## HW4 analysis runner

January 2, 2022

## 1 HW4\_analysis\_runner

```
[1]: import papermill as pm
import scrapbook as sb
import pandas as pd
from datetime import datetime, timedelta
import os
```

C:\ProgramData\Anaconda3\lib\site-packages\papermill\iorw.py:50: FutureWarning: pyarrow.HadoopFileSystem is deprecated as of 2.0.0, please use pyarrow.fs.HadoopFileSystem instead.

from pyarrow import HadoopFileSystem

Run the template notebook across all tokens in the database and the last 2 complete weeks in the data base.

```
[2]: #Read all data from data/data.db in order to get list of tokens and find the

→ last two complete weeks

db = pd.read_sql(f'''

SELECT

*

FROM ohlc

''', 'sqlite:///data/data.db')

db['ts'] = pd.to_datetime(db['ts'])

db.set_index('ts', inplace=True)

db.iloc[-1:]
```

- [2]: open high low close volume volumeUSD token chain ts
  2021-12-15 185.07 185.78 184.5 184.89 335.307 62142.63926 COMP ETH
- [3]: def last\_dow(d: datetime, dow: int) -> datetime:

  '''

  Gets the date of the last day of week, using datetime convention of Monday

  → is 0 and Sunday is 6

  '''

```
dow_d = d.weekday()
         days_delta = (7 - dow) + dow_d if dow_d < dow else dow_d - dow
         return d - timedelta(days=days_delta)
[4]: # The last two complete weeks are from 2021-11-29 to 2021-12-12
     run_dates = [(last_dow(datetime(2021, 12, 1), 2) + timedelta(days=14 * x)) for_u
     \rightarrowx in range(1)]
     tokens = db['token'].unique().tolist()
     connection_string = 'sqlite:///data/data.db'
[5]: # Outputs are saved in HW4_outputs
     base_output_folder = os.path.expanduser('HW4_outputs')
     if not os.path.exists(base_output_folder):
             os.makedirs(base_output_folder)
[6]: # There is a problem on my end to save file with names similar to those in_{\perp}
     \rightarrow lecture examples. As a result I simplified to token name.
     for run_date in run_dates:
         for token in tokens:
             print(f'running for {token} for date {run_date} to {run_date +__
      →timedelta(days=14)}', end='\r')
             res = pm.execute notebook(
                 'HW4_analysis_template.ipynb',
                 f'{base_output_folder}/{token}.ipynb',
                 parameters = {
                     'from_date': f'{run_date}',
                     'to_date': f'{run_date + timedelta(days=14)}',
                     'token': token,
                     'connection_string': connection_string
                 }
             )
    running for BTC for date 2021-12-01 00:00:00 to 2021-12-15 00:00:00
                               | 0/20 [00:00<?, ?cell/s]
                 0%1
    Executing:
    running for ETH for date 2021-12-01 00:00:00 to 2021-12-15 00:00:00
                               | 0/20 [00:00<?, ?cell/s]
    Executing:
                 0%1
    running for USDT for date 2021-12-01 00:00:00 to 2021-12-15 00:00:00
                 0%|
                               | 0/20 [00:00<?, ?cell/s]
    Executing:
    running for SOL for date 2021-12-01 00:00:00 to 2021-12-15 00:00:00
                 0%|
                               | 0/20 [00:00<?, ?cell/s]
    Executing:
```

| 0/20 [00:00<?, ?cell/s]

running for ADA for date 2021-12-01 00:00:00 to 2021-12-15 00:00:00

Executing:

0%1

```
| 0/20 [00:00<?, ?cell/s]
    Executing:
                  0%1
    running for AVAX for date 2021-12-01 00:00:00 to 2021-12-15 00:00:00
                               | 0/20 [00:00<?, ?cell/s]
    Executing:
                  0%1
    running for ATOM for date 2021-12-01 00:00:00 to 2021-12-15 00:00:00
                               | 0/20 [00:00<?, ?cell/s]
    Executing:
                  0%|
    running for CRV for date 2021-12-01 00:00:00 to 2021-12-15 00:00:00
                               | 0/20 [00:00<?, ?cell/s]
    Executing:
                  0%|
    running for AAVE for date 2021-12-01 00:00:00 to 2021-12-15 00:00:00
    Executing:
                  0%|
                               | 0/20 [00:00<?, ?cell/s]
    running for COMP for date 2021-12-01 00:00:00 to 2021-12-15 00:00:00
                               | 0/20 [00:00<?, ?cell/s]
    Executing:
                  0%1
    Summarize avg hourly return by run, avg volatility by run, maximum drawdown over the entire
    period of each run, range of last close price - first close price in a single DataFrame.
[7]: nbs = sb.read_notebooks('HW4_outputs')
[8]: pd.DataFrame([
         pd.Series({
             'token': scrap['token'].data,
             'from_date': scrap['from_date'].data,
             'to_date': scrap['to_date'].data,
             'hourly return': scrap['prices'].data['hourly return'].mean(),
             'hourly vol': scrap['prices'].data['hourly vol'].mean(),
             'MDD': scrap['prices'].data['MDD'][-1],
             'Range of last close price - first close price': scrap['prices'].

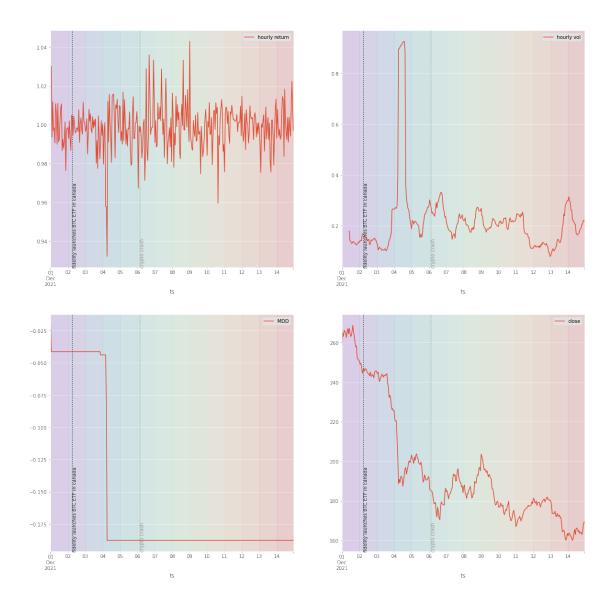
data['close'][-1] - scrap['prices'].data['close'][0]})

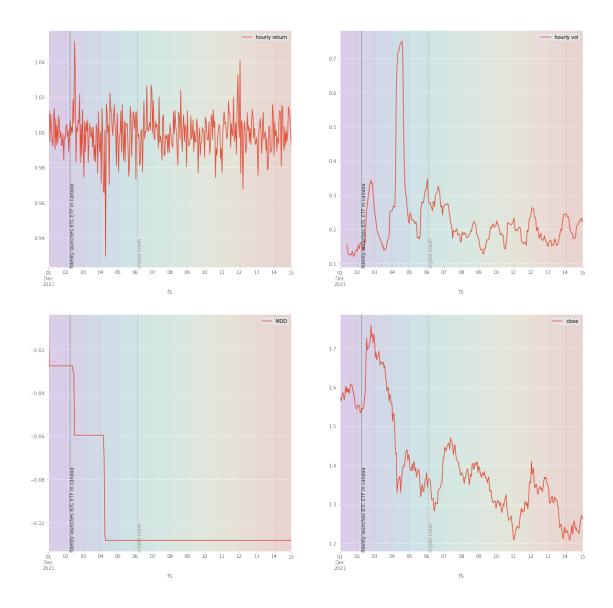
         for book, scrap in nbs.notebook_scraps.items()
     ])
                                                 to_date hourly return hourly vol \
        token
                          from_date
```

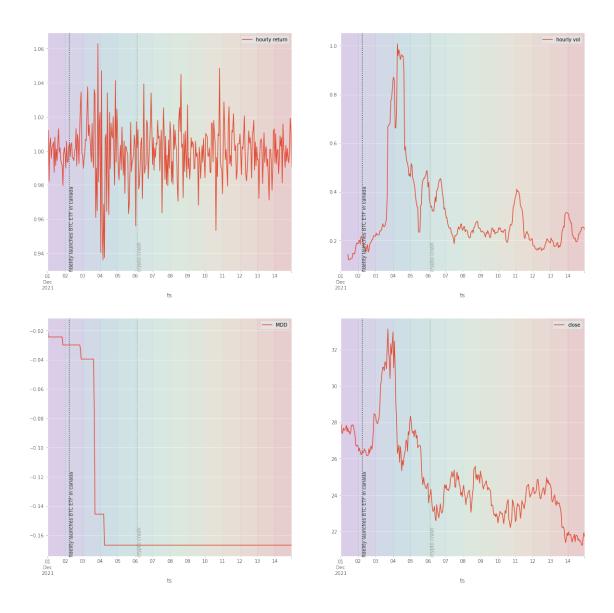
running for DOT for date 2021-12-01 00:00:00 to 2021-12-15 00:00:00

