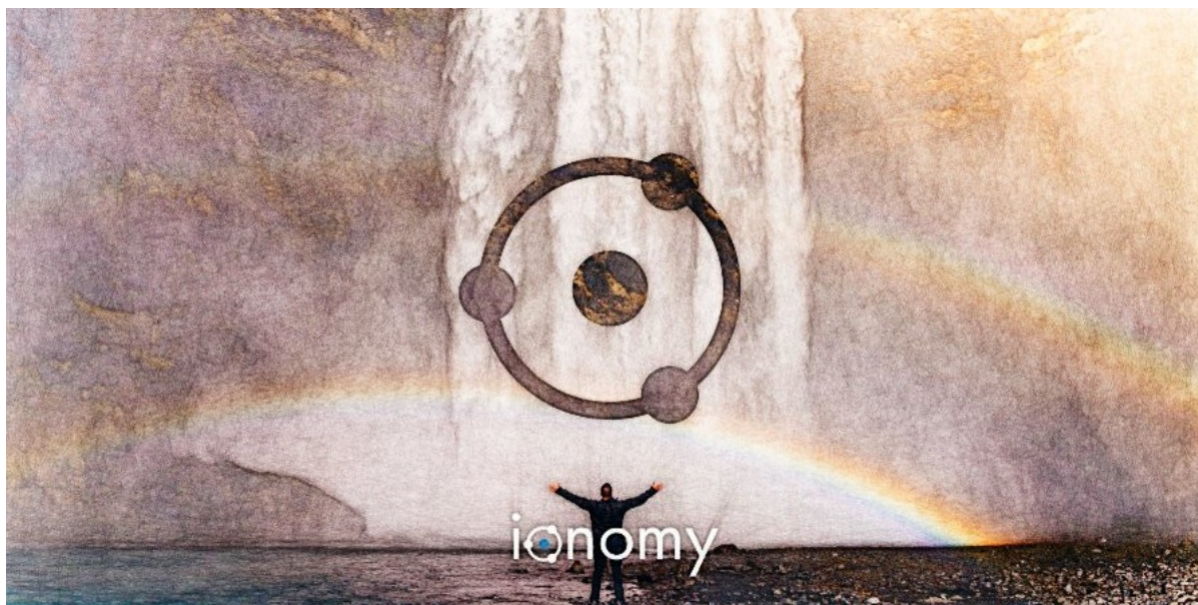

ionomy-python

Apr 24, 2021

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1.1 Quick Start

1.1.1 Playground

Start up a jupyter lab instance with all the tutorial notebooks.

```
docker run -p 8888:8888 ionomy-playground:latest
```

Copy and Paste the url with token from terminal output into your browser.

Open the tutorial notebook of interest or create your own to experiment with!

1.1.2 Install

```
pip install ionomy-python
```

1.1.3 Imports and Common Params

The primary classes for use are IonPanda and BitTA.

There are lower level classes if needed: Ionomy, BitTrex, and BitPanda

Ionomy is a raw API wrapper, no extras.

The same with BitTrex, which is a raw api wrapper.

BitPanda is the bittrex equivalent of IonPanda,

but without the TA methods.

```
from Ionomy import IonPanda, BitTA

# Common Params
MARKET = 'btc-hive'
CURRENCY = 'hive'
BASE = 'btc'
TIME = 'day'
```

1.1.4 Instantiate

You must provide your Ionomy or BitTrex API key/secret, respectively.

IMPORTANT: I opted for the package user to manually load and update ohlcv data

This allows for control of the number of API call, since there are limits (See Exchanges Websites).

```
ionpd = IonPanda(IONOMY_KEY, IONOMY_SECRET)
bta = BitTA(BITTREX_KEY, BITTREX_SECRET)

# IMPORTANT - call this method to load/update ohlcv data
bta.update(CURRENCY, BASE, TIME)
```

1.1.5 Common Methods

The “tutorials” provides examples of each method and their returns.

The “modules” provides the detailed code documentation.

Here, I will show one method from each classed being used by the highest order classes.

```
# returns a regular dictionary from the raw json
# I saw no benefit from having a single row dataframe returned
market_summary = ionpd.market_summary(MARKET)

# returns a pandas dataframe
order_book_df = ionpd.order_book(MARKET)

# returns a regular dictionary from the raw json
market_summary = bta.market_summary(MARKET)

# returns a pandas dataframe
order_book_df = bta.order_book(MARKET)

# TA method are returned as DataFrames or Series
rsi_series = rsi_series = bta.rsi(length=1, drift=1, offset=0)
```

The most common TA methods are implemented and available Due to python limitations, such as floating point arithmetic, etc, results will/may differ from the standard talib package and other TA implementations on the same data. This difference should be within a very small margin of error from a c++ implementation.

2.1 Installation

The minimal working python version is 3.7.x

2.1.1 Install with pip

```
pip install -U ionomy-python
```

or:

```
python -m pip install ionomy-python
```

2.1.2 Manual installation

You can install beam from this repository if you want the latest but possibly non-compiling version:

```
git clone https://github.com/Distortedlogic/ionomy-python.git
cd ionomy-python
```

then

```
python setup.py build
python setup.py install
```

or:

```
pip install -e .
```

2.2 Tutorials

2.2.1 BitTrex

Imports and Params

```
[10]: from Ionomy import BitTrex
      from decouple import config

      MARKET = 'btc-hive'
      CURRENCY = 'hive'
      BASE = 'btc'
      TIME = 'day'
      BTC_QUANTITY = 0.001
      HIVE_QUANTITY = 100
      ADDRESS = 'memehub'
      PAYMENTID_OPTIONAL = None
      LOW_RATE=0.00003
      HIGH_RATE=0.00004
      TIMEINFORCE='GTC'
```

Instantiation

```
[2]: bt = BitTrex(config('TREX_KEY'), config('TREX_SECRET'))
```

Public Endpoint Methods

Markets

```
[3]: markets = bt.markets()
      markets[0]
```

```
[3]: {'MarketCurrency': 'LTC',
      'BaseCurrency': 'BTC',
      'MarketCurrencyLong': 'Litecoin',
      'BaseCurrencyLong': 'Bitcoin',
      'MinTradeSize': 0.03297187,
      'MarketName': 'BTC-LTC',
      'IsActive': True,
      'IsRestricted': False,
      'Created': '2014-02-13T00:00:00',
      'Notice': None,
      'IsSponsored': None,
      'LogoUrl': 'https://bittrexblobstorage.blob.core.windows.net/public/6defbc41-582d-
      ↪47a6-bb2e-d0fa88663524.png'}
```

Currencies

```
[4]: currencies = bt.currencies()
      currencies[0]
```

```
[4]: {'Currency': 'BTC',
      'CurrencyLong': 'Bitcoin',
      'MinConfirmation': 2,
      'TxFee': 0.0005,
      'IsActive': True,
      'IsRestricted': False,
      'CoinType': 'BITCOIN',
      'BaseAddress': '1N52wHoVR79PMDishab2XmRHsbekCdGquK',
      'Notice': None}
```

Ticker

```
[5]: bt.ticker(MARKET)
```

```
[5]: {'Bid': 3.313e-05, 'Ask': 3.331e-05, 'Last': 3.345e-05}
```

Market Summaries

```
[7]: market_summaries = bt.market_summaries()
      market_summaries[0]
```

```
[7]: {'MarketName': 'BTC-STPT',
      'High': 1.1e-06,
      'Low': 1.06e-06,
      'Volume': 89896.60128034,
      'Last': 1.1e-06,
      'BaseVolume': 0.09687313,
      'TimeStamp': '2020-05-16T16:52:42.353',
      'Bid': 1.06e-06,
      'Ask': 1.09e-06,
      'OpenBuyOrders': 42,
      'OpenSellOrders': 273,
      'PrevDay': 1.06e-06,
      'Created': '2019-06-11T18:34:44.627'}
```

Market Summary

```
[9]: bt.market_summary(MARKET)
```

```
[9]: {'MarketName': 'BTC-HIVE',
      'High': 3.49e-05,
      'Low': 3.28e-05,
      'Volume': 251042.78009306,
      'Last': 3.381e-05,
      'BaseVolume': 8.52802204,
      'TimeStamp': '2020-05-15T18:05:54.107',
      'Bid': 3.382e-05,
      'Ask': 3.409e-05,
      'OpenBuyOrders': 388,
      'OpenSellOrders': 853,
      'PrevDay': 3.472e-05,
      'Created': '2020-03-21T20:13:46.243'}
```

Market History

```
[11]: market_history = bt.market_history(MARKET)
market_history[0]

[11]: {'Id': 65281835,
      'TimeStamp': '2020-05-15T18:02:15.01',
      'Quantity': 16.80099545,
      'Price': 3.381e-05,
      'Total': 0.0005680416561645,
      'FillType': 'FILL',
      'OrderType': 'BUY',
      'Uuid': '822aacf4-da88-4be8-bb2b-6a723671acf9'}
```

Order Book

```
[14]: order_book = bt.order_book(MARKET)
order_book["buy"][0]

[14]: {'Quantity': 1013.53349185, 'Rate': 3.382e-05}
```

OHLCV

```
[16]: ohlcv = bt.ohlcv(CURRENCY, BASE, TIME)
ohlcv[0]

[16]: {'time': 1586908800,
      'close': 1.637e-05,
      'high': 1.7e-05,
      'low': 1.577e-05,
      'open': 1.633e-05,
      'volumefrom': 270819.04,
      'volumeto': 4.473,
      'conversionType': 'force_direct',
      'conversionSymbol': ''}
```

Market Endpoint Methods

Limit Buy/Sell Orders

```
[16]: # buy_order_uuid = bt.buy_limit(MARKET, HIVE_QUANTITY, LOW_RATE, TIMEINFORCE)
sell_order_uuid = bt.sell_limit(MARKET, HIVE_QUANTITY, HIGH_RATE, TIMEINFORCE)
```

Cancel Order

```
[20]: order_uuid = bt.cancel(sell_order_uuid)
order_uuid
```

Order Status

```
[19]: order = bt.get_order(sell_order_uuid)
order

[19]: {'AccountId': None,
      'OrderUuid': 'ffd0af51-6195-442b-aab6-320851bc72cc',
      'Exchange': 'BTC-HIVE',
      'Type': 'LIMIT_SELL',
      'Quantity': 100.0,
      'QuantityRemaining': 100.0,
      'Limit': 4e-05,
      'Reserved': None,
      'ReserveRemaining': None,
      'CommissionReserved': None,
      'CommissionReserveRemaining': None,
      'CommissionPaid': 0.0,
      'Price': 0.0,
      'PricePerUnit': None,
      'Opened': '2020-05-17T20:53:56.25',
      'Closed': None,
      'IsOpen': True,
      'Sentinel': None,
      'CancelInitiated': False,
      'ImmediateOrCancel': False,
      'IsConditional': False,
      'Condition': 'NONE',
      'ConditionTarget': None}
```

Open Orders

```
[17]: open_orders = bt.open_orders(MARKET)
open_orders[0]

[17]: {'Uuid': None,
      'OrderUuid': 'ffd0af51-6195-442b-aab6-320851bc72cc',
      'Exchange': 'BTC-HIVE',
      'OrderType': 'LIMIT_SELL',
      'Quantity': 100.0,
      'QuantityRemaining': 100.0,
      'Limit': 4e-05,
      'CommissionPaid': 0.0,
      'Price': 0.0,
      'PricePerUnit': None,
      'Opened': '2020-05-17T20:53:56.25',
      'Closed': None,
      'CancelInitiated': False,
      'ImmediateOrCancel': False,
      'IsConditional': False,
      'Condition': 'NONE',
      'ConditionTarget': None}
```

Account Endpoint Methods

Balances

```
[18]: balances = bt.balances()
balances[0]
```

```
[18]: {'Currency': 'BTC',
      'Balance': 1e-08,
      'Available': 1e-08,
      'Pending': 0.0,
      'CryptoAddress': None}
```

Balance

```
[19]: bt.balance(CURRENCY)
```

```
[19]: {'Currency': 'HIVE',
      'Balance': 13728.03479181,
      'Available': 13728.03479181,
      'Pending': 0.0,
      'CryptoAddress': '30c05de7684c4bf1817'}
```

Order History

```
[22]: order_history = bt.order_history()
order_history[0]
```

```
[22]: {'OrderUuid': '74b7ccde-f6f5-4083-8d0e-e56c080015a1',
      'Exchange': 'BTC-HIVE',
      'TimeStamp': '2020-05-14T21:11:14.44',
      'OrderType': 'LIMIT_BUY',
      'Limit': 3.41e-05,
      'Quantity': 1128.49345297,
      'QuantityRemaining': 0.0,
      'Commission': 7.696e-05,
      'Price': 0.03848162,
      'PricePerUnit': 3.41e-05,
      'IsConditional': False,
      'Condition': '',
      'ConditionTarget': 0.0,
      'ImmediateOrCancel': False,
      'Closed': '2020-05-14T21:11:14.44'}
```

Deposit History

```
[23]: deposit_history = bt.deposit_history(CURRENCY)
deposit_history[0]
```

```
[23]: {'Id': 102691733,
      'Amount': 7194.41,
      'Currency': 'HIVE',
      'Confirmations': 59,
      'LastUpdated': '2020-05-14T21:06:48.5',
```

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```
'TxId': 'da01a679e0c7d7afd6de6e3753ca2beb7dea0caf',
'CryptoAddress': '30c05de7684c4bf1817'}
```

Deposit Address

```
[24]: bt.deposit_address(CURRENCY)
```

```
[24]: {'Currency': 'HIVE', 'Address': '30c05de7684c4bf1817'}
```

Withdrawal History

```
[28]: withdrawal_history = bt.withdrawal_history(CURRENCY)
       withdrawal_history[0]
```

```
[28]: {'PaymentUuid': '78943ce0-b6d4-4fcb-991a-7516e4323e82',
'Currency': 'HIVE',
'Amount': 13728.02479181,
'Address': 'memehub',
'Opened': '2020-05-15T18:13:44.513',
'Authorized': False,
'PendingPayment': False,
'TxCost': 0.01,
'TxId': None,
'Canceled': True,
'InvalidAddress': False}
```

Withdraw

```
[33]: bt.withdraw(CURRENCY, QUANTITY, ADDRESS, PAYMENTID_OPTIONAL)
```

```
[33]: {'uuid': 'cee4a123-5656-4938-a777-66a05df1b500'}
```

```
[ ]:
```

2.2.2 BitPanda

Imports and Params

```
[1]: from Ionomy import BitPanda
       from decouple import config
```

```
MARKET = 'btc-hive'
CURRENCY = 'hive'
BASE = 'btc'
TIME = 'day'
```

Instantiation

```
[3]: bp = BitPanda(config('TREX_KEY'), config('TREX_SECRET'))
```

Public Endpoint Methods

Markets

```
[5]: markets_pd = bp.markets()
markets_pd.head()
```

```
[5]:
```

	MarketCurrency	BaseCurrency	MarketCurrencyLong	BaseCurrencyLong	\
0	LTC	BTC	Litecoin	Bitcoin	
1	DOGE	BTC	Dogecoin	Bitcoin	
2	VTC	BTC	Vertcoin	Bitcoin	
3	PPC	BTC	Peercoin	Bitcoin	
4	FTC	BTC	Feathercoin	Bitcoin	

	MinTradeSize	MarketName	IsActive	IsRestricted	Created	Notice	\
0	0.032972	BTC-LTC	True	False	2014-02-13	None	
1	1000.000000	BTC-DOGE	True	False	2014-02-13	None	
2	8.830802	BTC-VTC	True	False	2014-02-13	None	
3	6.025548	BTC-PPC	True	False	2014-02-13	None	
4	213.675214	BTC-FTC	True	False	2014-02-13	None	

	IsSponsored	LogoUrl
0	None	https://bittrexblobstorage.blob.core.windows.n...
1	None	https://bittrexblobstorage.blob.core.windows.n...
2	None	https://bittrexblobstorage.blob.core.windows.n...
3	None	https://bittrexblobstorage.blob.core.windows.n...
4	None	https://bittrexblobstorage.blob.core.windows.n...

Currencies

```
[6]: currencies_pd = bp.currencies()
currencies_pd.head()
```

```
[6]:
```

	Currency	CurrencyLong	MinConfirmation	TxFee	IsActive	IsRestricted	\
0	BTC	Bitcoin	2	0.0005	True	False	
1	LTC	Litecoin	6	0.0100	True	False	
2	DOGE	Dogecoin	6	2.0000	True	False	
3	VTC	Vertcoin	600	0.0200	True	False	
4	PPC	Peercoin	12	0.0200	True	False	

	CoinType	BaseAddress	Notice
0	BITCOIN	1N52wHoVR79PMDishab2XmRHsbekCdGquK	None
1	BITCOIN16	LhyLNfBkoKshT7R8Pce6vkB9T2cP2o84hx	None
2	BITCOIN	D9GqmkgCpghtnXP7xMD78v9xfqeDkqBZBMT	None
3	BITCOIN16	VfukW89WKT9h3YjHZdSAAuGNVGELY31wyj	None
4	BITCOIN16	None	None

Ticker

