

```
'''
Author:      Weiyi Chen
Copyright:   Copyright (C) 2015 Baruch College, Modeling and Market Making in
            Forex Exchange
Description: This programming question will try to determine whether using a
            factor-based approach to reducing
            dimensionality is better than an ad hoc method. The result will show that setting
            hedge notionals based on the
            true factor shocks should provide a better hedge performance than based on the ad
            hoc triangle shocks.
Test:        hedger_test.py
'''
```

```
from numpy import *
from lazy import lazy
```

```
class Hedger(object):
    """
```

```
    In the toy market, we assume that we know the dynamics of the asset currency
    interest rate:
```

$$dQ = \sigma_1 e^{(-\beta_1 T)} dz_1 + \sigma_2 e^{(-\beta_2 T)} dz_2$$

$$E[dz_1 dz_2] = \rho dt$$

```
where  $\sigma_1 = 1\%/\sqrt{\text{yr}}$ ,  $\sigma_2 = 0.8\%/\sqrt{\text{yr}}$ ,  $\beta_1 = 0.5/\text{yr}$ ,  $\beta_2 = 0.1/\text{yr}$ , and  $\rho = -0.4$ .
    """
```

```
def __init__(self):
    super(Hedger, self).__init__()
```

```
@lazy
```

```
def Spot(self):
    ''' We start by assuming a toy market: spot = 1 '''
    return 1.
```

```
@lazy
```

```
def Q(self):
    ''' asset currency interest rate curve = Q(T) = flat at 3% '''
    return .03
```

```
@lazy
```

```
def R(self):
    ''' denominated currency interest rate curve = R(T) = flat at 0% '''
    return 0.
```

```
@lazy
```

```
def Sigma1(self):
    ''' vol of the 1st parameter '''
    return .01
```

```
@lazy
```

```
def Sigma2(self):
    ''' vol of the 2nd parameter '''
    return .008
```

```
@lazy
```

```
def Beta1(self):
```

```

    ''' mean reversion of 1st parameter '''
    return .5

@lazy
def Beta2(self):
    ''' mean reversion of 1st parameter '''
    return .1

@lazy
def Rho(self):
    ''' correlation between two brownian motions '''
    return -.4

@lazy
def T1(self):
    '''
    1st benchmark date, we will use forwards to settlement date to hedge the
    forward rate risk (or equivalently,
    the risk to moves in the asset currency interest rate) of our portfolio.
    '''
    return .25

@lazy
def T2(self):
    '''
    2nd benchmark date, we will use forwards to settlement date to hedge the
    forward rate risk (or equivalently,
    the risk to moves in the asset currency interest rate) of our portfolio.
    '''
    return 1.

@lazy
def Dt(self):
    '''
    the length of the time step the simulation advances over

    Simulate the portfolio forward a time dt=0.001y. That will result in the
    asset-currency rates moving according
    to the factor model described above, which shocks the benchmark rates for
    tenors T1 and T2, and for the portfolio's
    risk tenor T.
    '''
    return 1e-3

@lazy
def SqrtDt(self):
    ''' square root of dt '''
    return sqrt(self.Dt)

@lazy
def Nruns(self):
    ''' the number of Monte Carlo runs '''
    return 1e5

@lazy
def Tenor(self):

```

```

'''
The portfolio to hedge has one position: a unit asset-currency notional
of a forward contract settling at time T.
You'll try this for values of T in [0.1,0.25,0.5,0.75,1,2] to see how
performance changes for portfolios with
risk to different tenors.
'''
return .1

@lazy
def HedgingStrategy(self):
    '''
    You will try three different hedging strategies:

    Value 1: one where you choose the hedge notionals (of forwards settling
    at times T1 and T2) based on the triangle
    shock we discussed in class (though as there are only two benchmarks here
    , the T1 shock will be flat before T1 and
    the T2 shock will be flat after T2);
    Value 2: one where the notionals are set to hedge the actual two shocks
    from the factors described above;
    Value 3: lastly, one where you don't hedge at all.
    '''
    return 0

@lazy
def Dz1s(self):
    ''' first brownian motion '''
    return random.normal(0, self.SqrtDt, self.Nruns)

@lazy
def Dz2s(self):
    ''' second brownian motion '''
    return self.Rho * self.Dz1s + sqrt(1 - self.Rho ** 2) * random.normal(0,
    self.SqrtDt, self.Nruns)

@lazy
def DQTs(self):
    ''' the dynamics of the asset currency interest rate '''
    return self.Sigma1 * exp(-self.Beta1 * self.Tenor) * self.Dz1s + self.
    Sigma2 * exp(-self.Beta2 * self.Tenor) * self.Dz2s

@lazy
def DQ1s(self):
    ''' rate shocks for T1 '''
    return self.Sigma1 * exp(-self.Beta1 * self.T1) * self.Dz1s + self.Sigma2
    * exp(-self.Beta2 * self.T1) * self.Dz2s

@lazy
def DQ2s(self):
    ''' rate shocks for T2 '''
    return self.Sigma1 * exp(-self.Beta1 * self.T2) * self.Dz1s + self.Sigma2
    * exp(-self.Beta2 * self.T2) * self.Dz2s

@lazy
def DQT_dQ1(self):

