CS124 Programming Assignment 1

Alexander Lin, Jessica Wang March 25, 2016

Analytical Analysis

We found the recursive equation of Strassens to be $T(n) = 7T(\left\lceil \frac{n}{2}\right\rceil) + 18(\left\lceil \frac{n}{2}\right\rceil)^2$. This is because if we are given matrices of size nxn, Strassen's Algorithm will recurse by doing 7 multiplications on $\left\lceil \frac{n}{2}\right\rceil$ size matrices which gives us the recursive element of the equation. Strassen's will additionally do 18 additions of $\left\lceil \frac{n}{2}\right\rceil$ matrices which will cost $18*(\left\lceil \frac{n}{2}\right\rceil))^2$. Additionally we know that the conventional matrix mulitplication for matrices of size nxn will cost $n^2(2n-1)$ because for each of the n^2 elements it will cost 2n-1 to calculate the value.

To find the analytical cross over point we used python by iterating through cross over points (n_0) for various dimensions of matrices (n). The following pseudo code describes our algorithm.

```
function STRASSEN(n,n_0)

if n \leq n_0 then

n^2(2n-1)

else if n is even then

7*strassen(\frac{n}{2},n_0)+18*(\frac{n}{2})^2

else

7*strassen(\frac{n}{2}+1,n_0)+18*(\frac{n}{2}+1)^2

end if

end function

for n_0 \leq n do

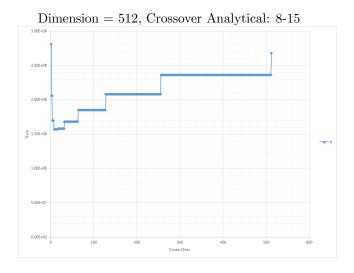
Return strassen(n_0,n)

end for
```

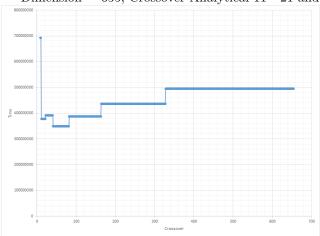
We ran this pseudocode on various values of n. When looking at the values that are returned we want to find the value of n_0 that returns the lowest cost. If there are multiple values of n_0 that all return the same minimum cost we can represent the crossover point as a range.

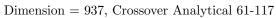
Analytical Results

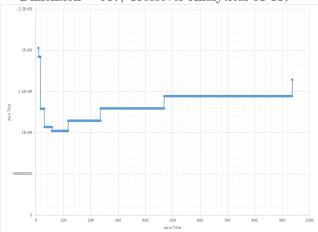
The graphs of the crossover numbers can be found below. Exact numbers that are displayed can be found within out code.

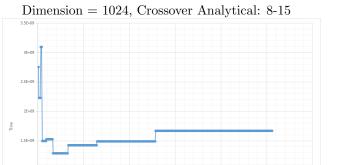




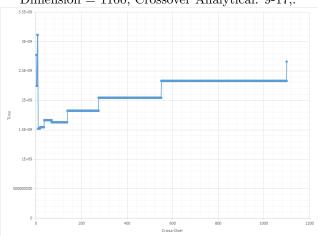












Analysis of Results

2^n Values

For dimensions of 2^n we find the crossover point to be 16. This is consistent through all the values of n that were powers of 2 that we tested.

Not 2^n Values

For dimensions that are not 2^n (particularly dimensions that are odd) there is no clear crossover point that covers all cases. This is most likely because these cases will involve padding the matrix with 0s at some point in the process which will increase the cost of calculating the multiplication. The amount the cost increases would vary with different dimension values. For even dimensions, the crossover point hovers around 16 possibly slightly higher. For odd dimensions we found it to be pretty consistent for n_0 to hover above 39.

