

MATH 773: Assignment 5

Due on Wednesday, December 22, 2010

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Problem 1

Proof We have

$$\psi(\theta) = a\theta + \frac{1}{2} \sum_{j=1}^m \left(\frac{\theta^2 b_j^2}{1 - 2\theta b_j} - \ln(1 - 2\theta \lambda_j) \right)$$

$$Q = a + \sum_{j=1}^m b_j Z_k + \sum_{j=1}^m \lambda_j Z_j^2$$

The CDF of $Z \sim N(0, I)$ under new measure \mathbb{P}_θ is

$$\begin{aligned} \mathbb{P}_\theta(Z_1 \leq z_1, \dots, Z_m \leq z_m) &= \mathbb{E}(e^{\theta Q - \psi(\theta)} \mathbb{1}(Z_1 \leq z_1, \dots, Z_m \leq z_m)) \\ &= \int \dots \int_{\mathcal{Z}_j = -\infty}^{z_j} e^{\theta Q - \psi(\theta)} \frac{e^{-\frac{1}{2} \bar{Z}' \Sigma^{-1} \bar{Z}}}{(2\pi)^{\frac{m}{2}} |\Sigma|^{\frac{1}{2}}} dZ_j \\ &= \frac{1}{(2\pi)^{\frac{m}{2}}} \int \dots \int_{\mathcal{Z}_j = -\infty}^{z_j} e^{\theta Q - \psi(\theta) - \frac{1}{2} \sum_{j=1}^m Z_j^2} dZ_j \\ &= \frac{1}{(2\pi)^{\frac{m}{2}}} \int \dots \int_{\mathcal{Z}_j = -\infty}^{z_j} e^{\theta(a + \sum_{j=1}^m b_j Z_k + \sum_{j=1}^m \lambda_j Z_j^2) - a\theta + \frac{1}{2} \sum_{j=1}^m \left(\frac{\theta^2 b_j^2}{1 - 2\theta b_j} - \ln(1 - 2\theta \lambda_j) \right) - \frac{1}{2} \sum_{j=1}^m Z_j^2} dZ_j \\ &= \frac{1}{(2\pi)^{\frac{m}{2}}} \int \dots \int_{\mathcal{Z}_j = -\infty}^{z_j} e^{\frac{1}{2} \sum_{j=1}^m \ln(1 - 2\theta \lambda_j)} e^{\sum_{j=1}^m (\theta \lambda_j - \frac{1}{2}) Z_j^2} e^{\sum_{j=1}^m \theta \lambda_j Z_j - \frac{1}{2} \sum_{j=1}^m \frac{\theta^2 b_j^2}{1 - 2\theta b_j}} \\ &= \frac{\left(\prod_{j=1}^m (1 - 2\theta \lambda_j) \right)^{\frac{1}{2}}}{(2\pi)^{\frac{m}{2}}} \int \dots \int_{\mathcal{Z}_j = -\infty}^{z_j} e^{\sum_{j=1}^m (\theta \lambda_j - \frac{1}{2}) Z_j^2 + \sum_{j=1}^m \theta \lambda_j Z_j - \frac{1}{2} \sum_{j=1}^m \frac{\theta^2 b_j^2}{1 - 2\theta b_j}} \end{aligned}$$

We want this to be the CDF of Z under measure \mathbb{P}_θ , i.e. $Z \sim N(\bar{\mu}, \Sigma')$ where

$$\mu_j(\theta) = \frac{\theta b_j}{1 - 2\theta \lambda_j},$$

and Σ' is a diagonal matrix with diagonal entries

$$\sigma_j^2(\theta) = \frac{1}{1 - 2\theta \lambda_j}.$$

Then the CDF of $Z \sim N(\bar{\mu}, \Sigma')$ is

$$\begin{aligned} \frac{e^{-\frac{1}{2}(\bar{z} - \bar{\mu})' \Sigma'^{-1}(\bar{z} - \bar{\mu})}}{(2\pi)^{\frac{m}{2}} |\Sigma'|^{\frac{1}{2}}} &= \frac{1}{(2\pi)^{\frac{m}{2}} \left(\prod_{j=1}^m \left(\frac{1}{1 - 2\theta \lambda_j} \right) \right)^{\frac{1}{2}}} e^{-\frac{1}{2} \sum_{j=1}^m \left(z_j - \frac{\theta b_j}{1 - 2\theta \lambda_j} \right) (1 - 2\theta \lambda_j) \left(z_j - \frac{\theta b_j}{1 - 2\theta \lambda_j} \right)} \\ &= \frac{\left(\prod_{j=1}^m (1 - 2\theta \lambda_j) \right)^{\frac{1}{2}}}{(2\pi)^{\frac{m}{2}}} e^{-\frac{1}{2} \sum_{j=1}^m \left((1 - 2\theta \lambda_j) z_j^2 - 2\theta b_j z_j + \frac{\theta^2 b_j^2}{1 - 2\theta b_j} \right)} \\ &= \frac{d}{d\bar{z}} \mathbb{P}_\theta(Z_1 \leq z_1, \dots, Z_m \leq z_m) \end{aligned}$$

Therefore under the new measure \mathbb{P}_θ Z is indeed still a multivariate normal random variable, with mean $\bar{\mu}$ and variance matrix Σ' . □

Problem 2

Proof From the text we know that

$$\mathbb{E}(e^{\theta Q_x} | Y) = e^{\alpha(\theta)Y} \prod_{j=1}^m \frac{1}{\sqrt{1 - 2\theta \lambda_j}}$$

where

$$\alpha(\theta) = (a-x) \frac{\theta}{\nu} + \frac{1}{2} \sum_{j=1}^m \frac{\theta^2 b_j^2 \nu}{1-2\theta\lambda_j}$$

We also know that

$$\phi_x(\theta) = \mathbb{E}(e^{\theta Q_x}) = \phi_Y(\alpha(\theta)) \prod_{j=1}^m \frac{1}{\sqrt{1-2\theta\lambda_j}}$$

and its cumulant generating function

$$\psi_x(\theta) = \ln \phi_x(\theta) = \ln \mathbb{E}(e^{\theta Q_x})$$

The density of Y under new measure \mathbb{P}_θ can be found by differentiating

$$\begin{aligned} \mathbb{P}_\theta(Y \leq y) &= \mathbb{E}\left(\frac{d\mathbb{P}_\theta}{d\mathbb{P}} \mathbb{1}(Y \leq y)\right) \\ &= \mathbb{E}(e^{\theta Q_x - \psi_x(\theta)} \mathbb{1}(Y \leq y)) \\ &= e^{-\psi_x(\theta)} \mathbb{E}(\mathbb{E}(e^{\theta Q_x} \mathbb{1}(Y \leq y) | Y)) \\ &= e^{-\psi_x(\theta)} \mathbb{E}\left(e^{\alpha(\theta)Y} \prod_{j=1}^m \frac{1}{\sqrt{1-2\theta\lambda_j}} \mathbb{1}(Y \leq y)\right) \\ &= \frac{1}{\mathbb{E}(e^{\theta Q_x})} \int_0^y e^{\alpha(\theta)\zeta} \prod_{j=1}^m \frac{1}{\sqrt{1-2\theta\lambda_j}} f_Y(\zeta) d\zeta \\ &= \frac{1}{\phi_Y(\alpha(\theta))} \int_0^y e^{\alpha(\theta)\zeta} f_Y(\zeta) d\zeta \\ \Rightarrow f_{Y,\theta}(y) &= \frac{d}{dy} \mathbb{P}_\theta(Y \leq y) = \frac{d}{dy} \left(\frac{1}{\phi_Y(\alpha(\theta))} \int_0^y e^{\alpha(\theta)\zeta} f_Y(\zeta) d\zeta \right) \\ &= \frac{e^{\alpha(\theta)y}}{\phi_Y(\alpha(\theta))} f_Y(y) \end{aligned}$$

□

Problem 3

Portfolio A: Standard Monte Carlo

```

clc;
clear;
rand('state',0);
%Portfolio A: Short positions in 10 calls and 5 puts on each of 10
%underlying assets, all options expiring in 0.1 years.
%% Initialize portfolio parameters
volatility=0.3;
rate=0.05;
maturity=0.1;
NCall=10;
NPut=5;
NAsset=10;
NRepl=120000;
QEBinN=600; %Bin size used for VaR quantile estimation.
QEIndex=2; %Estimated quantile=1-QEIndex/(NRepl/QEBinSize)=0.99
assetprice0=ones(NAsset,1);
assetprice=zeros(NAsset,NRepl);

```

```

strikes=[[0.55:0.1:1.45], [0.8:0.1:1.2]];
payoff=zeros(1,NRepl);
%% Calculate theoretical portfolio value
TPV=0;
for i=1:NAset
    call=zeros(1,NCall+NPut);
    put=zeros(1,NCall+NPut);
    for j=1:NCall+NPut
        [call(j),put(j)]=blsprice(assetprice0(i),strikes(j),rate,maturity,volatility,0);
    end
    TPV=TPV+sum([call(1:NCall),put(NCall+1:NCall+NPut)]);
end
%% Start Monte Carlo runs
assetprice=diag(assetprice0)*exp((rate-volatility^2/2)*maturity+sqrt(maturity)*...
    volatility*randn(size(assetprice)));
%% Calculate option prices from simulation results
for j=1:NRepl
    for i=1:NAset
        payoff(j)=payoff(j)+sum([max(assetprice(i,j)-strikes(1:NCall),0),...
            max(strikes(NCall+1:NCall+NPut)-assetprice(i,j),0)]);
    end
end
%% Calculate simulated portfolio loss
SPL=TPV-payoff;
%% Calculate VaR from quantile estimation
VaRQE=zeros(1,QEBinN);
QEBinSize=NRepl/QEBinN;
for i=1:QEBinSize
    VaRQE(i)=quantile(SPL(1+(i-1)*QEBinSize:i*QEBinSize),0.01);
end
[mean,sigma,CI]=normfit(VaRQE)

mean =

    -0.9222

sigma =

    1.3205

CI =

    -1.0280
    -0.8163

```

▷

Portfolio B: Standard Monte Carlo

```

clc;
clear;
rand('state',0);
%Portfolio B: Same as A, but with the number of puts increase to produce a

```

```

%net delta zero.
%% Initialize portfolio parameters
volatility=0.3;
rate=0.05;
maturity=0.1;
NCall=10;
NPut=5;%Increase number of puts in portfolio
NAsset=10;
NRepl=120000;
QEBinN=600; %Bin size used for VaR quantile estimation.
QEIndex=2; %Estimated quantile=1-QEIndex/(NRepl/QEBinSize)=0.99
assetprice0=ones(NAsset,1);
assetprice=zeros(NAsset,NRepl);
strikes=[[0.55:0.1:1.45], [0.8:0.1:1.2]];
payoff=zeros(1,NRepl);
%% Determine number of puts to be added to obtain 0 net delta
CallDelta=zeros(NAsset,1);
PutDelta=zeros(NAsset,1);
ExcessDelta=zeros(2,1);
for i=1:NAsset
    CallDelta(i)=sum(normcdf((log(assetprice0(i)./strikes(1:NCall))+...
        (rate+volatility^2/2)*maturity)/(volatility*sqrt(maturity)))));
    PutDelta(i)=sum(normcdf((log(assetprice0(i)./strikes(NCall+1:NCall+NPut))...
        +(rate+volatility^2/2)*maturity)/(volatility*sqrt(maturity)))-1);
end
ExcessDelta(1)=sum(CallDelta)+sum(PutDelta);
ExcessDelta(2)=ExcessDelta(1);
%ExcessDelta determined. Try to find countering put option number.
n=0;
AddPutStrike=[1-0.1*n:0.1:1+0.1*n];
AddPutDelta=FindPutDelta(assetprice0,AddPutStrike,rate,volatility,maturity);
while (n<10 && AddPutDelta+ExcessDelta(2)>0)
    n=n+0.5;
    AddPutStrike=[1-0.1*(n+0.5):0.1:1+0.1*(n+0.5)];
    AddPutDelta=FindPutDelta(assetprice0,AddPutStrike,rate,volatility,maturity);
end
AddPutStrike=[1-0.1*n:0.1:1+0.1*n];
AddPutDelta=FindPutDelta(assetprice0,AddPutStrike,rate,volatility,maturity);
ExcessDelta(2)=AddPutDelta+ExcessDelta(2);
%Used Maple to solve for a strike price 0.9891375095. MATLAB solving sucks!
K=0.9891375095;
AddPutStrike=[AddPutStrike,K];
AddPutDelta=FindPutDelta(assetprice0,AddPutStrike,rate,volatility,maturity);
ExcessDelta(2)=AddPutDelta+ExcessDelta(1);%Check the net delta is approximately 0
NPut=NPut+length(AddPutStrike);%Update Strike counts and values
strikes=[strikes, AddPutStrike];
%% Calculate theoretical portfolio value
TPV=0;
for i=1:NAsset
    call=zeros(1,NCall+NPut);
    put=zeros(1,NCall+NPut);
    for j=1:NCall+NPut
        [call(j),put(j)]=blsprice(assetprice0(i),strikes(j),rate,maturity,volatility,0);
    end
end

```

```

    TPV=TPV+sum([call(1:NCall),put(NCall+1:NCall+NPut)]);
end
%% Start Monte Carlo runs
assetprice=diag(assetprice0)*exp((rate-volatility^2/2)*maturity+sqrt(maturity)*...
    volatility*randn(size(assetprice)));
%% Calculate option prices from simulation results
for j=1:NRepl
    for i=1:NAsset
        payoff(j)=payoff(j)+sum([max(assetprice(i,j)-strikes(1:NCall),0),...
            max(strikes(NCall+1:NCall+NPut)-assetprice(i,j),0)]);
    end
end
%% Calculate simulated portfolio loss
SPL=TPV-payoff;
%% Calculate VaR from quantile estimation
VaRQE=zeros(1,QEBinN);
QEBinSize=NRepl/QEBinN;
for i=1:QEBinSize
    VaRQE(i)=quantile(SPL(1+(i-1)*QEBinSize:i*QEBinSize),0.01);
end
[mean,sigma,CI]=normfit(VaRQE)

mean =

    -0.7176

sigma =

    1.0302

CI =

    -0.8002
    -0.6350

```

The FindPutDelta function is coded as follows:

```

function delta=FindPutDelta(assetprice0,m,rate,volatility,maturity)
NAsset=length(assetprice0);
APD=zeros(NAsset,1);
for i=1:NAsset
    APD(i)=sum(normcdf((log(assetprice0(i)./m)+(rate+volatility^2/2)*maturity)/...
        (volatility*sqrt(maturity)))-1);
end
delta=sum(APD);

```

▷

Portfolio C: Standard Monte Carlo

This is basically A, just change the first part of the code:

```

%% Initialize portfolio parameters
volatility=0.3;
rate=0.05;

```

```

maturity=0.1;
NCall=10;
NPut=10;
NAsset=100;%<---change here
NRepl=120000;
QEBinN=600; %Bin size used for VaR quantile estimation.
QEIndex=2; %Estimated quantile=1-QEIndex/(NRepl/QEBinSize)=0.99
assetprice0=ones(NAsset,1);
assetprice=zeros(NAsset,NRepl);
strikes=[[0.55:0.1:1.45], [0.55:0.1:1.45]];%<---change here
payoff=zeros(1,NRepl);
.....
...

mean =

    -1.5974

sigma =

    2.2746

CI =

    -1.7798
    -1.4150

```

Summary of simulation:

Portfolio	x	$\hat{\mathbb{P}}^{MC}(L > x)$	$\ 99\% CI\ $	$\hat{\mathbb{P}}^{IS}(L > x)$	$\ 99\% CI\ $	Reduction Factor
A	-0.9222	1%	0.2117	n/a	n/a	n/a
B	-0.7176	1%	0.1652	n/a	n/a	n/a
C	-1.5974	1%	0.3648	n/a	n/a	n/a

□