

**IE523: Financial Computing**  
**Fall, 2012**  
**Programming Assignment 8: Repeated Squaring**  
**Algorithm**  
**Due Date: 10 November, 2017**  
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Later on in the course, when we deal with the problem of computing the price of an *European Option* efficiently using the method of *Dynamic Programming*, we will have to deal with the problem of taking an  $n \times n$  matrix  $\mathbf{A}$  and raising it to a large number. That is, we will have to compute  $\mathbf{A}^k$  for large values of  $k$ .

You could compute it by multiplying  $\mathbf{A}$  over with it self  $k$  times. That is, (with *NEWMAT*) something along the lines of

```
1: C = A;  
2: for int i = 1; i < k; i++ do  
3:   C = A*C;  
4: end for
```

Assuming you are doing straightforward matrix multiplication (nothing clever like Strassen's method, etc.) then the above procedure will take  $O(n^3k)$  steps, as each matrix multiplication is  $O(n^3)$  and there are  $k$ -many of them.

A candidate algorithm that is more efficient than the one shown above goes by the name of *Repeated Squaring*, and it described below.

### Repeated Squaring Algorithm

Suppose you want to compute  $\mathbf{A}^{11}$ , you write the exponent 11 in binary – which is  $\langle 1\ 0\ 1\ 1 \rangle_2$ . That is,  $11 = 1 \times 2^3 + 0 \times 2^2 + 1 \times 2^1 + 1 \times 2^0$ . You then compute all  $\lceil \log_2 11 \rceil$ -many powers of  $\mathbf{A}$ . That is, you compute  $\mathbf{A}, \mathbf{A}^2, \mathbf{A}^4, \mathbf{A}^8$ , then add the appropriate components as per the binary expansion of the exponent. In our case, we add  $\mathbf{A}^8 \times \mathbf{A}^2 \times \mathbf{A}$  to get the final product. It turns out that these constituent powers of  $\mathbf{A}$  (i.e. 8, 2 and 1) can be computed in a recursive manner quite efficiently. Here is something from the following [link](#) that computes  $n^k$  for integers  $n$  and  $k$ .

```
1: int power( int n, int k )  
2: if k == 0 then  
3:   return 1  
4: end if  
5: if k is odd then  
6:   return ( n * power ( n * n, (k-1)/2 ) )  
7: else  
8:   return ( power( n * n, k/2 ) )  
9: end if
```

A similar approach can be used to compute  $\mathbf{A}^k$  for a square matrix  $\mathbf{A}$ . I leave the nitty-gritty details for you to figure out. Keep in mind that with *NEWMAT* on your side, the matrix operations are structurally the same as that for scalars.

## Complexity of Repeated Squaring vs. Brute Force Multiplication

If you have to compute  $\mathbf{A}^k$ , and assuming each multiplication is  $O(b^3)$  (nothing clever here, straightforward matrix multiplication), and since we have  $\log_2 k$ -many of these to do (or, since  $\log_2 k = \frac{\log_{10} k}{\log_{10} 2}$  we can replace  $\log_2 k$  with  $O(\log k)$ ), this entire operation takes  $O(b^3 \log k)$ . This is in contrast to the  $O(b^3 k)$  procedure for multiplying  $\mathbf{A}$  with itself  $k$  times. Resulting in a total complexity of  $O(b^3 \ln k)$ . If we used *Strassen's* method for matrix multiplication we would have a procedure that is  $O(b^{2.81} \ln k)$ .

## The Programming Assignment

1. (Using *NEWMAT*) I want you to write a recursion routine

`Matrix repeated_squaring(Matrix A, int exponent, int no_rows)`

which takes a  $(\text{no\_rows} \times \text{no\_rows})$  matrix  $\mathbf{A}$  and computes  $\mathbf{A}^{\text{exponent}}$  using the *Repeated Squaring Algorithm*.

2. Your code should be able to take as input the size and exponent as input on the command line. That is, if we want to compute  $\mathbf{A}^k$ , where  $\mathbf{A}$  is an  $(n \times n)$  square matrix, I want to be able to read  $n$  and  $k$  on the command-line. It should fill the entries of the matrix  $\mathbf{A}$  with random entries in the interval  $(-5, 5)$ <sup>1</sup>.
3. The output should indicate: (1) The number of rows/columns in  $\mathbf{A}$  (that is read from the command line), (2) The exponent  $k$  (that is read from the command line), (3) The result and the amount of time it took to compute  $\mathbf{A}^k$  using repeated squaring, and (4) The result and the amount of time it took to compute  $\mathbf{A}^k$  using brute force multiplication. A sample output is shown in figure 1.
4. I want you to provide a plot of the computation time (in seconds) for the two methods as a function of the size of the matrix. That is, I am looking for something along the lines of figure 2. For this, you will have to place timer objects before and after appropriate portions of your code and do the needful – as the following lines of code illustrate.

```
time_before = clock();
B = repeated_squaring(A, exponent, dimension);
time_after = clock();
diff = ((float) time_after - (float) time_before);
cout << "It took " << diff/CLOCKS_PER_SEC << " seconds to
complete" << endl;
```

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<sup>1</sup>See `strassen.cpp` for ideas.

In terms of the value of the exponent, tell me the regions where one algorithm performs better than the other, as far as computation time is concerned.

```

MacBook-Air:Debug sreenivas$ ./Repeated\ Squaring 1000 10
The number of rows/columns in the square matrix is: 10
The exponent is: 1000
Repeated Squaring Result:
It took 0.000105 seconds to complete
Direct Multiplication Result:
It took 0.002044 seconds to complete

MacBook-Air:Debug sreenivas$ ./Repeated\ Squaring 1000 100
The number of rows/columns in the square matrix is: 100
The exponent is: 1000
Repeated Squaring Result:
It took 0.007249 seconds to complete
Direct Multiplication Result:
It took 0.436108 seconds to complete

MacBook-Air:Debug sreenivas$ ./Repeated\ Squaring 1000 1000
The number of rows/columns in the square matrix is: 1000
The exponent is: 1000
Repeated Squaring Result:
It took 11.669 seconds to complete
Direct Multiplication Result:
It took 725.793 seconds to complete

MacBook-Air:Debug sreenivas$

```

Figure 1: A sample output for different exponents and matrix-dimensions.

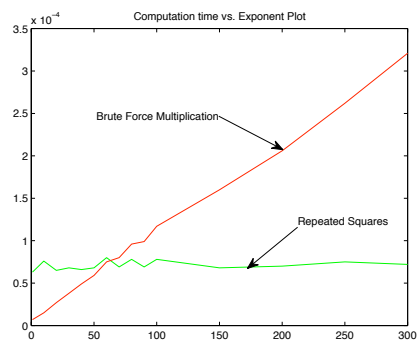


Figure 2: A comparison of the computation-time (obtained experimentally) for brute-force exponentiation and the method of repeated squares of a random  $5 \times 5$  matrix as a function of the exponent.