```
[7]: bp.ticker(MARKET)
```

```
[7]: {'Bid': 3.381e-05, 'Ask': 3.409e-05, 'Last': 3.38e-05}
```

Market Summaries

```
[9]: market_summaries_pd = bp.market_summaries()
market_summaries_pd.head()
```

```
[9]:
```

	MarketName	High	Low	Volume	Last	\
0	BTC-STPT	1.090000e-06	1.060000e-06	8.975255e+04	1.090000e-06	
1	ETH-WAXP	1.639700e-04	1.523000e-04	8.819855e+04	1.550100e-04	
2	ETH-FX	2.740000e-04	2.519900e-04	4.506822e+04	2.630000e-04	
3	BTC-PLA	5.000000e-08	4.000000e-08	1.844084e+06	4.000000e-08	
4	EUR-USDT	9.330000e-01	9.230000e-01	2.398502e+04	9.270000e-01	

	BaseVolume	TimeStamp	Bid	Ask	\
0	0.096687	2020-05-15 18:58:54.720	1.070000e-06	1.100000e-06	
1	13.731805	2020-05-15 18:58:54.720	1.540300e-04	1.597800e-04	
2	11.563228	2020-05-15 18:58:54.720	2.610000e-04	2.680000e-04	
3	0.074589	2020-05-15 18:58:54.720	3.000000e-08	4.000000e-08	
4	22295.859801	2020-05-15 18:58:54.720	9.230000e-01	9.290000e-01	

	OpenBuyOrders	OpenSellOrders	PrevDay	Created
0	40.0	271.0	1.060000e-06	2019-06-11 18:34:44.627
1	109.0	109.0	1.569600e-04	2018-02-15 01:09:15.167
2	72.0	135.0	2.590000e-04	2019-04-23 17:00:38.527
3	45.0	137.0	5.000000e-08	2019-03-28 17:12:11.213
4	73.0	72.0	9.300000e-01	2020-03-30 06:19:27.620

Market Summary

```
[12]: bp.market_summary(MARKET)
```

```
[12]: {'MarketName': 'BTC-HIVE',
'High': 3.453e-05,
'Low': 3.28e-05,
'Volume': 275969.44525378,
'Last': 3.365e-05,
'BaseVolume': 9.35729375,
'TimeStamp': '2020-05-15T18:59:34.75',
'Bid': 3.37e-05,
'Ask': 3.406e-05,
'OpenBuyOrders': 389,
'OpenSellOrders': 858,
'PrevDay': 3.4e-05,
'Created': '2020-03-21T20:13:46.243'}
```

Order Book

```
[13]: order_book_pd = bt.order_book(MARKET)
order_book_pd.head()
```

```
[13]:
```

	Quantity	Rate	type
0	833.815866	0.000034	bid
1	1700.000000	0.000034	bid
2	59.127338	0.000034	bid
3	1589.495316	0.000034	bid
4	557.397472	0.000034	bid

Market History

```
[15]: market_history_pd = bp.market_history(MARKET)
market_history_pd.head()
```

```
[15]:
```

	Id	TimeStamp	Quantity	Price	Total	\
0	65288419	2020-05-15 18:59:18.930	28737.484222	0.000034	0.967016	
1	65288418	2020-05-15 18:59:18.930	84.785096	0.000034	0.002854	
2	65288417	2020-05-15 18:59:18.930	464.468900	0.000034	0.015639	
3	65288416	2020-05-15 18:59:18.930	1700.000000	0.000034	0.057375	
4	65288415	2020-05-15 18:59:18.930	35.000000	0.000034	0.001181	

	FillType	OrderType	Uuid
0	FILL	SELL	5bb886f8-7f19-478c-97cf-9f1127f3a698
1	FILL	SELL	17bf5e26-8aa6-4eab-aca0-a87a84e601e0
2	FILL	SELL	4fd0d932-0f83-46f3-b34c-f55864c19199
3	FILL	SELL	6d30c744-537a-4b71-a93a-7194c004f932
4	FILL	SELL	a9a528d8-4d5d-4488-9314-4343a0dca7f8

Market Endpoint Methods

Limit Buy/Sell Order

```
[ ]: order_uuid = bp.buy_limit(MARKET, QUANTITY, RATE, TIMEINFORCE)
order_uuid = bp.sell_limit(MARKET, QUANTITY, RATE, TIMEINFORCE)
```

Cancel Order

```
[ ]: order_uuid = bp.cancel(UUID)
```

Order Status

```
[ ]: order = bp.get_order(UUID)
```

Account Endpoint Methods

Balances

```
[16]: balances_pd = bp.balances()
balances_pd.head()
```

```
[16]:
```

	Currency	Balance	Available	Pending	CryptoAddress
0	BTC	1.000000e-08	1.000000e-08	0.0	None
1	BTXCRD	8.489922e+00	8.489922e+00	0.0	None
2	HIVE	1.372603e+04	1.372603e+04	0.0	30c05de7684c4bf1817
3	STEEM	0.000000e+00	0.000000e+00	0.0	acc190939a9248d981f

Balance

```
[17]: bp.balance(CURRENCY)
```

```
[17]: {'Currency': 'HIVE',
'Balance': 13726.03479181,
'Available': 13726.03479181,
'Pending': 0.0,
'CryptoAddress': '30c05de7684c4bf1817'}
```

Order History

```
[18]: order_history_pd = bp.order_history()
order_history_pd.head()
```

```
[18]:
```

	OrderUid	Exchange	TimeStamp	\
0	74b7ccde-f6f5-4083-8d0e-e56c080015a1	BTC-HIVE	2020-05-14 21:11:14.440	
1	b1be09e1-c0b5-40fd-bae1-9ef061521483	BTC-HIVE	2020-05-14 21:09:07.890	
2	aef02e01-641f-4a0a-9223-05a846e47e4c	BTC-STEEM	2020-05-14 21:06:01.770	
3	aba84609-e842-4681-90b1-42fb24501174	BTC-HIVE	2020-05-13 14:27:35.190	
4	bf5b235d-ea76-4bd7-bf37-eebaaaea05dd	BTC-STEEM	2020-05-13 12:55:44.540	

	OrderType	Limit	Quantity	QuantityRemaining	Commission	\
0	LIMIT_BUY	0.000034	1128.493453	0.000000	0.000077	
1	LIMIT_BUY	0.000034	2828.824311	1128.824311	0.000116	
2	MARKET_SELL	NaN	5439.624000	0.000000	0.000193	
3	LIMIT_BUY	0.000035	485.000000	0.000000	0.000034	
4	MARKET_SELL	NaN	912.726000	0.000000	0.000034	

	Price	PricePerUnit	IsConditional	Condition	ConditionTarget	\
0	0.038482	0.000034	False		0.0	
1	0.057953	0.000034	False		0.0	
2	0.096666	0.000018	False		0.0	
3	0.016975	0.000035	False		0.0	
4	0.016958	0.000019	False		0.0	

	ImmediateOrCancel	Closed
0	False	2020-05-14 21:11:14.440
1	False	2020-05-14 21:09:34.990
2	True	2020-05-14 21:06:01.770

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```
3          False 2020-05-13 21:59:27.830
4          True  2020-05-13 12:55:44.540
```

Deposit History

```
[19]: deposit_history_pd = bp.deposit_history(CURRENCY)
      deposit_history_pd.head()
```

```
[19]:      Id      Amount Currency  Confirmations      LastUpdated  \
0  102691733  7194.41      HIVE                59 2020-05-14 21:06:48.500

      TxId      CryptoAddress
0  da01a679e0c7d7afd6de6e3753ca2beb7dea0caf  30c05de7684c4bf1817
```

Deposit History

```
[20]: bp.deposit_address(CURRENCY)
```

```
[20]: {'Currency': 'HIVE', 'Address': '30c05de7684c4bf1817'}
```

Withdrawal History

```
[21]: withdrawal_history_pd = bp.withdrawal_history(CURRENCY)
      withdrawal_history_pd.head()
```

```
[21]:      PaymentUuid Currency      Amount  Address  \
0  cee4a123-5656-4938-a777-66a05df1b500  HIVE    0.990000  memehub
1  35aca318-89b1-4aa5-9b1a-2ca702b8268f  HIVE    0.990000  memehub
2  78943ce0-b6d4-4fcb-991a-7516e4323e82  HIVE  13728.024792  memehub

      Opened  Authorized  PendingPayment  TxCost  \
0 2020-05-15 18:18:00.133          True          False    0.01
1 2020-05-15 18:17:31.920          True          False    0.01
2 2020-05-15 18:13:44.513          False          False    0.01

      TxId  Canceled  InvalidAddress
0  bd6a480971ab8fdf28bf46375551df9149b241dd  False          False
1  182ce6ce3c7e3b615c2558d046838dd904a85b7e  False          False
2                                     None          True          False
```

Withdraw

```
[ ]: uuid = bt.withdraw(CURRENCY, QUANTITY, ADDRESS, PAYMENTID_OPTIONAL)
```

2.2.3 BitTA

Imports and Params

```
[1]: from Ionomy import BitTA
    from decouple import config

    MARKET = 'btc-hive'
    CURRENCY = 'hive'
    BASE = 'btc'
    TIME = 'day'
```

Instantiation

```
[2]: bta = BitTA(config('TREX_KEY'), config('TREX_SECRET'))
```

Update the class ohlc dataframe with the latest data

```
[3]: bta.update(CURRENCY, BASE, TIME)
```

Momentum

Awesome Oscillator (AO)

```
[4]: ao_series = bta.ao(fast=5, slow=34, offset=0)
    ao_series.tail()
```

```
[4]: 26    -0.000005
    27    -0.000005
    28    -0.000006
    29    -0.000006
    30    -0.000007
    Name: AO_5_34, dtype: float64
```

Absolute Price Oscillator (APO)

```
[6]: apo_series = bta.apo(fast=12, slow=26, offset=0)
    apo_series.tail()
```

```
[6]: 26    -0.000003
    27    -0.000003
    28    -0.000003
    29    -0.000003
    30    -0.000003
    Name: APO_12_26, dtype: float64
```

Balance of Power (BOP)

```
[7]: bop_series = bta.bop(offset=0)
    bop_series.tail()
```

```
[7]: 26  -0.802281
      27  -0.332075
      28  -0.229630
      29  -0.594203
      30  -0.583333
      Name: BOP, dtype: float64
```

Commodity Channel Index (CCI)

```
[8]: cci_series = bta.cci(length=20, c=0.015, offset=0)
      cci_series.tail()
```

```
[8]: 26  -76.387008
      27  -77.379598
      28  -75.407177
      29  -85.992973
      30  -97.848760
      Name: CCI_20_0.015, dtype: float64
```

Center of Gravity (CG)

```
[9]: cg_series = bta.cg(length=20, offset=0)
      cg_series.tail()
```

```
[9]: 26  -12.041409
      27  -12.148211
      28  -12.138013
      29  -11.565258
      30  -11.332511
      Name: CG_20, dtype: float64
```

Chande Momentum Oscillator (CMO)

```
[11]: cmo_series = bta.cmo(length=14, drift=1, offset=0)
      cmo_series.tail()
```

```
[11]: 26  -39.251391
      27  -38.172458
      28  -24.536601
      29  -17.513736
      30  -67.250355
      Name: CMO_14, dtype: float64
```

Coppock Curve (COPC)

```
[12]: coppock_series = bta.coppock(length=10, fast=11, slow=14, offset=0)
      coppock_series.tail()
```

```
[12]: 26  -38.771187
      27  -50.955977
```

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```

28  -55.510796
29  -55.224571
30  -54.643435
Name: COPC_11_14_10, dtype: float64

```

Fisher Transform (FISHT)

```
[13]: fisher_series = bta.fisher(length=5, offset=0)
fisher_series.tail()
```

```
[13]: 26  -4.306868
27  -4.709653
28  -6.155028
29  -6.877715
30  -7.239059
Name: FISHERT_5, dtype: float64

```

Know Sure Thing' (KST)

```
[14]: # Need more ohlcv data to get results
default_kwargs = {
    "roc1": 10,
    "roc2": 15,
    "roc3": 20,
    "roc4": 30,
    "sma1": 10,
    "sma1": 10,
    "sma1": 10,
    "sma1": 15,
    "signal": 9,
    "drift": 1,
    "offset": 0
}
kst_series = bta.kst(**default_kwargs)
kst_series.tail()
```

```
[14]:      KST_10_15_20_30_15_10_10_15  KSTS_9
26                NaN                NaN
27                NaN                NaN
28                NaN                NaN
29                NaN                NaN
30                NaN                NaN

```

Moving Average Convergence Divergence (MACD)

```
[15]: macd_series = bta.macd(fast=12, slow=26, signal=9, offset=0)
macd_series.tail()
```

```
[15]:      MACD_12_26_9  MACDH_12_26_9  MACDS_12_26_9
26      -0.000003  -1.645118e-07  -0.000002
27      -0.000003  -2.936861e-07  -0.000002

```

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```
28    -0.000003  -3.568946e-07  -0.000003
29    -0.000003  -4.071771e-07  -0.000003
30    -0.000003  -4.292012e-07  -0.000003
```

Momentum (MOM)

```
[4]: mom_series = bta.mom(length=12, offset=0)
      mom_series.tail()
```

```
[4]: 26    -0.000007
      27    -0.000004
      28    -0.000013
      29    -0.000015
      30    -0.000012
      Name: MOM_12, dtype: float64
```

Percentage Price Oscillator (PPO)

```
[5]: ppo_series = bta.ppo(fast=12, slow=26, signal=9, offset=0)
      ppo_series.tail()
```

```
[5]:      PPO_12_26_9  PPOH_12_26_9  PPOS_12_26_9
26    -6.436170    -11.002642     4.566472
27    -8.523310    -10.471825     1.948516
28   -12.216228    -11.331795    -0.884433
29   -16.248230    -12.291037    -3.957193
30   -19.603026    -12.516667    -7.086359
```

Rate of Change (ROC)

```
[6]: roc_series = bta.roc(length=1, offset=0)
      roc_series.tail()
```

```
[6]: 26    -5.739935
      27    -2.539683
      28    -0.917975
      29    -2.450687
      30     1.041667
      Name: ROC_1, dtype: float64
```

Relative Strength Index (RSI)

```
[7]: rsi_series = bta.rsi(length=1, drift=1, offset=0)
      rsi_series.tail()
```

```
[7]: 26    16.266019
      27    10.589262
      28     8.499405
      29     4.158062
```

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```
30    32.675188
Name: RSI_1, dtype: float64
```

Relative Vigor Index (RVI)

```
[8]: rvi_series = bta.rvi(length=14, swma_length=4, offset=0)
     rvi_series.tail()
```

```
[8]:      RVI_14_4  RVI_14_4
26 -0.300914  -0.282395
27 -0.214800  -0.303975
28 -0.182882  -0.260117
29 -0.148107  -0.205258
30 -0.122529  -0.166287
```

Slope

```
[11]: slope_series = bta.slope(length=1, as_angle=False, to_degrees=False, offset=0)
     slope_series.tail()
```

```
[11]: 26    -2.110000e-06
     27    -8.800000e-07
     28    -3.100000e-07
     29    -8.200000e-07
     30     3.400000e-07
     Name: SLOPE_1, dtype: float64
```

Stochastic (STOCH)

```
[12]: stoch_series = bta.stoch(fast_k=14, slow_k=5, slow_d=3, offset=0)
     stoch_series.tail()
```

```
[12]:      STOCHF_14  STOCHF_3  STOCH_5  STOCH_3
26    6.709265    22.935425    15.241707    14.207310
27   13.065977    17.236516    17.021569    15.752323
28    9.055627     9.610290    18.185576    16.816284
29    3.249097     8.456901    12.802855    16.003333
30   23.015873    11.773533    11.019168    14.002533
```

Trix (TRIX)

```
[13]: trix_series = bta.trix(length=18, drift=1, offset=0)
     trix_series.tail()
```

```
[13]: 26    -0.242958
     27    -0.294361
     28    -0.347659
     29    -0.402758
     30    -0.456325
     Name: TRIX_18, dtype: float64
```

True Strength Index (TSI)

```
[14]: tsi_series = bta.tsi(fast=13, slow=25, drift=1, offset=0)
      tsi_series.tail()
```

```
[14]: 26      8.408619
      27      7.977190
      28      7.571647
      29      7.105209
      30      6.754438
      Name: TSI_13_25, dtype: float64
```

Ultimate Oscillator (UO)

```
[15]: uo_series = bta.uo(
      fast=7,
      medium=14,
      slow=28,
      fast_w=4.0,
      medium_w=2.0,
      slow_w=1.0,
      drift=1,
      offset=0,
      )
      uo_series.tail()
```

```
[15]: 26      NaN
      27      32.055171
      28      32.909266
      29      34.102532
      30      36.042116
      Name: UO_7_14_28, dtype: float64
```

William's Percent R (WILLR)

```
[17]: willr_series = bta.willr(length=20, offset=0)
      willr_series.tail()
```

