

MMAT Homework 5

Pittsburgh Campus

Jingyi Guo (jingyig1), Jiawen Zhang(jiawenz2)

1. Market Making Strategy

We implement the market making strategy for ZNU7 on Aug. 23.

Code:

```
from qpython import qconnection
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import datetime

def qSym(sym):
    return "`" + sym

def tsfmt(x):
    if x >= 0:
        return '0D%02d:00:00'% (x)
    else:
        return '-0D%02d:00:00'% (-x)

class MarketMakingStrategy(object):
    def __init__(self, db, d, sym, tL=-7, tR=16, maxPos=100, nmax = 0):
        self.db = db
        self.d = d
        self.sym = sym
        self.tL = tL
        self.tR = tR
        self.maxPos = maxPos
        self.nmax = nmax

    def queryData(self):
        query = "select from tqmergeT[" + self.d.strftime("%Y.%m.%d") + ";" +
        qSym(self.sym) + ";" + tsfmt(self.tL) + ';' + tsfmt(self.tR) + "]" where (differ
        bid)|(differ ask)"
        print(query)
        self.marketData = self.db(query)
        query = 'select minpxincr, dispfactor, notional from instinfo where inst
        =sym2inst[' + qSym(self.sym) + ']'
        print(query)
```

```

contractSpec = self.db(query)
self.pTick = contractSpec.minpxincr[0]
self.dispfactor = contractSpec.dispfactor[0]
self.notional = contractSpec.notional[0]
self.x = 0
self.P = 0
self.M = 0.5 * (self.marketData.bid[0] + self.marketData.ask[0])
self.bPrice = np.nan
self.aPrice = np.nan
self.bSize = np.nan
self.aSize = np.nan
self.tradeTime = [min(self.marketData.t)]
self.curPos = [self.x]
self.curCashFlow = [self.P]
self.curMidPrice = [self.M]

def trade(self, tseq, tTime, ssiz, mid, prc):
    if pd.isnull(prc): # quote
        if ssiz > 0:
            prc = self.bPrice
            self.bPrice = np.nan
        else:
            prc = self.aPrice
            self.aPrice = np.nan
        self.x += ssiz
        self.P -= ssiz * prc
        self.tradeTime.append(tTime)
        self.curPos.append(self.x)
        self.curCashFlow.append(self.P)
        self.curMidPrice.append(mid)

def runStrategy(self):
    bPrev, bsPrev, aPrev, asPrev = np.nan, np.nan, np.nan, np.nan
    for curdata in self.marketData.itertuples():
        if pd.isnull(curdata.prc):
            pmid = 0.5 * (curdata.bid + curdata.ask)

            if not pd.isnull(self.bPrice):
                if curdata.bid < self.bPrice:
                    self.trade(curdata.seq, curdata.t, +1, pmid, np.nan)
                elif curdata.bid > self.bPrice:
                    self.bPrice = np.nan

```

```

        if not pd.isnull(self.aPrice):
            if curdata.ask > self.aPrice:
                self.trade(curdata.seq, curdata.t, -1, pmid, np.nan)
            elif curdata.ask < self.aPrice:
                self.aPrice = np.nan
            bPrev, bsPrev, aPrev, asPrev = curdata.bid, curdata.bsiz,
curdata.ask, curdata.asiz

    else:
        pmid = 0.5 * (bPrev + aPrev)

        if not pd.isnull(self.bPrice):
            if curdata.prc < self.bPrice:
                self.trade(curdata.seq, curdata.t, +1, pmid, np.nan)
            elif curdata.prc == self.bPrice:
                self.bSize = self.bSize - curdata.siz
            if self.bSize < 0:
                self.trade(curdata.seq, curdata.t, +1, pmid, np.nan)

        if not pd.isnull(self.aPrice):
            if curdata.prc > self.aPrice:
                self.trade(curdata.seq, curdata.t, -1, pmid, np.nan)
            elif curdata.prc == self.aPrice:
                self.aSize = self.aSize - curdata.siz
            if self.aSize < 0:
                self.trade(curdata.seq, curdata.t, -1, pmid, np.nan)

    if pd.isnull(self.bPrice) and self.x < self.maxPos:
        self.bPrice, self.bSize = bPrev, bsPrev

    if pd.isnull(self.aPrice) and self.x > - self.maxPos:
        self.aPrice, self.aSize = aPrev, asPrev

    self.trade(max(self.marketData.seq), max(self.marketData.t), -
self.x, bPrev if self.x > 0 else aPrev, 0.5 * (bPrev + aPrev))
    self.res = pd.DataFrame({'tradeTime': self.tradeTime, 'position':
self.curPos, 'cashFlow': self.curCashFlow, 'midPrice': self.curMidPrice})
    self.res['PnL'] = self.res.cashFlow + self.res.position *
self.res.midPrice

    def plot(self):
        plt.figure(figsize = (8,6))