```

```

''' the resulting shock to Q(T); i.e. dQ(T)/dQ1 '''
dz1 = -1./self.Sigma1 * exp(self.Beta1*self.T2-self.Beta2*(self.T2-self.
    T1)) / (1.-exp((self.Beta1-self.Beta2)*(self.T2-self.T1)))
dz2 = 1./self.Sigma2 * exp(self.Beta2*self.T1) / (1.-exp((self.Beta1-
    self.Beta2)*(self.T2-self.T1)))
return self.Sigma1 * exp(-self.Beta1*self.Tenor)*dz1 + self.Sigma2 * exp
    (-self.Beta2*self.Tenor)*dz2

@lazy
def DQT_dQ2(self):
    ''' second shock, i.e. dQ(T)/dQ2 '''
    dz1 = -1./self.Sigma1 * exp(self.Beta1*self.T1+self.Beta2*(self.T2-self.
        T1)) / (1.-exp((self.Beta2-self.Beta1)*(self.T2-self.T1)))
    dz2 = 1./self.Sigma2 * exp(self.Beta2*self.T2) / (1.-exp((self.Beta2-
        self.Beta1)*(self.T2-self.T1)))
    return self.Sigma1 * exp(-self.Beta1*self.Tenor)*dz1 + self.Sigma2*exp(-
        self.Beta2*self.Tenor)*dz2

@lazy
def HedgingNotional1(self):
    '''
    Add in the hedges: two forwards, settling at times T1 and T2, with
    notionals set to hedge the portfolio (either
    against the two triangle shocks or against the two factor shocks).
    '''
    if self.HedgingStrategy == 0:
        return 0.
    elif self.HedgingStrategy == 1:
        if self.Tenor <= self.T1:
            return self.Tenor / self.T1 * exp(-self.Q*(self.Tenor-self.T1))
        elif self.Tenor >= self.T2:
            return 0.
        else:
            return (self.T2-self.Tenor) / (self.T2-self.T1) * self.Tenor /
                self.T1 * exp(-self.Q*(self.Tenor-self.T1))
    elif self.HedgingStrategy == 2:
        return self.DQT_dQ1 * self.Tenor / self.T1 * exp(-self.Q*(self.Tenor-
            self.T1))
    else:
        raise TypeError('Hedging Strategy can only be 0, 1 or 2')

@lazy
def HedgingNotional2(self):
    '''
    Add in the hedges: two forwards, settling at times T1 and T2, with
    notionals set to hedge the portfolio (either
    against the two triangle shocks or against the two factor shocks).
    '''
    if self.HedgingStrategy == 0:
        return 0.
    elif self.HedgingStrategy == 1:
        if self.Tenor <= self.T1:
            return 0.
        elif self.Tenor >= self.T2:
            return self.Tenor/self.T2*exp(-self.Q*(self.Tenor-self.T2))
        else:

```

```

        return (self.Tenor-self.T1)/(self.T2-self.T1)*self.Tenor/self.T2*
               exp(self.Q*(self.T2-self.Tenor))
    elif self.HedgingStrategy == 2:
        return self.DQT_dQ2 * self.Tenor / self.T2 * exp(self.Q*(self.T2-self
            .Tenor))
    else:
        raise TypeError('Hedging Strategy can only be 0, 1 or 2')

@lazy
def PNLs(self):
    ''' Construct the PNL distributions for the three hedging approaches.
        Determine the PNL realized. '''
    pnls = self.Spot * (exp(-(self.Q+self.DQTs)*self.Tenor) - exp(-self.Q*
        self.Tenor))
    pnls -= self.HedgingNotional1 * self.Spot * (exp(-(self.Q+self.DQ1s)*self
        .T1) - exp(-self.Q*self.T1))
    pnls -= self.HedgingNotional2 * self.Spot * (exp(-(self.Q+self.DQ2s)*self
        .T2) - exp(-self.Q*self.T2))
    return pnls

@lazy
def PNL_std(self):
    ''' standard deviation of simulation PNL '''
    return self.PNLs.std()

```