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# **Gemini Documentation**

*Release 1.0.0*

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### 1.1 Repository Clone

Currently the best way to using Gemini is to clone the Github repository.

```
git clone https://github.com/anfederico/Gemini
```

### 1.2 Quick Example

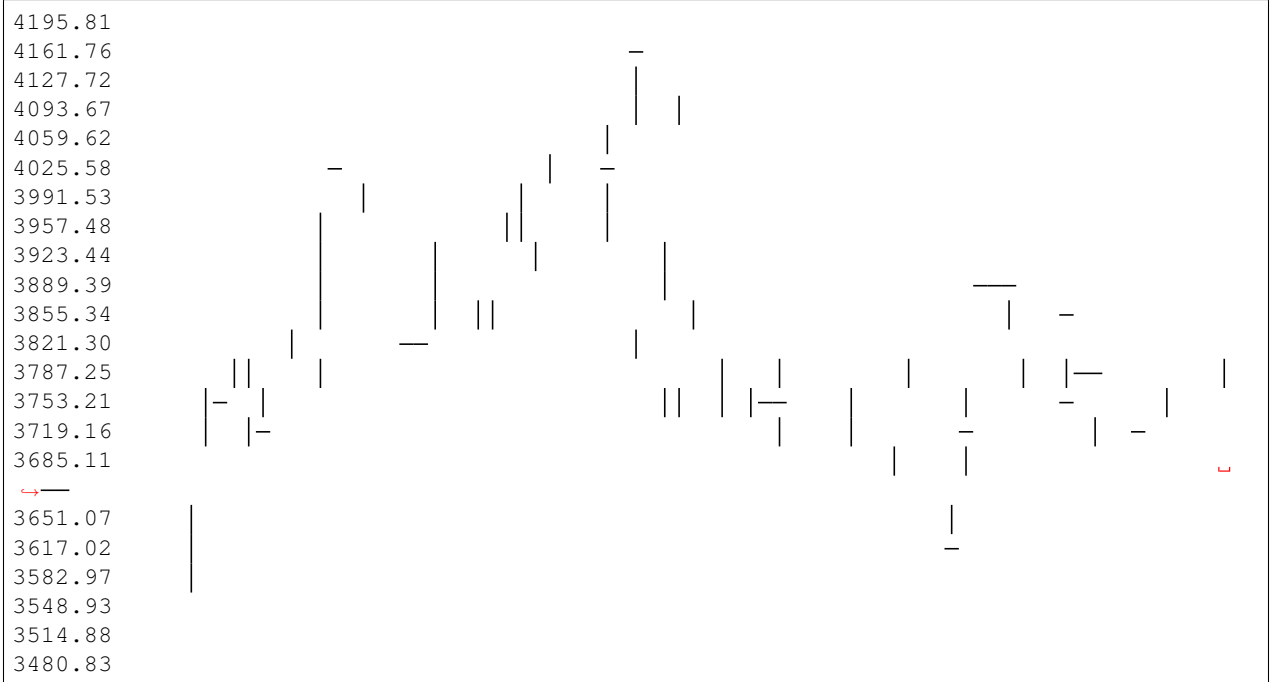
For a quick example that shows how one would run a backtest using a clone of Gemini, run the following.

```
cd Gemini/examples  
python mean_reversion.py
```









### 2.1 Importing Data

If you have your own data that has/hasn't been processed, you should conform to the following structure. Basically, load your data into a Pandas dataframe object and be sure to convert the dates to datetime format and include the following lowercase column titles.

	date	high	low	open	close
0	2017-07-08 11:00:00	2480.186778	2468.319314	2477.279567	2471.314030
1	2017-07-08 11:30:00	2471.314030	2455.014057	2471.202796	2458.073602
2	2017-07-08 12:00:00	2480.000000	2456.000000	2458.073602	2480.000000
3	2017-07-08 12:30:00	2489.004639	2476.334333	2479.402768	2481.481258
4	2017-07-08 13:00:00	2499.000000	2476.621873	2481.458643	2491.990000
5	2017-07-08 13:30:00	2503.503479	2490.314610	2492.440289	2496.005562
6	2017-07-08 14:00:00	2525.000000	2491.062741	2494.449524	2520.775500
7	2017-07-08 14:30:00	2521.500036	2510.000000	2520.775500	2518.450645
8	2017-07-08 15:00:00	2519.817394	2506.054360	2518.451000	2514.484009

