

# Intro to Financial Engineering IEOR W4700

## Homework 6

Linan Qiu  
lq2137

October 25, 2015

Problem 1.

Problem 1(a).

To find  $q$ ,

```
1 q_prob = function(r, delta_t, sigma) {  
2   u = exp(sigma*sqrt(delta_t))  
3   d = exp(-sigma*sqrt(delta_t))  
4  
5   return((exp(r*delta_t) - d)/(u-d))  
6 }
```

To build the stock tree,

```
1 build_stock_tree = function(S, sigma, delta_t, N) {  
2   tree = matrix(0, nrow=N+1, ncol=N+1)  
3  
4   u = exp(sigma*sqrt(delta_t))  
5   d = exp(-sigma*sqrt(delta_t))  
6  
7   for (i in 1:(N+1)) {  
8     for (j in 1:i) {  
9       tree[i,j] = S * u^(j-1) * d^((i-1)-(j-1))  
10    }  
11  }  
12  return(tree)  
13 }
```

To value the binomial option using the stock tree generated,

```
1 value_binomial_option = function(tree, sigma, delta_t, r, X,
2   type) {
3   q = q_prob(r, delta_t, sigma)
4   option_tree = matrix(0, nrow=nrow(tree), ncol=ncol(tree))
5   if(type == 'put') {
6     option_tree[nrow(option_tree),] = pmax(X - tree[nrow(
7       tree),], 0)
8   } else {
9     option_tree[nrow(option_tree),] = pmax(tree[nrow(tree),]
10      - X, 0)
11   }
12   for (i in (nrow(tree)-1):1) {
13     for(j in 1:i) {
14       option_tree[i, j] = ((1-q)*option_tree[i+1,j] + q*
15         option_tree[i+1,j+1])/exp(r*delta_t)
16     }
17   }
18   return(option_tree)
19 }
```

Putting this all together,

```
1 binomial_option = function(type, sigma, T, r, X, S, N) {
2   q = q_prob(r=r, delta_t=T/N, sigma=sigma)
3   tree = build_stock_tree(S=S, sigma=sigma, delta_t=T/N, N=N
4     )
5   option = value_binomial_option(tree, sigma=sigma, delta_t=
6     T/N, r=r, X=X, type=type)
7   delta = (option[2,2]-option[2,1])/(tree[2,2]-tree[2,1])
8   return(list(q=q, stock=tree, option=option, price=option
9     [1,1], delta=delta))
10 }
```

I coded this manually because none of the R packages (`fOption`, `m4fe`) seem to work (and be able to replicate the numbers on the slides). They also don't show the entire tree. So I coded my own. This code for Binomial European Option Pricing is (I made it open source) at <https://github.com/linanqiu/binomial-european-option-r>.

Using the variables in the question,

```
1 > binomial_option(type='put', sigma=0.33, T=1/4, r=0.05, X
2   =48, S=50, N=3)
```

```

2 $q
3 [1] 0.4980841
4
5 $stock
6      [,1]      [,2]      [,3]      [,4]
7 [1,] 50.00000  0.00000  0.00000  0.00000
8 [2,] 45.45670 54.99739  0.00000  0.00000
9 [3,] 41.32623 50.00000 60.49427  0.00000
10 [4,] 37.57108 45.45670 54.99739 66.54054
11
12 $option
13      [,1]      [,2] [,3] [,4]
14 [1,]  2.247762 0.0000000  0    0
15 [2,]  3.866524 0.6353901  0    0
16 [3,]  6.474186 1.2712151  0    0
17 [4,] 10.428919 2.5433006  0    0
18
19 $price
20 [1] 2.247762
21
22 $delta
23 [1] -0.3386687

```

The option price is 2.247762

Problem 1(b).

Shortcut function to calculate  $\Delta$  from the tree produced by `binomial_option`:

```

1 delta = function(binomial_option, row, col) {
2   stock_tree = binomial_option$stock
3   option_tree = binomial_option$option
4   return((option_tree[row+1, col+1] - option_tree[row+1, col]
5     )/(stock_tree[row+1, col+1] - stock_tree[row+1, col]))
6 }

```

- At start,  $S = 50$ , so  $\Delta = \frac{C_U - C_D}{S_U - S_D} = \frac{0.6353901 - 3.866524}{54.99739 - 45.45670} = -0.3386687$ .  
However, this  $\Delta$  is for buying a put. If we are selling (writing) a put, we use  $-\Delta$  stocks, hence 0.3386687 stocks.

- If stock went up, we are at  $S = 54.99739$  and  $C = 0.6353901$ . Then,  $\Delta = \frac{0 - 1.2712151}{60.49427 - 50.00000} = -0.1211343$ . Again, We use  $-\Delta$  stocks, hence need 0.1211343 stocks.
- Now that stock went down, we are at  $S = 50$ ,  $C = 1.2712151$ . Then  $\Delta = \frac{0 - 2.5433006}{54.99739 - 45.45670} = -0.266574$ . We use  $-\Delta$  stocks, hence need 0.266574 stocks.

## Problem 2.

### Problem 2(a).

```

1 > binomial_option(type='call', sigma=0.33, T=1, r=0.1, X
  =100, S=100, N=6)
2 $q
3 [1] 0.5285562
4
5 $stock
6      [,1]      [,2]      [,3]      [,4]      [,5]
7 [1,] 100.00000  0.00000  0.00000  0.0000  0.0000
   0.0000  0.0000
8 [2,]  87.39589 114.42186  0.00000  0.0000  0.0000
   0.0000  0.0000
9 [3,]  76.38041 100.00000 130.92361  0.0000  0.0000
   0.0000  0.0000
10 [4,]  66.75334  87.39589 114.42186 149.8052  0.0000
   0.0000  0.0000
11 [5,]  58.33968  76.38041 100.00000 130.9236 171.4099
   0.0000  0.0000
12 [6,]  50.98648  66.75334  87.39589 114.4219 149.8052
   196.1304  0.0000
13 [7,]  44.56009  58.33968  76.38041 100.0000 130.9236
   171.4099 224.4161
14
15 $option
16      [,1]      [,2]      [,3]      [,4]      [,5]
17 [1,] 17.275347  0.000000  0.000000  0.00000  0.00000
   0.00000  0.0000
18 [2,]  7.944699 26.147082  0.000000  0.00000  0.00000
   0.00000  0.0000

```

```

19 [3,]  2.257886 13.269646 38.464454  0.00000  0.00000
    0.00000  0.0000
20 [4,]  0.000000  4.343593 21.653136 54.68229  0.00000
    0.00000  0.0000
21 [5,]  0.000000  0.000000  8.355956 34.20200 74.68832
    0.00000  0.0000
22 [6,]  0.000000  0.000000  0.000000 16.07471 51.45809
    97.78328  0.0000
23 [7,]  0.000000  0.000000  0.000000  0.00000 30.92361
    71.40993 124.4161
24
25 $price
26 [1] 17.27535
27
28 $delta
29 [1] 0.6735146

```

Problem 2(b).

```

1 > binomial_option(type='put', sigma=0.33, T=1, r=0.1, X=100,
    S=100, N=6)
2 $q
3 [1] 0.5285562
4
5 $stock
6      [,1]      [,2]      [,3]      [,4]      [,5]
7      [,6]      [,7]
8 [1,] 100.00000  0.00000  0.00000  0.0000  0.0000
    0.0000  0.0000
9 [2,]  87.39589 114.42186  0.00000  0.0000  0.0000
    0.0000  0.0000
10 [3,]  76.38041 100.00000 130.92361  0.0000  0.0000
    0.0000  0.0000
11 [4,]  66.75334  87.39589 114.42186 149.8052  0.0000
    0.0000  0.0000
12 [5,]  58.33968  76.38041 100.00000 130.9236 171.4099
    0.0000  0.0000
13 [6,]  50.98648  66.75334  87.39589 114.4219 149.8052
    196.1304  0.0000
14 [7,]  44.56009  58.33968  76.38041 100.0000 130.9236
    171.4099 224.4161

```

```

15 $option
16      [,1]      [,2]      [,3]      [,4] [,5] [,6]
      [,7]
17 [1,]  7.759088  0.000000  0.000000 0.000000e+00  0  0
      0
18 [2,] 12.553251  3.729666  0.000000 0.000000e+00  0  0
      0
19 [3,] 19.428170  6.820344  1.091538 0.000000e+00  0  0
      0
20 [4,] 28.369599 12.070646  2.354221 1.416430e-15  0  0
      0
21 [5,] 38.381932 20.341195  5.077566 3.054946e-15  0  0
      0
22 [6,] 47.360665 31.593802 10.951256 6.588884e-15  0  0
      0
23 [7,] 55.439912 41.660322 23.619585 1.421085e-14  0  0
      0
24
25 $price
26 [1]  7.759088
27
28 $delta
29 [1] -0.3264854

```

Problem 2(c).

To satisfy put call parity, let  $C_C$  be price of call option and  $C_P$  be price of put option. Then, buying a call and selling a put should give us the same cash flow as a forward on the underlier.

$$C_C - C_P = S - \frac{X}{e^r}$$

$$17.27535 - 7.759088 = 100 - \frac{100}{e^{0.1}}$$

$$9.516258 = 9.516258$$

Problem 3.

Program written in above sections. Code is available at <https://github.com/linanqiu/binomial-european-option-r>.

```

1 periods = seq(100, 120)
2 option_price_vary_period = function(period) {
3   print(period)
4   option = binomial_option(type='call', sigma=0.2, T=1, r=0,
5     X=100, S=100, N=period)
6   return(option$price)
7 }
8 values = sapply(periods, option_price_vary_period)
9 library(ggplot2)
10 data = as.data.frame(list(periods=periods, values=values))
11 plot = ggplot(data=data) + geom_line(aes(x=periods, y=values
12   )) + labs(title="Call Value", x="Periods", y="Value")
13 plot

```

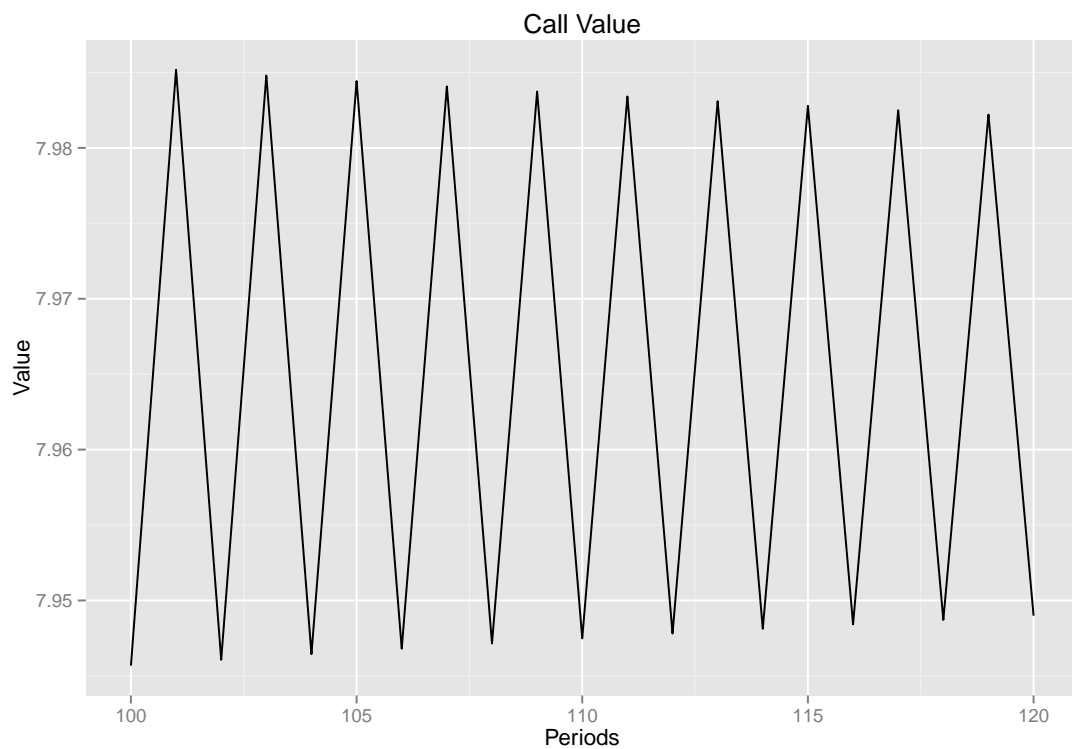


Figure 1: Plot of call option value calculated using the function above against periods (from 100 to 120)

Why do 100 to 120 when one can do more!

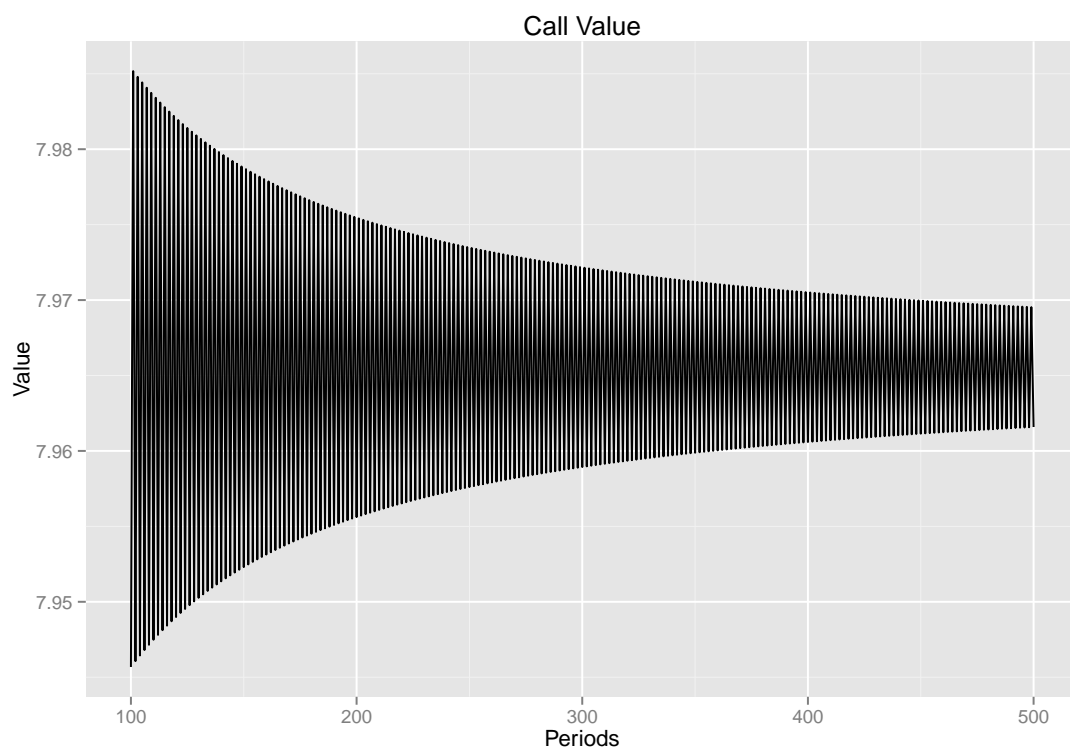


Figure 2: Plot of call option value calculated using the function above against periods (from 100 to 500). Also this is pretty pretty. After 500 the code becomes a little slow since I'm generating  $N^2$  matrices.

Okay I managed to speed it up by parallelizing it. Let's try a 1000 periods now (this took a minute on 8 CPU cores). The code is

```
1 library(parallel)
2 cl = makeCluster(8)
3 clusterEvalQ(cl, source('binomial.R'))
4 periods = seq(100, 1000)
5 periods = sample(periods)
6 valuesPar = parSapply(cl=cl, periods, option_price_vary_
  period)
7 data = as.data.frame(list(periods=periods, values=valuesPar)
  )
8 plot = ggplot(data=data) + geom_line(aes(x=periods, y=values
  )) + labs(title="Call Value", x="Periods", y="Value")
9 plot
10 stopCluster(cl)
```



6

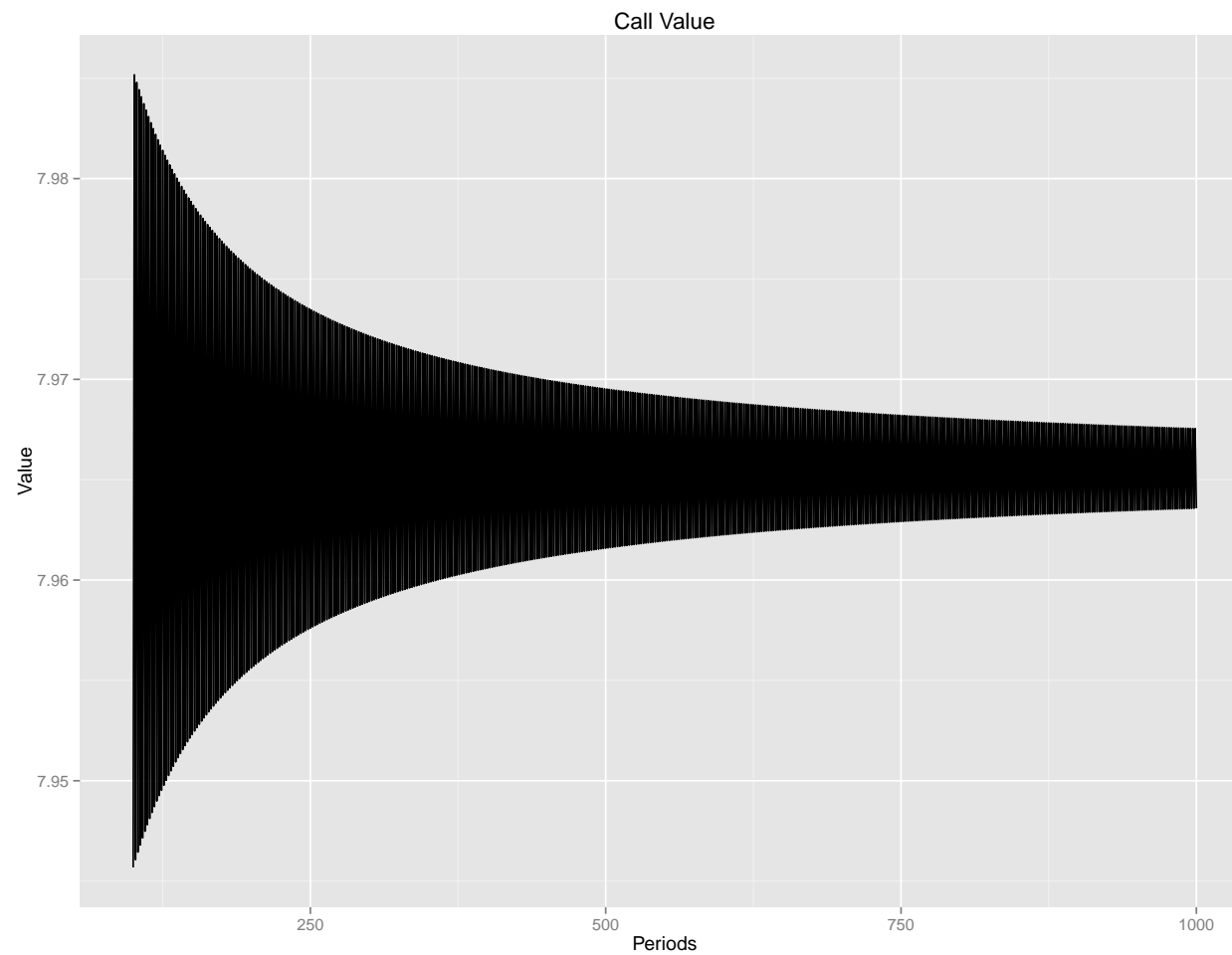


Figure 3: Plot of call option value calculated using the function above against periods (from 100 to 1000). This is almost beautiful.

#### Problem 4.

Build the tree manually. Let's find  $q$  the risk neutral "probability" first.

$$S = 50 = \frac{qS_U + (1 - q)S_D}{R} = \frac{q55 + (1 - q)45}{e^{0.1*0.5}}$$

Solving for  $q$ ,  $q = 0.7563555$

```
1 > uniroot(function(q) {(q*55 + (1-q)*45)/exp(0.1*0.5) - 50},
2   interval=c(-1, 1))
3 $root
4 [1] 0.7563555
5 $f.root
6 [1] 0
7
8 $iter
9 [1] 1
10
11 $init.it
12 [1] NA
13
14 $estim.prec
15 [1] 1.756355
```

Then,

$$P = \frac{qP_U + (1 - q)P_D}{R} = \frac{0 + (1 - 0.7563555)(50 - 45)}{e^{0.1*0.5}} = 1.158809$$

To verify this (since I'm revising for the midterm anyway), let's replicate the riskless bond. Consider a portfolio with  $\Delta$  stocks and 1 put.

- When  $S_U = 55$ ,  $P_U = 0$ . Portfolio is worth  $\Delta 55$ .
- When  $S_D = 45$ ,  $P_D = 5$ . Portfolio is worth  $\Delta 45 + 5$ .

$$\Delta 55 = \Delta 45 + 5$$

Then,  $\Delta = 0.5$ , ie. we must long 0.5 stocks. The value of both portfolios are

- When  $S_U = 55$ ,  $P_U = 0$ . Portfolio is worth  $\Delta 55 = 27.5$ .
- When  $S_D = 45$ ,  $P_D = 5$ . Portfolio is worth  $\Delta 45 + 5 = 27.5$ .

That means the portfolio must be worth  $\frac{27.5}{e^{0.1 \cdot 0.5}} = 26.15881$  presently. That means

$$P + \Delta S = 26.15881 = P + 0.5(50)$$

$$P = 1.158809$$

The value of the put option is 1.158809 which verifies the answer from using binomial trees.

## Problem 5.

Let  $D$  be the value of the derivative.

$$S = 50 = \frac{qS_U + (1 - q)S_D}{R} = \frac{q27 + (1 - q)23}{e^{0.1/6}}$$

Solving for  $q$ ,  $q = 0.6050396$

```

1 > uniroot(function(q) {(q*27 + (1-q)*23)/exp(0.1/6) - 25},
2   interval=c(-1, 1))
3 $root
4 [1] 0.6050396
5 $f.root
6 [1] 0
7
8 $iter
9 [1] 1
10
11 $init.it
12 [1] NA
13
14 $estim.prec
15 [1] 1.60504

```

Then,

$$D = \frac{qD_U + (1 - q)D_D}{R} = \frac{(0.6050396)27^2 + (1 - 0.6050396)23^2}{e^{0.1/6}} = 639.2642$$

Can be verified via portfolio replication method.

Suppose portfolio comprises  $\Delta$  stocks and 1  $D$ .

- When  $S_U = 27$ ,  $D_U = 27^2$ . Portfolio is worth  $\Delta 27 + D_U$
- When  $S_D = 23$ ,  $D_D = 23^2$ . Portfolio is worth  $\Delta 23 + D_D$

$$\Delta 27 + D_U = \Delta 23 + D_D$$

$$\Delta 27 + 27^2 = \Delta 23 + 23^2$$

$$\Delta = -50$$

We short 50 stocks. Then in both states, portfolio is worth  $\Delta 27 + 27^2 = -621 = \Delta 23 + 23^2$

Then the value of both portfolios before two months is  $\frac{\Delta 27 + 27^2}{e^{0.1/6}} = -610.7358$ .

$$D + \Delta S = D - 50(25) = -610.7358$$

$$D = 639.2642$$

Verifies the answer above.