Experimental Analysis

We implemented Strassen's Algorithm in C. We stored our matrices as an array of arrays. We decided to do this because it was the most intuitive to us and easier to index into correctly. The array of arrays take

an extra step to allocate memory for and thus a little more time, but the memory space will be the same. To do this we created the helper function addMatrix(matA, ai, aj, matB, bi, bj, dim, sub, returnmat) which returned returnmat which added the dimxdim size of matA starting at matA[ai][aj] and the dimxdim size of matB starting at matB[bi][bj] if sub = 1 and subtracted the matrices if sub = -1. Helper function copyMatrix(mat, i, j, dim) which will copy the dimxdim size of mat starting at mat[i][j] into returnmat. The last helper function combineMatrix(matA, matB, matC, matD, dim) will combine four dimxdim matrices such that the returnmat will be

 $\left(\begin{array}{cc} matA & matB \\ matC & matD \end{array}\right)$

The first step within our recursive Strassen function is to determine whether the dimension of the current matrices attempting to be multiplied is less than or equal to n_0 . If it is, then we have passed the cross over point and Strassen will revert to conventional multiplication. Otherwise Strassens will recurse and calculate the multiplication using the 7 multiplications of Strassen's algorithm. Strassen's algorithm is calculated by first calculating the both matrices of each multiplication, multiplying those seven pairs together then add the 7 multiplications to get the final matrix.

```
Our code can be described with the following pseudocode
function ADDMATRIX(matA, ai, aj, matB, bi, bj, dim, sub, returnmat)
   Described above
end function
function COPYMATRIX(mat, i, j, dim, returnmat)
   Described above
end function
function COMBINEMATRIX (matA, matB, matC, matD, dim, returnmat)
   Described above
end function
function STRASSEN(matA, matB, dim, n_0, returnmat)
   if dim \leq n_0 then
      Return matA * matB using conventional multiplication
   end if
   addMatrix(matA, 0, 0, matA, dim/2, dim/2, dim/2, 1, m1a);
   addMatrix(matB, 0, 0, matB, dim/2, dim/2, dim/2, 1, m1b);
   strassen(m1a, m1b, dim/2, n0, m1);
   addMatrix(matA, dim/2, 0, matA, dim/2, dim/2, dim/2, 1, m2a);
   copyMatrix(matB, 0, 0, dim/2, m2b);
   strassen(m2a, m2b, dim/2, n0, m2);
   copyMatrix(matA, 0, 0, dim/2, m3a);
   addMatrix(matB, 0, dim/2, matB, dim/2, dim/2, dim/2, -1, m3b);
   strassen(m3a, m3b, dim/2, n0, m3);
   copyMatrix(matA, dim/2, dim/2, dim/2, m4a);
   addMatrix(matB, dim/2, 0, matB, 0, 0, dim/2, -1, m4b);
   strassen(m4a, m4b, dim/2, n0, m4);
   addMatrix(matA, 0, 0, matA, 0, dim/2, dim/2, 1, m5a);
   copyMatrix(matB, dim/2, dim/2, dim/2, m5b);
   strassen(m5a, m5b, dim/2, n0, m5);
   addMatrix(matA, dim/2, 0, matA, 0, 0, dim/2, -1, m6a);
   addMatrix(matB, 0, 0, matB, 0, dim/2, dim/2, 1, m6b);
   strassen(m6a, m6b, dim/2, n0, m6);
```

```
addMatrix(matA,0,dim/2,matA,dim/2,dim/2,dim/2,-1,m7a);\\ addMatrix(matB,dim/2,0,matB,dim/2,dim/2,dim/2,1,m7b);\\ strassen(m7a,m7b,dim/2,n0,m7);\\ addMatrix(m1,0,0,m4,0,0,dim/2,1,c00);\\ addMatrix(c00,0,0,m5,0,0,dim/2,-1,c00);\\ addMatrix(c00,0,0,m7,0,0,dim/2,1,c00);\\ addMatrix(m3,0,0,m5,0,0,dim/2,1,c01);\\ addMatrix(m2,0,0,m4,0,0,dim/2,1,c10);\\ addMatrix(m1,0,0,m2,0,0,dim/2,1,c11);\\ addMatrix(c11,0,0,m3,0,0,dim/2,1,c11);\\ addMatrix(c11,0,0,m6,0,0,dim/2,1,c11);\\ combineMatrix(c00,c01,c10,c11,dim/2,returnmat);\\ end function
```

Allocating Space

The biggest difficulty we faced was space allocation. We originally had copied each the pair of matrices we wanted to multiply into a_{00} , a_{01} , a_{10} , a_{11} , b_{00} , b_{01} , b_{10} , b_{11} such that matA =

$$\begin{pmatrix} a_{00} & a_{01} \\ a_{10} & a_{11} \end{pmatrix}$$
 and $matB=\begin{pmatrix} b_{00} & b_{01} \\ b_{10} & b_{11} \end{pmatrix}$

This process took a lot of time to allocate and deallocate. To avoid allocating this we changed our add function, which originally simply added two matrices of equal size, to take in indices value. This allows you to dictate $a_{00}, a_{01}, a_{10}, a_{11}, b_{00}, b_{01}, b_{10}, b_{11}$ when you add them by specifying the start indices of each into the addition function instead of allocating space to store them.

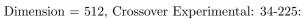
We additionally tried to change our code so we could allocate space for m1a, m1b, m2a, m2b, ..., m7a, m7b at the beginning to avoid allocating space each time Strassen's occurs by implementing the same indices technique into the strassen function. We were able to make it work for 2x2 matrices, but ran into override issues for larger matrices as Strassen's tried to recurse.

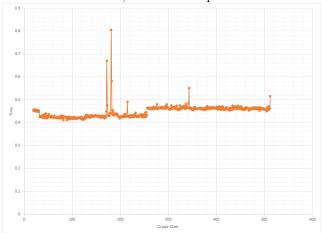
Dimensions not 2^n

To deal with odd dimensions we decided to alter the original matrices that would initialized the strassen function to be of dimension 2^n (the closest power of 2 to the initial dimension) with 0s filling the extra space. We decided to do this before calling the strassen function because it allows us to not check the dimension of the matrices within the strassen function.

Experimental Results

We tested experimental results using an external BASH script to time our Strassen implementation. For values that were feasible, we tested n_0 with $\Delta = 1$, on larger values we changed Δ to be some larger value to maximize time efficiency. Attached to our pset we have included a pdf that includes all our experimental datapoints; and below are side by side comparisons of the analytical calculations with experimental calculations.

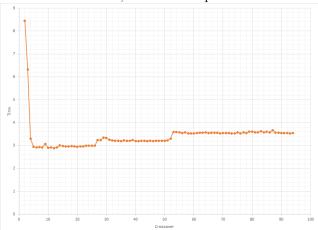


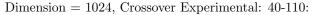


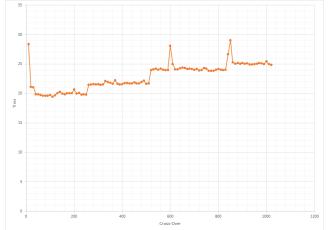
$\label{eq:Dimension} \mbox{Dimension} = 655, \mbox{Crossover Experimental } 55 \mbox{ - } 255$



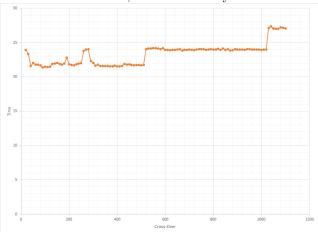
Dimension = 937, Crossover Experimental: 40-250:







Dimension = 1100, Crossover Analytical: 90-120:



2^n Values

The analytical results indicate that the crossover point should be around 16. The experimental results indicate that this value should be higher, which indicates that Strassen takes a longer time in comparison to the conventional multiplication. This is most likely because the implementation to Strassen not only involves calculating values but also allocating and deallocating space. This takes a good deal of time, which would slow down Strassen's Algorithm in comparison to the conventional algorithm, which does not allocate or deallocate space. Additionally as values of 2^n gets larger, the crossover point increases because there will be more memory being allocated and deallocated as 2^n grows. Experimental results were harder to define a threshold for n_0 since memory allocation shifted results quite a bit; what is certain is that experimental n_0 analytical n_0 , however from our observations of dimensions 512 and 1024, it seems $34 < n_0 < 250$ with our best estimations.

Not 2^n Values

The same increase the crossover point in comparison to the analytical result as described for the 2^n values is shown in not 2^n values. Similar to the 2^n values, this has to do with memory allocation and deallocation. The best estimate we can provide given the data that is present is that for not 2^n values, $55 < n_0 < 255$