## 2.2 Downloading Data

If you don't have your own data, we've included useful functions for grabbing low and high timeframe historical data from crypto exchanges. These helper functions will automatically resample your datasets to any desired timeframe and return a Gemini-compatible dataframe.

```
import data

# Higher timeframes (>= daily)
df = data.get_htf_candles("BTC_USD", "Bitfinex", "3-DAY", "2019-01-12 00:00:00",
↳"2019-02-01 00:00:00")

# Lower timeframes (< daily)
df = data.get_ltf_candles("USDC_BTC", "30-MIN", "2019-01-12 00:00:00", "2019-02-01_
↳00:00:00")
```



### 3.1 Initialize Backtest

Once your import or download data, you must initialize that backtesting engine with your data. This will simply create a backtesting class with your data pre-loaded.

```
import data

# Higher timeframes (>= daily)
df = data.get_hmf_candles("BTC_USD", "Bitfinex", "3-DAY", "2019-01-12 00:00:00",
    ↪ "2019-02-01 00:00:00")

# Lower timeframes (< daily)
df = data.get_ltf_candles("USDC_BTC", "30-MIN", "2019-01-12 00:00:00", "2019-02-01_
    ↪ 00:00:00")

# Loading data into the backtester
import engine

backtest = engine.backtest(df)
```

### 3.2 Define Strategy

In addition to loading the data, you must define the strategy you want to test. To do this, we'll create a logic function that can be passed to the backtester when you start. The backtester will proceed step-wise through the dataset, copying the current/past datapoints into a variable called "Lookback" to prevent lookahead bias. If the data hasn't already been processed, you may process it within the logic function (this makes the simulation more accurate but significantly increases runtime). You can then use the helper class called "Period" to conveniently reference current and past datapoints. With those, you may execute long, sell, short, and cover positions directly on the "Account" class based on your strategy.

### 3.2.1 Basic

A basic mock strategy example

```
def logic(account, lookback):
    try:
        # Process dataframe to collect signals
        lookback = helpers.get_signals(lookback)

        # Load into period class to simplify indexing
        lookback = helpers.period(lookback)

        today = lookback.loc(0) # Current candle
        yesterday = lookback.loc(-1) # Previous candle

        if today['signal'] == "down":
            if yesterday['signal'] == "down":
                exit_price = today['close']
                for position in account.positions:
                    if position.type == 'long':
                        account.close_position(position, 0.5, exit_price)

        if today['signal'] == "up":
            if yesterday['signal'] == "up":
                risk = 0.03
                entry_price = today['close']
                entry_capital = account.buying_power*risk
                if entry_capital >= 0:
                    account.enter_position('long', entry_capital, entry_price)

    except ValueError:
        pass # Handles lookback errors in beginning of dataset
```

### 3.2.2 Advanced

A real mean reversion strategy example

```
import pandas as pd
import numpy as np
from talib.abstract import *

def bands(df, timeperiod=26, nbdevup=2.6, nbdevdn=2.6, matype=0):
    cols = ['high', 'low', 'open', 'close', 'volume']
    HLOCV = {key: df[key].values for key in df if key in cols}
    u, m, l = BBANDS(HLOCV, timeperiod=timeperiod, nbdevup=nbdevup, nbdevdn=nbdevdn,
    ↪matype=matype)
    df['upper'] = u
    df['middle'] = m
    df['lower'] = l
    return df

def touches(df):
    df['touch_upper'] = df.high >= df.upper
    df['touch_lower'] = df.low <= df.lower
    df['crossing_dn'] = (df.close < df.middle) & (df.open > df.middle)
    df['crossing_up'] = (df.close > df.middle) & (df.open < df.middle)
```