```
[17]: 26      -90.027701
      27      -98.949449
      28      -99.271895
      29      -99.804645
      30      -97.614151
      Name: WILLR_20, dtype: float64
```

Overlap

Double Exponential Moving Average (DEMA)

```
[18]: dema_series = bta.dema(length=10, offset=0)
      dema_series.tail()
```

```
[18]: 26    0.000036
      27    0.000035
      28    0.000034
      29    0.000033
      30    0.000033
      Name: DEMA_10, dtype: float64
```

Exponential Moving Average (EMA)

```
[20]: ema_series = bta.ema(length=10, offset=0)
      ema_series.tail()
```

```
[20]: 26    0.000038
      27    0.000038
      28    0.000037
      29    0.000036
      30    0.000035
      Name: EMA_10, dtype: float64
```

Fibonacci's Weighted Moving Average (FWMA)

```
[21]: fwma_series = bta.fwma(length=10, asc=True, offset=0)
      fwma_series.tail()
```

```
[21]: 26    0.000036
      27    0.000035
      28    0.000035
      29    0.000034
      30    0.000033
      Name: FWMA_10, dtype: float64
```

Hull Moving Average (HMA)

```
[22]: hma_series = bta.hma(length=10, offset=0)
      hma_series.tail()
```

```
[22]: 26    0.000035
      27    0.000034
      28    0.000034
      29    0.000033
      30    0.000032
      Name: HMA_10, dtype: float64
```

Ichimoku Kinkō Hyō (ichimoku)

```
[24]: ichimoku_pd_1, ichimoku_pd_2 = bta.ichimoku(tenkan=9, kijun=26, senkou=52, offset=0)
```

```
[25]: ichimoku_pd_1.tail()
```

```
[25]:
```

	ISA_9	ISB_26	ITS_9	IKS_26	ICS_26
26	NaN	NaN	0.000042	0.000072	NaN
27	NaN	NaN	0.000039	0.000072	NaN
28	NaN	NaN	0.000038	0.000072	NaN
29	NaN	NaN	0.000036	0.000072	NaN
30	NaN	NaN	0.000036	0.000072	NaN

```
[26]: ichimoku_pd_2.tail()
```

```
[26]:
```

	ISA_9	ISB_26
52	0.000057	NaN
53	0.000055	NaN
54	0.000055	NaN
55	0.000054	NaN
56	0.000054	NaN

Kaufman's Adaptive Moving Average (KAMA)

```
[4]: kama_series = bta.kama(length=10, fast=2, slow=30, drift=1, offset=0)
kama_series.tail()
```

```
[4]: 26    0.000030
27    0.000031
28    0.000032
29    0.000032
30    0.000032
Name: KAMA_10_2_30, dtype: float64
```

Linear Regression Moving Average (linreg)

```
[5]: linreg_series = bta.linreg(length=10, offset=0)
linreg_series.tail()
```

```
[5]: 26    0.000035
27    0.000034
28    0.000034
29    0.000034
30    0.000033
Name: LR_10, dtype: float64
```

Pascal's Weighted Moving Average (PWMA)

```
[6]: pwma_series = bta.pwma(length=10, asc=True, offset=0)
pwma_series.tail()
```

```
[6]: 26    0.000037
27    0.000037
28    0.000036
29    0.000036
30    0.000035
Name: PWMA_10, dtype: float64
```

wildeR's Moving Average (RMA)

```
[7]: rma_series = bta.rma(length=10, offset=0)
rma_series.tail()
```

```
[7]: 26    0.000041
      27    0.000040
      28    0.000040
      29    0.000039
      30    0.000038
      Name: RMA_10, dtype: float64
```

Sine Weighted Moving Average (SINWMA)

```
[10]: sinwma_series = bta.sinwma(length=10, offset=0)
sinwma_series.tail()
```

```
[10]: 26    0.000038
      27    0.000037
      28    0.000036
      29    0.000036
      30    0.000035
      Name: SINWMA_10, dtype: float64
```

Simple Moving Average (SMA)

```
[9]: sma_series = bta.sma(length=10, offset=0)
sma_series.tail()
```

```
[9]: 26    0.000038
      27    0.000038
      28    0.000037
      29    0.000036
      30    0.000035
      Name: EMA_10, dtype: float64
```

Symmetric Weighted Moving Average (SWMA)

```
[11]: swma_series = bta.swma(length=10, offset=0)
swma_series.tail()
```

```
[11]: 26    0.000037
      27    0.000037
      28    0.000036
      29    0.000036
      30    0.000035
      Name: SWMA_10, dtype: float64
```

Tim Tillson's T3 Moving Average (T3)

```
[12]: t3_series = bta.t3(length=10, a=0.7, offset=0)
      t3_series.tail()
```

```
[12]: 26    0.000041
      27    0.000040
      28    0.000039
      29    0.000038
      30    0.000037
      Name: T3_10_0.7, dtype: float64
```

Triple Exponential Moving Average (TEMA)

```
[13]: tema_series = bta.tema(length=10, offset=0)
      tema_series.tail()
```

```
[13]: 26    0.000034
      27    0.000033
      28    0.000033
      29    0.000032
      30    0.000031
      Name: TEMA_10, dtype: float64
```

Triangular Moving Average (TRIMA)

```
[14]: trima_series = bta.trima(length=10, offset=0)
      trima_series.tail()
```

```
[14]: 26    0.000039
      27    0.000038
      28    0.000037
      29    0.000036
      30    0.000035
      Name: TRIMA_10, dtype: float64
```

Volume Weighted Average Price (VWAP)

```
[15]: vwap_series = bta.vwap(offset=0)
      vwap_series.tail()
```

```
[15]: 26    0.000068
      27    0.000068
      28    0.000068
      29    0.000068
      30    0.000068
      Name: VWAP, dtype: float64
```

Volume Weighted Moving Average (VWMA)

```
[16]: vwma_series = bta.vwma(length=10, offset=0)
      vwma_series.tail()
```

```
[16]: 26    0.000040
      27    0.000038
      28    0.000037
      29    0.000036
      30    0.000035
      Name: VWMA_10, dtype: float64
```

Weighted Moving Average (WMA)

```
[17]: wma_series = bta.wma(length=10, asc=True, offset=0)
      wma_series.tail()
```

```
[17]: 26    0.000037
      27    0.000036
      28    0.000035
      29    0.000035
      30    0.000034
      Name: WMA_10, dtype: float64
```

Zero Lag Moving Average (ZLMA)

```
[19]: zlma_series = bta.zlma(length=10, offset=0, mamode='ema')
      zlma_series.tail()
```

```
[19]: 26    0.000035
      27    0.000035
      28    0.000034
      29    0.000033
      30    0.000032
      Name: ZLEMA_10, dtype: float64
```

Log Return

```
[20]: log_return_series = bta.log_return(length=10, cumulative=False, percent=False,
      ↪ offset=0)
      log_return_series.tail()
```

```
[20]: 26    -0.289628
      27    -0.344520
      28    -0.294087
      29    -0.258776
      30    -0.175491
      Name: LOGRET_10, dtype: float64
```

```
[21]: percent_return_series = bta.percent_return(length=10, cumulative=False, percent=False,
      ↪ offset=0)
      percent_return_series.tail()
```

```
[21]: 26    -0.251458
      27    -0.291439
      28    -0.254788
      29    -0.228004
      30    -0.160955
      Name: PCTRET_10, dtype: float64
```

Trend Return

```
[29]: # trend_return_series = bta.trend_return(trend="close", log=True, cumulative=False,
      ↪offset=0, trend_reset=0)
      # trend_return_series.tail()
```

Rolling Kurtosis

```
[30]: kurtosis_series = bta.kurtosis(length=10, offset=0, mamode='ema')
      kurtosis_series.tail()
```

```
[30]: 26    0.065587
      27    1.024344
      28    1.768071
      29   -1.197006
      30   -1.618680
      Name: KURT_10, dtype: float64
```

Rolling Mean Absolute Deviation

```
[31]: mad_series = bta.mad(length=10, offset=0)
      mad_series.tail()
```

```
[31]: 26    0.000004
      27    0.000003
      28    0.000002
      29    0.000002
      30    0.000002
      Name: MAD_10, dtype: float64
```

Rolling Median

```
[32]: median_series = bta.median(length=10, offset=0)
      median_series.tail()
```

```
[32]: 26    0.000037
      27    0.000037
      28    0.000037
      29    0.000036
      30    0.000035
      Name: MEDIAN_10, dtype: float64
```


Rolling Quantile

```
[33]: quantile_series = bta.quantile(length=30, q=0.5, offset=0)
      quantile_series.tail()
```

```
[33]: 26      NaN
      27      NaN
      28      NaN
      29    0.000037
      30    0.000037
      Name: QTL_30_0.5, dtype: float64
```

Rolling Skew

```
[34]: skew_series = bta.skew(length=30, offset=0)
      skew_series.tail()
```

```
[34]: 26      NaN
      27      NaN
      28      NaN
      29    1.928790
      30    2.001403
      Name: SKEW_30, dtype: float64
```

Rolling Standard Deviation

```
[35]: stdev_series = bta.stdev(length=30, offset=0)
      stdev_series.tail()
```

```
[35]: 26      NaN
      27      NaN
      28      NaN
      29    0.000021
      30    0.000020
      Name: STDEV_30, dtype: float64
```

Rolling Variance

```
[36]: variance_series = bta.variance(length=30, offset=0)
      variance_series.tail()
```

```
[36]: 26      NaN
      27      NaN
      28      NaN
      29    4.263450e-10
      30    4.121269e-10
      Name: VAR_30, dtype: float64
```

Rolling Z Score

```
[37]: zscore_series = bta.zscore(length=30, std=1, offset=0)
zscore_series.tail()
```

```
[37]: 26      NaN
27      NaN
28      NaN
29    -0.386298
30    -0.430835
Name: Z_30, dtype: float64
```

Trend

Average Directional Movement (ADX)

```
[38]: adx_series = bta.adx(length=14, drift=1, offset=0)
adx_series.tail()
```

```
[38]:      ADX_14      DMP_14      DMN_14
26      NaN    43.174417    24.456502
27  33.907743    43.244219    26.401392
28  32.872561    44.458695    27.142852
29  31.783427    45.526168    28.934007
30  30.660034    46.291902    30.529761
```

Aroon (AROON)

```
[40]: aroon_series = bta.aroon(length=1, offset=0)
aroon_series.tail()
```

```
[40]:      AROOND_1      AROONU_1
26      100.0      100.0
27      100.0      100.0
28      100.0      100.0
29      100.0      100.0
30      100.0      100.0
```

Decreasing

```
[44]: decreasing_series = bta.decreasing(length=1, asint=True, offset=0)
decreasing_series.tail()
```

```
[44]: 26    1
27    1
28    1
29    1
30    1
Name: DEC_1, dtype: int64
```

Detrend Price Oscillator (DPO)

```
[45]: dpo_series = bta.dpo(length=1, centered=True, offset=0)
      dpo_series.tail()
```

```
[45]: 26    8.800000e-07
      27    3.100000e-07
      28    8.200000e-07
      29    3.200000e-07
      30             NaN
      Name: DPO_1, dtype: float64
```

Increasing

```
[46]: increasing_series = bta.increasing(length=1, asint=True, offset=0)
      increasing_series.tail()
```

```
[46]: 26    0
      27    0
      28    0
      29    0
      30    0
      Name: INC_1, dtype: int64
```

Linear Decay

```
[47]: linear_decay_series = bta.linear_decay(length=1, offset=0)
      linear_decay_series.tail()
```

```
[47]: 26    0.000035
      27    0.000034
      28    0.000033
      29    0.000033
      30    0.000032
      Name: LDECAY_1, dtype: float64
```

Q Stick

```
[48]: qstick_series = bta.qstick(length=1, offset=0)
      qstick_series.tail()
```

```
[48]: 26   -2.110000e-06
      27   -8.800000e-07
      28   -3.100000e-07
      29   -8.200000e-07
      30   -3.200000e-07
      Name: QS_1, dtype: float64
```

Vortex

```
[49]: vortex_series = bta.vortex(length=1, offset=0)
vortex_series.tail()
```

```
[49]:      VTXP_1      VTXM_1
26  0.311787  1.433460
27  0.445283  1.547170
28  1.311111  1.651852
29  0.478261  1.500000
30  0.573333  1.346667
```

Acceleration Bands (ACCBANDS)

```
[52]: vortex_series = bta.accbands(
      length=10,
      c=4,
      drift=1,
      mamode='ema',
      offset=0
    )
vortex_series.tail()
```

```
[52]:      ACCBL_10  ACCBM_10  ACCBU_10
26  0.000027  0.000039  0.000055
27  0.000027  0.000038  0.000052
28  0.000028  0.000037  0.000049
29  0.000028  0.000036  0.000047
30  0.000028  0.000036  0.000045
```

Average True Range (ATR)

```
[53]: atr_series = bta.atr(length=1, mamode='ema', offset=0)
atr_series.tail()
```

```
[53]: 26    0.000003
27    0.000003
28    0.000001
29    0.000001
30    0.000002
Name: ATR_1, dtype: float64
```

Bollinger Bands (BBANDS)

```
[54]: bbands_series = bta.bbands(length=20, std=2, mamode='ema', offset=0)
bbands_series.tail()
```

```
[54]:      BBL_20      BBM_20      BBU_20
26  6.995546e-07  0.000040  0.000079
27 -2.899943e-07  0.000039  0.000079
28 -2.216349e-07  0.000039  0.000077
29  1.852932e-05  0.000038  0.000058
30  2.409578e-05  0.000038  0.000051
```

Donchian Channels (DC)

```
[55]: donchian_series = bta.donchian(lower_length=10, upper_length=20, offset=0)
      donchian_series.tail()
```

```
[55]:   DCL_10_20  DCM_10_20  DCU_10_20
      26    0.000035  0.000077  0.000118
      27    0.000034  0.000076  0.000118
      28    0.000033  0.000076  0.000118
      29    0.000033  0.000053  0.000073
      30    0.000032  0.000044  0.000056
```

Keltner Channels (KC)

```
[56]: kc_series = bta.kc(length=20, scalar=2, mamode='ema', offset=0)
      kc_series.tail()
```

```
[56]:   KCL_20  KCB_20  KCU_20
      26  0.000027  0.000041  0.000055
      27  0.000028  0.000040  0.000053
      28  0.000028  0.000040  0.000051
      29  0.000029  0.000039  0.000049
      30  0.000029  0.000038  0.000048
```

Mass Index (MASSI)

```
[60]: massi_series = bta.massi(fast=9, slow=25, offset=0)
      massi_series.tail()
```

```
[60]: 26  NaN
      27  NaN
      28  NaN
      29  NaN
      30  NaN
      Name: MASSI_9_25, dtype: float64
```

Normalized Average True Range (NATR)

```
[61]: natr_series = bta.natr(length=20, mamode='ema', drift=1, offset=0)
      natr_series.tail()
```

```
[61]: 26    24.382287
      27    23.276758
      28    21.531668
      29    20.284296
      30    18.905182
      Name: NATR_20, dtype: float64
```

True Range

```
[62]: true_range_series = bta.true_range(drift=1, offset=0)
      true_range_series.tail()
```

```
[62]: 26    0.000003
      27    0.000003
      28    0.000001
      29    0.000001
      30    0.000002
      Name: TRUERANGE_1, dtype: float64
```

Volume

Accumulation/Distribution (AD)

```
[63]: ad_series = bta.ad(offset=0)
      ad_series.tail()
```

```
[63]: 26   -10.025277
      27   -13.443919
      28   -17.399771
      29   -21.928423
      30   -23.634089
      Name: AD, dtype: float64
```

Accumulation/Distribution Oscillator or Chaikin Oscillator

```
[64]: adosc_series = bta.adosc(fast=12, slow=26, offset=0)
      adosc_series.tail()
```

```
[64]: 26   -53.875972
      27   -53.644655
      28   -53.167654
      29   -52.549297
      30   -51.602041
      Name: ADOSC_12_26, dtype: float64
```

Chaikin Money Flow (CMF)

```
[65]: cmf_series = bta.cmf(length=20, offset=0)
      cmf_series.tail()
```

```
[65]: 26   -0.024778
      27   -0.164204
      28   -0.215310
      29   -0.625480
      30   -0.483050
      Name: CMF_20, dtype: float64
```

Elder's Force Index (EFI)

```
[66]: efi_series = bta.efi(length=13, drift=1, mamode='ema', offset=0)
      efi_series.tail()
```

```
[66]: 26    0.000105
      27    0.000088
      28    0.000075
      29    0.000064
      30    0.000054
      Name: EFI_13, dtype: float64
```

Ease of Movement (EOM)

```
[67]: eom_series = bta.eom(length=14, drift=1, offset=0)
      eom_series.tail()
```

```
[67]: 26   -0.000038
      27   -0.000036
      28   -0.000031
      29   -0.000025
      30   -0.000029
      Name: EOM_14_100000000, dtype: float64
```

Money Flow Index (MFI)

