```
In [1]: import os
    os.chdir('../')
```

## **Package Overview**

## Pip install the package (git clone)

After installation the user needs to configure the API keys in the config.ini file.

- The user needs to register an app in twitter applications inorder to use tweepy.
- Needs to create an account in Quandl to be able to retreive various day level data.

```
In [4]: !cat 'pybcoin/config/config.ini'
        [Twitter]
        consumer_key =
        consumer secret =
        access token =
        access token secret =
        tweet-count-url = https://bitinfocharts.com/comparison/tweets-btc.html
        [Reddit]
        api-uri = https://elastic.pushshift.io/rc/comments/ search?source={"query":{"bool":{"must"
        :[{"simple query string":{"query":"bitcoin|btc|crypto","fields":["body"],"default operator
        ":"and"}}], "filter":[{"range":{"created utc":{"gte":START UTC,"lte":END UTC}}}, {"terms":{"
        subreddit":["bitcoin","btc","cryptocurrency"]}}],"should":[],"must not":[]}},"size":10000,
        "sort":{"created utc":"desc"}}
        data path = ./data/latest/
        [Collector]
        in_path_btc = ./data/btc/
        in path comm = ./data/commodity/
        in path gtrends = ./data/gtrends/
        out path = ./data/
        [Forecast]
        in_path_btc = ./data/btc/
        in path comm = ./data/commodity/
        in path gtrends = ./data/gtrends/
        out path = ./data/
        in path social = ./data/latest/
        path time pred = ./data/pred/
        [Quandl]
        quandl-key =
        [Sentiment]
        text csv path = ./data/latest/
        wc path = ./pybcoin/static/
```

## **Triggering data collection**

This will start the entire data collection pipeline. You can also trigger individual data collection modules as needed. For example you can use BtcDataCollector module to collect day level data for btc prices, number of btc related tweets and the total volume of btc transactions in USD.

```
In [10]: from pybcoin.DataCollector.controller_collector import BtcDataCollector
In [12]: collector = BtcDataCollector(config)
In [13]: collector.fetch_btc_price()
In [14]: collector.fetch_transaction_volume()
In [15]: collector.fetch_tweet_counts()
```

The data will be appended to the corresponding files in data/btc/ folder.

Similarly you can retrieve data from various other sources. See documentation for details.

## **Analysis overview**

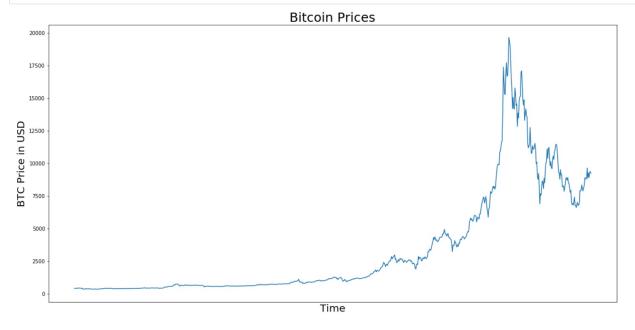
```
In [17]: # import packages
    import matplotlib.pyplot as plt
    import pandas as pd
    import warnings
    from fbprophet import Prophet
    import seaborn as sns

In [18]: # suppressing warnings for now
    warnings.filterwarnings("ignore")

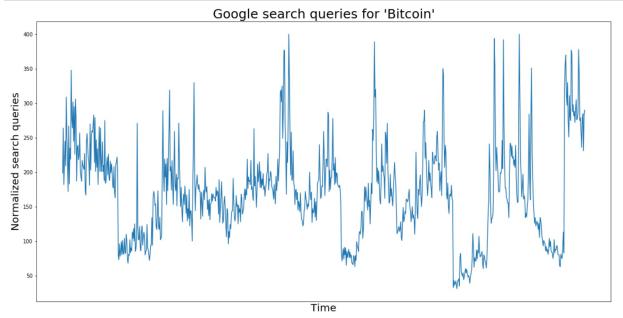
In [19]: # reading data
    btc_data = pd.read_csv('data/btc/btc_prices.csv')
    oil_data = pd.read_csv('data/commodity/oil_price.csv')
    google_data = pd.read_csv('data/gtrends/GTrendsData.csv')
    twitter_data = pd.read_csv('data/latest/tweets_sentiment.csv')
    reddit_data = pd.read_csv('data/latest/reddit_comments_sentiment.csv')
```

```
In [4]: # getting relevant sum from the google trends data
    google_data['google_hits'] = google_data['btc'] + google_data['btcoin'] + google_data['btc
    usd'] + google_data['btcusd']
```

```
In [5]: # visualizing the various prices
# plotting BTC prices
plt.figure(figsize=(20,10))
plt.plot(btc_data['Date'], btc_data['btc_price'])
plt.xticks([])
plt.title('Bitcoin Prices', fontsize=25)
plt.xlabel('Time', fontsize=20);
plt.ylabel('BTC Price in USD', fontsize=20);
```



```
In [6]: # plotting google search queries
   plt.figure(figsize=(20,10))
   plt.plot(google_data['Date'], google_data['google_hits'])
   plt.xticks([])
   plt.title("Google search queries for Bitcoin", fontsize=25)
   plt.xlabel('Time', fontsize=20);
   plt.ylabel('Normalized search queries', fontsize=20);
```



```
In [7]: # demonstration to use the time series package: fbprophet
    # prophet module requires columns ds (date) and y (value)
    time_series = btc_data.rename(columns={'Date': 'ds', 'btc_price': 'y'})

# parameters can be changed based on the seasonality that needs to be incorporated
    # the parameter changepoint_prior_scale can be tuned to give the best results for specific u
    se-cases
    m = Prophet(yearly_seasonality=True, daily_seasonality=False, changepoint_prior_scale=0.001)

# fitting the time series time model
    m.fit(time_series);
```

```
In [8]: # making the future predictions for 1 day excluding history
future = m.make_future_dataframe(periods=1, include_history=False)
future = m.predict(future)
```

```
In [9]: # the result contains various trend and seasonality components
# the result also contains the confidence levels for the various components
future
```

Out[9]:

	ds	trend	yhat_lower	yhat_upper	trend_lower	trend_upper	seasonal	seasonal_lower	se
0	2018-05-01	7763.114318	4535.96216	10037.568007	7763.114318	7763.114318	-544.975864	-544.975864	-5

```
In [10]: # computing the correlation-matrix between the variables
         # joining the various data-frames
         df = pd.merge(btc data, oil data, on='Date')
         del df['dates s']
         df = pd.merge(df, google data, on='Date')
         del df['date']
         df = pd.merge(df, twitter data, on='Date')
         del df['Date']
         # renaming columns name for twitter
         df = df.rename(columns = {'Negative':'twitter negative'})
         df = df.rename(columns = {'Positive':'twitter positive'})
         df = pd.merge(df, reddit data, left on = 'utc time', right on = 'Date')
         del df['Date']
         # renaming columns name for reddit
         df = df.rename(columns = {'Negative':'reddit negative'})
         df = df.rename(columns = {'Positive':'reddit positive'})
         # creating various ratios
         df['twitter'] = df['twitter positive']/(df['twitter positive'] + df['twitter negative'])
         df['reddit'] = df['reddit positive']/(df['reddit positive'] + df['reddit negative'])
```