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```

return df

def logic(account, lookback):
    try:
        lookback = bands(lookback)
        lookback = touches(lookback)
        lookback = helpers.period(lookback)
        today = lookback.loc(0)

        # Selling
        if today.touch_upper:
            exit_price = today.upper
            for position in account.positions:
                if position.type_ == 'long':
                    account.close_position(position, 1, exit_price)

        if today.crossing_up:
            exit_price = today.close
            for position in account.positions:
                if position.type_ == 'long':
                    account.close_position(position, 1, exit_price)

        # Buying
        if today.touch_lower | today.crossing_dn:
            risk = 1
            entry_price = today.lower
            entry_capital = account.buying_power*risk
            if entry_capital > 0:
                account.enter_position('long', entry_capital, entry_price)

        if today.crossing_dn:
            risk = 1
            entry_price = today.close
            entry_capital = account.buying_power*risk
            if entry_capital > 0:
                account.enter_position('long', entry_capital, entry_price)

    except Exception as e:
        print(e)
    pass # Handles lookback errors in beginning of dataset

```

### 3.3 Run Backtest

Running the backtest is as simple as running the following one-liner.

```
backtest.start(1000, logic)
```

### 3.4 Performance Statistics

After the backtest, you can analyze your strategy by printing the results to console. As of now, these include simple statistics of your run but we plan to implement more complicated metrics for a stronger understanding of performance.

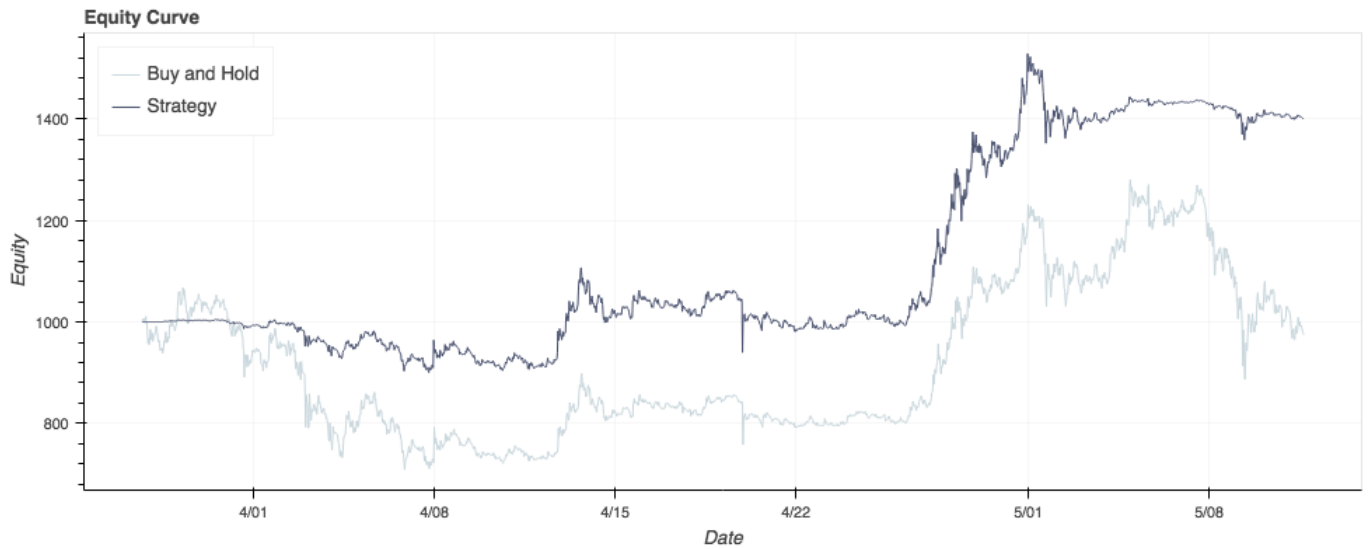
```
backtest.results()
```

```
----- Results -----  
Buy and Hold : 534.15%  
Net Profit   : 5341.47  
Strategy     : 3811.75%  
Net Profit   : 38117.46  
Longs        : 30  
Sells        : 31  
Shorts       : 0  
Covers       : 0  
-----  
Total Trades : 61  
-----
```

### 3.5 Equity Curve

You can visualize the performance of your strategy by comparing the equity curve with a buy and hold baseline. The equity curve simply tracks your account value throughout the backtest and will optionally show where your algorithm made its trades including longs, sells, shorts, and covers.