```
[68]: mfi_series = bta.mfi(length=14, drift=1, offset=0)
      mfi_series.tail()
```

```
[68]: 26    45.056603
      27    48.168097
      28    54.635662
      29    58.310922
      30    46.966315
      Name: MFI_14, dtype: float64
```

Negative Volume Index (NVI)

```
[69]: nvi_series = bta.nvi(length=13, initial=1_000, offset=0)
      nvi_series.tail()
```

```
[69]: 26    1447.481869
      27    1429.308862
      28    1417.968110
      29    1388.480099
      30    1356.293779
      Name: NVI_13, dtype: float64
```

On Balance Volume (OBV)

```
[70]: obv_series = bta.obv(offset=0)
      obv_series.tail()
```

```
[70]: 26    400.875
      27    386.495
      28    379.735
      29    373.608
      30    368.491
      Name: OBV, dtype: float64
```

Positive Volume Index (PVI)

```
[71]: pvi_series = bta.pvi(length=13, initial=1_000, offset=0)
      pvi_series.tail()
```

```
[71]: 26    1247.036814
      27    1247.036814
      28    1247.036814
      29    1247.036814
      30    1247.036814
      Name: PVI_13, dtype: float64
```

Price-Volume (PVOL)

```
[72]: pvol_series = bta.pvol(offset=0)
      pvol_series.tail()
```

```
[72]: 26    0.000650
      27    0.000486
      28    0.000226
      29    0.000200
      30    0.000165
      Name: PVOL, dtype: float64
```

Price-Volume Trend (PVT)

```
[73]: pvt_series = bta.pvt(drift=1, offset=0)
      pvt_series.tail()
```

```
[73]: 26    60957.106450
      27    60920.585815
      28    60914.380307
      29    60899.364945
      30    60894.348279
      Name: PVT, dtype: float64
```


Volume Profile (VP)

```
[74]: vp_series = bta.vp(width=10)
      vp_series.tail()
```

```
[74]:
```

	low_close	mean_close	high_close	pos_volume	neg_volume	total_volume
5	0.000045	0.000046	0.000048	174.200	44.120	218.320
6	0.000037	0.000039	0.000042	8.082	27.090	35.172
7	0.000035	0.000036	0.000037	63.750	0.000	63.750
8	0.000034	0.000035	0.000037	18.770	22.503	41.273
9	0.000032	0.000033	0.000033	0.000	18.004	18.004

```
[ ]:
```

2.2.4 Ionomy

Imports and Params

```
[1]: from Ionomy import Ionomy
      from decouple import config
```

```
MARKET = 'btc-hive'
CURRENCY = 'hive'
BASE = 'btc'
TIME = 'day'
BTC_QUANTITY = 0.001
HIVE_QUANTITY = 100
LOW_RATE=0.00003
HIGH_RATE=0.00004
ADDRESS = 'memehub'
```

Instantiation

```
[2]: ion = Ionomy(config('IONOMY_KEY'), config('IONOMY_SECRET'))
```

Public Endpoint Methods

Markets

```
[3]: markets = ion.markets()
      markets[0]
```

```
[3]: {'market': 'btc-eth',
      'title': 'Bitcoin:Ethereum',
      'currencyBase': 'btc',
      'currencyMarket': 'eth',
      'orderMinSize': '0.00001000',
      'buyFee': '0.20000000',
      'sellFee': '0.20000000',
      'inMaintenance': False}
```

Currencies

```
[6]: currencies = ion.currencies()
currencies[0]

[6]: {'currency': 'dash',
      'title': 'Dash',
      'withdrawMinSize': '0.00100000',
      'withdrawFee': '0.00200000',
      'inMaintenance': False,
      'canDeposit': 1,
      'canWithdraw': 1}
```

Order Book

```
[11]: order_book = ion.order_book(MARKET)
order_book["bids"][0]

[11]: {'size': '207.00000000', 'price': '0.00003301'}
```

Market Summaries

```
[12]: market_summaries = ion.market_summaries()
market_summaries[0]

[12]: {'market': 'btc-eth',
      'high': '0.02150000',
      'low': '0.01900001',
      'volume': '0.31467562',
      'price': '0.01900001',
      'change': '-11.63',
      'baseVolume': '0.00597884',
      'bidsOpenOrders': '10',
      'bidsLastPrice': '0.01900003',
      'highestBid': '0.01900003',
      'asksOpenOrders': '14',
      'asksLastPrice': '0.02250000',
      'lowestAsk': '0.02250000'}
```

Market Summary

```
[14]: market_summary = ion.market_summary(MARKET)
market_summary

[14]: {'market': 'btc-hive',
      'high': '0.00003538',
      'low': '0.00003300',
      'volume': '3856.61483680',
      'price': '0.00003370',
      'change': '2.06',
      'baseVolume': '0.12996792',
      'bidsOpenOrders': '39',
```

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```
'bidsLastPrice': '0.00003301',
'highestBid': '0.00003301',
'asksOpenOrders': '47',
'asksLastPrice': '0.00003447',
'lowestAsk': '0.00003370'}
```

Market History

```
[15]: market_history = ion.market_history(MARKET)
market_history[0]
```

```
[15]: {'type': 'MARKET_BUY',
'amount': '140.93003798',
'price': '0.00003370',
'total': '0.00474934',
'createdAt': '2020-05-16T09:28:58Z'}
```

Market Endpoint Methods

Limit Buy/Sell Order

```
[3]: order = ion.limit_buy(HIVE_QUANTITY, LOW_RATE, MARKET)
# order = ion.limit_sell(HIVE_QUANTITY, HIGH_RATE, MARKET)
```

```
-----
Exception                                 Traceback (most recent call last)
<ipython-input-3-8453aalba66a> in <module>
      1 order = ion.limit_buy(HIVE_QUANTITY, LOW_RATE, MARKET)
----> 2 order = ion.limit_sell(HIVE_QUANTITY, HIGH_RATE, MARKET)

/notebooks/packages/Ionomy/ionomy.py in limit_sell(self, amount, price, market)
     92         'price': f'{price:.8f}'
     93     }
----> 94     return self._request('market/sell-limit', params)
     95
     96     def cancel_order(self, orderId: str) -> bool:

/notebooks/packages/Ionomy/ionomy.py in _request(self, endpoint, params)
     47         data = json.loads(response.content)
     48         if not data['success']:
----> 49             raise Exception(data['message'])
     50         return data['data']
     51

Exception: 40017
```

Cancel Order

```
[9]: success = ion.cancel_order(order['orderId'])
success
```

```
[9]: True
```

Order Status

```
[8]: order_status = ion.order_status(order['orderId'])  
order_status
```

```
[8]: {'orderId': '5ec1a9d18b196d3e336cf6d2',  
      'status': 'OPEN',  
      'market': 'btc-hive',  
      'type': 'LIMIT_BUY',  
      'amount': '100.00000000',  
      'price': '0.00003000',  
      'filled': '0.00000000',  
      'opened': '2020-05-17T21:17:05Z',  
      'closed': None}
```

Account Endpoint Methods

Open Orders

```
[4]: open_orders = ion.open_orders(MARKET)  
open_orders[0]
```

```
[4]: [{'orderId': '5ec1a9d18b196d3e336cf6d2',  
      'market': 'btc-hive',  
      'type': 'LIMIT_BUY',  
      'amount': '100.00000000',  
      'price': '0.00003000',  
      'filled': '0.00000000',  
      'createdAt': '2020-05-17T21:17:05Z'}]
```

Balances

```
[18]: balances = ion.balances()  
balances[0]
```

```
[18]: {'currency': 'btc', 'available': '0.11574671', 'reserved': '0.00000001'}
```

Balance

```
[19]: balance = ion.balance(CURRENCY)  
balance
```

```
[19]: {'currency': 'hive', 'available': '0.54814539', 'reserved': '0.00000000'}
```

Deposit Address

```
[20]: deposit_address = ion.deposit_address(CURRENCY)
      deposit_address
[20]: {'currency': 'hive', 'address': '5e8f6cd1a2a3e2080524eb42'}
```

Deposit History

```
[29]: deposit_history = ion.deposit_history("Steem")
      deposit_history["deposits"][0]
[29]: {'transactionId': '5e8f6db5149be70cd35ab8e2',
      'state': 'PROCESSED',
      'currency': 'steem',
      'amount': '1.00000000',
      'createdAt': '2020-04-09T18:47:17Z'}
```

Withdrawal History

```
[25]: withdrawal_history = ion.withdrawal_history(CURRENCY)
      withdrawal_history[0]
[25]: {'transactionId': '5e8f726da2a3e220e6566b72',
      'state': 'PROCESSED',
      'currency': 'hive',
      'amount': '7186.01000000',
      'feeAmount': '0.01000000',
      'createdAt': '2020-04-09T19:07:25Z'}
```

2.2.5 IonPanda

Imports and Params

```
[1]: from Ionomy import IonPanda
      from decouple import config

      MARKET = 'btc-hive'
      CURRENCY = 'hive'
      BASE = 'btc'
      TIME = 'day'
      AMOUNT = 1
      PRICE = 0.0000001
```

Instantiation

```
[2]: ionpd = IonPanda(config('IONOMY_KEY'), config('IONOMY_SECRET'))
```

Public Endpoint Methods

Markets

```
[3]: markets_pd = ionpd.markets()
markets_pd.head()
```

```
[3]:
```

	market	title	currencyBase	currencyMarket	orderMinSize	\
0	btc-eth	Bitcoin:Ethereum	btc	eth	0.00001	
1	btc-ion	Bitcoin:Ion	btc	ion	0.00001	
2	btc-dash	Bitcoin:Dash	btc	dash	0.00001	
3	btc-pivx	Bitcoin:Pivx	btc	pivx	0.00001	
4	btc-atoms	Bitcoin:Atoms	btc	atoms	0.00001	

	buyFee	sellFee	inMaintenance
0	0.2	0.2	False
1	0.2	0.2	False
2	0.2	0.2	False
3	0.2	0.2	False
4	0.0	0.4	False

Currencies

```
[4]: currencies_pd = ionpd.currencies()
currencies_pd.head()
```

```
[4]:
```

	currency	title	withdrawMinSize	withdrawFee	inMaintenance	canDeposit	\
0	dash	Dash	0.001	0.0020	False	True	
1	pivx	Pivx	0.001	0.1000	False	True	
2	ion	Ion	0.250	0.5000	False	True	
3	btc	Bitcoin	0.001	0.0005	False	True	
4	eth	Ethereum	0.001	0.0060	False	True	

	canWithdraw
0	True
1	True
2	True
3	True
4	True

Order Book

```
[6]: order_book_pd = ionpd.order_book(MARKET)
order_book_pd.head()
```

```
[6]:
```

	size	price	type
0	207.0	0.000033	bid
1	3.2	0.000033	bid
2	160.0	0.000032	bid
3	143.0	0.000032	bid
4	100.0	0.000032	bid

Market Summaries

```
[7]: market_summaries_pd = ionpd.market_summaries()
market_summaries_pd.head()
```

```
[7]:
```

	market	high	low	volume	price	change	baseVolume	\
0	btc-ion	0.000003	0.000003	3488.067194	0.000003	-17.05	0.008825	
1	btc-eth	0.021500	0.019000	0.314676	0.019000	-11.63	0.005979	
2	btc-ltc	0.004750	0.000560	11.346268	0.000560	-13.85	0.006354	
3	btc-dash	0.007992	0.007949	0.130268	0.007949	-0.54	0.001035	
4	btc-pivx	0.000043	0.000028	345.795984	0.000028	-21.96	0.009644	

	bidsOpenOrders	bidsLastPrice	highestBid	asksOpenOrders	asksLastPrice	\
0	9	0.000003	0.000003	163	0.000003	
1	10	0.019000	0.019000	14	0.022500	
2	10	0.004500	0.004500	43	0.005300	
3	10	0.008000	0.007464	42	0.007949	
4	7	0.000029	0.000028	19	0.000036	

	lowestAsk
0	0.000003
1	0.022500
2	0.005300
3	0.007949
4	0.000043

Market Summary

```
[8]: market_summary = ionpd.market_summary(MARKET)
market_summary
```

```
[8]: {'market': 'btc-hive',
'high': '0.00003538',
'low': '0.00003300',
'volume': '3856.61483680',
'price': '0.00003370',
'change': '2.06',
'baseVolume': '0.12996792',
'bidsOpenOrders': '39',
'bidsLastPrice': '0.00003301',
'highestBid': '0.00003301',
'asksOpenOrders': '47',
'asksLastPrice': '0.00003447',
'lowestAsk': '0.00003370'}
```

Market History

```
[9]: market_history = ionpd.market_history(MARKET)
market_history.head()
```

```
[9]:
```

	type	amount	price	total	createdAt
0	MARKET_BUY	140.930038	0.000034	0.004749	2020-05-16 09:28:58
1	MARKET_BUY	71.402354	0.000034	0.002463	2020-05-16 06:14:42
2	MARKET_BUY	88.000000	0.000034	0.002979	2020-05-16 06:14:42

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```
3  LIMIT_BUY  50.000000  0.000034  0.001692  2020-05-16 06:01:16
4  MARKET_BUY 74.982622  0.000034  0.002579  2020-05-16 03:27:41
```

Market Endpoint Methods

Limit Buy/Sell Orders

```
[ ]: order = ionpd.limit_buy(AMOUNT, PRICE, MARKET)
     order = ionpd.limit_sell(AMOUNT, PRICE, MARKET)
```

Success

```
[ ]: success = ionpd.cancel_order(ORDERID)
```

Order Status

```
[ ]: order_status = ionpd.order_status(ORDERID)
```

Account Endpoint Methods

Open Orders

```
[10]: open_orders_pd = ionpd.open_orders(MARKET)
      open_orders_pd
```

```
[10]: Empty DataFrame
      Columns: [orderId, market, type, amount, price, filled, createdAt]
      Index: []
```

Balances

```
[11]: balances_pd = ionpd.balances()
      balances_pd.head()
```

```
[11]:
```

	currency	available	reserved
0	btc	0.115747	1.000000e-08
1	ion	0.000000	0.000000e+00
2	steem	0.560000	0.000000e+00
3	hive	0.548145	0.000000e+00
4	gravity-live	0.000000	0.000000e+00

Balance


```
[12]: balance = ionpd.balance(CURRENCY)
balance
```

```
[12]: {'currency': 'hive', 'available': '0.54814539', 'reserved': '0.00000000'}
```

Deposit Address

```
[13]: deposit_address = ionpd.deposit_address(CURRENCY)
deposit_address
```

```
[13]: {'currency': 'hive', 'address': '5e8f6cd1a2a3e2080524eb42'}
```

Deposit History

```
[14]: deposit_history_pd = ionpd.deposit_history(CURRENCY)
deposit_history_pd.head()
```

```
[14]: Empty DataFrame
Columns: [currency, deposits]
Index: []
```

Withdrawal History

```
[15]: withdrawal_history_pd = ionpd.withdrawal_history(CURRENCY)
withdrawal_history_pd.head()
```

```
[15]:
```

	transactionId	state	currency	amount	feeAmount	\
0	5e8f726da2a3e220e6566b72	PROCESSED	hive	7186.010000	0.01	
1	5e8f9934a2a3e24bda6b9962	PROCESSED	hive	579.563903	0.01	
2	5e9a6fe4a2a3e2016530b782	PROCESSED	hive	6194.010000	0.01	
3	5ea24752a2a3e2245f19e352	PROCESSED	hive	4308.010000	0.01	
4	5eacd1792249e5764b7323d2	PROCESSED	hive	124.050000	0.05	

```

      createdAt
0 2020-04-09 19:07:25
1 2020-04-09 21:52:52
2 2020-04-18 03:11:32
3 2020-04-24 01:56:34
4 2020-05-02 01:48:41
```