```

```

plt.subplot(311)
plt.plot(self.res.tradeTime, self.res.midPrice, linewidth=.5, color='b')
plt.ylabel('Mid Price')
plt.title('{} on {}, maxpos {}, notional {:.0f}'.format(self.sym,
self.d, self.maxPos, self.notional))
plt.subplot(312)
plt.plot(self.res.tradeTime, self.res.position, linewidth=.5, color='r')
plt.plot(self.res.tradeTime, np.zeros(self.res.tradeTime.shape), linewidth
=.5, color='k')
plt.ylabel('Position')
plt.subplot(313)
plt.plot(self.res.tradeTime, self.res.PnL, linewidth=.5, color='g')
plt.plot(self.res.tradeTime, np.zeros(self.res.tradeTime.shape), linewidth
=.5, color='k')
plt.ylabel('Mark-to-market P&L')
plt.savefig('MMResult.png')
plt.show()

```

```

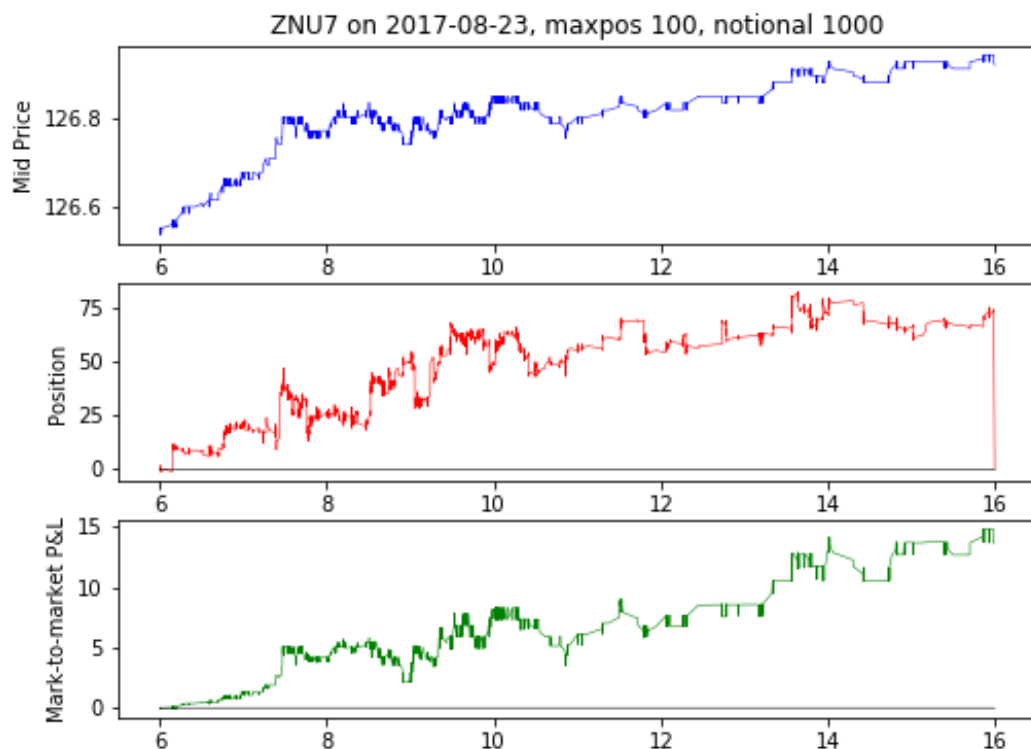
def main():
    curDT = datetime.datetime.now()
    db_IR = qconnection.QConnection(host='kx', port=6000, pandas = True)

    # initialize connection
    db_IR.open()
    d = datetime.date(2017, 8, 23)
    sym = 'ZNU7'
    tL = 6
    tR = 16
    maxPos = 100
    nmax = 0
    mms = MarketMakingStrategy(db_IR, d, sym, tL, tR, maxPos, nmax)
    mms.queryData()
    mms.runStrategy()
    mms.plot()
    print(datetime.datetime.now() - curDT)

if __name__ == '__main__':
    main()

```

Output:



- Reasons why this strategy might be profitable or not profitable:
According to the P&L plot, this strategy is generally profitable on Aug 23, this might because that market making strategy is generally profitable on mean reverting time series, and the ZNU7 prices somehow follows a mean-reversion pattern on Aug 23.
- How the strategy could be improved:
We can take into consideration the bid/ask size in the limit order book when buliding our inventory.
In addition, we could try dynamic inventory management instead of fixed constraints.

2. Given $dP_t = \sigma dB_t + \alpha dt + \langle \text{impact} \rangle$, where impact can be permanent and ~~per~~ temporary.

Also, the price trajectory: $\tilde{P}_t = P_0 + \alpha t + \sigma W_t + \nu(X_t - X_0) + H(\theta_0)$

3. The trading cost: $Z = \int_0^T \tilde{P}_t dX_t = \int_0^T [P_0 + \alpha t + \sigma W_t + \nu(X_t - X_0) + H(\theta_0)] dX_t$.

$$P_0 X + \frac{1}{2} \nu X^2 + \eta \int_0^T \theta_t^2 dt + \alpha \int_0^T t \theta_t dt + \sigma \int_0^T W(t) dX(t) = \text{non-random part} + \text{random part}$$

To calculate the statistics:

$$E(Z) = P_0 X + \frac{1}{2} \nu X^2 + \eta \int_0^T \theta_t^2 dt + \alpha \int_0^T t \theta_t dt \quad V(Z) = \sigma^2 \int_0^T (X - X_t)^2 dt$$

4. the execution cost: $C = Z - P_0 X$.

5. we form the optimization problem: $\max_{\substack{X(0)=0 \\ X(T)=X}} E(Z) + \lambda V(Z)$; we want to reform the problem:

$$\text{Let } z(t) = X(t) - \bar{x}, \quad (*) \Leftrightarrow F(z) = \eta \int_0^T z(t)^2 dt + \alpha \int_0^T t z'(t) dt + \lambda \sigma^2 \int_0^T (X - (z(t) + \bar{x}))^2 dt.$$

To have $z(0) = z(T) = 0$ with perturbation $z(t)$, adding up to optimal $y(t)$ is equivalent to have the linear part equaling 0 for any $z(t)$.

~~Fig 2.2.2.2~~

$$6. F(z+\bar{z}) - F(z) \sim \alpha \int_0^T z'(t) \bar{z}(t) dt + 2\eta \int_0^T z'(t) \bar{z}(t) dt + 2\lambda \sigma^2 \int_0^T T^2 (t(X(z(t) + \bar{x}) - X)) dt + O(\bar{z}^2).$$

$$\text{Here, } \int_0^T z'(t) \bar{z}(t) dt = - \int_0^T z(t) \bar{z}'(t) dt.$$

$$\int_0^T z(t) dt = - \int_0^T t z'(t) dt.$$

$$\therefore F(z+\bar{z}) - F(z) \approx 2 \int_0^T [\lambda \sigma^2 (z(t) + \bar{x} - X) - \eta z''(t) - \frac{1}{2} \alpha] z(t) dt + O(\bar{z}^2).$$

7. keep the linear term ~~Fig 2.2.2.2~~ 0: $\lambda \sigma^2 (z(t) + \bar{x} - X) - \eta z''(t) - \frac{1}{2} \alpha = 0$.

Just for intuition, we choose $\bar{x} = X + \frac{\alpha}{2\lambda \sigma^2}$ s.t. $\lambda \sigma^2 (\bar{x} - X) - \frac{1}{2} \alpha = 0$, and $z(t) = \bar{z}^2 z(t)$.

The solution for $z(t)$ that satisfies the boundary condition $z(0) = -\bar{x}$, $z(T) = X - \bar{x}$.

$$z(t) = -\frac{\alpha}{2\lambda \sigma^2} \frac{\sinh(k(T-t)) + \sinh(kt)}{\sinh(kT)} - X \frac{\sinh(k(T-t))}{\sinh(kT)}$$

$$x(t) = z(t) - \bar{x} = \frac{\alpha}{2\lambda \sigma^2} \left(1 - \frac{\sinh(k(T-t)) + \sinh(kt)}{\sinh(kT)}\right) + X \left(1 - \frac{\sinh(k(T-t))}{\sinh(kT)}\right)$$

8. When the static optimal portfolio $\alpha^* = \frac{\alpha}{2\lambda \sigma^2}$ is larger than the target position X , we can likely have the trajectory temporarily exceeding the target position.

⑨ Some explanation:

A. $\bar{x} = X + x^*$, Here, \bar{x} is the optimal level the optimal trajectory trying to "reach".

And the $\alpha(t)$ is the actual distance between the "optimal level" and current price level.

Also, \bar{x} is ~~the part we~~ introduced to reduce the computational cost.

B. for $\alpha(t) - \bar{x} > 0$. $x^*(\sinh(kT) - \sinh(k(T-t)) - \sinh(kt)) > X \sinh(k(T-t))$

$$\Leftrightarrow \frac{x^*}{X} > \frac{\sinh(k(T-t))}{(\sinh(kT) - \sinh(k(T-t)) - \sinh(kt))}$$

$$\therefore \sinh(kT) = \sinh(k(T-t)) \cosh(kt) + \cosh(k(T-t)) \sinh(kt) \geq \sinh(k(T-t)) + \sinh(kt)$$

For this equation, the RHS is positive. Therefore, if x is small enough, the optimal trajectory may exceed X and reverse the direction of trading.

C. If $\alpha < 0$, the optimal position is then ~~is~~ negative. Thus, the optimal trajectory may firstly short the security and then buy it back.