```
backtest.chart()
```

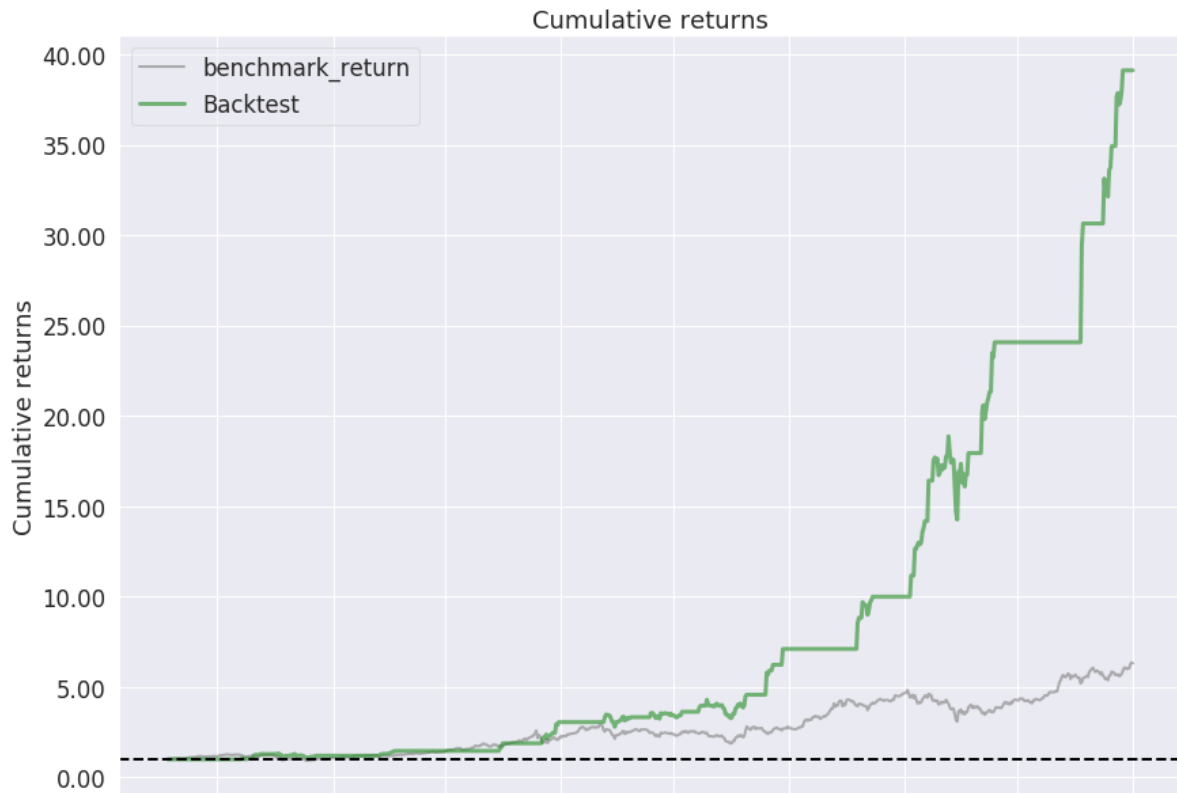




### 4.1 Pyfolio Example

### 4.2 Raw Statistics

### 4.3 Tearsheet Charts





## **5.1 gemini**

### **5.1.1 gemini package**

#### **5.1.1.1 Submodules**

**gemini.data module**

**gemini.engine module**

**gemini.exchange module**

**gemini.helpers module**

**gemini.ptable module**

**gemini.settings module**

#### **5.1.1.2 Module contents**