```
[ ]:
```

2.3 Modules

2.3.1 BitTrex

```
class Ionomy.BitTrex(api_key: str, secret_key: str)
    Bases: object
    BitTrex API Wrapper
```

Parameters

- **{str}** -- **BitTrex API Key** (*api_key*) –
- **{str}** -- **BitTrex API Secret** (*secret_key*) –

balance (*currency: str*) → List[Dict[str, Union[str, float, None]]]
balances () → List[Dict[str, Union[str, float, None]]]
buy_limit (*market: str, quantity: float, rate: float, timeInForce: str*) → str
cancel (*uuid: str*) → bool
currencies () → List[Dict[str, Union[bool, str, float, None]]]
deposit_address (*currency: str*)
deposit_history (*currency: str*) → List[Dict[str, Union[str, int, float, None]]]
get_order (*uuid: str*)
market_history (*market: str*) → List[Dict[str, Any]]
market_summaries () → List[Dict[str, Union[bool, int, float, str, None]]]
market_summary (*market: str*) → Dict[str, Union[str, int, float]]
markets () → List[Dict[str, Union[bool, str, float]]]
ohlcv (*currency: str, base: str, time: str, limit: int*) → List[Dict[str, Union[str, int, float]]]
open_orders (*market: str*) → List[Dict[str, Union[bool, str, float, None]]]
order_book (*market: str*) → Dict[str, List[Dict[str, Any]]]
order_history () → List[Dict[str, Any]]
sell_limit (*market: str, quantity: float, rate: float, timeInForce: Optional[str]*) → str
ticker (*market: str*) → Dict[str, float]
withdraw (*currency: str, quantity: float, address: str, paymentid: Optional[str]*)
withdrawal_history (*currency: str*) → List[Dict[str, Union[bool, str, float, None]]]

2.3.2 BitPanda

class Ionomy.**BitPanda** (*api_key: str, secret_key: str*)

Bases: Ionomy.bittrex.BitTrex

Pandas DataFrame Wrapper for BitTrex Base Class

Parameters

- **{str}** -- **BitTrex API Key** (*api_key*) –
- **{str}** -- **BitTrex API Secret** (*secret_key*) –

balances () → pandas.core.frame.DataFrame
currencies () → pandas.core.frame.DataFrame
deposit_history (*currency: str*) → pandas.core.frame.DataFrame
market_history (*market: str*) → pandas.core.frame.DataFrame
market_summaries () → pandas.core.frame.DataFrame

markets () → pandas.core.frame.DataFrame

ohlcv (currency: str, base: str, time: str, limit: int) → pandas.core.frame.DataFrame

order_book (market: str) → pandas.core.frame.DataFrame

order_history () → pandas.core.frame.DataFrame

withdrawal_history (currency: str) → pandas.core.frame.DataFrame

2.3.3 BitTA

class Ionomy.BitTA (api_key: str, secret_key: str)

Bases: Ionomy.bit_panda.BitPanda

Technical Analysis Wrapper for BitPanda

Parameters

- {str} -- **BitTrex API Key** (api_key) –
- {str} -- **BitTrex API Secret** (secret_key) –

accbands (length=None, c=None, drift=None, mamode=None, offset=None, **kwargs)
Acceleration Bands (ACCBANDS)

Acceleration Bands created by Price Headley plots upper and lower envelope bands around a simple moving average.

Sources: <https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/acceleration-bands-abands/>

Calculation:

Default Inputs: length=10, c=4

EMA = Exponential Moving Average

SMA = Simple Moving Average

$HL_RATIO = c * (high - low) / (high + low)$

$LOW = low * (1 - HL_RATIO)$

$HIGH = high * (1 + HL_RATIO)$

if 'ema': LOWER = EMA(LOW, length)

MID = EMA(close, length)

UPPER = EMA(HIGH, length)

else: LOWER = SMA(LOW, length)

MID = SMA(close, length)

UPPER = SMA(HIGH, length)

Parameters

- **high** (pd.Series) – Series of 'high's
- **low** (pd.Series) – Series of 'low's
- **close** (pd.Series) – Series of 'close's
- **length** (int) – It's period. Default: 10

- **c** (*int*) – Multiplier. Default: 4
- **mode** (*str*) – Two options: None or 'ema'. Default: 'ema'
- **drift** (*int*) – The difference period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns lower, mid, upper columns.

Return type `pd.DataFrame`

ad (*offset=None, **kwargs*)

Accumulation/Distribution (AD)

Accumulation/Distribution indicator utilizes the relative position of the close to its High-Low range with volume. Then it is cumulated.

Sources: <https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/accumulationdistribution-ad/>

Calculation: CUM = Cumulative Sum if 'open':

$AD = close - open$

else: $AD = 2 * close - high - low$

$hl_range = high - low$
 $AD = AD * volume / hl_range$
 $AD = CUM(AD)$

Parameters

- **high** (*pd.Series*) – Series of 'high's
- **low** (*pd.Series*) – Series of 'low's
- **close** (*pd.Series*) – Series of 'close's
- **volume** (*pd.Series*) – Series of 'volume's
- **open** (*pd.Series*) – Series of 'open's
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)` fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

adosc (*fast=None, slow=None, offset=None, **kwargs*)

Accumulation/Distribution Oscillator or Chaikin Oscillator

Accumulation/Distribution Oscillator indicator utilizes Accumulation/Distribution and treats it similarly to MACD or APO.

Sources: <https://www.investopedia.com/articles/active-trading/031914/understanding-chaikin-oscillator.asp>

Calculation:

Default Inputs: fast=12, slow=26

AD = Accum/Dist

ad = AD(high, low, close, open)

fast_ad = EMA(ad, fast)

slow_ad = EMA(ad, slow)

ADOSC = fast_ad - slow_ad

Parameters

- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **open** (*pd.Series*) – Series of ‘open’s
- **volume** (*pd.Series*) – Series of ‘volume’s
- **fast** (*int*) – The short period. Default: 12
- **slow** (*int*) – The long period. Default: 26
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): *pd.DataFrame.fillna(value)*

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type *pd.Series*

adx (*length=None, drift=None, offset=None, **kwargs*)

Average Directional Movement (ADX)

Average Directional Movement is meant to quantify trend strength by measuring the amount of movement in a single direction.

Sources: <https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/average-directional-movement-adx/>

Calculation:

DMI ADX TREND 2.0 by @TraderR0BERT, NETWORTHIE.COM //Created by @TraderR0BERT, NETWORTHIE.COM, last updated 01/26/2016 //DMI Indicator //Resolution input option for higher/lower time frames study(title="DMI ADX TREND 2.0", shorttitle="ADX TREND 2.0")

adxlen = input(14, title="ADX Smoothing")

dilen = input(14, title="DI Length")

thold = input(20, title="Threshold")

threshold = thold

//Script for Indicator dirmov(len) =>

```
up = change(high)
down = -change(low)
truerange = rma(tr, len)
plus = fixnan(100 * rma(up > down and up > 0 ? up : 0, len) / truerange)
minus = fixnan(100 * rma(down > up and down > 0 ? down : 0, len) / truerange)
[plus, minus]
adx(dilen, adxlen) =>
[plus, minus] = dirmov(dilen)
sum = plus + minus
adx = 100 * rma(abs(plus - minus) / (sum == 0 ? 1 : sum), adxlen)
[adx, plus, minus]
[sig, up, down] = adx(dilen, adxlen)
osob=input(40,title="Exhaustion Level for ADX, default = 40")
col = sig >= sig[1] ? green : sig <= sig[1] ? red : gray
//Plot Definitions Current Timeframe p1 = plot(sig, color=col, linewidth = 3, title="ADX")
p2 = plot(sig, color=col, style=circles, linewidth=3, title="ADX")
p3 = plot(up, color=blue, linewidth = 3, title="+DI")
p4 = plot(up, color=blue, style=circles, linewidth=3, title="+DI")
p5 = plot(down, color=fuchsia, linewidth = 3, title="-DI")
p6 = plot(down, color=fuchsia, style=circles, linewidth=3, title="-DI")
h1 = plot(threshold, color=black, linewidth =3, title="Threshold")
trender = (sig >= up or sig >= down) ? 1 : 0
bgcolor(trender>0?black:gray, transp=85)
//Alert Function for ADX crossing Threshold
Up_Cross = crossover(up, threshold)
alertcondition(Up_Cross, title="DMI+ cross", message="DMI+ Crossing Threshold")
Down_Cross = crossover(down, threshold)
alertcondition(Down_Cross, title="DMI- cross", message="DMI- Crossing Threshold")
```

Parameters

- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 14
- **drift** (*int*) – The difference period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns adx, dmp, dmn columns.

Return type pd.DataFrame

amat (fast=None, slow=None, mamode=None, lookback=None, offset=None, **kwargs)

Indicator: Archer Moving Averages Trends (AMAT)

ao (fast: int = None, slow: int = None, offset: int = None, **kwargs) → pandas.core.series.Series

Awesome Oscillator (AO)

The Awesome Oscillator is an indicator used to measure a security's momentum. AO is generally used to affirm trends or to anticipate possible reversals.

Sources: [https://www.tradingview.com/wiki/Awesome_Oscillator_\(AO\)](https://www.tradingview.com/wiki/Awesome_Oscillator_(AO)) <https://www.ifcm.co.uk/ntx-indicators/awesome-oscillator>

Calculation:

Default Inputs: fast=5, slow=34

SMA = Simple Moving Average

median = (high + low) / 2

AO = SMA(median, fast) - SMA(median, slow)

Parameters

- **high** (pd.Series) – Series of 'high's
- **low** (pd.Series) – Series of 'low's
- **fast** (int) – The short period. Default: 5
- **slow** (int) – The long period. Default: 34
- **offset** (int) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

aobv (fast=None, slow=None, mamode=None, max_lookback=None, min_lookback=None, offset=None, **kwargs)

Indicator: Archer On Balance Volume (AOBV)

apo (fast=None, slow=None, offset=None, **kwargs) → pandas.core.series.Series

Absolute Price Oscillator (APO)

The Absolute Price Oscillator is an indicator used to measure a security's momentum. It is simply the difference of two Exponential Moving Averages (EMA) of two different periods. Note: APO and MACD lines are equivalent.

Sources: <https://www.investopedia.com/terms/p/ppo.asp>

Calculation:

Default Inputs: fast=12, slow=26

EMA = Exponential Moving Average

APO = EMA(close, fast) - EMA(close, slow)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **fast** (*int*) – The short period. Default: 12
- **slow** (*int*) – The long period. Default: 26
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): *pd.DataFrame.fillna(value)*

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type *pd.Series*

aroon (*length=None, offset=None, **kwargs*)
Aroon (AROON)

Aroon attempts to identify if a security is trending and how strong.

Sources: <https://www.tradingview.com/wiki/Aroon> <https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/aroon-ar/>

Calculation:

Default Inputs: length=1

```
def maxidx(x): return 100 * (int(np.argmax(x)) + 1) / length
```

```
def minidx(x): return 100 * (int(np.argmin(x)) + 1) / length
```

```
_close = close.rolling(length, min_periods=min_periods)
```

```
aroon_up = _close.apply(maxidx, raw=True)
```

```
aroon_down = _close.apply(minidx, raw=True)
```

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): *pd.DataFrame.fillna(value)*

fill_method (value, optional): Type of fill method

Returns aroon_up, aroon_down columns.

Return type *pd.DataFrame*

atr (*length=None, mamode=None, offset=None, **kwargs*) → pandas.core.series.Series
Average True Range (ATR)

Average True Range is used to measure volatility, especially volatility caused by gaps or limit moves.

Sources: [https://www.tradingview.com/wiki/Average_True_Range_\(ATR\)](https://www.tradingview.com/wiki/Average_True_Range_(ATR))

Calculation:

Default Inputs: length=14, drift=1

SMA = Simple Moving Average

EMA = Exponential Moving Average

TR = True Range

tr = TR(high, low, close, drift)

if 'ema': ATR = EMA(tr, length)

else: ATR = SMA(tr, length)

Parameters

- **high** (*pd.Series*) – Series of 'high's
- **low** (*pd.Series*) – Series of 'low's
- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 14
- **mamode** (*str*) – Two options: None or 'ema'. Default: 'ema'
- **drift** (*int*) – The difference period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

bbands (*length=None, std=None, mamode=None, offset=None, **kwargs*)
Bollinger Bands (BBANDS)

A popular volatility indicator.

Sources: [https://www.tradingview.com/wiki/Bollinger_Bands_\(BB\)](https://www.tradingview.com/wiki/Bollinger_Bands_(BB))

Calculation:

Default Inputs: length=20, std=2

EMA = Exponential Moving Average

SMA = Simple Moving Average

STDEV = Standard Deviation

stdev = STDEV(close, length)

if 'ema': MID = EMA(close, length)

else: MID = SMA(close, length)

LOWER = MID - std * stdev

UPPER = MID + std * stdev

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – The short period. Default: 20
- **std** (*int*) – The long period. Default: 2
- **mamode** (*str*) – Two options: None or ‘ema’. Default: ‘ema’
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns lower, mid, upper columns.

Return type `pd.DataFrame`

bop (*offset=None, **kwargs*) → `pandas.core.series.Series`
Balance of Power (BOP)

Balance of Power measure the market strength of buyers against sellers.

Sources: http://www.worden.com/TeleChartHelp/Content/Indicators/Balance_of_Power.htm

Calculation: $BOP = (close - open) / (high - low)$

Parameters

- **open** (*pd.Series*) – Series of ‘open’s
- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)` fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

cci (*length=None, c=None, offset=None, **kwargs*) → `pandas.core.series.Series`
Commodity Channel Index (CCI)

Commodity Channel Index is a momentum oscillator used to primarily identify overbought and oversold levels relative to a mean.

Sources: [https://www.tradingview.com/wiki/Commodity_Channel_Index_\(CCI\)](https://www.tradingview.com/wiki/Commodity_Channel_Index_(CCI))

Calculation:

Default Inputs: length=20, c=0.015

SMA = Simple Moving Average
MAD = Mean Absolute Deviation
 $tp = \text{typical_price} = \text{hlc3} = (\text{high} + \text{low} + \text{close}) / 3$
 $\text{mean_tp} = \text{SMA}(tp, \text{length})$
 $\text{mad_tp} = \text{MAD}(tp, \text{length})$
 $\text{CCI} = (tp - \text{mean_tp}) / (c * \text{mad_tp})$

Parameters

- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 20
- **c** (*float*) – Scaling Constant. Default: 0.015
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

cg (*length=None, offset=None, **kwargs*) → `pandas.core.series.Series`
Center of Gravity (CG)

The Center of Gravity Indicator by John Ehlers attempts to identify turning points while exhibiting zero lag and smoothing.

Sources: <http://www.mesasoftware.com/papers/TheCGOscillator.pdf>

Calculation:

Default Inputs: length=10

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – The length of the period. Default: 10
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

cmf (*length=None, offset=None, **kwargs*)
Chaikin Money Flow (CMF)

Chaikin Money Flow measures the amount of money flow volume over a specific period in conjunction with Accumulation/Distribution.

Sources: [https://www.tradingview.com/wiki/Chaikin_Money_Flow_\(CMF\)](https://www.tradingview.com/wiki/Chaikin_Money_Flow_(CMF)) https://stockcharts.com/school/doku.php?id=chart_school:technical_indicators:chaikin_money_flow_cmf

Calculation:

Default Inputs: length=20

if 'open': ad = close - open

else: ad = 2 * close - high - low

hl_range = high - low

ad = ad * volume / hl_range

CMF = SUM(ad, length) / SUM(volume, length)

Parameters

- **high** (*pd.Series*) – Series of 'high's
- **low** (*pd.Series*) – Series of 'low's
- **close** (*pd.Series*) – Series of 'close's
- **open** (*pd.Series*) – Series of 'open's
- **volume** (*pd.Series*) – Series of 'volume's
- **length** (*int*) – The short period. Default: 20
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): *pd.DataFrame.fillna(value)*

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type *pd.Series*

cmo (*length=None, drift=None, offset=None, **kwargs*) → *pandas.core.series.Series*
Chande Momentum Oscillator (CMO)

Attempts to capture the momentum of an asset with overbought at 50 and oversold at -50.

Sources: <https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/chande-momentum-oscillator-cmo/>

Calculation:

Default Inputs: drift=1

if close.diff(drift) > 0: PSUM = SUM(close - prev_close)

else: NSUM = ABS(SUM(close - prev_close))

CMO = 100 * (PSUM - NSUM) / (PSUM + NSUM)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – The length of the period. Default: 14
- **drift** (*int*) – The short period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

coppock (*length=None, fast=None, slow=None, offset=None, **kwargs*) → `pandas.core.series.Series`
Coppock Curve (COPC)

Coppock Curve (originally called the “Trendex Model”) is a momentum indicator is designed for use on a monthly time scale. Although designed for monthly use, a daily calculation over the same period can be made, converting the periods to 294-day and 231-day rate of changes, and a 210-day weighted moving average.

Sources: https://en.wikipedia.org/wiki/Coppock_curve

Calculation:

Default Inputs: length=10, fast=11, slow=14

SMA = Simple Moving Average

MAD = Mean Absolute Deviation

tp = typical_price = hlc3 = (high + low + close) / 3

mean_tp = SMA(tp, length)

mad_tp = MAD(tp, length)

CCI = (tp - mean_tp) / (c * mad_tp)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – WMA period. Default: 10
- **fast** (*int*) – Fast ROC period. Default: 11
- **slow** (*int*) – Slow ROC period. Default: 14
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

decreasing (*length=None, asint=True, offset=None, **kwargs*)

Decreasing

Returns True or False if the series is decreasing over a periods. By default, it returns True and False as 1 and 0 respectively with kwarg 'asint'.

Sources:

Calculation: `decreasing = close.diff(length) < 0`

if asint: `decreasing = decreasing.astype(int)`

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 1
- **asint** (*bool*) – Returns as binary. Default: True
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: `fillna (value, optional): pd.DataFrame.fillna(value)`

`fill_method (value, optional): Type of fill method`

Returns New feature generated.

Return type `pd.Series`

dema (*length=None, offset=None, **kwargs*)

Double Exponential Moving Average (DEMA)

The Double Exponential Moving Average attempts to a smoother average with less lag than the normal Exponential Moving Average (EMA).

Sources: <https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/double-exponential-moving-average-dema/>

Calculation:

Default Inputs: `length=10`

`EMA = Exponential Moving Average`

`ema1 = EMA(close, length)`

`ema2 = EMA(ema1, length)`

`DEMA = 2 * ema1 - ema2`

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 10
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: `fillna (value, optional): pd.DataFrame.fillna(value)`

`fill_method (value, optional): Type of fill method`

Returns New feature generated.

Return type pd.Series

donchian (*lower_length=None, upper_length=None, offset=None, **kwargs*)
Donchian Channels (DC)

Donchian Channels are used to measure volatility, similar to Bollinger Bands and Keltner Channels.

Sources: [https://www.tradingview.com/wiki/Donchian_Channels_\(DC\)](https://www.tradingview.com/wiki/Donchian_Channels_(DC))

Calculation:

Default Inputs: length=20

LOWER = close.rolling(length).min()

UPPER = close.rolling(length).max()

MID = 0.5 * (LOWER + UPPER)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **lower_length** (*int*) – The short period. Default: 10
- **upper_length** (*int*) – The long period. Default: 20
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns lower, mid, upper columns.

Return type pd.DataFrame

dpo (*length=None, centered=True, offset=None, **kwargs*)
Detrend Price Oscillator (DPO)

Is an indicator designed to remove trend from price and make it easier to identify cycles.

Sources: http://stockcharts.com/school/doku.php?id=chart_school:technical_indicators:detrended_price_osci

Calculation:

Default Inputs: length=1, centered=True

SMA = Simple Moving Average

drift = int(0.5 * length) + 1

DPO = close.shift(drift) - SMA(close, length)

if centered: DPO = DPO.shift(-drift)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 1
- **centered** (*bool*) – Shift the dpo back by int(0.5 * length) + 1. Default: True
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

efi (*length=None, drift=None, mamode=None, offset=None, **kwargs*)

Elder's Force Index (EFI)

Elder's Force Index measures the power behind a price movement using price and volume as well as potential reversals and price corrections.

Sources: [https://www.tradingview.com/wiki/Elder%27s_Force_Index_\(EFI\)](https://www.tradingview.com/wiki/Elder%27s_Force_Index_(EFI)) https://www.motivewave.com/studies/elders_force_index.htm

Calculation:

Default Inputs: length=20, drift=1, mamode=None

EMA = Exponential Moving Average

SMA = Simple Moving Average

pv_diff = close.diff(drift) * volume if mamode == 'sma':

EFI = SMA(pv_diff, length)

else: EFI = EMA(pv_diff, length)

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **volume** (*pd.Series*) – Series of 'volume's
- **length** (*int*) – The short period. Default: 13
- **drift** (*int*) – The diff period. Default: 1
- **mamode** (*str*) – Two options: None or 'sma'. Default: None
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

ema (*length=None, offset=None, **kwargs*) → pandas.core.series.Series

Exponential Moving Average (EMA)

The Exponential Moving Average is more responsive moving average compared to the Simple Moving Average (SMA). The weights are determined by alpha which is proportional to it's length. There are several different methods of calculating EMA. One method uses just the standard definition of EMA and another uses the SMA to generate the initial value for the rest of the calculation.

Sources: https://stockcharts.com/school/doku.php?id=chart_school:technical_indicators:moving_averages <https://www.investopedia.com/ask/answers/122314/what-exponential-moving-average-ema-formula-and-how-ema-calculated.asp>

Calculation:**Default Inputs:** length=10

SMA = Simple Moving Average if kwargs['presma']:

initial = SMA(close, length) rest = close[length:] close = initial + rest

EMA = close.ewm(span=length, adjust=adjust).mean()

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 10
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: adjust (bool, optional): Default: True sma (bool, optional): If True, uses SMA for initial value.
 fillna (value, optional): pd.DataFrame.fillna(value) fill_method (value, optional): Type of fill method

Returns New feature generated.**Return type** pd.Series

eom (*length=None, divisor=None, drift=None, offset=None, **kwargs*)
 Ease of Movement (EOM)

Ease of Movement is a volume based oscillator that is designed to measure the relationship between price and volume fluctuating across a zero line.

Sources: [https://www.tradingview.com/wiki/Ease_of_Movement_\(EOM\)](https://www.tradingview.com/wiki/Ease_of_Movement_(EOM)) https://www.motivewave.com/studies/ease_of_movement.htm https://stockcharts.com/school/doku.php?id=chart_school:technical_indicators:ease_of_movement_emv

Calculation:**Default Inputs:** length=14, divisor=100000000, drift=1

SMA = Simple Moving Average

hl_range = high - low

distance = 0.5 * (high - high.shift(drift) + low - low.shift(drift))

box_ratio = (volume / divisor) / hl_range

eom = distance / box_ratio

EOM = SMA(eom, length)

Parameters

- **high** (*pd.Series*) – Series of 'high's
- **low** (*pd.Series*) – Series of 'low's
- **close** (*pd.Series*) – Series of 'close's
- **volume** (*pd.Series*) – Series of 'volume's
- **length** (*int*) – The short period. Default: 14
- **drift** (*int*) – The diff period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

fisher (*length=None, offset=None, **kwargs*) → pandas.core.series.Series
Fisher Transform (FISHT)

Attempts to identify trend reversals.

Sources: <https://tulipindicators.org/fisher>

Calculation:

Default Inputs: length=5

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – WMA period. Default: 5
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

fwma (*length=None, asc=None, offset=None, **kwargs*)
Fibonacci’s Weighted Moving Average (FWMA)

Fibonacci’s Weighted Moving Average is similar to a Weighted Moving Average (WMA) where the weights are based on the Fibonacci Sequence.

Source: Kevin Johnson

Calculation:

Default Inputs: length=10,

def weights(w):

def _compute(x): return np.dot(w * x)

 return _compute

fibs = utils.fibonacci(length - 1)

FWMA = close.rolling(length)._apply(weights(fibs), raw=True)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 10
- **asc** (*bool*) – Recent values weigh more. Default: True
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

h12 (*offset=None, **kwargs*)

Indicator: HL2

h1c3 (*offset=None, **kwargs*)

Indicator: HLC3

hma (*length=None, offset=None, **kwargs*)

Hull Moving Average (HMA)

The Hull Exponential Moving Average attempts to reduce or remove lag in moving averages.

Sources: <https://alanhull.com/hull-moving-average>

Calculation:

Default Inputs: length=10

WMA = Weighted Moving Average

half_length = int(0.5 * length)

sqrt_length = int(math.sqrt(length))

wmaf = WMA(close, half_length)

wmas = WMA(close, length)

HMA = WMA(2 * wmaf - wmas, sqrt_length)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 10
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

ichimoku (*tenkan=None, kijun=None, senkou=None, offset=None, **kwargs*)

Ichimoku Kinkō Hyō (ichimoku)

Developed Pre WWII as a forecasting model for financial markets.

Sources: <https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/ichimoku-ich/>

Calculation:

Default Inputs: tenkan=9, kijun=26, senkou=52

```
MIDPRICE = Midprice
TENKAN_SEN = MIDPRICE(high, low, close, length=tenkan)
KIJUN_SEN = MIDPRICE(high, low, close, length=kijun)
CHIKOU_SPAN = close.shift(-kijun)
SPAN_A = 0.5 * (TENKAN_SEN + KIJUN_SEN)
SPAN_A = SPAN_A.shift(kijun)
SPAN_B = MIDPRICE(high, low, close, length=senkou)
SPAN_B = SPAN_B.shift(kijun)
```

Parameters

- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **tenkan** (*int*) – Tenkan period. Default: 9
- **kijun** (*int*) – Kijun period. Default: 26
- **senkou** (*int*) – Senkou period. Default: 52
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: `fillna (value, optional): pd.DataFrame.fillna(value)`

`fill_method (value, optional): Type of fill method`

Returns

Two DataFrames.

For the visible period: `spanA`, `spanB`, `tenkan_sen`, `kijun_sen`, and `chikou_span` columns

For the forward looking period: `spanA` and `spanB` columns

Return type `pd.DataFrame`

increasing (*length=None, asint=True, offset=None, **kwargs*)

Increasing

Returns True or False if the series is increasing over a periods. By default, it returns True and False as 1 and 0 respectively with kwarg ‘asint’.

Sources:

Calculation: `increasing = close.diff(length) > 0` if `asint`:

`increasing = increasing.astype(int)`

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 1
- **asint** (*bool*) – Returns as binary. Default: True

- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)` fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

kama (*length=None, fast=None, slow=None, drift=None, offset=None, **kwargs*)
Kaufman's Adaptive Moving Average (KAMA)

Developed by Perry Kaufman, Kaufman's Adaptive Moving Average (KAMA) is a moving average designed to account for market noise or volatility. KAMA will closely follow prices when the price swings are relatively small and the noise is low. KAMA will adjust when the price swings widen and follow prices from a greater distance. This trend-following indicator can be used to identify the overall trend, time turning points and filter price movements.

Sources: https://stockcharts.com/school/doku.php?id=chart_school:technical_indicators:kaufman_s_adaptive_moving_average

Calculation:

Default Inputs: length=10

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 10
- **fast** (*int*) – The short period. Default: 2
- **slow** (*int*) – The long period. Default: 30
- **drift** (*int*) – The difference period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

kc (*length=None, scalar=None, mamode=None, offset=None, **kwargs*)
Keltner Channels (KC)

A popular volatility indicator similar to Bollinger Bands and Donchian Channels.

Sources: [https://www.tradingview.com/wiki/Keltner_Channels_\(KC\)](https://www.tradingview.com/wiki/Keltner_Channels_(KC))

Calculation:

Default Inputs: length=20, scalar=2

ATR = Average True Range

EMA = Exponential Moving Average

SMA = Simple Moving Average

```
if 'ema': BASIS = EMA(close, length)
    BAND = ATR(high, low, close)
else: hl_range = high - low
    tp = typical_price = hlc3(high, low, close)
    BASIS = SMA(tp, length)
    BAND = SMA(hl_range, length)
LOWER = BASIS - scalar * BAND
UPPER = BASIS + scalar * BAND
```

Parameters

- **high** (*pd.Series*) – Series of 'high's
- **low** (*pd.Series*) – Series of 'low's
- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – The short period. Default: 20
- **scalar** (*float*) – A positive float to scale the bands. Default: 2
- **mode** (*str*) – Two options: None or 'ema'. Default: 'ema'
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): *pd.DataFrame.fillna(value)*

fill_method (value, optional): Type of fill method

Returns lower, basis, upper columns.

Return type *pd.DataFrame*

kst (*roc1=None, roc2=None, roc3=None, roc4=None, sma1=None, sma2=None, sma3=None, sma4=None, signal=None, drift=None, offset=None, **kwargs*) → *pandas.core.series.Series* 'Know Sure Thing' (KST)

The 'Know Sure Thing' is a momentum based oscillator and based on ROC.

Sources: [https://www.tradingview.com/wiki/Know_Sure_Thing_\(KST\)](https://www.tradingview.com/wiki/Know_Sure_Thing_(KST)) <https://www.incrediblecharts.com/indicators/kst.php>

Calculation:

Default Inputs: roc1=10, roc2=15, roc3=20, roc4=30, sma1=10, sma2=10, sma3=10, sma4=15, signal=9, drift=1

ROC = Rate of Change

SMA = Simple Moving Average

rocsma1 = SMA(ROC(close, roc1), sma1)

rocsma2 = SMA(ROC(close, roc2), sma2)

rocsma3 = SMA(ROC(close, roc3), sma3)

rocsma4 = SMA(ROC(close, roc4), sma4)

KST = 100 * (rocsma1 + 2 * rocsma2 + 3 * rocsma3 + 4 * rocsma4)

KST_Signal = SMA(KST, signal)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **roc1** (*int*) – ROC 1 period. Default: 10
- **roc2** (*int*) – ROC 2 period. Default: 15
- **roc3** (*int*) – ROC 3 period. Default: 20
- **roc4** (*int*) – ROC 4 period. Default: 30
- **sma1** (*int*) – SMA 1 period. Default: 10
- **sma2** (*int*) – SMA 2 period. Default: 10
- **sma3** (*int*) – SMA 3 period. Default: 10
- **sma4** (*int*) – SMA 4 period. Default: 15
- **signal** (*int*) – It’s period. Default: 9
- **drift** (*int*) – The difference period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns kst and kst_signal columns

Return type `pd.DataFrame`

kurtosis (*length=None, offset=None, **kwargs*)

Rolling Kurtosis

Sources:

Calculation:

Default Inputs: length=30

`KURTOSIS = close.rolling(length).kurt()`

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 30
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

linear_decay (*length=None, offset=None, **kwargs*)

Linear Decay

Adds a linear decay moving forward from prior signals like crosses.

Sources: <https://tulipindicators.org/decay>

Calculation:

Default Inputs: length=5

$\max(\text{close}, \text{close}[-1] - (1 / \text{length}), 0)$

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): *pd.DataFrame.fillna(value)*

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type *pd.Series*

linreg (*length=None, offset=None, **kwargs*)

Linear Regression Moving Average (linreg)

Linear Regression Moving Average

Source: TA Lib

Calculation:

Default Inputs: length=14

$x = [1, 2, \dots, n]$

$x_sum = 0.5 * \text{length} * (\text{length} + 1)$

$x2_sum = \text{length} * (\text{length} + 1) * (2 * \text{length} + 1) / 6$

$\text{divisor} = \text{length} * x2_sum - x_sum * x_sum$

lr(series): $y_sum = \text{series.sum}()$

$y2_sum = (\text{series} * \text{series}).\text{sum}()$

$xy_sum = (x * \text{series}).\text{sum}()$

$m = (\text{length} * xy_sum - x_sum * y_sum) / \text{divisor}$

$b = (y_sum * x2_sum - x_sum * xy_sum) / \text{divisor}$

$\text{return } m * (\text{length} - 1) + b$

$\text{linreg} = \text{close.rolling}(\text{length}).\text{apply}(\text{lr})$

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 10

- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

angle (bool, optional): Default: False. If True, returns the angle of the slope in radians

degrees (bool, optional): Default: False. If True, returns the angle of the slope in degrees

intercept (bool, optional): Default: False. If True, returns the angle of the slope in radians

r (bool, optional): Default: False. If True, returns it's correlation 'r'

slope (bool, optional): Default: False. If True, returns the slope

tsf (bool, optional): Default: False. If True, returns the Time Series Forecast value.

Returns New feature generated.

Return type pd.Series

log_return (*length=None, cumulative=False, percent=False, offset=None, **kwargs*) → pandas.core.series.Series

Log Return

Calculates the logarithmic return of a Series. See also: help(df.ta.log_return) for additional ****kwargs** a valid 'df'.

Sources: <https://stackoverflow.com/questions/31287552/logarithmic-returns-in-pandas-dataframe>

Calculation:

Default Inputs: length=1, cumulative=False

LOGRET = log(close.diff(periods=length))

CUMLOGRET = LOGRET.cumsum() if cumulative

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 20
- **cumulative** (*bool*) – If True, returns the cumulative returns. Default: False
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

long_run (*fast: pandas.core.series.Series, slow: pandas.core.series.Series, length=None, offset=None, **kwargs*)

Indicator: Long Run

macd (*fast=None, slow=None, signal=None, offset=None, **kwargs*) → pandas.core.series.Series
Moving Average Convergence Divergence (MACD)

The MACD is a popular indicator to that is used to identify a security's trend. While APO and MACD are the same calculation, MACD also returns two more series called Signal and Histogram. The Signal is an EMA of MACD and the Histogram is the difference of MACD and Signal.

Sources: [https://www.tradingview.com/wiki/MACD_\(Moving_Average_Convergence/Divergence\)](https://www.tradingview.com/wiki/MACD_(Moving_Average_Convergence/Divergence))

Calculation:

Default Inputs: fast=12, slow=26, signal=9

EMA = Exponential Moving Average

MACD = EMA(close, fast) - EMA(close, slow)

Signal = EMA(MACD, signal)

Histogram = MACD - Signal

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **fast** (*int*) – The short period. Default: 12
- **slow** (*int*) – The long period. Default: 26
- **signal** (*int*) – The signal period. Default: 9
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns macd, histogram, signal columns.

Return type `pd.DataFrame`

mad (*length=None, offset=None, **kwargs*)

Rolling Mean Absolute Deviation

Sources:

Calculation:

Default Inputs: length=30

mad = close.rolling(length).mad()

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 30
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

massi (*fast=None, slow=None, offset=None, **kwargs*)

Mass Index (MASSI)

The Mass Index is a non-directional volatility indicator that utilizes the High-Low Range to identify trend reversals based on range expansions.

Sources: https://stockcharts.com/school/doku.php?id=chart_school:technical_indicators:mass_index `mi = sum(ema(high - low, 9) / ema(ema(high - low, 9), 9), length)`

Calculation:

Default Inputs: fast: 9, slow: 25

EMA = Exponential Moving Average

hl = high - low

hl_ema1 = EMA(hl, fast)

hl_ema2 = EMA(hl_ema1, fast)

hl_ratio = hl_ema1 / hl_ema2

MASSI = SUM(hl_ratio, slow)

Parameters

- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **fast** (*int*) – The short period. Default: 9
- **slow** (*int*) – The long period. Default: 25
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

median (*length=None, offset=None, **kwargs*)

Rolling Median

Rolling Median of over ‘n’ periods. Sibling of a Simple Moving Average.

Sources: https://www.incrediblecharts.com/indicators/median_price.php

Calculation:

Default Inputs: length=30

MEDIAN = `close.rolling(length).median()`

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 30
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

mfi (*length=None, drift=None, offset=None, **kwargs*)

Money Flow Index (MFI)

Money Flow Index is an oscillator indicator that is used to measure buying and selling pressure by utilizing both price and volume.

Sources: [https://www.tradingview.com/wiki/Money_Flow_\(MFI\)](https://www.tradingview.com/wiki/Money_Flow_(MFI))

Calculation:

Default Inputs: length=14, drift=1

tp = typical_price = hlc3 = (high + low + close) / 3

rmf = raw_money_flow = tp * volume

pmf = pos_money_flow = SUM(rmf, length) if tp.diff(drift) > 0 else 0

nmf = neg_money_flow = SUM(rmf, length) if tp.diff(drift) < 0 else 0

MFR = money_flow_ratio = pmf / nmf

MFI = money_flow_index = 100 * pmf / (pmf + nmf)

Parameters

- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **volume** (*pd.Series*) – Series of ‘volume’s
- **length** (*int*) – The sum period. Default: 14
- **drift** (*int*) – The difference period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

midpoint (*length=None, offset=None, **kwargs*)

Indicator: Midpoint

midprice (*length=None, offset=None, **kwargs*)

Indicator: Midprice

mom (*length=None, offset=None, **kwargs*) → pandas.core.series.Series
Momentum (MOM)

Momentum is an indicator used to measure a security's speed (or strength) of movement. Or simply the change in price.

Sources: <http://www.onlinetradingconcepts.com/TechnicalAnalysis/Momentum.html>

Calculation:

Default Inputs: length=1

MOM = close.diff(length)

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

natr (*length=None, mamode=None, drift=None, offset=None, **kwargs*)
Normalized Average True Range (NATR)

Normalized Average True Range attempt to normalize the average true range.

Sources: <https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/normalized-average-true-range-natr/>

Calculation:

Default Inputs: length=20

ATR = Average True Range

NATR = (100 / close) * ATR(high, low, close)

Parameters

- **high** (*pd.Series*) – Series of 'high's
- **low** (*pd.Series*) – Series of 'low's
- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – The short period. Default: 20
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature

Return type pd.Series

nvi (*length=None, initial=None, offset=None, **kwargs*)
Negative Volume Index (NVI)

The Negative Volume Index is a cumulative indicator that uses volume change in an attempt to identify where smart money is active.

Sources: https://stockcharts.com/school/doku.php?id=chart_school:technical_indicators:negative_volume_inde https://www.motivewave.com/studies/negative_volume_index.htm

Calculation:

Default Inputs: length=1, initial=1000

ROC = Rate of Change

roc = ROC(close, length)

signed_volume = signed_series(volume, initial=1)

nvi = signed_volume[signed_volume < 0].abs() * roc_

nvi.fillna(0, inplace=True)

nvi.iloc[0]= initial

nvi = nvi.cumsum()

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **volume** (*pd.Series*) – Series of ‘volume’s
- **length** (*int*) – The short period. Default: 13
- **initial** (*int*) – The short period. Default: 1000
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): *pd.DataFrame.fillna(value)*

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type *pd.Series*

obv (*offset=None, **kwargs*)
On Balance Volume (OBV)

On Balance Volume is a cumulative indicator to measure buying and selling pressure.

Sources: [https://www.tradingview.com/wiki/On_Balance_Volume_\(OBV\)](https://www.tradingview.com/wiki/On_Balance_Volume_(OBV)) <https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/on-balance-volume-obv/>
https://www.motivewave.com/studies/on_balance_volume.htm

Calculation: signed_volume = signed_series(close, initial=1) * volume

obv = signed_volume.cumsum()

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **volume** (*pd.Series*) – Series of ‘volume’s

- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

ohlc4 (*offset=None, **kwargs*)

Indicator: OHLC4

percent_return (*length=None, cumulative=False, percent=False, offset=None, **kwargs*) → pandas.core.series.Series

Percent Return

Calculates the percent return of a Series. See also: help(df.ta.percent_return) for additional ****kwargs** a valid 'df'.

Sources: <https://stackoverflow.com/questions/31287552/logarithmic-returns-in-pandas-dataframe>

Calculation:

Default Inputs: length=1, cumulative=False

PCTRET = close.pct_change(length)

CUMPCTRET = PCTRET.cumsum() if cumulative

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 20
- **cumulative** (*bool*) – If True, returns the cumulative returns. Default: False
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

ppo (*fast=None, slow=None, signal=None, offset=None, **kwargs*)

Percentage Price Oscillator (PPO)

The Percentage Price Oscillator is similar to MACD in measuring momentum.

Sources: [https://www.tradingview.com/wiki/MACD_\(Moving_Average_Convergence/Divergence\)](https://www.tradingview.com/wiki/MACD_(Moving_Average_Convergence/Divergence))

Calculation:

Default Inputs: fast=12, slow=26

SMA = Simple Moving Average

EMA = Exponential Moving Average

fast_sma = SMA(close, fast)

slow_sma = SMA(close, slow)

```
PPO = 100 * (fast_sma - slow_sma) / slow_sma
```

```
Signal = EMA(PPO, signal)
```

```
Histogram = PPO - Signal
```

Parameters

- **close** (*pandas.Series*) – Series of ‘close’s
- **fast** (*int*) – The short period. Default: 12
- **slow** (*int*) – The long period. Default: 26
- **signal** (*int*) – The signal period. Default: 9
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns ppo, histogram, signal columns

Return type `pd.DataFrame`

pvi (*length=None, initial=None, offset=None, **kwargs*)
Positive Volume Index (PVI)

The Positive Volume Index is a cumulative indicator that uses volume change in an attempt to identify where smart money is active. Used in conjunction with NVI.

Sources: <https://www.investopedia.com/terms/p/pvi.asp>

Calculation:

Default Inputs: length=1, initial=1000

```
ROC = Rate of Change
```

```
roc = ROC(close, length)
```

```
signed_volume = signed_series(volume, initial=1)
```

```
pvi = signed_volume[signed_volume > 0].abs() * roc_
```

```
pvi.fillna(0, inplace=True)
```

```
pvi.iloc[0]= initial
```

```
pvi = pvi.cumsum()
```

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **volume** (*pd.Series*) – Series of ‘volume’s
- **length** (*int*) – The short period. Default: 13
- **initial** (*int*) – The short period. Default: 1000
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

pvol (*offset=None, **kwargs*)

Price-Volume (PVOL)

Returns a series of the product of price and volume.

Calculation:

if signed: $pvol = signed_series(close, 1) * close * volume$

else: $pvol = close * volume$

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **volume** (*pd.Series*) – Series of ‘volume’s
- **signed** (*bool*) – Keeps the sign of the difference in ‘close’s. Default: True
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

pvt (*drift=None, offset=None, **kwargs*)

Price-Volume Trend (PVT)

The Price-Volume Trend utilizes the Rate of Change with volume to and it’s cumulative values to determine money flow.

Sources: [https://www.tradingview.com/wiki/Price_Volume_Trend_\(PVT\)](https://www.tradingview.com/wiki/Price_Volume_Trend_(PVT))

Calculation:

Default Inputs: drift=1

ROC = Rate of Change

$pv = ROC(close, drift) * volume$

$PVT = pv.cumsum()$

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **volume** (*pd.Series*) – Series of ‘volume’s
- **drift** (*int*) – The diff period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

pwma (*length=None, asc=None, offset=None, **kwargs*)

Pascal's Weighted Moving Average (PWMA)

Pascal's Weighted Moving Average is similar to a symmetric triangular window except PWMA's weights are based on Pascal's Triangle.

Source: Kevin Johnson

Calculation:

Default Inputs: length=10

def weights(w):

def _compute(x): return np.dot(w * x)

 return _compute

triangle = utils.pascals_triangle(length + 1)

PWMA = close.rolling(length)._apply(weights(triangle), raw=True)

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 10
- **asc** (*bool*) – Recent values weigh more. Default: True
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

qstick (*length=None, offset=None, **kwargs*)

Q Stick

The Q Stick indicator, developed by Tushar Chande, attempts to quantify and identify trends in candlestick charts.

Sources: <https://library.tradingtechnologies.com/trade/chrt-ti-qstick.html>

Calculation:

Default Inputs: length=10

xMA is one of: sma (default), dema, ema, hma, rma

qstick = xMA(close - open, length)

Parameters

- **open** (*pd.Series*) – Series of 'open's
- **close** (*pd.Series*) – Series of 'close's

- **length** (*int*) – It's period. Default: 1
- **ma** (*str*) – The type of moving average to use. Default: None, which is 'sma'
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

quantile (*length=None, q=None, offset=None, **kwargs*)

Rolling Quantile

Sources:

Calculation:

Default Inputs: length=30, q=0.5

QUANTILE = close.rolling(length).quantile(q)

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 30
- **q** (*float*) – The quantile. Default: 0.5
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

rma (*length=None, offset=None, **kwargs*)

wildeR's Moving Average (RMA)

The WildeR's Moving Average is simply an Exponential Moving Average (EMA) with a modified alpha = 1 / length.

Sources: <https://alanhull.com/hull-moving-average>

Calculation:

Default Inputs: length=10

EMA = Exponential Moving Average

alpha = 1 / length

RMA = EMA(close, alpha=alpha)

Parameters

- **close** (*pd.Series*) – Series of 'close's

- **length** (*int*) – It's period. Default: 10
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

roc (*length=None, offset=None, **kwargs*) → pandas.core.series.Series
Rate of Change (ROC)

Rate of Change is an indicator is also referred to as Momentum (yeah, confusingly). It is a pure momentum oscillator that measures the percent change in price with the previous price 'n' (or length) periods ago.

Sources: [https://www.tradingview.com/wiki/Rate_of_Change_\(ROC\)](https://www.tradingview.com/wiki/Rate_of_Change_(ROC))

Calculation:

Default Inputs: length=1

MOM = Momentum

ROC = 100 * MOM(close, length) / close.shift(length)

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

rsi (*length=None, drift=None, offset=None, **kwargs*) → pandas.core.series.Series
Relative Strength Index (RSI)

The Relative Strength Index is popular momentum oscillator used to measure the velocity as well as the magnitude of directional price movements.

Sources: [https://www.tradingview.com/wiki/Relative_Strength_Index_\(RSI\)](https://www.tradingview.com/wiki/Relative_Strength_Index_(RSI))

Calculation:

Default Inputs: length=14, drift=1

ABS = Absolute Value

EMA = Exponential Moving Average

positive = close if close.diff(drift) > 0 else 0

negative = close if close.diff(drift) < 0 else 0

pos_avg = EMA(positive, length)

```
neg_avg = ABS(EMA(negative, length))
RSI = 100 * pos_avg / (pos_avg + neg_avg)
```

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 1
- **drift** (*int*) – The difference period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

rvi (*length=None, swma_length=None, offset=None, **kwargs*)
Relative Vigor Index (RVI)

The Relative Vigor Index attempts to measure the strength of a trend relative to its closing price to its trading range. It is based on the belief that it tends to close higher than they open in uptrends or close lower than they open in downtrends.

Sources: https://www.investopedia.com/terms/r/relative_vigor_index.asp

Calculation:

Default Inputs: length=14, swma_length=4

SWMA = Symmetrically Weighted Moving Average

numerator = SUM(SWMA(close - open, swma_length), length)

denominator = SUM(SWMA(high - low, swma_length), length)

RVI = numerator / denominator

Parameters

- **open** (*pd.Series*) – Series of ‘open’s
- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 14
- **swma_length** (*int*) – It’s period. Default: 4
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

short_run (*fast: pandas.core.series.Series, slow: pandas.core.series.Series, length=None, offset=None, **kwargs*)

Indicator: Short Run

sinwma (*length=None, offset=None, **kwargs*)

Sine Weighted Moving Average (SWMA)

A weighted average using sine cycles. The middle term(s) of the average have the highest weight(s).

Source: <https://www.tradingview.com/script/6MWFvnPO-Sine-Weighted-Moving-Average/> **Author:** Everget (<https://www.tradingview.com/u/everget/>)

Calculation:

Default Inputs: length=10

def weights(w):

def _compute(x): return np.dot(w * x)

 return _compute

sines = Series([sin((i + 1) * pi / (length + 1)) for i in range(0, length)])

w = sines / sines.sum()

SINWMA = close.rolling(length, min_periods=length).apply(weights(w), raw=True)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 10
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

skew (*length=None, offset=None, **kwargs*)

Rolling Skew

Sources:

Calculation:

Default Inputs: length=30

SKEW = close.rolling(length).skew()

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 30
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

slope (*length=None, as_angle=None, to_degrees=None, offset=None, **kwargs*)

Slope

Returns the slope of a series of length n. Can convert the slope to angle, default as radians or degrees.

Sources: Algebra I

Calculation:

Default Inputs: length=1

slope = close.diff(length) / length

if as_angle: slope = slope.apply(atan) if to_degrees:

slope *= 180 / PI

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: as_angle (value, optional): Converts slope to an angle. Default: False

to_degrees (value, optional): Converts slope angle to degrees. Default: False

fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

sma (*length=None, offset=None, **kwargs*) → pandas.core.series.Series

Simple Moving Average (SMA)

The Simple Moving Average is the classic moving average that is the equally weighted average over n periods.

Sources: <https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/simple-moving-average-sma/>

Calculation:

Default Inputs: length=10

SMA = SUM(close, length) / length

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 10
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: adjust (bool): Default: True

presma (bool, optional): If True, uses SMA for initial value.

fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

stdev (*length=None, offset=None, **kwargs*)

Rolling Standard Deviation

Sources:

Calculation:

Default Inputs: length=30

VAR = Variance

STDEV = `variance(close, length).apply(np.sqrt)`

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 30
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)` fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

stoch (*fast_k=None, slow_k=None, slow_d=None, offset=None, **kwargs*)

Stochastic (STOCH)

Stochastic Oscillator is a range bound momentum indicator. It displays the location of the close relative to the high-low range over a period.

Sources: [https://www.tradingview.com/wiki/Stochastic_\(STOCH\)](https://www.tradingview.com/wiki/Stochastic_(STOCH))

Calculation:

Default Inputs: fast_k=14, slow_k=5, slow_d=3

SMA = Simple Moving Average

lowest_low = low for last fast_k periods

highest_high = high for last fast_k periods

FASTK = $100 * (close - lowest_low) / (highest_high - lowest_low)$

FASTD = `SMA(FASTK, slow_d)`

SLOWK = `SMA(FASTK, slow_k)`

SLOWD = `SMA(SLOWK, slow_d)`

Parameters

- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **fast_k** (*int*) – The Fast %K period. Default: 14
- **slow_k** (*int*) – The Slow %K period. Default: 5
- **slow_d** (*int*) – The Slow %D period. Default: 3
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns fastk, fastd, slowk, slowd columns.

Return type `pd.DataFrame`

swma (*length=None, asc=None, offset=None, **kwargs*)

Symmetric Weighted Moving Average (SWMA)

Symmetric Weighted Moving Average where weights are based on a symmetric triangle. For example: `n=3` -> `[1, 2, 1]`, `n=4` -> `[1, 2, 2, 1]`, etc... This moving average has variable length in contrast to TradingView’s fixed length of 4.

Source: https://www.tradingview.com/study-script-reference/#fun_swma

Calculation:

Default Inputs: `length=10`

def weights(w):

def _compute(x): `return np.dot(w * x)`

`return _compute`

`triangle = utils.symmetric_triangle(length - 1)` SWMA = `close.rolling(length)._apply(weights(triangle), raw=True)`

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 10
- **asc** (*bool*) – Recent values weigh more. Default: True
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)` fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

t3 (*length=None, a=None, offset=None, **kwargs*)
Tim Tillson's T3 Moving Average (T3)

Tim Tillson's T3 Moving Average is considered a smoother and more responsive moving average relative to other moving averages.

Sources: <http://www.binarytribune.com/forex-trading-indicators/t3-moving-average-indicator/>

Calculation:

Default Inputs: length=10, a=0.7

$c1 = -a^3$ $c2 = 3a^2 + 3a^3 = 3a^2 * (1 + a)$ $c3 = -6a^2 - 3a - 3a^3$ $c4 = a^3 + 3a^2 + 3a + 1$

ema1 = EMA(close, length) ema2 = EMA(ema1, length) ema3 = EMA(ema2, length) ema4 = EMA(ema3, length) ema5 = EMA(ema4, length) ema6 = EMA(ema5, length) T3 = c1 * ema6 + c2 * ema5 + c3 * ema4 + c4 * ema3

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 10
- **a** (*float*) – $0 < a < 1$. Default: 0.7
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: adjust (bool): Default: True presma (bool, optional): If True, uses SMA for initial value. fillna (value, optional): pd.DataFrame.fillna(value) fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

tema (*length=None, offset=None, **kwargs*)
Triple Exponential Moving Average (TEMA)

A less laggy Exponential Moving Average.

Sources: <https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/triple-exponential-moving-average-tema/>

Calculation:

Default Inputs: length=10

EMA = Exponential Moving Average

ema1 = EMA(close, length)

ema2 = EMA(ema1, length)

ema3 = EMA(ema2, length)

TEMA = 3 * (ema1 - ema2) + ema3

Parameters

- **close** (*pd.Series*) – Series of 'close's
- **length** (*int*) – It's period. Default: 10
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: adjust (bool): Default: True

presma (bool, optional): If True, uses SMA for initial value.

fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

trend_return (trend: pandas.core.series.Series, log: bool = True, cumulative: bool = None, offset: int = None, trend_reset: int = 0, **kwargs)

Trend Return

Calculates the (Cumulative) Returns of a Trend as defined by some conditional. By default it calculates log returns but can also use percent change.

Sources: Kevin Johnson

Calculation:

Default Inputs: trend_reset=0, log=True, cumulative=False

sum = 0

returns = log_return if log else percent_return # These are not cumulative

returns = (trend * returns).apply(zero)

for i, in range(0, trend.size):

if item == trend_reset: sum = 0

else: return_ = returns.iloc[i] if cumulative:

sum += **return_**

else: sum = **return_**

trend_return.append(sum)

if cumulative and variable: trend_return += returns

Parameters

- **close** (pd.Series) – Series of ‘close’s
- **trend** (pd.Series) – Series of ‘trend’s. Preferably 0’s and 1’s.
- **trend_reset** (value) – Value used to identify if a trend has ended. Default: 0
- **log** (bool) – Calculate logarithmic returns. Default: True
- **cumulative** (bool) – If True, returns the cumulative returns. Default: False
- **offset** (int) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

variable (bool, optional): Whether to include if return fluxuations in the cumulative returns.

Returns New feature generated.

Return type pd.Series

trima (*length=None, offset=None, **kwargs*)
Triangular Moving Average (TRIMA)

A weighted moving average where the shape of the weights are triangular and the greatest weight is in the middle of the period.

Sources: <https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/triangular-moving-average-trima/> tma = sma(sma(src, ceil(length / 2)), floor(length / 2) + 1) # Tradingview trima = sma(sma(x, n), n) # Tradingview

Calculation:

Default Inputs: length=10

SMA = Simple Moving Average

half_length = math.round(0.5 * (length + 1))

SMA1 = SMA(close, half_length)

TRIMA = SMA(SMA1, half_length)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 10
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: adjust (bool): Default: True

fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

trix (*length=None, drift=None, offset=None, **kwargs*)
Trix (TRIX)

TRIX is a momentum oscillator to identify divergences.

Sources: <https://www.tradingview.com/wiki/TRIX>

Calculation:

Default Inputs: length=18, drift=1

EMA = Exponential Moving Average

ROC = Rate of Change

ema1 = EMA(close, length)

ema2 = EMA(ema1, length)

ema3 = EMA(ema2, length)

TRIX = 100 * ROC(ema3, drift)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 18
- **drift** (*int*) – The difference period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

true_range (*drift=None, offset=None, **kwargs*)

True Range

An method to expand a classical range (high minus low) to include possible gap scenarios.

Sources: <https://www.macroption.com/true-range/>

Calculation:

Default Inputs: drift=1

ABS = Absolute Value

prev_close = close.shift(drift)

TRUE_RANGE = ABS([high - low, high - prev_close, low - prev_close])

Parameters

- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **drift** (*int*) – The shift period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature

Return type `pd.Series`

tsi (*fast=None, slow=None, drift=None, offset=None, **kwargs*)

True Strength Index (TSI)

The True Strength Index is a momentum indicator used to identify short-term swings while in the direction of the trend as well as determining overbought and oversold conditions.

Sources: <https://www.investopedia.com/terms/t/tsi.asp>

Calculation:

Default Inputs: fast=13, slow=25, drift=1

```
EMA = Exponential Moving Average
diff = close.diff(drift)
slow_ema = EMA(diff, slow)
fast_slow_ema = EMA(slow_ema, slow)
abs_diff_slow_ema = absolute_diff_ema = EMA(ABS(diff), slow)
abema = abs_diff_fast_slow_ema = EMA(abs_diff_slow_ema, fast)
TSI = 100 * fast_slow_ema / abema
```

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **fast** (*int*) – The short period. Default: 13
- **slow** (*int*) – The long period. Default: 25
- **drift** (*int*) – The difference period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

uo (*fast=None, medium=None, slow=None, fast_w=None, medium_w=None, slow_w=None, drift=None, offset=None, **kwargs*)
Ultimate Oscillator (UO)

The Ultimate Oscillator is a momentum indicator over three different periods. It attempts to correct false divergence trading signals.

Sources: [https://www.tradingview.com/wiki/Ultimate_Oscillator_\(UO\)](https://www.tradingview.com/wiki/Ultimate_Oscillator_(UO))

Calculation:

Default Inputs: `fast=7, medium=14, slow=28, fast_w=4.0, medium_w=2.0, slow_w=1.0, drift=1`

```
min_low_or_pc = close.shift(drift).combine(low, min)
```

```
max_high_or_pc = close.shift(drift).combine(high, max)
```

```
bp = buying pressure = close - min_low_or_pc
```

```
tr = true range = max_high_or_pc - min_low_or_pc
```

```
fast_avg = SUM(bp, fast) / SUM(tr, fast)
```

```
medium_avg = SUM(bp, medium) / SUM(tr, medium)
```

```
slow_avg = SUM(bp, slow) / SUM(tr, slow)
```

```
total_weight = fast_w + medium_w + slow_w
```

```
weights = (fast_w * fast_avg) + (medium_w * medium_avg) + (slow_w * slow_avg)
```

```
UO = 100 * weights / total_weight
```

Parameters

- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **fast** (*int*) – The Fast %K period. Default: 7
- **medium** (*int*) – The Slow %K period. Default: 14
- **slow** (*int*) – The Slow %D period. Default: 28
- **fast_w** (*float*) – The Fast %K period. Default: 4.0
- **medium_w** (*float*) – The Slow %K period. Default: 2.0
- **slow_w** (*float*) – The Slow %D period. Default: 1.0
- **drift** (*int*) – The difference period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): *pd.DataFrame.fillna(value)*

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type *pd.Series*

update (*currency: str, base: str, time: str, limit: int*)

Set the current ohlcv dataframe to the latest data with the given args

Parameters

- {str} (*time*) –
- {str} –
- {str} –

variance (*length=None, offset=None, **kwargs*)

Rolling Variance

Sources:

Calculation:

Default Inputs: length=30

VARIANCE = close.rolling(length).var()

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 30
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): *pd.DataFrame.fillna(value)* fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type *pd.Series*

vortex (*length=None, drift=None, offset=None, **kwargs*)

Vortex

Two oscillators that capture positive and negative trend movement.

Sources: https://stockcharts.com/school/doku.php?id=chart_school:technical_indicators:vortex_indicator

Calculation:

Default Inputs: length=14, drift=1

TR = True Range

SMA = Simple Moving Average

tr = TR(high, low, close)

tr_sum = tr.rolling(length).sum()

vmp = (high - low.shift(drift)).abs()

vmn = (low - high.shift(drift)).abs()

VIP = vmp.rolling(length).sum() / tr_sum

VIM = vmn.rolling(length).sum() / tr_sum

Parameters

- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – ROC 1 period. Default: 14
- **drift** (*int*) – The difference period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): *pd.DataFrame.fillna(value)*

fill_method (value, optional): Type of fill method

Returns vip and vim columns

Return type *pd.DataFrame*

vp (*width=None, **kwargs*)

Volume Profile (VP)

Calculates the Volume Profile by slicing price into ranges. Note: Value Area is not calculated.

Sources: https://stockcharts.com/school/doku.php?id=chart_school:technical_indicators:volume_by_price https://www.tradingview.com/wiki/Volume_Profile <http://www.ranchodinero.com/volume-tpo-essentials/> <https://www.tradingtechnologies.com/blog/2013/05/15/volume-at-price/>

Calculation:

Default Inputs: width=10

vp = *pd.concat([close, pos_volume, neg_volume], axis=1)*

vp_ranges = *np.array_split(vp, width)*


```

result = ({high_close, low_close, mean_close, neg_volume, pos_volume} foreach range in vp_ranges)
vpdf = pd.DataFrame(result)
vpdf['total_volume'] = vpdf['pos_volume'] + vpdf['neg_volume']

```

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **volume** (*pd.Series*) – Series of ‘volume’s
- **width** (*int*) – How many ranges to distribute price into. Default: 10

Kwargs: fillna (value, optional): *pd.DataFrame.fillna(value)*

fill_method (value, optional): Type of fill method

sort_close (value, optional): Whether to sort by close before splitting into ranges. Default: False

Returns New feature generated.

Return type *pd.DataFrame*

vwap (*offset=None, **kwargs*)

Volume Weighted Average Price (VWAP)

The Volume Weighted Average Price that measures the average typical price by volume. It is typically used with intraday charts to identify general direction.

Sources: [https://www.tradingview.com/wiki/Volume_Weighted_Average_Price_\(VWAP\)](https://www.tradingview.com/wiki/Volume_Weighted_Average_Price_(VWAP))
<https://www.tradingtechnologies.com/help/x-study/technical-indicator-definitions/volume-weighted-average-price-vwap/>

Calculation: $tp = \text{typical_price} = \text{hlc3}(\text{high}, \text{low}, \text{close})$

$tpv = tp * \text{volume}$

$\text{VWAP} = \text{tpv.cumsum()} / \text{volume.cumsum()}$

Parameters

- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **volume** (*pd.Series*) – Series of ‘volume’s
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): *pd.DataFrame.fillna(value)*

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type *pd.Series*

vwma (*length=None, offset=None, **kwargs*) → *pandas.core.series.Series*

Volume Weighted Moving Average (VWMA)

Volume Weighted Moving Average.

Sources: https://www.motivewave.com/studies/volume_weighted_moving_average.htm

Calculation:

Default Inputs: length=10

SMA = Simple Moving Average

pv = close * volume

VWMA = SMA(pv, length) / SMA(volume, length)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **volume** (*pd.Series*) – Series of ‘volume’s
- **length** (*int*) – It’s period. Default: 10
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

willr (*length=None, offset=None, **kwargs*)

William’s Percent R (WILLR)

William’s Percent R is a momentum oscillator similar to the RSI that attempts to identify overbought and oversold conditions.

Sources: [https://www.tradingview.com/wiki/Williams_%25R_\(%25R\)](https://www.tradingview.com/wiki/Williams_%25R_(%25R))

Calculation:

Default Inputs: length=20

lowest_low = low.rolling(length).min()

highest_high = high.rolling(length).max()

WILLR = 100 * ((close - lowest_low) / (highest_high - lowest_low) - 1)

Parameters

- **high** (*pd.Series*) – Series of ‘high’s
- **low** (*pd.Series*) – Series of ‘low’s
- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 14
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

wma (*length=None, asc=None, offset=None, **kwargs*)
Weighted Moving Average (WMA)

The Weighted Moving Average where the weights are linearly increasing and the most recent data has the heaviest weight.

Sources: https://en.wikipedia.org/wiki/Moving_average#Weighted_moving_average

Calculation:

Default Inputs: length=10, asc=True

total_weight = 0.5 * length * (length + 1)

weights_ = [1, 2, ..., length + 1] # Ascending

weights = weights if asc else weights[::-1]

def linear_weights(w):

def _compute(x): return (w * x).sum() / total_weight

return _compute

WMA = close.rolling(length)._apply(linear_weights(weights), raw=True)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 10
- **asc** (*bool*) – Recent values weigh more. Default: True
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): pd.DataFrame.fillna(value)

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type pd.Series

z1ma (*length=None, offset=None, mamode=None, **kwargs*)
Zero Lag Moving Average (ZLMA)

The Zero Lag Moving Average attempts to eliminate the lag associated with moving averages. This is an adaption created by John Ehler and Ric Way.

Sources: https://en.wikipedia.org/wiki/Zero_lag_exponential_moving_average

Calculation:

Default Inputs: length=10, mamode=EMA

EMA = Exponential Moving Average

lag = int(0.5 * (length - 1))

source = 2 * close - close.shift(lag)

ZLMA = EMA(source, length)

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 10
- **mode** (*str*) – Two options: None or ‘ema’. Default: ‘ema’
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

zscore (*length=None, std=None, offset=None, **kwargs*)

Rolling Z Score

Sources:

Calculation:

Default Inputs: length=30, std=1

SMA = Simple Moving Average

STDEV = Standard Deviation

std = std * STDEV(close, length)

mean = SMA(close, length)

ZSCORE = (close - mean) / std

Parameters

- **close** (*pd.Series*) – Series of ‘close’s
- **length** (*int*) – It’s period. Default: 30
- **std** (*float*) – It’s period. Default: 1
- **offset** (*int*) – How many periods to offset the result. Default: 0

Kwargs: fillna (value, optional): `pd.DataFrame.fillna(value)`

fill_method (value, optional): Type of fill method

Returns New feature generated.

Return type `pd.Series`

2.3.4 Ionomy

class `Ionomy.Ionomy` (*api_key: str, api_secret: str*)

Bases: `object`

Base Ionomy API Wrapper

Parameters

- **{str}** -- **Ionomy API key** (*api_key*) –

- **{str} -- Ionomy API Secret** (*api_secret*) –

balance (*currency: str*) → Dict[str, Union[str, float]]

balances () → List[Dict[str, Union[str, float]]]

cancel_order (*orderId: str*) → bool

currencies () → List[Dict[str, Union[str, bool, int, float]]]

deposit_address (*currency: str*) → Dict[str, str]

deposit_history (*currency: str*) → List[Dict[str, Union[str, float]]]

limit_buy (*amount: Union[int, float], price: Union[int, float], market: str*) → dict

limit_sell (*amount: Union[int, float], price: Union[int, float], market: str*) → dict

market_history (*market: str*) → List[Dict[str, Union[str, float]]]

market_summaries () → List[Dict[str, Union[str, int, float]]]

market_summary (*market: str*) → Dict[str, Union[str, int, float]]

markets () → List[Dict[str, Union[str, float, bool]]]

open_orders (*market: str*) → List[Dict[str, str]]

order_book (*market: str*) → Dict[str, List[Dict[str, float]]]

order_status (*orderId: str*) → Dict[str, Optional[str]]

withdraw (*currency, amount, address*)

withdrawal_history (*currency: str*) → List[Dict[str, Union[str, float]]]

2.3.5 IonPanda

class Ionomy.IonPanda (*api_key: str, api_secret: str*)

Bases: Ionomy.ionomy.Ionomy

Pandas DataFrame Wrapper for Ionomy Base Class

Parameters

- **{str} -- Ionomy API key** (*api_key*) –
- **{str} -- Ionomy API Secret** (*api_secret*) –

balances () → pandas.core.frame.DataFrame

currencies () → pandas.core.frame.DataFrame

deposit_history (*currency: str*) → pandas.core.frame.DataFrame

market_history (*market: str*) → pandas.core.frame.DataFrame

market_summaries () → pandas.core.frame.DataFrame

markets () → pandas.core.frame.DataFrame

open_orders (*market: str*) → pandas.core.frame.DataFrame

order_book (*market: str*) → pandas.core.frame.DataFrame

withdrawal_history (*currency: str*) → pandas.core.frame.DataFrame

3.1 Contributing to ionomy-python

We welcome your contributions to our project.

3.1.1 Repository

The repository of ionomy-python is currently located at:

<https://github.com/Distortedlogic/ionomy-python>

3.1.2 Flow

This project makes heavy use of [git flow](#). If you are not familiar with it, then the most important thing for your to understand is that:

pull requests need to be made against the develop branch

3.1.3 How to Contribute

0. Familiarize yourself with [contributing on github](#)
1. Fork or branch from the develop branch.
2. Create commits following the commit style
3. Start a pull request to the develop branch
4. Wait for a [@distortedlogic](#) or another member to review

3.1.4 Issues

Feel free to submit issues and enhancement requests.

3.1.5 Contributing

Please refer to each project's style guidelines and guidelines for submitting patches and additions. In general, we follow the “fork-and-pull” Git workflow.

1. **Fork** the repo on GitHub
2. **Clone** the project to your own machine
3. **Commit** changes to your own branch
4. **Push** your work back up to your fork
5. Submit a **Pull request** so that we can review your changes

Note: Be sure to merge the latest from “develop” before making a pull request!

3.1.6 Copyright and Licensing

This library is open sources under the MIT license. We require you to release your code under that license as well.

3.2 Support and Questions

Help and discussion channel for beem can be found here:

- <https://discord.gg/ZvD9yXD>

3.3 Indices and tables

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