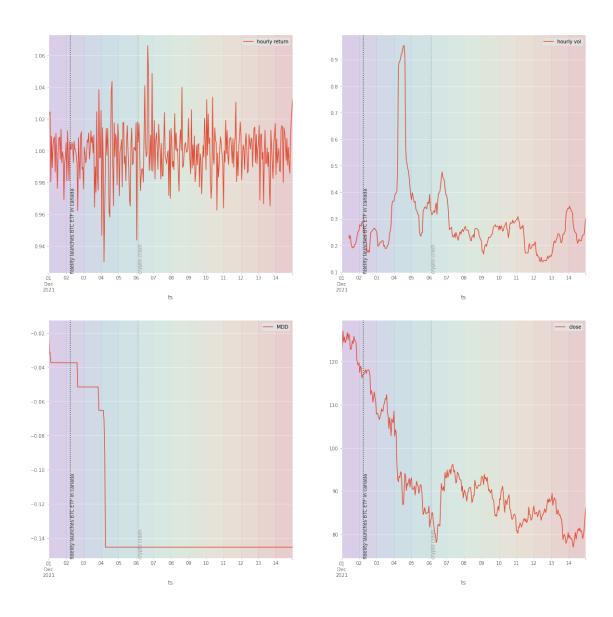
```
[8]:
                                                                           0.216067
     0
         AAVE
               2021-12-01 00:00:00
                                    2021-12-15 00:00:00
                                                               0.998815
     1
          ADA
               2021-12-01 00:00:00
                                    2021-12-15 00:00:00
                                                               0.999450
                                                                           0.224460
     2
         MOTA
               2021-12-01 00:00:00
                                    2021-12-15 00:00:00
                                                               0.999428
                                                                           0.303370
     3
               2021-12-01 00:00:00
                                    2021-12-15 00:00:00
         AVAX
                                                               0.999118
                                                                           0.283772
               2021-12-01 00:00:00
                                    2021-12-15 00:00:00
     4
          BTC
                                                               0.999534
                                                                           0.157194
               2021-12-01 00:00:00
     5
         COMP
                                    2021-12-15 00:00:00
                                                               0.998865
                                                                           0.234725
          CRV
               2021-12-01 00:00:00
                                    2021-12-15 00:00:00
                                                                           0.324726
     6
                                                               0.998994
     7
          DOT
               2021-12-01 00:00:00
                                    2021-12-15 00:00:00
                                                               0.998974
                                                                           0.248345
     8
          ETH 2021-12-01 00:00:00 2021-12-15 00:00:00
                                                               0.999479
                                                                           0.177147
     9
               2021-12-01 00:00:00 2021-12-15 00:00:00
          SOL
                                                               0.999281
                                                                           0.248466
```

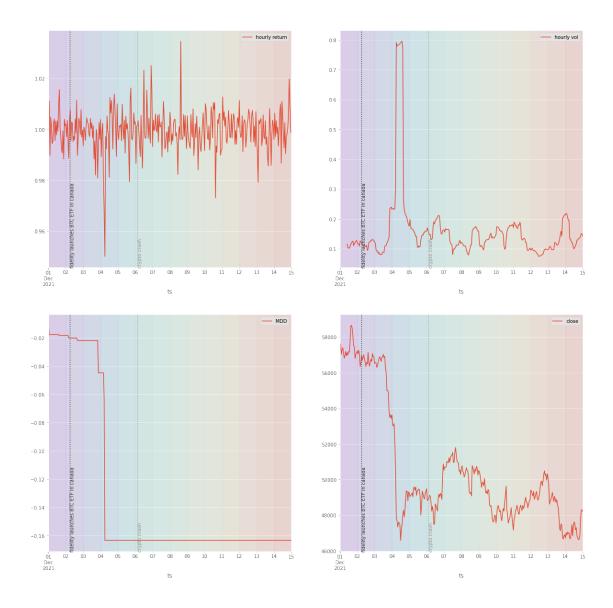
```
10 USDT 2021-12-01 00:00:00 2021-12-15 00:00:00
                                                           0.999999
                                                                        0.010131
             MDD
                  Range of last close price - first close price
    0 -0.187685
                                                       -88.3180
    1 -0.108138
                                                       -0.2947
    2 -0.166814
                                                       -5.8280
    3 -0.145482
                                                       -35.1400
    4 -0.163526
                                                    -8768.8800
    5 -0.152001
                                                       -93.6200
    6 -0.195640
                                                       -1.6295
    7 -0.165718
                                                       -11.8820
    8 -0.114836
                                                      -796.9200
    9 -0.112899
                                                      -49.3260
    10 -0.023415
                                                       -0.0003
[9]: [nbs[nb_name].reglue('summary_plot') for nb_name in nbs]
```

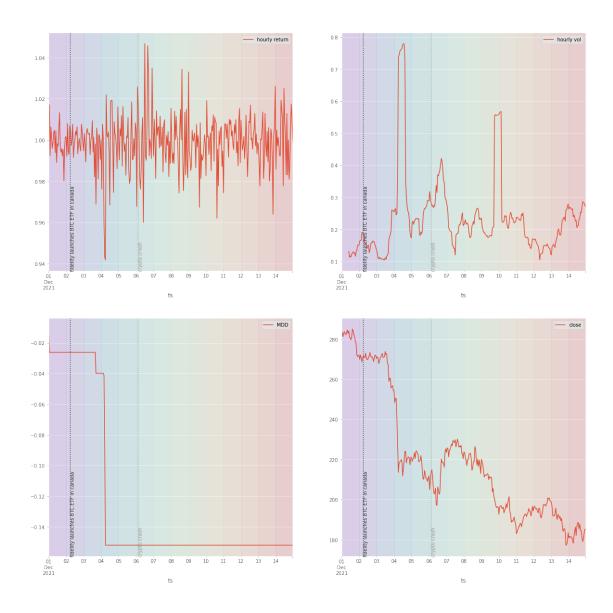


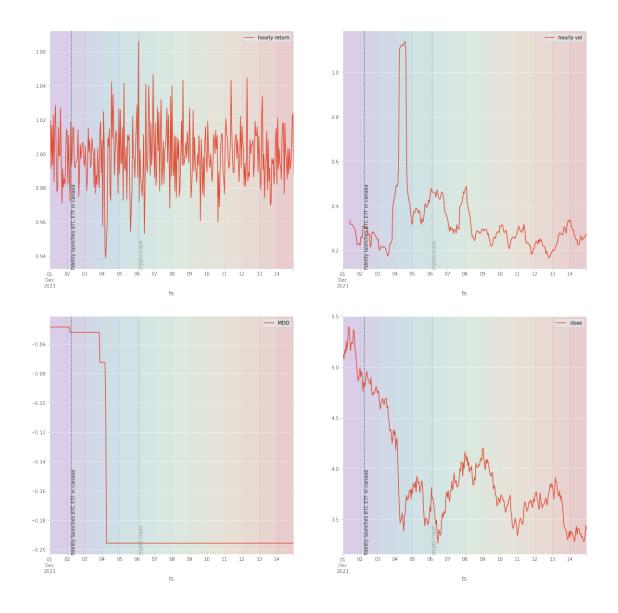


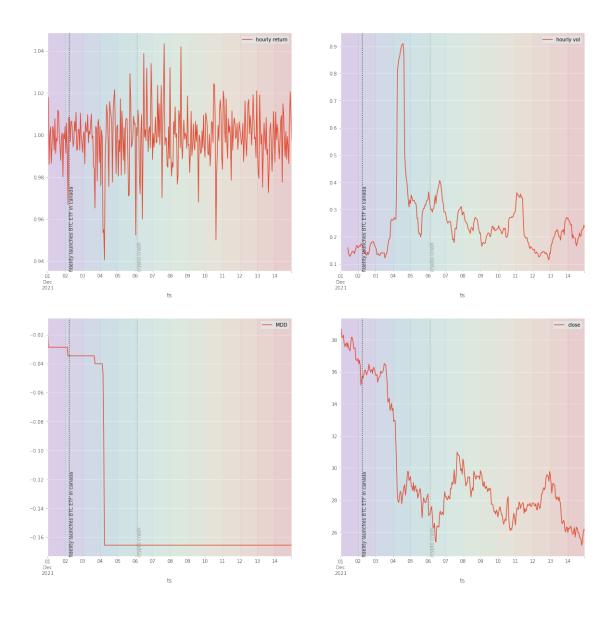


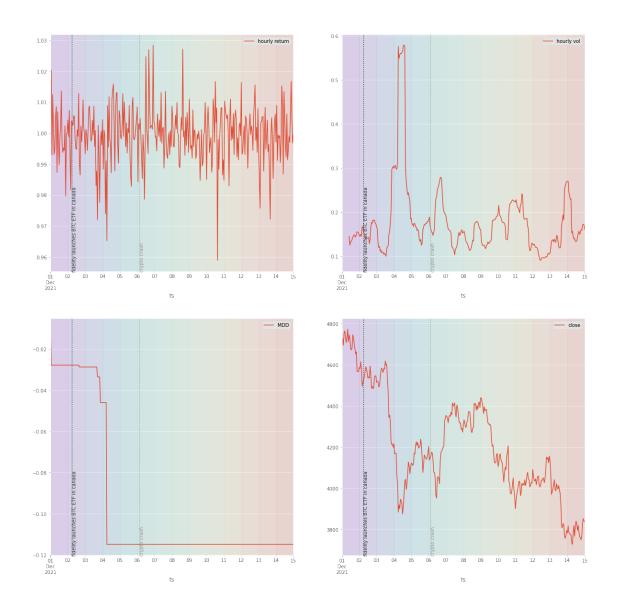


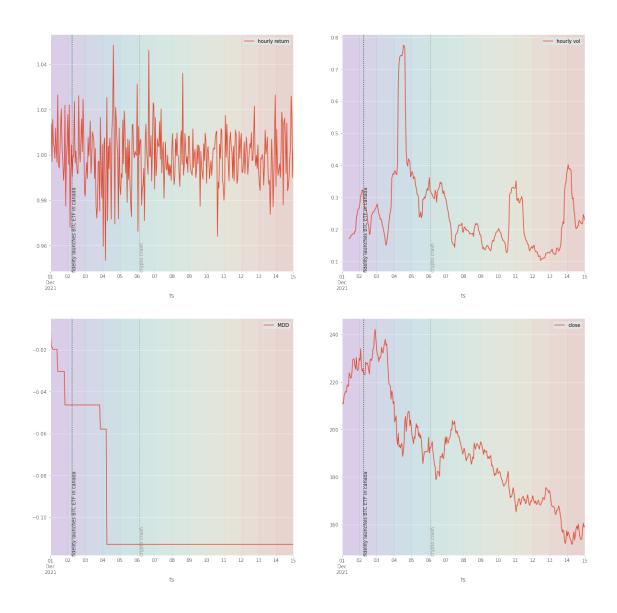


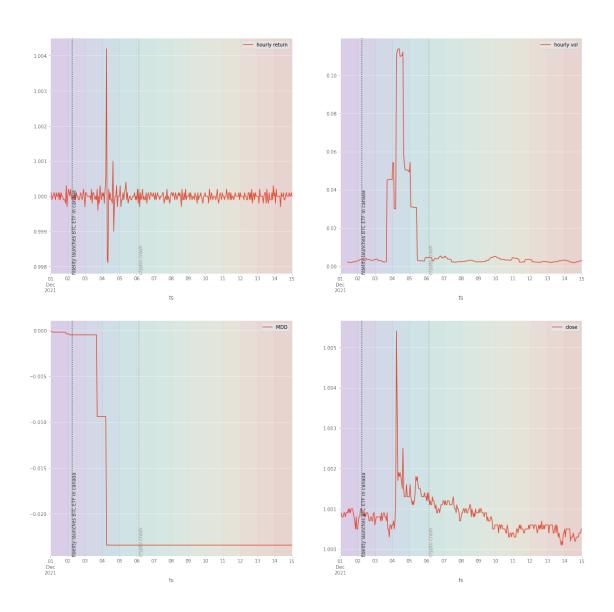












[9]: [None, None, None, None, None, None, None, None, None, None]

[]: