d',_
      →limit=1000):
        exchange_class = getattr(ccxt, exchange_name)
        exchange = exchange_class()
        ohlcv = exchange.fetch_ohlcv(symbol, timeframe=timeframe, limit=limit)
        data = pd.DataFrame(ohlcv, columns=['timestamp', 'open', 'high', 'low', |
      data['timestamp'] = pd.to_datetime(data['timestamp'], unit='ms')
        return data
    # Fetch data for Bitcoin and Ethereum
    btc_data = fetch_crypto_data('BTC/USD')
```

```
eth_data = fetch_crypto_data('ETH/USD')
     # Display first few rows
    print("Bitcoin Data:")
    print(btc_data.head())
    print("\nEthereum Data:")
    print(eth_data.head())
    Bitcoin Data:
                                              close
       timestamp
                    open
                             high
                                       low
                                                          volume
    0 2023-01-01 16528.7 16618.7 16490.0 16614.9 1389.076949
    1 2023-01-02 16615.0 16777.7 16550.0 16669.2
                                                      969.461261
    2 2023-01-03 16668.3 16770.2 16599.6 16669.3
                                                     2014.398974
    3 2023-01-04 16669.7 16987.4 16650.1 16848.9
                                                     2316.939638
    4 2023-01-05 16849.3 16873.3 16752.7 16824.9
                                                     1457.987495
    Ethereum Data:
       timestamp
                    open
                             high
                                       low
                                              close
                                                           volume
    0 2023-01-01 1195.00 1203.73 1190.86 1199.76
                                                      9172.167712
    1 2023-01-02 1199.85 1223.95 1192.58 1213.71
                                                      8749.184504
    2 2023-01-03 1213.61 1219.29 1204.07 1213.92 13209.923566
    3 2023-01-04 1214.00 1271.27 1212.25 1256.32
                                                     30445.187644
    4 2023-01-05 1256.39 1259.05 1242.06 1250.50 11128.564124
[]: # Function to add technical indicators
    def add_technical_indicators(data):
        data['SMA_20'] = ta.sma(data['close'], length=20)
        data['EMA 20'] = ta.ema(data['close'], length=20)
        data['RSI_14'] = ta.rsi(data['close'], length=14)
        return data
     # Add indicators to each dataset
    apple_data = add_technical_indicators(apple_data)
    tesla_data = add_technical_indicators(tesla_data)
    btc_data = add_technical_indicators(btc_data)
    eth_data = add_technical_indicators(eth_data)
     # Confirm that indicators are added
    print("Apple Data with Indicators:\n", apple_data.head())
    print("Tesla Data with Indicators:\n", tesla_data.head())
    print("BTC Data with Indicators:\n", btc_data.head())
    print("ETH Data with Indicators:\n", eth_data.head())
    Apple Data with Indicators:
```

open

timestamp

high

low

close \

```
0 2020-01-02 00:00:00-05:00
                              71.799858
                                          72.856598
                                                      71.545372
                                                                  72.796005
1 2020-01-03 00:00:00-05:00
                              72.020432
                                          72.851761
                                                      71.862892
                                                                  72.088295
2 2020-01-06 00:00:00-05:00
                              71.206077
                                          72.701500
                                                      70.954010
                                                                  72.662720
3 2020-01-07 00:00:00-05:00
                              72.672402
                                          72.929314
                                                      72.100410
                                                                  72.320969
4 2020-01-08 00:00:00-05:00
                              72.022843
                                          73.787300
                                                      72.022843
                                                                  73.484337
               SMA_20
                       EMA 20
                                RSI 14
      volume
0
   135480400
                  NaN
                          NaN
                                   NaN
   146322800
                          NaN
1
                  NaN
                                   NaN
2
  118387200
                  NaN
                          NaN
                                   NaN
3
   108872000
                                   NaN
                  NaN
                          NaN
4
   132079200
                  NaN
                          NaN
                                   NaN
Tesla Data with Indicators:
                                                              low
                                                                       close
                    timestamp
                                     open
                                                 high
0 2020-01-02 00:00:00-05:00
                              28.299999
                                          28.713333
                                                      28.114000
                                                                  28.684000
                                                                  29.534000
1 2020-01-03 00:00:00-05:00
                               29.366667
                                          30.266666
                                                      29.128000
2 2020-01-06 00:00:00-05:00
                               29.364668
                                          30.104000
                                                      29.333332
                                                                  30.102667
3 2020-01-07 00:00:00-05:00
                               30.760000
                                                      30.224001
                                          31.441999
                                                                  31.270666
4 2020-01-08 00:00:00-05:00
                               31.580000
                                          33.232666
                                                      31.215334
                                                                  32.809334
      volume
               SMA_20
                       EMA 20
                                RSI 14
0
  142981500
                  NaN
                          NaN
                                   NaN
1
   266677500
                  NaN
                          NaN
                                   NaN
2
                          NaN
                                   NaN
  151995000
                  NaN
3
   268231500
                          NaN
                                   NaN
                  NaN
   467164500
                                   NaN
                  NaN
                          NaN
BTC Data with Indicators:
    timestamp
                   open
                            high
                                       low
                                               close
                                                            volume
                                                                    SMA_20
                                                                             EMA_20
0 2023-01-01
               16528.7
                        16618.7
                                  16490.0
                                            16614.9
                                                     1389.076949
                                                                      NaN
                                                                               NaN
1 2023-01-02
               16615.0
                        16777.7
                                  16550.0
                                           16669.2
                                                      969.461261
                                                                      NaN
                                                                               NaN
2 2023-01-03
               16668.3
                        16770.2
                                  16599.6
                                           16669.3
                                                     2014.398974
                                                                      NaN
                                                                               NaN
3 2023-01-04
               16669.7
                        16987.4
                                  16650.1
                                           16848.9
                                                     2316.939638
                                                                      NaN
                                                                               NaN
4 2023-01-05
               16849.3
                        16873.3
                                  16752.7
                                           16824.9
                                                     1457.987495
                                                                      NaN
                                                                               NaN
   RSI_14
0
      NaN
1
      NaN
2
      NaN
3
      NaN
4
      NaN
ETH Data with Indicators:
    timestamp
                   open
                            high
                                       low
                                               close
                                                             volume
                                                                     SMA_20
0 2023-01-01
              1195.00
                        1203.73
                                  1190.86
                                           1199.76
                                                      9172.167712
                                                                       NaN
1 2023-01-02
              1199.85
                        1223.95
                                  1192.58
                                           1213.71
                                                      8749.184504
                                                                       NaN
2 2023-01-03
               1213.61
                        1219.29
                                  1204.07
                                            1213.92
                                                     13209.923566
                                                                       NaN
3 2023-01-04
               1214.00
                        1271.27
                                  1212.25
                                           1256.32
                                                     30445.187644
                                                                       NaN
4 2023-01-05
              1256.39
                        1259.05
                                  1242.06
                                           1250.50
                                                     11128.564124
                                                                       NaN
```

```
EMA_20 RSI_14
    0
          NaN
                  NaN
    1
          NaN
                  NaN
    2
          NaN
                  NaN
    3
          NaN
                  NaN
    4
          NaN
                  NaN
[]: # Function to preprocess data for LSTM
     def prepare_lstm_data(data, target_col='close', seq_length=60):
         data = data.dropna()
         features = data[['close', 'SMA_20', 'EMA_20', 'RSI_14']]
         scaler = MinMaxScaler()
         features_scaled = scaler.fit_transform(features)
         X, y = [], []
         for i in range(seq_length, len(features_scaled)):
             X.append(features_scaled[i-seq_length:i])
             y.append(features_scaled[i, 0])
         X, y = np.array(X), np.array(y)
         split = int(0.8 * len(X))
         return X[:split], X[split:], y[:split], y[split:], scaler
     # Prepare data for stocks and cryptocurrencies
     X_train_apple, X_test_apple, y_train_apple, y_test_apple, apple_scaler = __
      →prepare_lstm_data(apple_data)
     X_train_tesla, X_test_tesla, y_train_tesla, y_test_tesla, tesla_scaler =_
      prepare_lstm_data(tesla_data)
     X_train_btc, X_test_btc, y_train_btc, y_test_btc, btc_scaler =_
      ⇒prepare_lstm_data(btc_data)
     X_train_eth, X_test_eth, y_train_eth, y_test_eth, eth_scaler =_u
      prepare_lstm_data(eth_data)
     # Display shapes of training data
     print("Apple Training Data Shape:", X_train_apple.shape)
     print("Tesla Training Data Shape:", X_train_tesla.shape)
     print("BTC Training Data Shape:", X_train_btc.shape)
     print("ETH Training Data Shape:", X_train_eth.shape)
    Apple Training Data Shape: (937, 60, 4)
    Tesla Training Data Shape: (937, 60, 4)
    BTC Training Data Shape: (512, 60, 4)
    ETH Training Data Shape: (512, 60, 4)
[]: # Function to build and train LSTM model
```

```
def build_and_train_lstm(X_train, y_train, X_test, y_test, epochs=20,__
  ⇔batch_size=32):
    model = Sequential([
        LSTM(50, return_sequences=True, input_shape=(X_train.shape[1], X_train.
  \hookrightarrowshape[2])),
        Dropout(0.2),
        LSTM(50, return_sequences=False),
        Dropout(0.2),
        Dense(25),
        Dense(1)
    ])
    model.compile(optimizer='adam', loss='mean squared error')
    early_stop = EarlyStopping(monitor='val_loss', patience=5,__
  →restore_best_weights=True)
    history = model.fit(X_train, y_train, epochs=epochs, batch_size=batch_size,_
  ⇔validation_data=(X_test, y_test), callbacks=[early_stop])
    return model, history
# Train models for stocks and cryptocurrencies
apple_model, _ = build_and_train_lstm(X_train_apple, y_train_apple,_u
 →X_test_apple, y_test_apple)
tesla_model, _ = build_and_train_lstm(X_train_tesla, y_train_tesla,_u
 →X_test_tesla, y_test_tesla)
btc_model, _ = build_and_train_lstm(X_train_btc, y_train_btc, X_test_btc,__
 →y_test_btc)
eth model, = build and train lstm(X train eth, y train eth, X test eth, u

y_test_eth)

/usr/local/lib/python3.10/dist-packages/keras/src/layers/rnn/rnn.py:204:
UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When
using Sequential models, prefer using an `Input(shape)` object as the first
layer in the model instead.
  super().__init__(**kwargs)
Epoch 1/20
30/30
                 7s 107ms/step -
loss: 0.0713 - val_loss: 0.0078
Epoch 2/20
30/30
                  2s 58ms/step -
loss: 0.0052 - val_loss: 0.0066
Epoch 3/20
30/30
                  2s 56ms/step -
loss: 0.0040 - val loss: 0.0031
Epoch 4/20
30/30
                  2s 79ms/step -
loss: 0.0029 - val_loss: 0.0047
Epoch 5/20
```

```
30/30
                  3s 86ms/step -
loss: 0.0030 - val_loss: 0.0018
Epoch 6/20
30/30
                  5s 80ms/step -
loss: 0.0028 - val_loss: 0.0026
Epoch 7/20
30/30
                  3s 105ms/step -
loss: 0.0025 - val_loss: 0.0019
Epoch 8/20
30/30
                  2s 59ms/step -
loss: 0.0020 - val_loss: 0.0037
Epoch 9/20
30/30
                  2s 56ms/step -
loss: 0.0020 - val_loss: 0.0027
Epoch 10/20
30/30
                  3s 57ms/step -
loss: 0.0018 - val_loss: 0.0013
Epoch 11/20
30/30
                  3s 95ms/step -
loss: 0.0015 - val_loss: 0.0024
Epoch 12/20
30/30
                  2s 70ms/step -
loss: 0.0016 - val_loss: 0.0014
Epoch 13/20
30/30
                  2s 56ms/step -
loss: 0.0015 - val_loss: 0.0020
Epoch 14/20
30/30
                  2s 57ms/step -
loss: 0.0016 - val_loss: 0.0015
Epoch 15/20
30/30
                  3s 56ms/step -
loss: 0.0017 - val_loss: 0.0028
Epoch 1/20
30/30
                  7s 77ms/step -
loss: 0.0360 - val_loss: 0.0070
Epoch 2/20
30/30
                  2s 57ms/step -
loss: 0.0056 - val_loss: 0.0046
Epoch 3/20
30/30
                  2s 59ms/step -
loss: 0.0048 - val_loss: 0.0032
Epoch 4/20
30/30
                  2s 56ms/step -
loss: 0.0035 - val_loss: 0.0028
Epoch 5/20
30/30
                  3s 56ms/step -
loss: 0.0031 - val_loss: 0.0020
Epoch 6/20
```

```
30/30
                  4s 97ms/step -
loss: 0.0031 - val_loss: 0.0019
Epoch 7/20
30/30
                  2s 57ms/step -
loss: 0.0030 - val_loss: 0.0031
Epoch 8/20
30/30
                  3s 56ms/step -
loss: 0.0032 - val_loss: 0.0017
Epoch 9/20
30/30
                  2s 57ms/step -
loss: 0.0027 - val_loss: 0.0017
Epoch 10/20
30/30
                  3s 59ms/step -
loss: 0.0021 - val_loss: 0.0015
Epoch 11/20
30/30
                  3s 80ms/step -
loss: 0.0021 - val_loss: 0.0018
Epoch 12/20
30/30
                  3s 89ms/step -
loss: 0.0020 - val_loss: 0.0015
Epoch 13/20
                  5s 76ms/step -
30/30
loss: 0.0022 - val_loss: 0.0014
Epoch 14/20
30/30
                  2s 72ms/step -
loss: 0.0026 - val_loss: 0.0019
Epoch 15/20
30/30
                  2s 58ms/step -
loss: 0.0020 - val_loss: 0.0021
Epoch 16/20
30/30
                  4s 92ms/step -
loss: 0.0020 - val_loss: 0.0013
Epoch 17/20
30/30
                  2s 81ms/step -
loss: 0.0019 - val_loss: 0.0012
Epoch 18/20
30/30
                  2s 56ms/step -
loss: 0.0017 - val_loss: 0.0012
Epoch 19/20
30/30
                  3s 57ms/step -
loss: 0.0018 - val_loss: 0.0013
Epoch 20/20
30/30
                  2s 57ms/step -
loss: 0.0014 - val_loss: 0.0018
Epoch 1/20
16/16
                  5s 123ms/step -
loss: 0.0263 - val_loss: 0.0273
Epoch 2/20
```

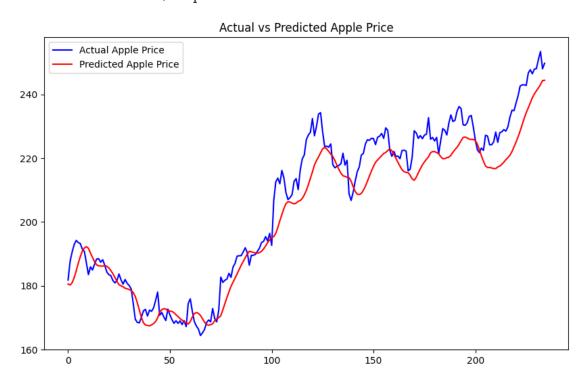
```
16/16
                  1s 86ms/step -
loss: 0.0042 - val_loss: 0.0262
Epoch 3/20
16/16
                  2s 50ms/step -
loss: 0.0040 - val_loss: 0.0086
Epoch 4/20
16/16
                  1s 51ms/step -
loss: 0.0029 - val_loss: 0.0069
Epoch 5/20
16/16
                  1s 49ms/step -
loss: 0.0022 - val_loss: 0.0065
Epoch 6/20
16/16
                  1s 50ms/step -
loss: 0.0019 - val_loss: 0.0105
Epoch 7/20
16/16
                  1s 51ms/step -
loss: 0.0019 - val_loss: 0.0075
Epoch 8/20
16/16
                  1s 49ms/step -
loss: 0.0018 - val_loss: 0.0034
Epoch 9/20
                  1s 50ms/step -
16/16
loss: 0.0020 - val_loss: 0.0077
Epoch 10/20
16/16
                  1s 49ms/step -
loss: 0.0019 - val_loss: 0.0034
Epoch 11/20
16/16
                  1s 55ms/step -
loss: 0.0014 - val_loss: 0.0073
Epoch 12/20
16/16
                  1s 87ms/step -
loss: 0.0018 - val_loss: 0.0080
Epoch 13/20
16/16
                  1s 85ms/step -
loss: 0.0014 - val_loss: 0.0038
Epoch 14/20
16/16
                  2s 50ms/step -
loss: 0.0014 - val_loss: 0.0076
Epoch 15/20
16/16
                  1s 51ms/step -
loss: 0.0018 - val_loss: 0.0081
Epoch 1/20
16/16
                  4s 72ms/step -
loss: 0.0823 - val_loss: 0.0320
Epoch 2/20
16/16
                  1s 48ms/step -
loss: 0.0153 - val_loss: 0.0147
Epoch 3/20
```

```
16/16
                  1s 49ms/step -
loss: 0.0091 - val_loss: 0.0102
Epoch 4/20
16/16
                  1s 50ms/step -
loss: 0.0089 - val loss: 0.0072
Epoch 5/20
16/16
                  1s 79ms/step -
loss: 0.0086 - val_loss: 0.0064
Epoch 6/20
16/16
                  1s 86ms/step -
loss: 0.0065 - val_loss: 0.0058
Epoch 7/20
16/16
                  1s 88ms/step -
loss: 0.0067 - val_loss: 0.0065
Epoch 8/20
16/16
                  2s 49ms/step -
loss: 0.0062 - val_loss: 0.0068
Epoch 9/20
16/16
                  1s 50ms/step -
loss: 0.0050 - val_loss: 0.0056
Epoch 10/20
16/16
                  1s 48ms/step -
loss: 0.0053 - val_loss: 0.0054
Epoch 11/20
16/16
                  1s 49ms/step -
loss: 0.0048 - val_loss: 0.0052
Epoch 12/20
16/16
                  1s 49ms/step -
loss: 0.0047 - val_loss: 0.0051
Epoch 13/20
16/16
                  1s 50ms/step -
loss: 0.0047 - val_loss: 0.0050
Epoch 14/20
16/16
                  1s 49ms/step -
loss: 0.0046 - val_loss: 0.0045
Epoch 15/20
16/16
                  1s 48ms/step -
loss: 0.0054 - val_loss: 0.0053
Epoch 16/20
16/16
                  1s 58ms/step -
loss: 0.0050 - val_loss: 0.0047
Epoch 17/20
16/16
                  1s 83ms/step -
loss: 0.0042 - val_loss: 0.0050
Epoch 18/20
16/16
                  1s 88ms/step -
loss: 0.0051 - val_loss: 0.0054
Epoch 19/20
```

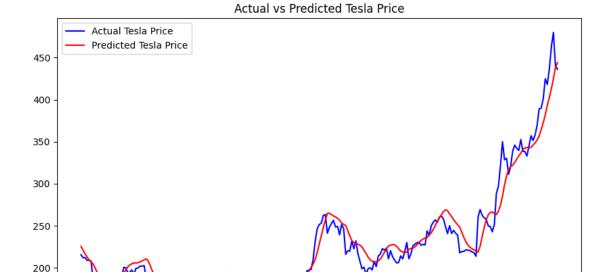
```
16/16 2s 49ms/step - loss: 0.0043 - val_loss: 0.0046
```

```
[]: # Function to evaluate and plot results for LSTM models
     def evaluate_lstm(model, X_test, y_test, scaler, name="Asset"):
        predictions = model.predict(X_test)
        predictions = scaler.inverse_transform(np.concatenate([predictions, np.
      seros((predictions.shape[0], 3))], axis=1))[:, 0]
         actual = scaler.inverse_transform(np.concatenate([v_test.reshape(-1, 1), np.
      ⇒zeros((y_test.shape[0], 3))], axis=1))[:, 0]
        plt.figure(figsize=(10, 6))
        plt.plot(actual, color='blue', label=f'Actual {name} Price')
        plt.plot(predictions, color='red', label=f'Predicted {name} Price')
        plt.title(f'Actual vs Predicted {name} Price')
        plt.legend()
        plt.show()
     # Evaluate models for stocks
     evaluate_lstm(apple_model, X_test_apple, y_test_apple, apple_scaler, "Apple")
     evaluate_lstm(tesla_model, X_test_tesla, y_test_tesla, tesla_scaler, "Tesla")
     # Evaluate models for cryptocurrencies
     evaluate_lstm(btc_model, X_test_btc, y_test_btc, btc_scaler, "Bitcoin")
     evaluate lstm(eth model, X test eth, y test eth, eth scaler, "Ethereum")
```

8/8 3s 181ms/step



8/8 1s 54ms/step



WARNING:tensorflow:5 out of the last 17 calls to <function
TensorFlowTrainer.make_predict_function.<locals>.one_step_on_data_distributed at
0x7a2876dead40> triggered tf.function retracing. Tracing is expensive and the
excessive number of tracings could be due to (1) creating @tf.function
repeatedly in a loop, (2) passing tensors with different shapes, (3) passing
Python objects instead of tensors. For (1), please define your @tf.function
outside of the loop. For (2), @tf.function has reduce_retracing=True option that
can avoid unnecessary retracing. For (3), please refer to
https://www.tensorflow.org/guide/function#controlling_retracing and
https://www.tensorflow.org/api_docs/python/tf/function for more details.

100

150

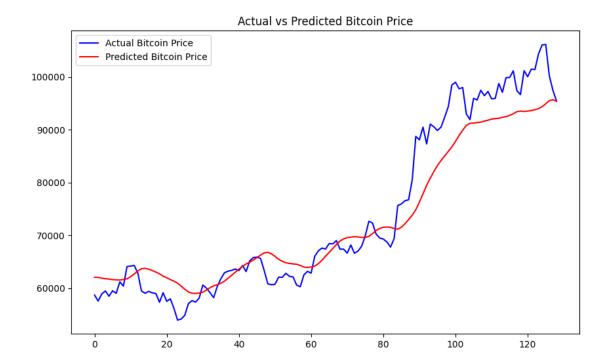
200

5/5 1s 83ms/step

150

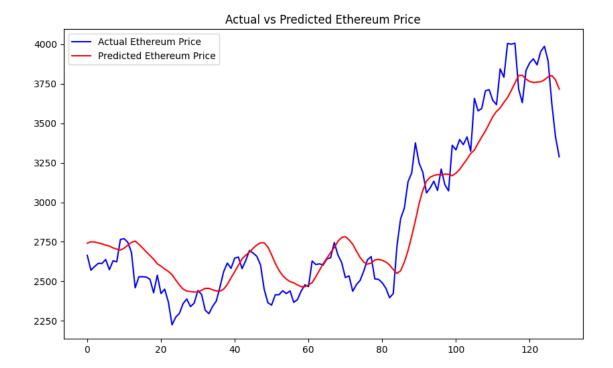
Ó

50



WARNING:tensorflow:5 out of the last 14 calls to <function
TensorFlowTrainer.make_predict_function.<locals>.one_step_on_data_distributed at
0x7a287e9e77f0> triggered tf.function retracing. Tracing is expensive and the
excessive number of tracings could be due to (1) creating @tf.function
repeatedly in a loop, (2) passing tensors with different shapes, (3) passing
Python objects instead of tensors. For (1), please define your @tf.function
outside of the loop. For (2), @tf.function has reduce_retracing=True option that
can avoid unnecessary retracing. For (3), please refer to
https://www.tensorflow.org/guide/function#controlling_retracing and
https://www.tensorflow.org/api_docs/python/tf/function for more details.

5/5 1s 86ms/step



[]: