Metrics and Analytics Case Study

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Overview

Learning outcome: Metrics creation process in MQD and creating research case studies using Jupyter.

Pre-requisite: Uptick training, Market Microstructure knowledge, MQD and Workflow training session 1 to 6.

Contents Summary:

Introduction

What are MQD Metrics?

- Generate Financial data catering to research areas such as analysis of Market Efficiency and Fairness, Systemic risk, Market Fragmentation, Fundamental Analysis, Fixed Income Analysis, Energy Market Analysis.

- Based on Python language and uptick API.

- Available at the security level, a group of securities level, index level, a decile of securities level or a market wide level.

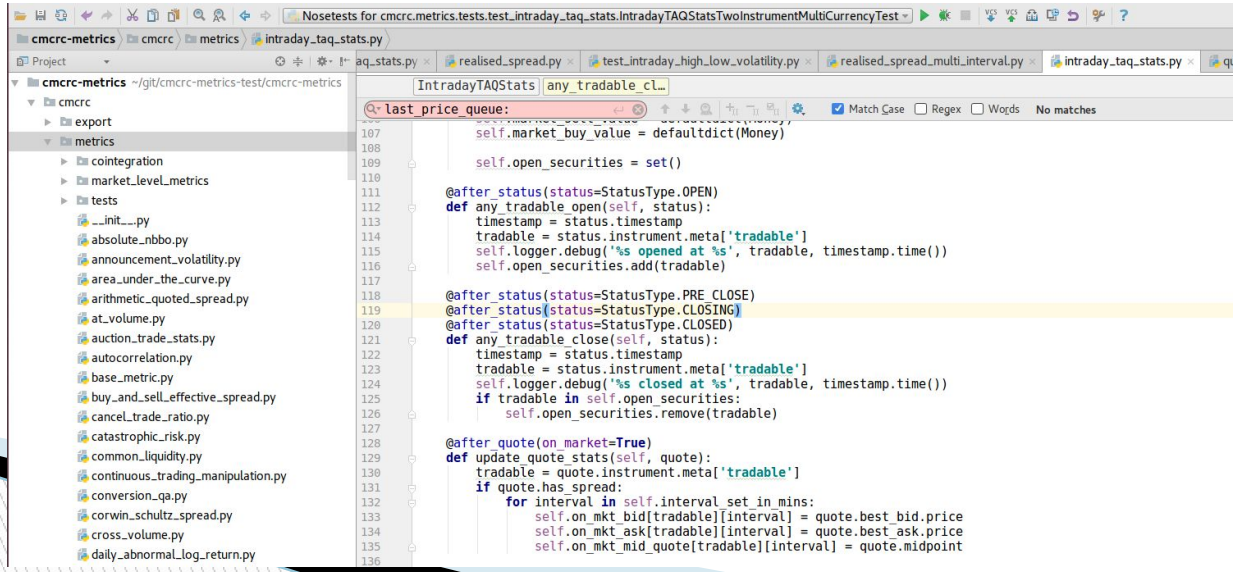
Click here for MQD Link for metric details

Click here for MQD metrics spec (needs login)

Metrics Creation Process

Tools and Environment

- Tools and Languages you need
  - Python – MQD Uptick Metrics is based on Python 3.6
  - Setup and build CMCRC–Metrics Repo https://wiki.cmcrc.com/display/MQ/Work+Environment+Set+Up+for+Uptick+metric
  - Where do we write metrics? Pycharm IDE
    - Easy–to–use and development friendly tool to write code in Python. One stop place to write, verify, test and submit the code for review.

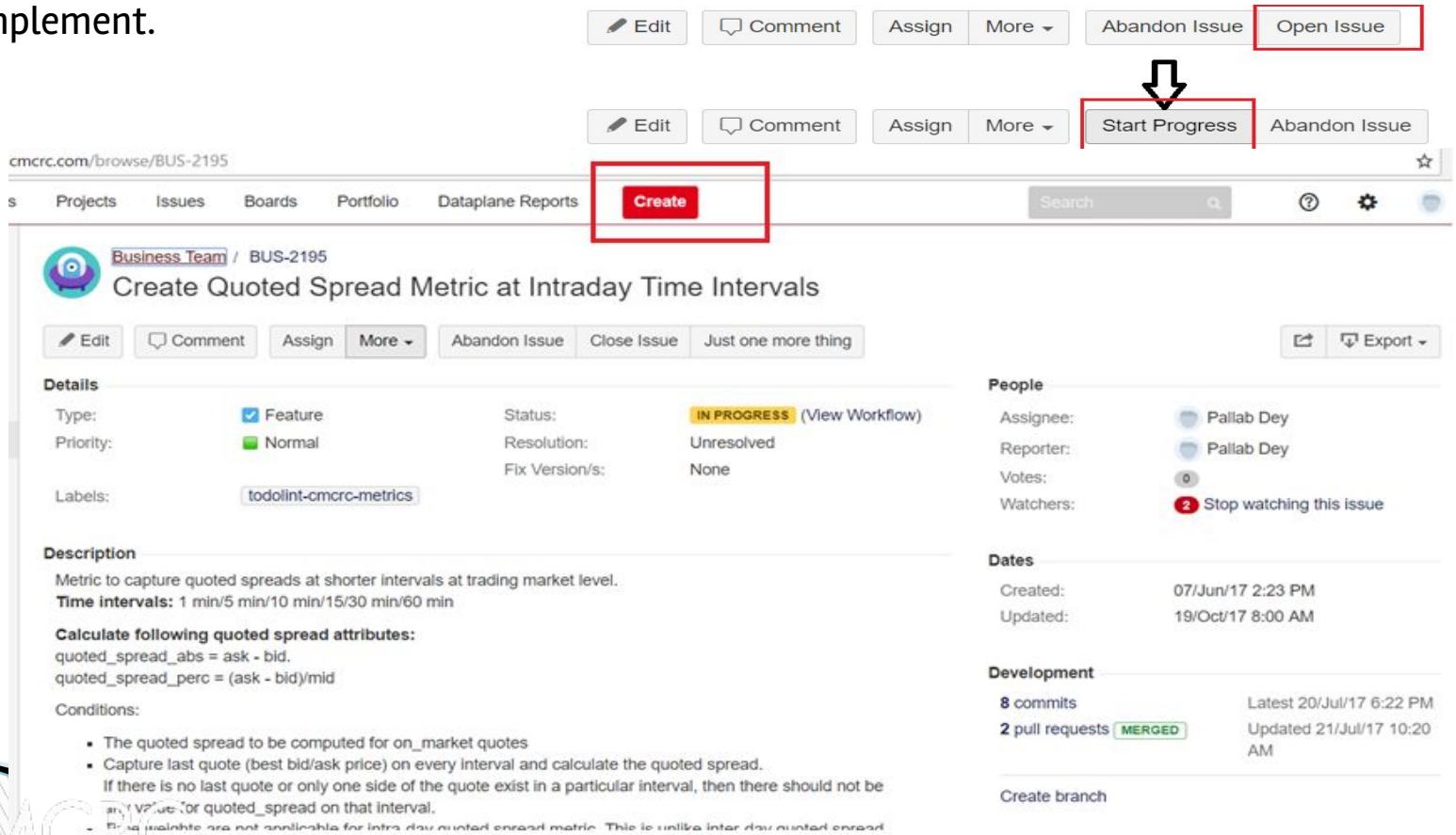


Steps to create and deploy a metric

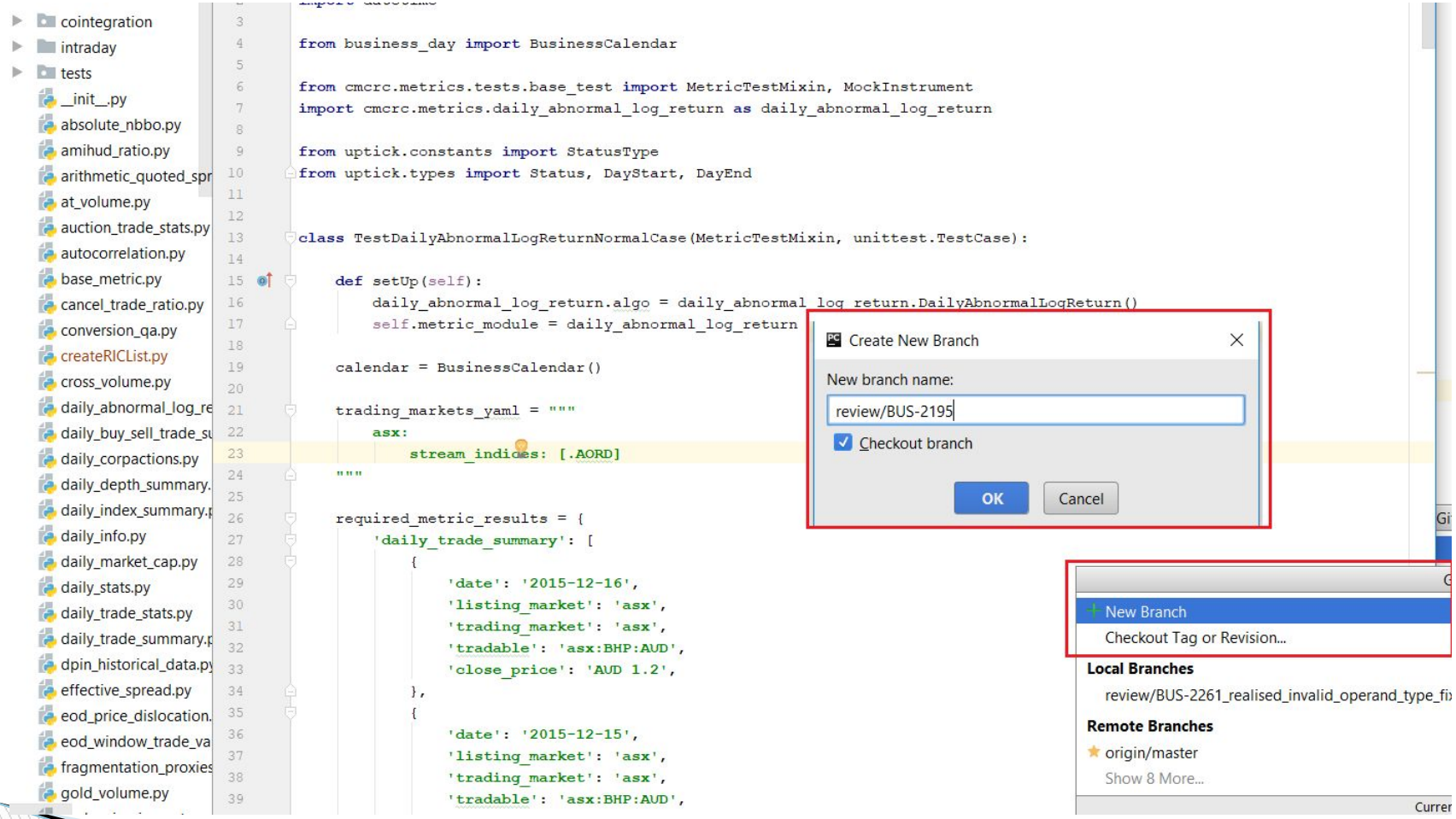
- Create JIRA for the metric
- Ensure you are working on latest code repository (update master branch using “git pull”
- Create new branch from Master (JIRA name should be part of branch name)
- Create New Metric class file
- Create Test case file
- Run the test case file to test for success
- Run the metric on real trading data
- Submit the metric code and test case code for review
- Changes to display in MQD UI

Create JIRA (Overview)

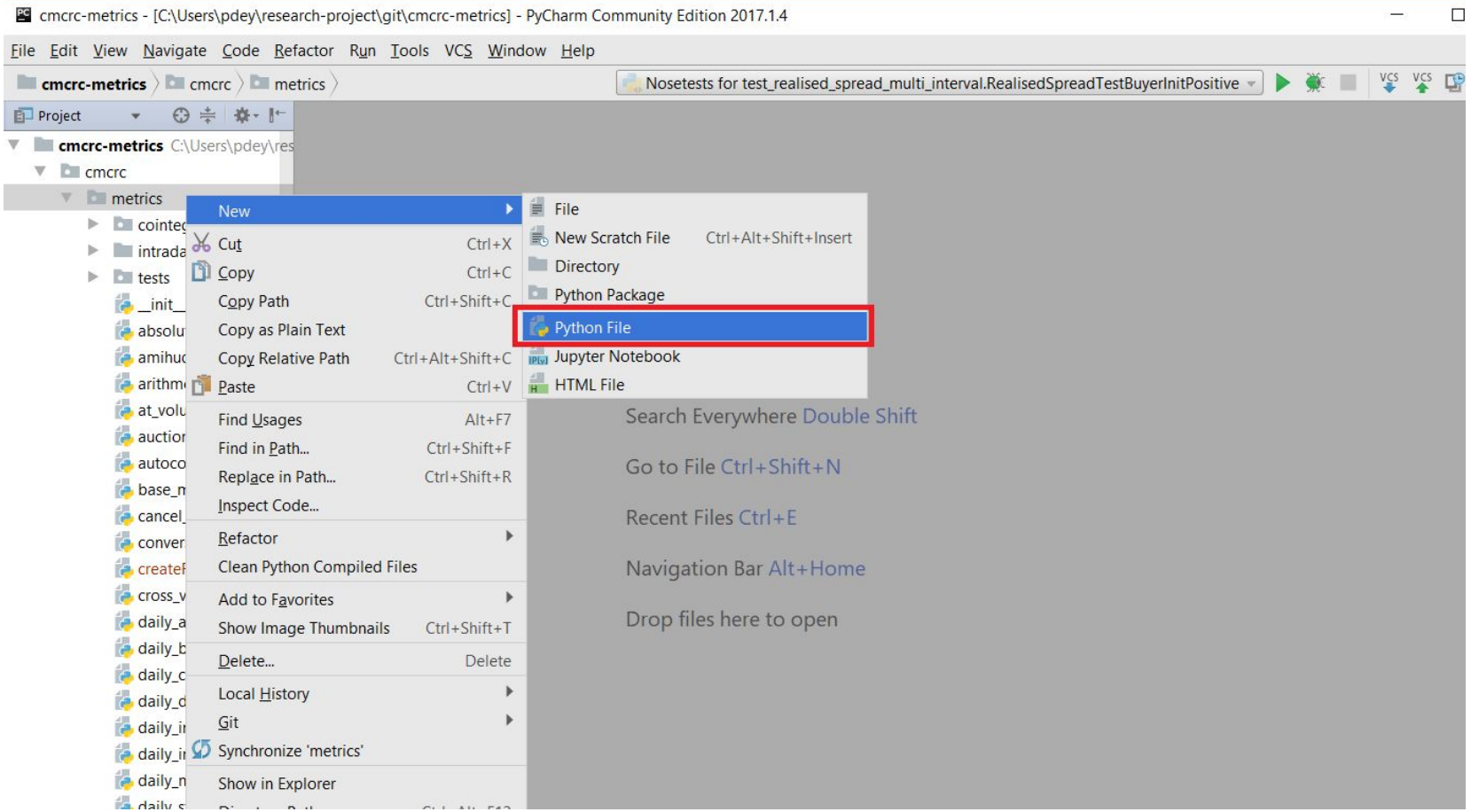
- Create JIRA using “Create” Button
- Fill–in JIRA title, description and add any attachments (if applicable)
- Assign the JIRA to the “Assignee” who will work on the JIRA
- Update the status to “Open” followed by “Start Progress” before starting to implement.



Create New Branch



Create New MQD Metric File



- Create new metric and test case file using this option.
- Test case file should start with “test” and then metric file name. For example, if metric is named as quoted_spread, test case should be named as ‘test_quoted_spread’.

Key Metric Elements

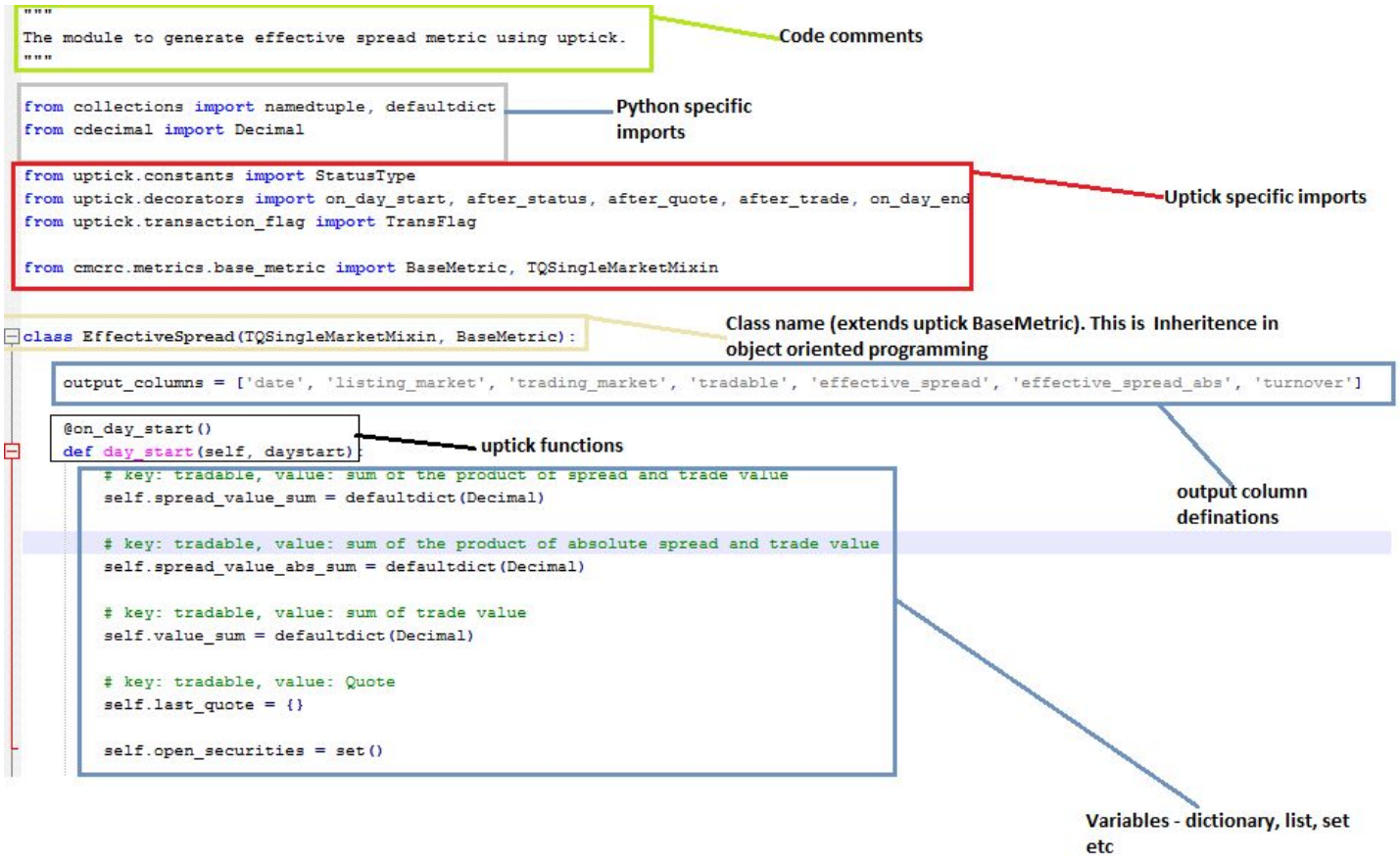
- Import packages – uptick and python specific packages
- Python variables (dictionary, list, set, Boolean etc.)
- Output fields definition: attributes that will be in output csv file.
- How do we construct the instrument list for which metric needs to generate the result?
  - **Scenario 1:** Using the instruments opened for trading.
  - **Scenario 2:** Using the list of instruments from a metric output (click here for example)
- Uptick functions –
  - Event Triggers. For example: @on_day_start, @after_status (flag input as StatusType.OPEN/PRE_OPEN/PRE_CLOSE/CLOSING/CLOSED), after_quote, after_trade (takes input such as on_market = True/False, @on_day_end

    In python terms these are called as 'decorator'.

  - Uptick pre-defined input objects(for e.g. trade, quote, status). Provides information such as: Security information, trade/quote , timestamp, trade price and volume, bid/ask price and volume information.

    For details on uptick pre-defined objects & variables, click here

Key Metric Elements (Cont...)



Key Metric Elements (Cont…)

```
@after_status(status=StatusType.OPEN)
def any_tradable_open(self, status):
    timestamp = status.timestamp
    tradable = status.instrument.meta['tradable']
    self.logger.debug('%s opened at %s', tradable, timestamp.time())
    self.open_securities.add(tradable)

@after_status(status=StatusType.PRE_CLOSE)
@after_status(status=StatusType.CLOSING)
@after_status(status=StatusType.CLOSED)
def any_tradable_close(self, status):
    timestamp = status.timestamp
    tradable = status.instrument.meta['tradable']
    self.logger.debug('%s closed at %s', tradable, timestamp.time())
    if tradable in self.open_securities:
        self.open_securities.remove(tradable)

@after_quote(on_market=True)
def update_last_price(self, quote):
    tradable = quote.instrument.meta['tradable']
    self.last_quote[tradable] = quote

@after_trade(on_market=True)
def update_spread_value_sum(self, trade):
    tradable = trade.instrument.meta['tradable']

    if tradable not in self.open_securities:
        self.logger.debug('security not opened yet, ignored')
        return

    if tradable in self.last_quote and self.last_quote[tradable].has_spread:
        last_quote = self.last_quote[tradable]
        price = trade.price
        volume = trade.volume
        flags = trade.flags
        is_dark = TransFlag.FLAG_DARK_TRADE in flags
        spread = self.spread_in_percent(last_quote.best_ask.price, last_quote.best_bid.price, price, is_dark)
        self.spread_value_sum[tradable] += spread * price * volume
        abs_spread = self.abs_spread(last_quote.best_ask.price, last_quote.best_bid.price, price, is_dark)
        self.spread_value_abs_sum[tradable] += 2 * abs_spread * price.amount * volume
        self.value_sum[tradable] += abs(price * volume)
    else:
        if tradable in self.last_quote:
            self.logger.warn('the quote price (%s) of %s is invalid, ignored',
                             self.last_quote[tradable], tradable)
        else:
```

Uptick function executes when trading phase of security changes to Open

This python "set" variable captures all the open securities universe on which metrics will be computed

Uptick function executes when trading phase of security changes to Pre-Close/Closing/Closed state

provides information such as bid/ask quote price and volume, security details

provides information such as trade price, volume, security details

uptick function executes when there is an on_market quote event for a security.

Uptick function executes when there is a on_market trade event for a security.

Users can choose their own function names (which starts with "def …). However, for good naming style, this should ideally be self-explanatory in terms of purpose of the function

Key Metric Elements (Cont…)

```
@staticmethod
def abs_spread(best_ask_price, best_bid_price, trade_price, is_dark):
    midpoint = (best_ask_price + best_bid_price) * Decimal(0.5)
    if is_dark:
        # return zero, retaining currency and decimal places
        return 0 * trade_price
    else:
        return abs(trade_price - midpoint)


@staticmethod
def spread_in_percent(best_ask_price, best_bid_price, trade_price, is_dark):
    abs_spread = EffectiveSpread.abs_spread(best_ask_price, best_bid_price,
                                            trade_price, is_dark)
    midpoint = (best_ask_price + best_bid_price) * Decimal(0.5)
    # multiply by 100 so that the spreads are in "percent"
    return Decimal('200.00') * abs_spread / midpoint


@staticmethod
def results_generator(listing_market, trading_market, date,
                      spread_value_sum, abs_spread_value_sum,
                      value_sum):
    for tradable in value_sum:
        if value_sum[tradable] > 0:
            row = {
                'date': date,
                'listing_market': listing_market,
                'trading_market': trading_market,
                'tradable': tradable,
                'effective_spread':
                    spread_value_sum[tradable] / value_sum[tradable],
                'effective_spread_abs':
                    repr(abs_spread_value_sum[tradable] / value_sum[tradable].amount),
                'turnover': value_sum[tradable]
            }
            yield row
```

User defined functions

```
@on_day_end()
def day_end(self, dayend):
    for row in self.results_generator(self.listing_market, self.trading_market,
                                      dayend.timestamp.date(),
                                      self.spread_value_sum, self.spread_value_abs_sum,
                                      self.value_sum):
        self.write_row(row)
```

Function executes on end-of-the day. This is mainly used for final computation and generation of results for metrics meant for daily frequency. The final output has one record row per security.

```
metric = EffectiveSpread()
```

Create instance of metric class. Refer to Python object oriented programming for more details.

User defined functions does not have uptick decorators (like after_quote , after_trade, after_status etc) and take any set of user defined input variables. When would you usually like to define your own functions?

When do we use variables or functions with self

- Self refers to current instance of the “Class”
- Variables starting with “self” means the variables values belong to instance of the class. The values can be accessed anywhere within any functions having access to self.
- In function, it refers to the instance whose method is called.

So do all functions and variables need to have self?

Quick facts: *Concept of using “self” is common to object oriented program (OOP) languages (Java has something called “this”).*

Definition of Class: *A class is a blueprint for creating objects (a particular data structure), providing initial values for state (member variables a.k.a attributes), and implementations of behavior (member functions a.k.a methods) (reference: https://brilliant.org/wiki/classes-oop/)*

Coding Best Practices and Due Diligence

- Is the data structure appropriate? For example correct usage of dictionary definitions. Common dictionary key is tradable. Dictionary could be multi-dimensional (i.e. having multiple keys).

  Exercise: in which case do you think we will use interval as another dictionary key along with tradable? Click here for an example. Can you provide example of keys?

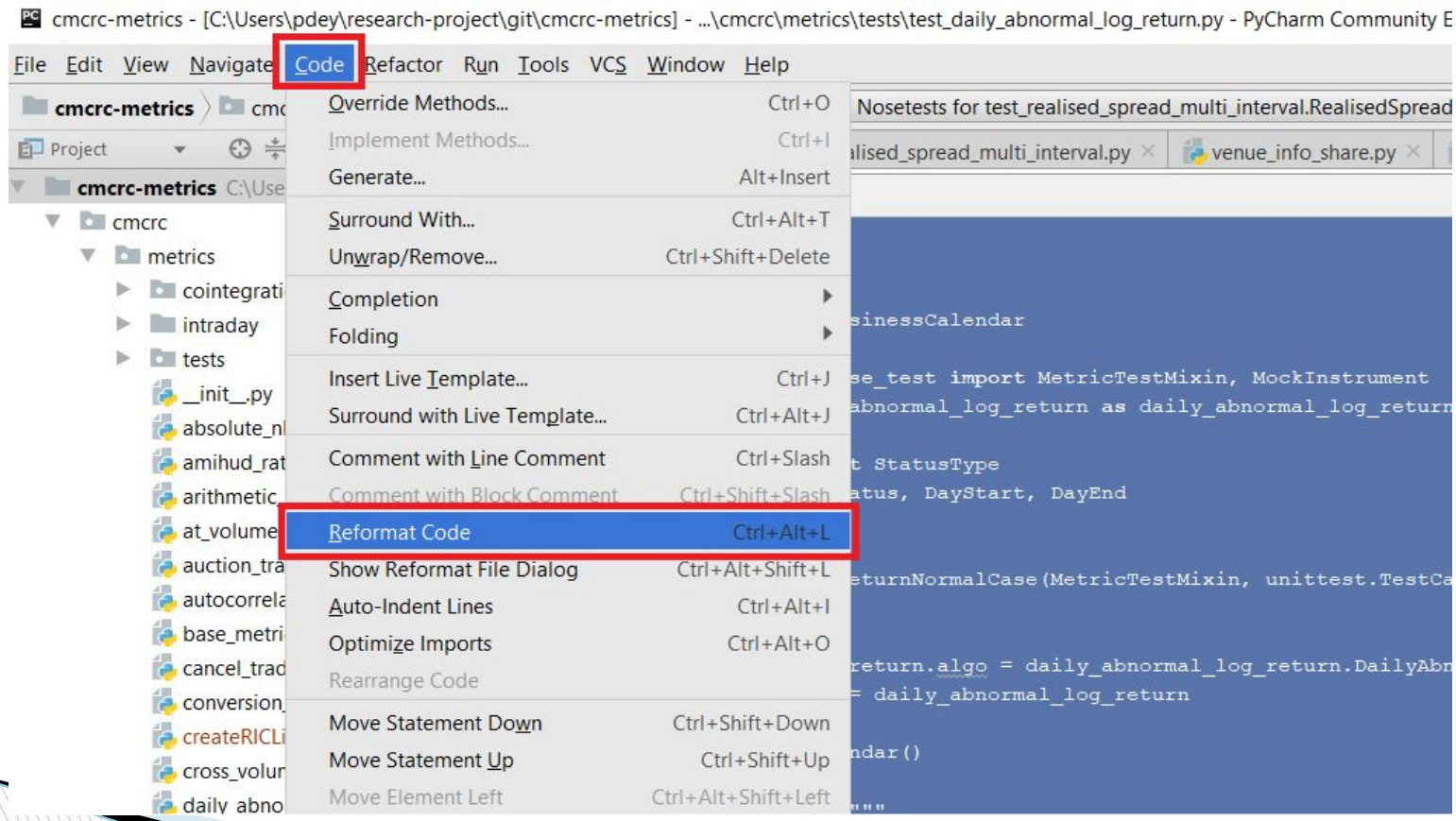
- Are we missing any trading phases?

- Are we using multiple nested loops causing potential performance bottleneck?

- What if any data is not available – exceptions scenarios? Is metric code robust to both liquid and illiquid stocks.

Coding Best Practices and Due Diligence (Cont…)

Are we following appropriate code style/formatting (Pycharm IDE can help you!)



Metrics Case Study Examples

Metric to measure Quote Volatility and Effective Spread:

- Which uptick functions we need to use?
- What are the key user defined variables and uptick pre-defined variables that are required?
- What are the key considerations for using trade and quote?
  Handling one sided quote, invalid spread.
- Which trading phases are involved in the computation?
- What happens when there is trade halt/suspension?
- How do we perform the computations? For example, we "online algorithms" for standard deviations, autocorrelation could be faster.

Example Exception checks:

- Is the quote valid (ask >bid)
- What happens if a stock has few sample sizes?
- Singular matrix problems (for example price series for illiquid stocks)

Quote Volatility code, Effective Spread Code

Metrics Test Cases

Metrics Test Cases

- Why unit test matters?
  - Minimize bugs in the code.
  - Make code robust in handling various input scenarios.
  - Improves quality of code.

- Creating metrics test cases and key elements of the test case:
  - Create new test case starting with test_<metric_file_name>.py.
  - Define test case python class per test scenario.
  - What should test scenario cover?
    - Given test data feed, what should be the expected output?
  - How many scenarios? Break up metric code into logical subparts – does test case at least covers key scenarios handled by the code? Less unit test coverage could lead to more issues during integration testing and end output.

Test Case Elements

```
class RealisedSpreadTestBuyerInitPositive(unittest.TestCase, MetricTestMixin):
    """
    Unit tests for uptick realised spread metric
    (trade is buyer-initiated and the 10-min later mid_price is lower than trade price)
    """
    metric_module = realised_spread
    feed = [
        DayStart(
            timestamp=datetime.datetime(2014, 1, 1, 0, 0),
            venue='asx',
        ),
        Status(
            instrument=MockInstrument('asx:BHP:AUD'),
            timestamp=datetime.datetime(2014, 1, 1, 10, 0),
            venue='asx',
            status_name=StatusType.OPEN,
        ),
        # mid point price (mp1): 10.00
        Quote(
            instrument=MockInstrument('asx:BHP:AUD'),
            timestamp=datetime.datetime(2014, 1, 1, 10, 5),
            ask_price=Money.loads('AUD 12.00'),
            ask_volume=1,
            bid_price=Money.loads('AUD 8.00'),
            bid_volume=1,
            venue='asx',
            on_market=True,
        ),
        # trade price (tp): 12.00, direction (d): +1
        Trade(
            instrument=MockInstrument('asx:BHP:AUD'),
            timestamp=datetime.datetime(2014, 1, 1, 10, 6),
            price=Money.loads('AUD 12.00'),
            volume=1,
            venue='asx',
            on_market=True,
            flags=[],
        ),
        # mid point price (mp2): 11.00
        Quote(
            instrument=MockInstrument('asx:BHP:AUD'),
            timestamp=datetime.datetime(2014, 1, 1, 10, 16),
            ask_price=Money.loads('AUD 12.00'),
            ask_volume=1,
            bid_price=Money.loads('AUD 10.00'),
            bid_volume=1,
            venue='asx',
            on_market=True,
        ),
        DayEnd(
            timestamp=datetime.datetime(2014, 1, 1, 23, 59, 59, 999999),
            venue='asx',
        ),
    ]
```

Test case 'Class'. Represents Test Scenario. Extends unittest.TestCase (Python inheritence)

Metric file name where the metric is coded. This test case will test this metric.

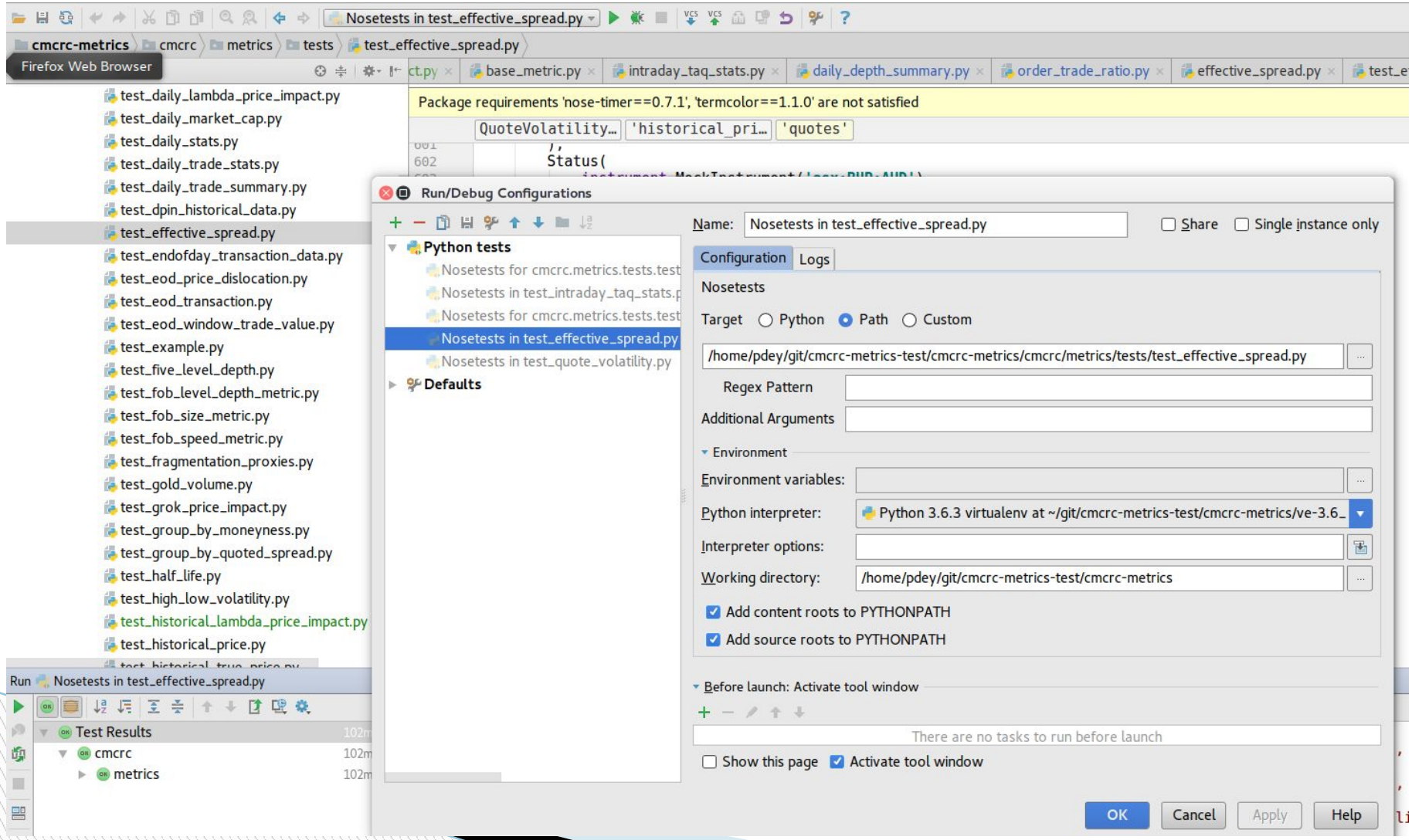
Input data feed when which metric will be tested

Expected output

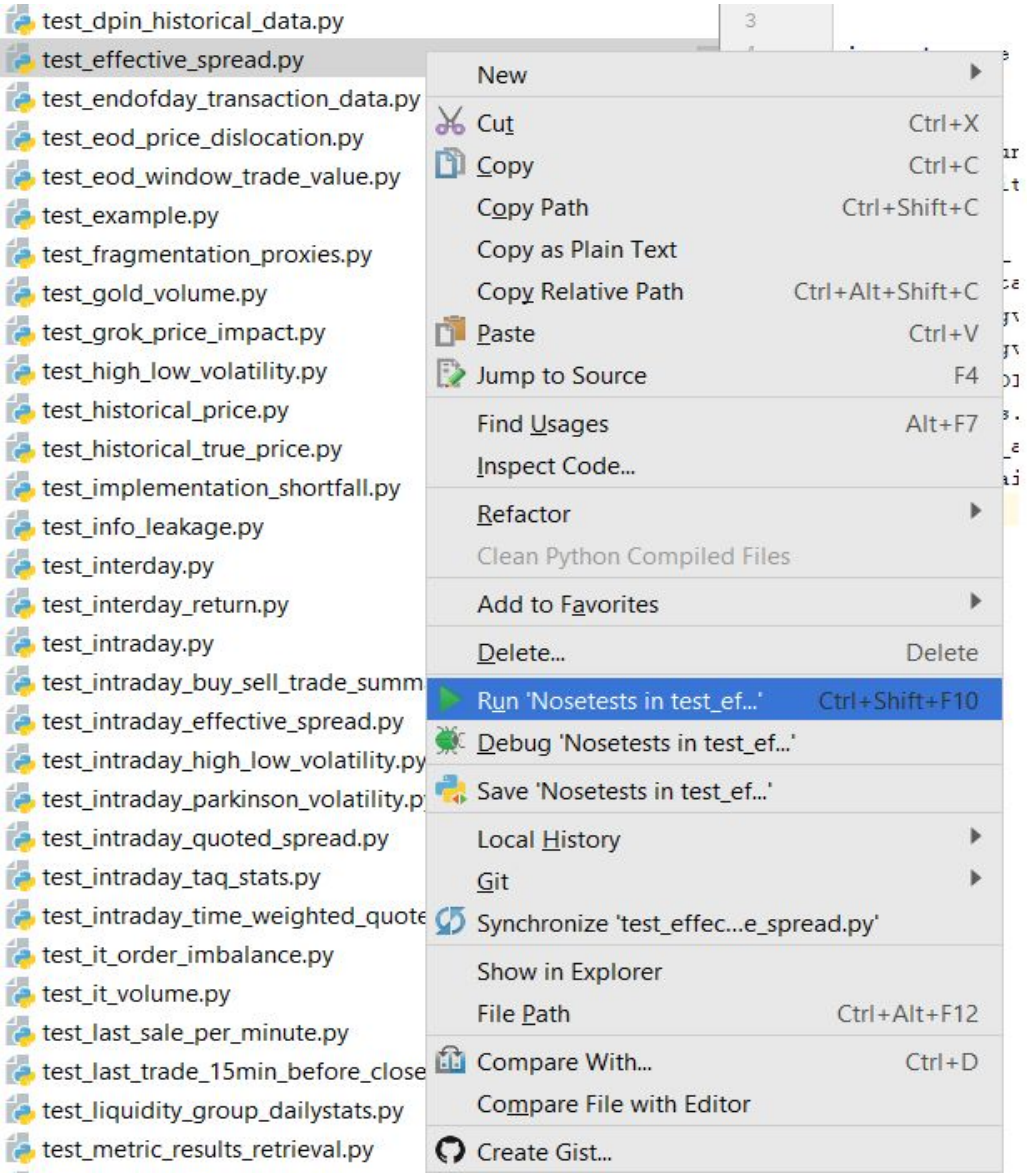
```
# realised spread abs: 2*(tp - mp2)*d  = 2.00
expected_metric = [
    {u'date': u'2014-01-01',
     u'tradable': u'asx:BHP:AUD',
     u'listing_market': u'asx',
     u'trading_market': u'asx',
     u'realised_spread': u'20.0',
     u'realised_spread_abs': u'AUD 2.000'},
]
```

Always manually verify the expected output values based on your test data!

Running Test Case



Running Test Case (Cont...)



Test Scenarios case study (Effective Spread)

- What is Effective Spread MQD Metric? It measures how much the actual trade price deviates from the mid-point. It represents the actual, round-trip cost of trading to the liquidity demander.

$$EffSpr_{v,s,d} = \sum 200 * D_{v,s,d,t} * \frac{P_{v,s,d,t} - Mid_{v,s,d,t}}{Mid_{v,s,d,t}} * W_{v,s,d,t}$$

P = trade price, M = mid point price, W = value weights, D = direction of trade

- What scenarios you can think of?

  What should be the Expected output when market has:
  - Both buy and sell trades
  - Only buy trades or only sell trades
  - No trades
  - Holiday
  - What could be other scenario?

- Input feed

  a) Check if it is dependant on any other metrics?

  b) Create test data feed

- What could be output Attribute?

Click here to go to the test code

Test Scenarios case study (Quote Volatility)

- What is Quote Volatility MQD Metric?
  Standard deviation of 1 minute midpoint price return.

- What scenarios you can think of?
  What should be the expected output when market has:
  - Series of on-market quotes (simple case)
  - Invalid spread (example: bid>=ask)
  - Bid or Ask side is unavailable
  - What could be other scenarios?
- Input feed
  - Check if it is dependant on any other metrics?
  - Create test data feed
- What could be output Attribute?

Click here to go to the test code

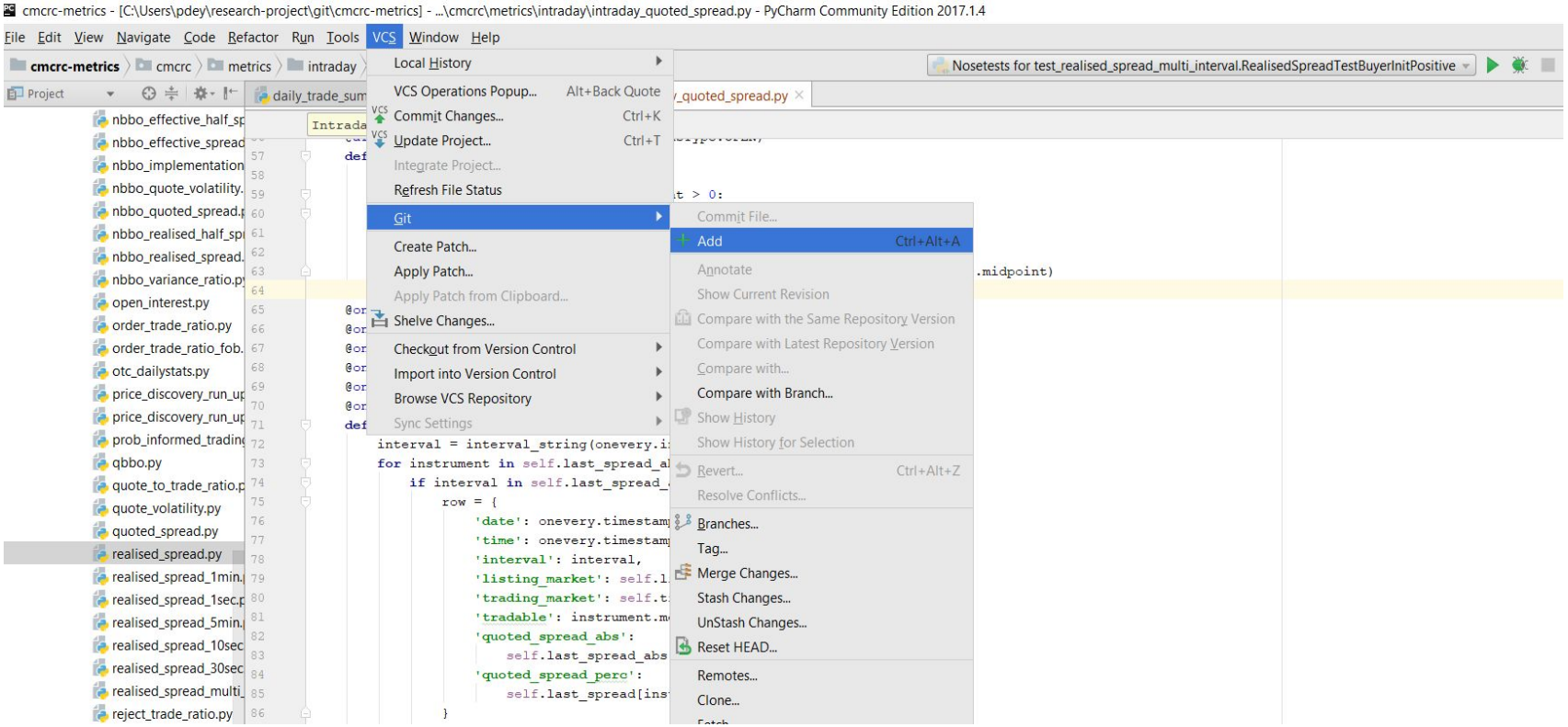
Review and Deployment Process

- Ensure code is complete and unit tested.
- Mark JIRA status “In Progress” before committing the changes” to git repository.
- Submitting changes to code repository for review. Tutorial about learning about Git tool in pycharm and git (in general), please refer to “https://www.jetbrains.com/help/pycharm/using-git-tutorial.html#d387025e2”
  - Use git to commit and push the changes
  - Add appropriate code comments including the JIRA number.
- Code is submitted for review in bit-bucket tool. The changes will be merged to Master repository once reviewed/accepted by all code reviewers.

Bit bucket link

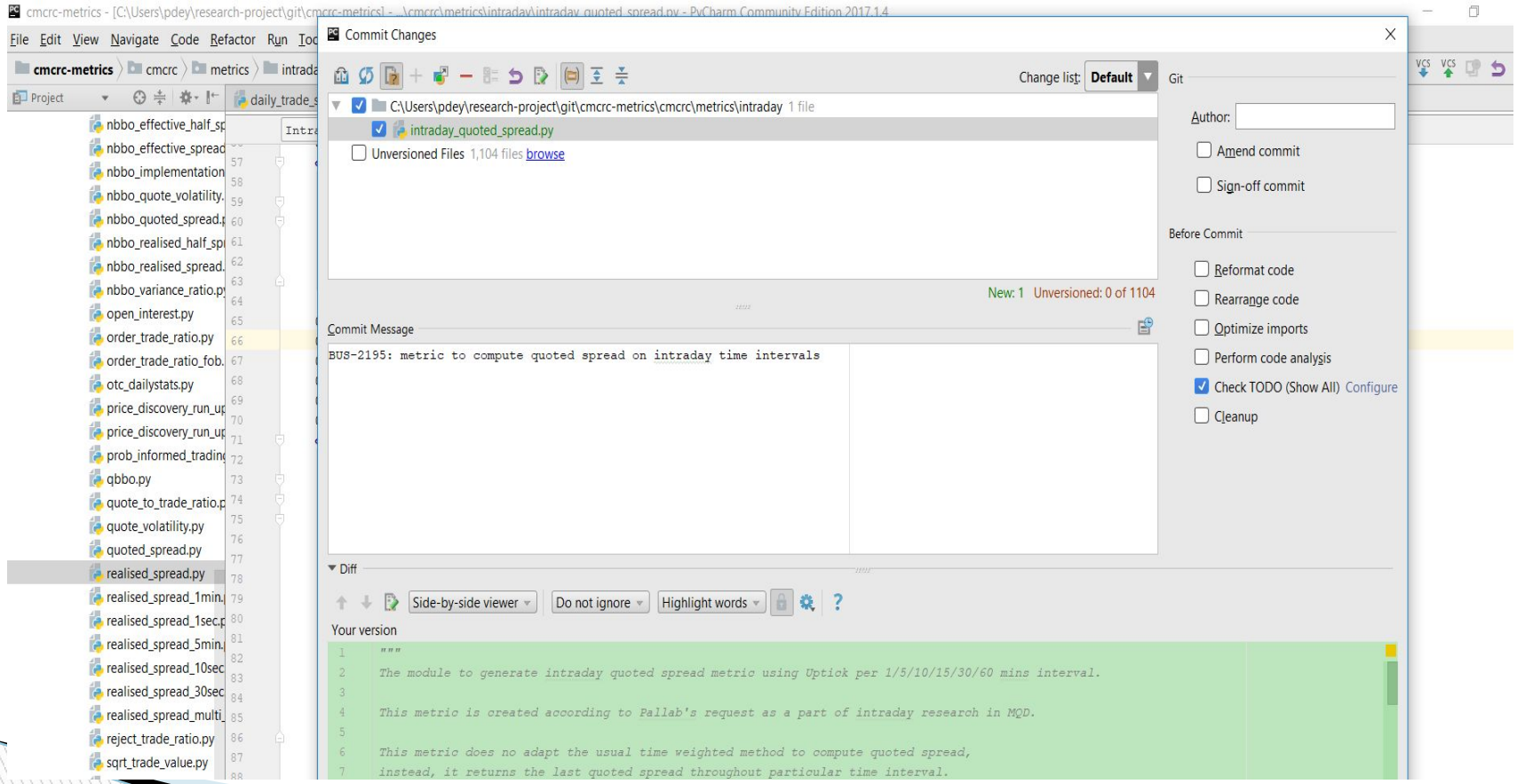
Submitting Metric Changes

If new file, then perform git–>add, else directly perform git “commit” or “git push” so that changes are submitted to remote repository



Submitting Metric Changes (Cont..)

Sample git commit operation. The commit message should start with JIRA name, colon ":" followed by user defined commit message. Commit message should briefly highlights the change



Mqdashboard Analytics

MQD Analytics

What is mqd analytics?

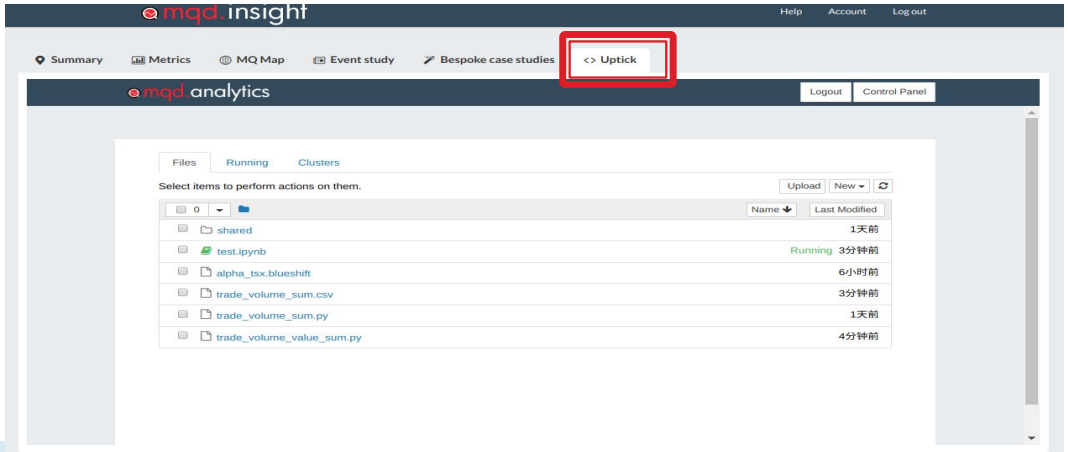
It is a python framework in a jupyter platform for financial metric calculation and result display

What is Jupyter?

The Jupyter Notebook is an open-source web application that allows you to create and share documents that contain live code

https://jupyter.readthedocs.io/en/latest/

How to use mqd analytics?



How can you do research in mqd analytic?

1. Read and investigate converted data
2. Write New Metric or obtain existing metric result directly
3. Customized research report

1. Read and Investigate Data

- 1) Day Start

daystart: DayStart(meta={}, timestamp=2003-01-06 00:00:00+11:00, venue=asx)

status: Status(flags=frozenset({''}), instrument=asx:BHP:AUD, meta={'trth_ric': 'BHP.AX', 'trth_qualifiers': ()}, on_market=True, status_name=PRE_OPEN, timestamp=2003-01-06 07:00:00+11:00, venue=asx)

- 2) Open

status: Status(flags=frozenset({''}), instrument=asx:BHP:AUD, meta={'trth_ric': 'BHP.AX', 'trth_qualifiers': ()} , on_market=True, status_name=OPEN, timestamp=2003-01-06 10:09:15+11:00, venue=asx)

- 3) Closed

status: Status(flags=frozenset({''}), instrument=asx:BHP:AUD, meta={'trth_ric': 'BHP.AX', 'trth_qualifiers': ()}, on_market=True, status_name=CLOSED, timestamp=2003-01-06 16:16:15+11:00, venue=asx)

1. Read and Investigate Data (Cont..)

- 1) Day Start and Pre-Open

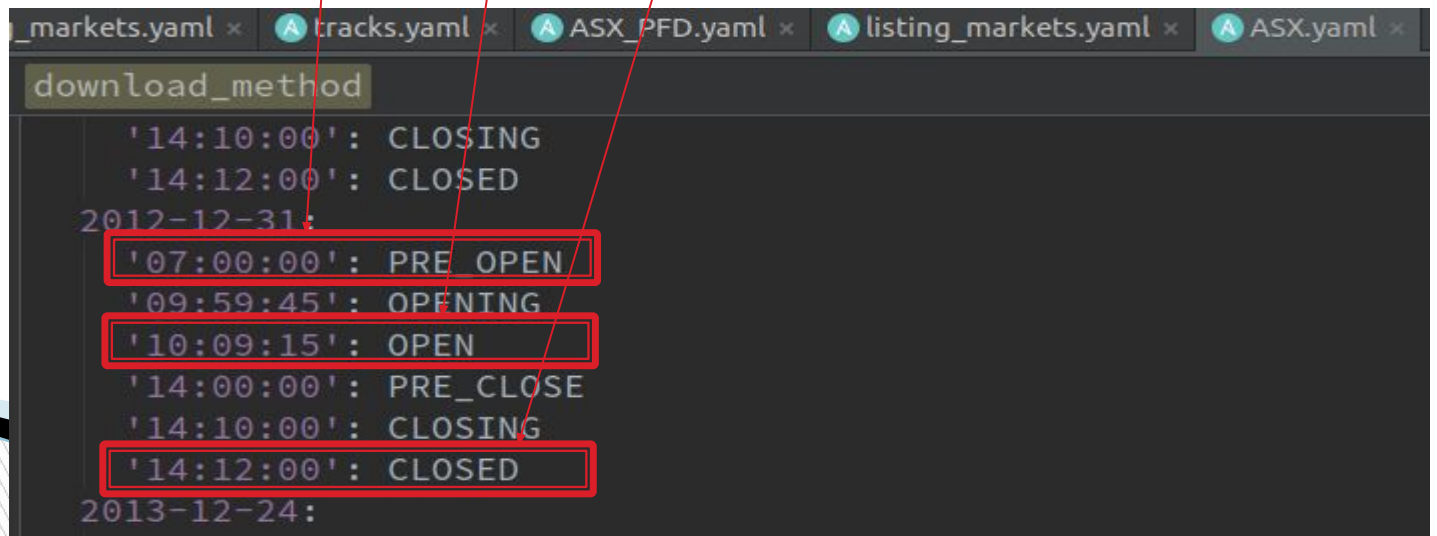
```
daystart: DayStart(meta={}, timestamp=2003-01-06 00:00:00+11:00, venue=asx)
status: Status(flags=frozenset({''}), instrument=asx:BHP:AUD, meta={'trth_ric': 'BHP.AX', 'trth_qualifiers': ()},
 on_market=True, status_name=PRE_OPEN, timestamp=2003-01-06 07:00:00+11:00, venue=asx)
```

- 2) Open

```
status: Status(flags=frozenset({''}), instrument=asx:BHP:AUD, meta={'trth_ric': 'BHP.AX', 'trth_qualifiers': ()},
 on_market=True, status_name=OPEN, timestamp=2003-01-06 10:09:15+11:00, venue=asx)
```

- 3) Closed

```
status: Status(flags=frozenset({''}), instrument=asx:BHP:AUD, meta={'trth_ric': 'BHP.AX', 'trth_qualifiers': ()},
 on_market=True, status_name=CLOSED, timestamp=2003-01-06 16:16:15+11:00, venue=asx)
```



1. Read and Investigate Data (Cont..)

- 4) Trade
- A) From TRTH

RIC	Date[L]	Time[L]	Type	Ex	Price	Volume	Market VWAP	Bu	Se	Qualifiers
BHP.AX	20170803	09:11:37.632300	Trade		25.5367	100000				LTXT [GV4_TEXT]

- B) From blueshift

trade: Trade(buy_order_id=None, currency=AUD, flags=frozenset({'Xi', 'OF', 'DT'}), instrument=asx:BHP:AUD, meta={'trth_ric': 'BHP.AX', 'trth_qualifiers': (('LTXT ', 'GV4_TEXT'),)}, on_market=False, price=25.5367, sell_order_id=None, status_name=PRE_OPEN, timestamp=2017-08-03 09:11:37.632300+10:00, venue=asx, volume=100000)

1. Read and Investigate Data (Cont.)

- 4) Trade
- A) From TRTH

```
g_markets.yaml × | tracks.yaml × | ASX_PFD.yaml × | listing_markets
download_method
request_split_regex: '^[^!].*\.AX$'
timezone: 'Australia/Sydney'
trading_market: 'asx'
listing_market: 'asx'
currency: 'AUD'
```

RIC	Date[L]	Time[L]	Type	Ex	Price	Volume	Market VWAP	Bid	Ask	Qualifiers
BHP.AX	20170803	09:11:37.632300	Trade		25.5367	100000				LTXT [GV4_TEXT]

- B) From blueshift

trade: Trade(buy_order_id=None, currency=AUD, flags=frozenset({'Xi', 'OF', 'DT'}), instrument=asx:BHP:AUD, meta={'trth_ric': 'BHP.AX', 'trth_qualifiers': (('LTXT ', 'GV4_TEXT'),)}, on_market=False, price=25.5367, sell_order_id=None, status_name=PRE_OPEN, timestamp=2017-08-03 09:11:37.632300+10:00, venue=asx, volume=100000)

```
GV4_TEXT:
  /^\s*LTXT\s*$/: [FLAG_DARK_TRADE, FLAG_CROSSING, FLAG_OFF_MARKET]
```

1. Read and Investigate Data (Cont..)

- 5) Quote
- A) From TRTH

RIC	Date[L]	Time[L]	Type	Bi	Bid Price	Bid Size	No. Buyers	Si	Ask Price	Ask Size	Ni	Qu	Si	Bi	As	Tr	Quote Time
BHP.AX	20030106	10:09:20.732222	Quote			150663	4										23:09:00.000

- B) From blueshift

quote: Quote(LevelsProxy=<class 'uptick.types.Quote.LevelsProxy'>, ask_prices=(AUD 9.95,), ask_volumes=(Decimal('138708'),), bid_prices=(AUD 9.94,), bid_volumes=(Decimal('150663'),), currency=AUD, flags=frozenset({''}), instrument=asx:BHP:AUD, lot_size=1, meta={'trth_type': 'Quote', 'trth_ric': 'BHP.AX', 'trth_qualifiers': ()}, on_market=True, set_asks=<bound method Quote.set_asks of <uptick.types.Quote object at 0x7f7ce0398dd8>>, set_bids=<bound method Quote.set_bids of <uptick.types.Quote object at 0x7f7ce0398dd8>>, status_name=OPEN, timestamp=2003-01-06 10:09:20.732222+11:00, venue=asx)

1. Read and Investigate Data (Cont..)

- 5) Quote
- A) From TRTH

RIC	Date[L]	Time[L]	Type	Bi	Bid Price	Bid Size	Ne	Se	Ask Price	Ask Size	Ne	Qu	Se	Bi	As	Tr	Quote Time
BHP.AX	20030106	10:09:02.567096	Quote						9.94	14337	1						23:09:00.000
BHP.AX	20030106	10:09:16.357502	Quote		9.94	85663	1										23:09:00.000
BHP.AX	20030106	10:09:16.357502	Quote						9.95	138708	2						23:09:00.000
BHP.AX	20030106	10:09:20.732222	Quote			150663	4										23:09:00.000

- B) From blueshift

```
quote: Quote(LevelsProxy=<class 'uptick.types.Quote.LevelsProxy'>, ask_prices=(AUD 9.95,), ask_volumes=(Decimal('1
38708'),), bid_prices=(AUD 9.94,), bid_volumes=(Decimal('150663'),), currency=AUD, flags=frozenset({''}), instrume
nt=asx:BHP:AUD, lot_size=1, meta={'trth_type': 'Quote', 'trth_ric': 'BHP.AX', 'trth_qualifiers': ()}, on_market=Tr
ue, set_asks=<bound method Quote.set_asks of <uptick.types.Quote object at 0x7f7ce0398dd8>>, set_bids=<bound metho
d Quote.set_bids of <uptick.types.Quote object at 0x7f7ce0398dd8>>, status_name=OPEN, timestamp=2003-01-06 10:09:2
0.732222+11:00, venue=asx)
```

2. Obtain Metric Result

1. Write a new metric on blueshift or your own data

2. Obtain existing metric result stored in mqdashboard data base

2. Obtain Metric Result (Cont..)

1) Write and run your own metric

```
%%writefile trade_volume_value_sum.py

from uptick import *


class TradeVolumeValueSum(Metric):

    @on_day_start
    def daystart(self, daystart):
        self.volume_sum = 0
        self.value_sum = 0

    @on_trade
    def trade(self, trade):
        self.volume_sum += trade.volume
        self.value_sum += (trade.volume * trade.price)

    @on_day_end
    def dayend(self, dayend):
        print('Calculated trade volume sum on %s: %s' % (dayend.timestamp.date(), self.volume_sum))
        self.output.write_row({
            'date': dayend.timestamp.date(),
            'volume': self.volume_sum,
            'value': self.value_sum,
        })

metric = TradeVolumeValueSum()
```

Python Metric Name

Python Class Name

Python Output Fields

2. Obtain Metric Result (Cont..)

1) Write and run your own metric

```
!uptick trade_volume_value_sum --feed-reader-class uptick_blueshift.reader.BlueshiftFeedReader --feed
./alpha_tsx.blueshift --output=CsvOutput --output-filename=trade_volume_value_sum.csv --loglevel ERROR
```

```
Calculated trade volume sum on 2015-08-21: 40926825
Calculated trade volume sum on 2015-08-24: 59260233
Calculated trade volume sum on 2015-08-25: 45905244
Calculated trade volume sum on 2015-08-26: 44738387
Calculated trade volume sum on 2015-08-27: 49806928
Calculated trade volume sum on 2015-08-28: 39051216
Calculated trade volume sum on 2015-08-31: 43349833
Calculated trade volume sum on 2015-09-01: 43822265
Calculated trade volume sum on 2015-09-02: 42096859
Calculated trade volume sum on 2015-09-03: 35825406
Calculated trade volume sum on 2015-09-04: 27542465
Calculated trade volume sum on 2015-09-08: 32198844
Calculated trade volume sum on 2015-09-09: 39381927
Calculated trade volume sum on 2015-09-10: 35435772
```

Python Metric Name

Csv Metric Output

Blueshift Input

2. Obtain Metric Result (Cont..)

2) Obtain existing metric result in database

```
y_variable = ['eff_spread']
x_variables = ['intraday_volatility', 'log_total_trade_volume']
variables = y_variable + x_variables
date_start = '2017-03-29'
date_end = '2017-03-31'
market = 'ASX'
market_id = str(market_id_by_name[market])
country = country_by_name[market]
level = 'venue'
#level = 'instrument'
instrument_list = [32771, 24582]
reg_formula= y_variable[0] + ' ~ ' + " + ".join(x_variables)
print ('regression formula:', reg_formula)
print ('market:', market, 'country:', country)
print ('date from:', date_start, 'date to:', date_end)

variable_dict = store_data(variables, level, date_start, date_end, market_id, country, instrument_list)
```

Variables Input

Date Range and Market

Particular Instrument or all instruments

3. Customize Your Report

1. Display the result in dataframe

https://pandas.pydata.org/pandas-docs/stable/generated/pandas.DataFrame.html

2. Show descriptive statistics

http://hamelg.blogspot.com.au/2015/11/python-for-data-analysis-part-21.html

3. Show regression result

https://www.statsmodels.org/stable/index.html

4. Construct charts

https://www.highcharts.com/

4. Display Your Research Results

- 1) DataFrame

```
In [31]: import pandas as pd
         # 1) Display in dataframe
         df=pd.read_csv('regression_sample.csv')
         df
```

Out[31]:

	date	year	security	eff_spread	intraday_volatility	log_total_trade_volume
0	2017-03-29	2017	1AG	757.344492	0.021990	14.731269
1	2017-03-30	2017	1AG	282.648852	0.008309	12.659305
2	2017-03-31	2017	1AG	298.507463	0.001783	12.589735
3	2017-03-29	2017	3DP	427.559054	0.001787	12.063495
4	2017-03-30	2017	3DP	279.922924	0.008257	13.949449
5	2017-03-31	2017	3DP	612.835494	0.001879	13.477900
6	2017-03-29	2017	3PL	57.301775	0.000622	11.672959
7	2017-03-30	2017	3PL	65.559570	0.002820	10.914743
8	2017-03-31	2017	3PL	107.450813	0.001617	11.767909
9	2017-03-29	2017	4CE	468.817218	0.012338	13.949923

4. Display Your Research Results (Cont..)

- 2) Descriptive Statistics

```
In [29]: df[['eff_spread', 'intraday_volatility', 'log_total_trade_volume']].describe()
```

Out[29]:

	eff_spread	intraday_volatility	log_total_trade_volume
count	4440.000000	4440.000000	4440.000000
mean	365.064871	0.003893	12.028473
std	661.840783	0.006197	2.358921
min	0.000000	0.000000	0.000000
25%	32.510300	0.000590	10.637423
50%	159.998077	0.001787	12.218357
75%	425.531915	0.004750	13.666133
max	9189.189189	0.091141	19.171917

4. Display Your Research Results (Cont..)

- 3) Regression

```
In [33]: import statsmodels.formula.api as smf
         # 2) Generate OLS regression result
         y_variable = ['eff_spread']
         x_variables = ['intraday_volatility', 'log_total_trade_volume']
         reg_formula= y_variable[0] + ' ~ ' + " + ".join(x_variables)
         lm = smf.ols(formula=reg_formula, data=df)
         result = lm.fit()
         print ('OLS Regression:', reg_formula)
         print (result.summary())
```

```
OLS Regression: eff_spread ~ intraday_volatility + log_total_trade_volume
                            OLS Regression Results
==============================================================================
Dep. Variable:              eff_spread   R-squared:                       0.087
Model:                             OLS   Adj. R-squared:                  0.086
Method:                  Least Squares   F-statistic:                     211.0
Date:                 Sat, 24 Mar 2018   Prob (F-statistic):           2.88e-88
Time:                         01:44:06   Log-Likelihood:                -34936.
No. Observations:                4440   AIC:                          6.988e+04
Df Residuals:                    4437   BIC:                          6.990e+04
Df Model:                           2
Covariance Type:            nonrobust
================================================================================================
                                   coef    std err          t      P>|t|      [0.025      0.975]
------------------------------------------------------------------------------------------------
Intercept                      400.3430     49.694      8.056      0.000     302.918     497.768
intraday_volatility         3.232e+04   1579.253     20.465      0.000    2.92e+04    3.54e+04
log_total_trade_volume        -13.3941      4.149     -3.228      0.001     -21.528      -5.260
==============================================================================
Omnibus:                     5073.720   Durbin-Watson:                   1.282
Prob(Omnibus):                  0.000   Jarque-Bera (JB):           496674.903
Skew:                           5.953   Prob(JB):                         0.00
Kurtosis:                      53.428   Cond. No.                     2.05e+03
==============================================================================

Warnings:
[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
[2] The condition number is large, 2.05e+03. This might indicate that there are
strong multicollinearity or other numerical problems.
```

4. Display Your Research Results (Cont..)

- 4) Chart

- Required Packages

```
import datetime
import csv
import math
import statsmodels.api as sm
import statsmodels.formula.api as smf
import pandas as pd
import glob
from highcharts import Highchart, Highstock
```

- General Option

```
# 1. General Set Up
# 1) General Options
OPTIONS_PIE = {
    'chart': {
        'plotBackgroundColor': None,
        'plotBorderWidth': None,
        'plotShadow': False
    },
    'tooltip': {
        'pointFormat': '{series.name}: <b>{point.percentage:.1f}%</b>'
    },
}

OPTIONS_LINE = {
    'tooltip': {'valueDecimals': 2, 'crosshairs': [True, True]},
    'legend': {'enabled': True},
    'chart': {'zoomType': 'x'},
}

OPTIONS_SPLINE = {
    'chart': {'zoomType': 'x'},
    'rangeSelector': {'enabled': False},
    'navigator': {'enabled': False},
    'scrollbar': {'enabled': False},
}
```

4. Display Your Research Results (Cont..)

- a) Line Chart

In [49]:
```
# 2. Draw line chart
# 1) Set up line chart option
options = OPTIONS_SPLINE
options['title'] = {'text': 'trade volume versus trade value'}
options['yAxis'] = {'opposite': False,
                    }
```

In [50]:
```
# 2) Add data
LINE_CHART = Highstock()
LINE_CHART.set_dict_options(options)
series = read_time_series_csv('trade_volume_value_sum.csv', 'date')
for serie in series:
    LINE_CHART.add_data_set(series[serie], name=serie)
LINE_CHART
```

4. Display Your Research Results (Cont..)

- a) Line Chart

4. Display Your Research Results (Cont..)

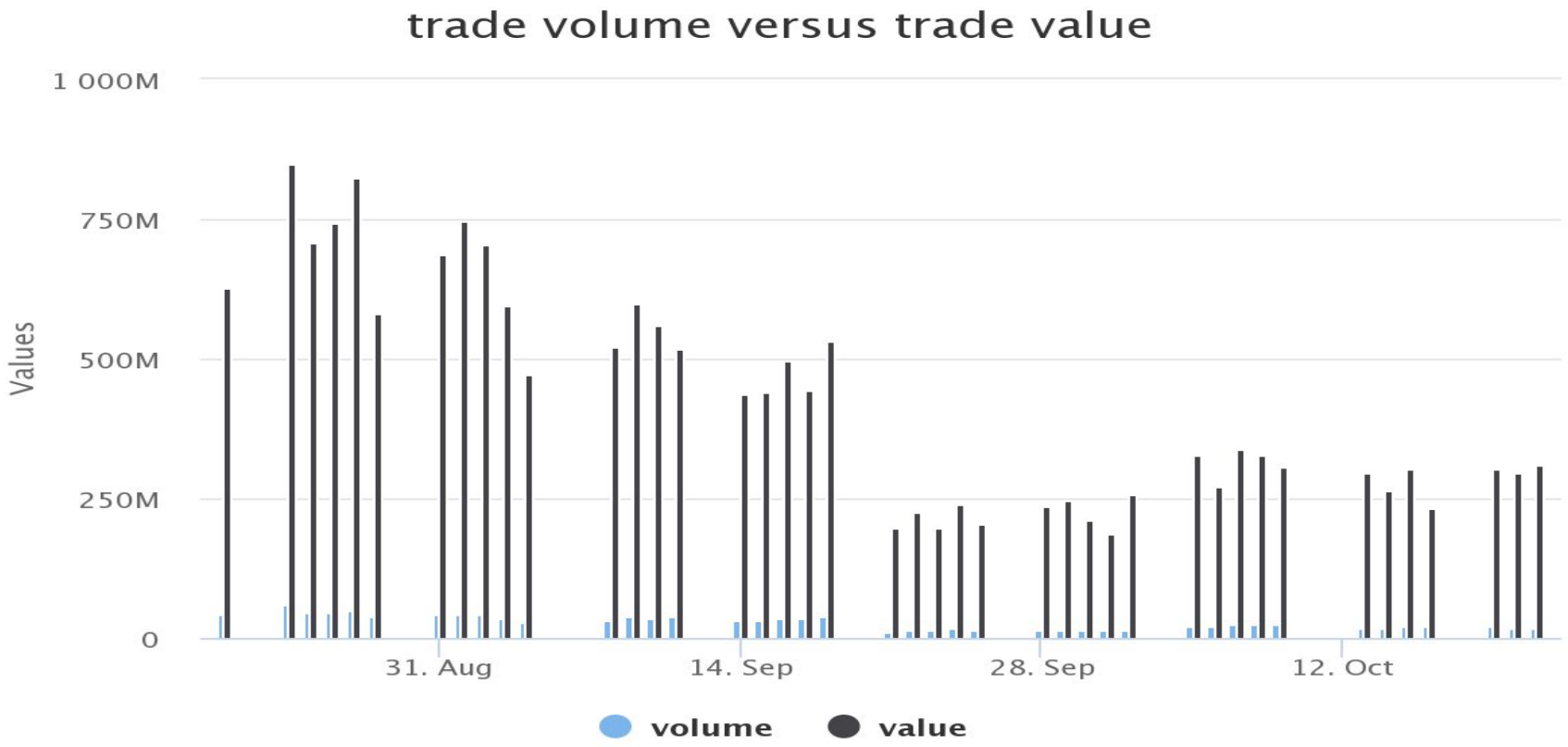
- b) Column Chart

In [51]:
```
# 2. Draw column chart
# 1) Set up column chart option
options = OPTIONS_LINE
options['title'] = {'text': 'trade volume versus trade value'}
options['yAxis'] = {'opposite': False,
                    }
options['xAxis'] = {'type': 'datetime',
                    }
```

In [53]:
```
# 2) Add data
COLUMN_CHART = Highchart()
COLUMN_CHART.set_dict_options(options)
for serie in series:
    COLUMN_CHART.add_data_set(series[serie], name=serie, type='column')
COLUMN_CHART
```

4. Display Your Research Results (Cont..)

- b) Column Chart



4. Display Your Research Results (Cont..)

- c) Pie Chart

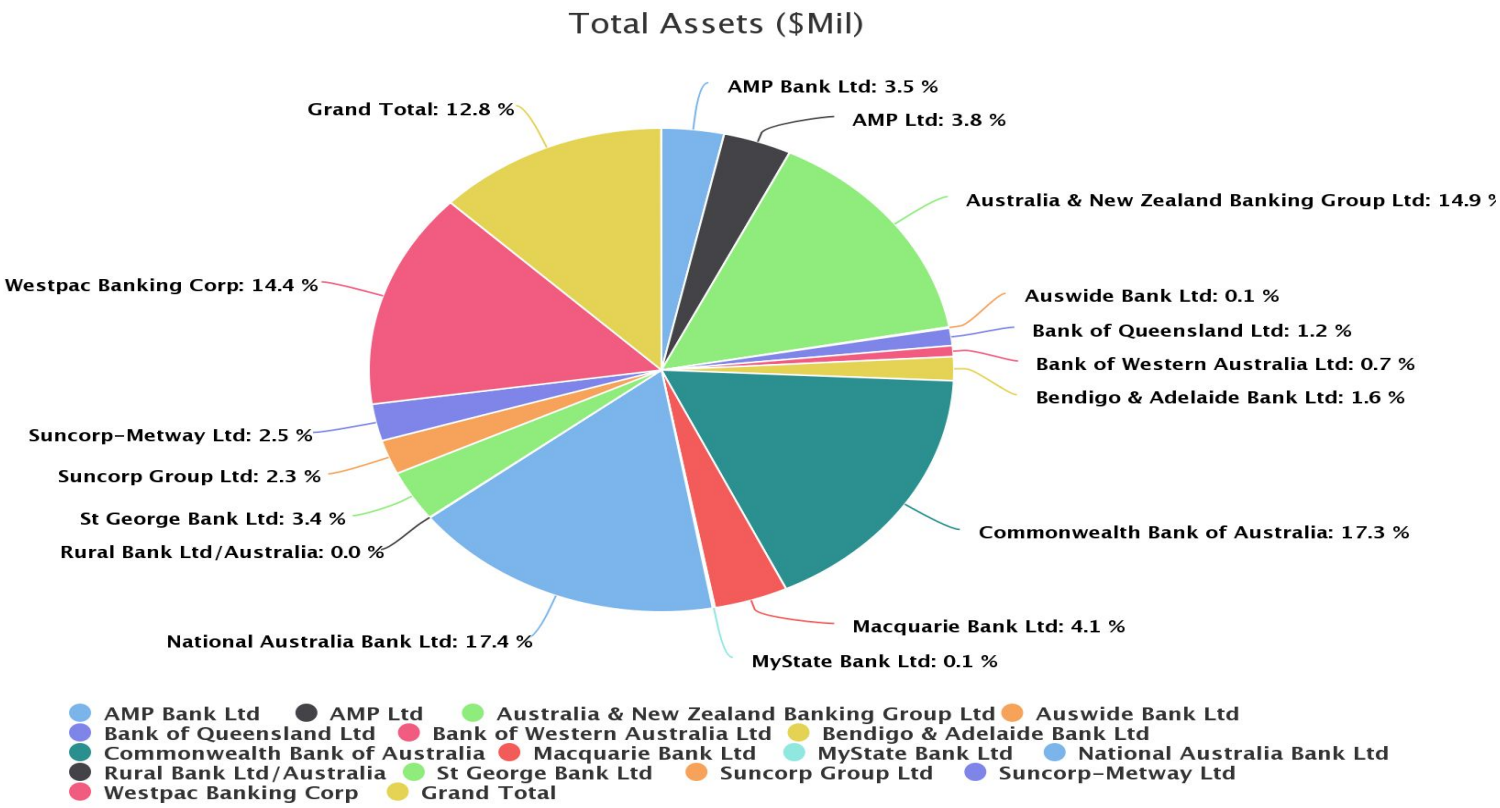
In [60]:
```
# 3. Draw pie chart
# 1) Read CSV and store them in a list
TA_reader = csv.DictReader(open('TA.csv', 'r'))
TA_data = []
for row in TA_reader:
    TA_data.append({'name': row['Bank'], 'y': float(row['TA'])})
```

In [61]:
```
# 2) Set Option
options = OPTIONS_PIE
options['title'] = {'text': 'Total Assets ($Mil)'}
```

In [62]:
```
# 3) Add Data
PIE_CHART = Highchart(width=850, height=550)
PIE_CHART.set_dict_options(options)
PIE_CHART.add_data_set(
                TA_data,
               'pie',
                '% total assets',
                allowPointSelect=True,
                cursor='pointer',
                showInLegend=True,
                dataLabels={
                    'enabled': True,
                    'format': '<b>{point.name}</b>: {point.percentage:.1f} %',
                    'style': {
                        'color': "(Highcharts.theme && Highcharts.theme.contrastTextColor) || 'black'"
                    }
                }
            )
PIE_CHART
```

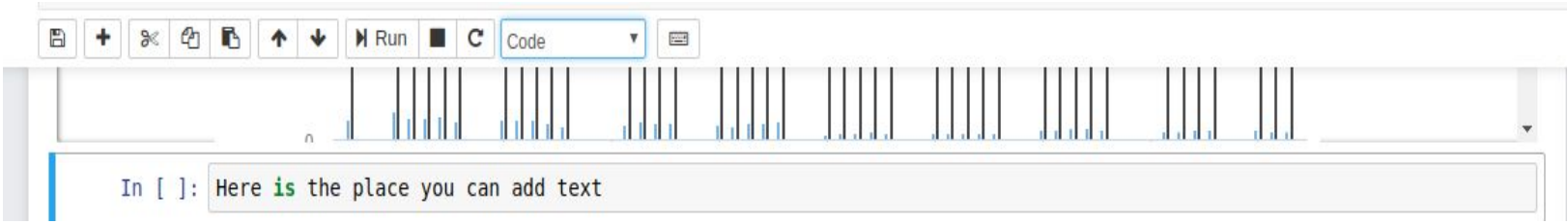
4. Display Your Research Results (Cont..)

- c) Pie Chart



4. Display Your Research Results (Cont..)

- 5) Add text
- Before adding markdown text



- After adding markdown text

Questions?