

# Matrix product and Power Algorithms

CSCI 694 Presentation

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Theory

# **GENERAL MATRIX-MATRIX PRODUCT**

# Theory

- The product of two matrices  $\mathbf{A} \in \mathbf{R}^{m \times n}$  and  $\mathbf{B} \in \mathbf{R}^{p \times q}$  is possible, if and only if,  $n = p$ .
- The number of columns of first matrix must be equal to number of rows of second matrix.
- This is called conformability.

Algorithm

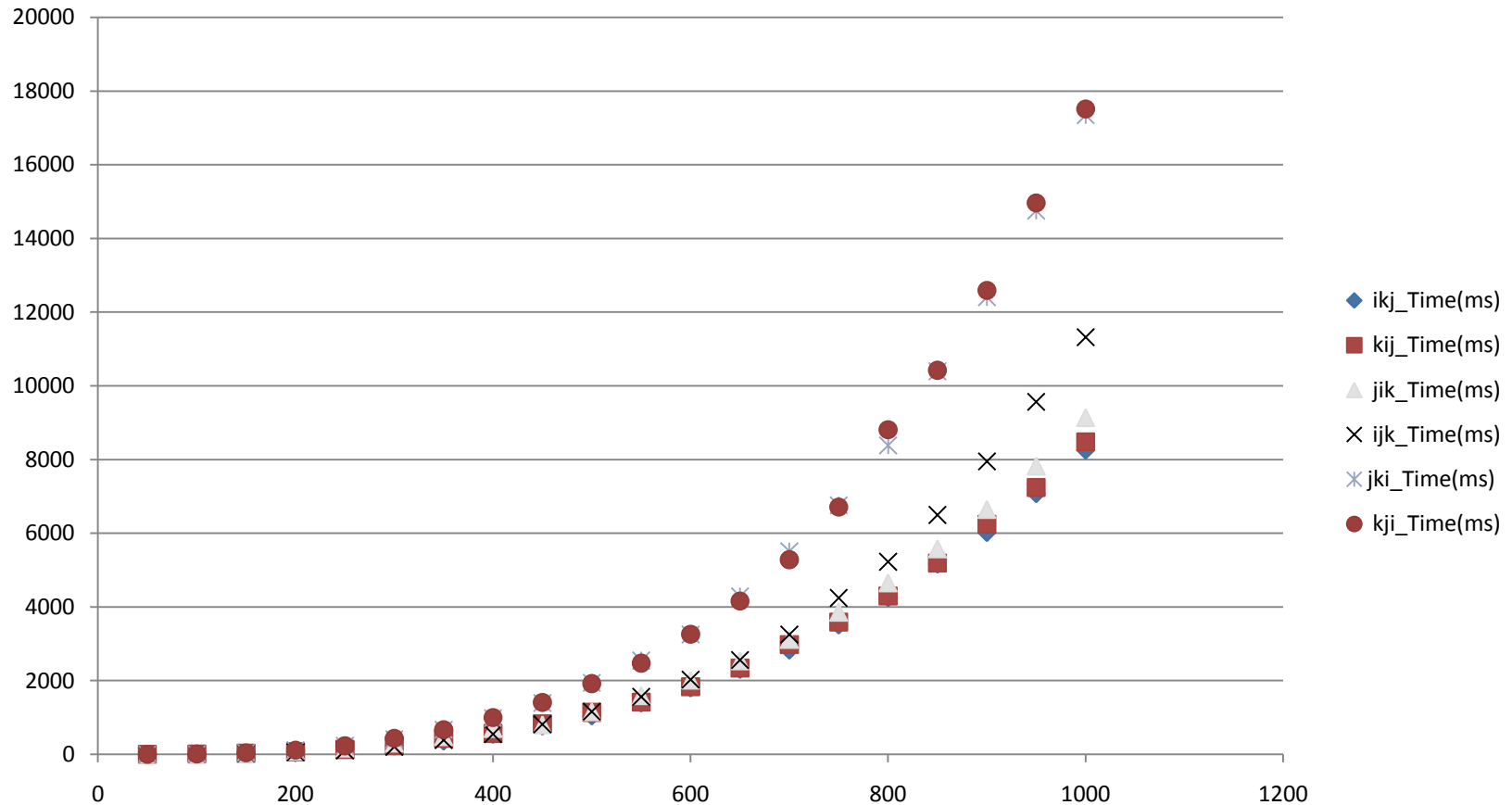
# **GENERAL MATRIX-MATRIX PRODUCT**

# (T – method) Algorithm

```
function: C = matmat.ijk(A,B)  
m = rows(A); r = cols(A); n = cols(B)  
C(1:m, 1:n) = 0  
for i = 1:m  
  for j = 1:n  
    for k = 1:r  
      C(i,j) = C(i,j) + A(i,k)B(k,j)  
    end  
  end  
end  
end matmat.ijk
```

Loop Order	Inner Loop	Middle Loop	Inner Loop Data Access
matmat.ijk()	Dot	Vector * Matrix	A by row, B by column
matmat.ikj()	Dot	Matrix * vector	A by row, B by column
matmat.jik()	saxpy	Row gaxpy	B by row
matmat.jki()	saxpy	Column gaxpy	A by column
matmat.kij()	saxpy	Row outer product	B by row
matmat.kji()	saxpy	Column outer product	A by column

# Performance



Processor :Intel® Core(TM)2 Duo p8600 @ 2.40 GHz 2.40 GHz

Compiler: Microsoft® Visual C++ professional (student edition)

Language: ANSI C

Theory

# **BLAS - 3**



# ?gemm() BLAS

- *Stands for GEneral Mat-Mat multiplication.*
- *Computes a scalar-matrix-matrix product and adds the result to a scalar-matrix product.*
- $c := \alpha * \text{op}(a) * \text{op}(b) + \beta * c,$
- ? Could be s, d, c and z.

Notation	Precision
s	Single
d	Double
c	Complex Single
z	Complex Double

Theory

# **3M METHOD**

why complex mat-mat product is complex ?

- Two complex numbers multiplication costs 4 multiplication and 2 additions.
- Product of two square complex matrices of order  $N$  results in  $2*N^2$  multiplications, on contrast to  $2*N$  operations.
- Solution: We can trade multiplications with additions !

# 3M Method

- Technique to multiply two complex scalars using three multiplication and five real addition.
- The Number of operation drops to  $\frac{3}{4}$  of the usual operation
- Use it when :  
time (3 \* SAXPY) < time (1\*SGEMM)  
time (3 \* DAXPY) < time (1\*DGEMM)

Algorithm

# **3M METHOD**

# Algorithm

*Function:  $Cc = 3M(Ac, Bc)$*

*//precondition:  $Ac = (Ar, Ai), Bc = (Br, Bi)$*

*//post condition:  $Cc = (Cr, Ci)$*

*{*

*$S1 = Br - Bi$*

*$S2 = Ar + Ai$*

*$S3 = Ar - Ai$*

*$R1 = Ar * S1$*

*$R2 = Br * S2$*

*$R3 = Bi * S3$*

*$(Cr, Ci) = (Cr + R1 + R3, Ci + R2 - R1)$*

*}*

Implementation

# **3M METHOD**

C:\Windows\system32\cmd.exe

\*\*\*\*\* MATRIX A \*\*\*\*\*

41.000000 + i18467.000000	19169.000000 + i15724.000000	26962.000000 + i24464.000000	23281.000000 + i16827.000000
2995.000000 + i11942.000000	32391.000000 + i14604.000000	292.000000 + i12382.000000	19718.000000 + i19895.000000
14771.000000 + i11538.000000	25667.000000 + i26299.000000	28703.000000 + i23811.000000	17673.000000 + i4664.000000
28253.000000 + i6868.000000	32662.000000 + i32757.000000	8723.000000 + i9741.000000	12316.000000 + i3035.000000

\*\*\*\*\* MATRIX B \*\*\*\*\*

6334.000000 + i26500.000000	11478.000000 + i29358.000000	5705.000000 + i28145.000000	9961.000000 + i491.000000
4827.000000 + i5436.000000	3902.000000 + i153.000000	17421.000000 + i18716.000000	5447.000000 + i21726.000000
1869.000000 + i19912.000000	17035.000000 + i9894.000000	31322.000000 + i30333.000000	15141.000000 + i7711.000000
25547.000000 + i27644.000000	20037.000000 + i12859.000000	27529.000000 + i778.000000	22190.000000 + i1842.000000

\*\*\*\*\* MATRIX C \*\*\*\*\*

1265405169.000000 + i2397814600.000000	591507298.000000 + i2463619696.000000	1066078203.000000 + i3711232267.000000
1117354530.000000 + i3481785859.000000	1066078203.000000 + i3711232267.000000	1265405169.000000 + i2397814600.000000
1265405169.000000 + i2397814600.000000	1117354530.000000 + i3481785859.000000	1117354530.000000 + i3481785859.000000
1117354530.000000 + i3481785859.000000	1265405169.000000 + i2397814600.000000	1265405169.000000 + i2397814600.000000
1265405169.000000 + i2397814600.000000		

Elapsed time is: 0.000000

Press any key to continue . . .



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Gmail - Compose M...



C:\Windows\system...



Theory

# **STRASSEN'S ALGORITHM**

# Theory

- Strassen (1969) devised a algorithm which computes the coefficients of the product of the matrix  $A \in \mathbb{R}^{n \times n}$  and  $B \in \mathbb{R}^{n \times n}$  in less than  $4.7 (n^{\log_2 7}) = n^{2.807}$
- Consider a 2-by-2 matrix multiplication
- T-method requires 8 multiplications and 4 addition
- S – method requires 7 multiplications and 18 additions

Algorithm

# **STRASSEN'S ALGORITHM**

```

function: C = Strassen.ijk(A,B)
//precondition: Matrices A and B of order m*2 k & k = blksize
n = row(A)
for i = 1:n
    for j = 1:n
        S11[i][j] = A11[i][j] + A22[i][j]
        S12[i][j] = B11[i][j] + B22[i][j]
        S21[i][j] = A21[i][j] + B22[i][j]
        S32[i][j] = B12[i][j] - B22[i][j]
        S42[i][j] = B21[i][j] - B11[i][j]
        S51[i][j] = A11[i][j] + A12[i][j]
        S61[i][j] = A21[i][j] - A11[i][j]
        S62[i][j] = B11[i][j] + B12[i][j]
        S71[i][j] = A12[i][j] - A22[i][j]
        S72[i][j] = B21[i][j] + B22[i][j]

```



*for i = 1:n*  
*for j = 1:n*

*for k = 1:n*

$$S_1[i][j] = S_1[i][j] + S_{11}[i][k] * S_{12}[k][j]$$

$$S_2[i][j] = S_2[i][j] + S_{21}[i][k] * B_{11}[k][j]$$

$$S_3[i][j] = S_3[i][j] + A_{11}[i][k] * S_{32}[k][j]$$

$$S_4[i][j] = S_4[i][j] + A_{22}[i][k] * S_{42}[k][j]$$

$$S_5[i][j] = S_5[i][j] + S_{51}[i][k] * B_{22}[k][j]$$

$$S_6[i][j] = S_6[i][j] + S_{61}[i][k] * S_{62}[k][j]$$

$$S_7[i][j] = S_7[i][j] + S_{71}[i][k] * S_{72}[k][j]$$

*for i = 1:n*

*for j = 1:n*

$$C_{11}[i][j] = S_1[i][j] + S_4[i][j] - S_5[i][j] + S_7[i][j]$$

$$C_{12}[i][j] = S_3[i][j] + S_5[i][j]$$

$$C_{21}[i][j] = S_2[i][j] + S_4[i][j]$$

$$C_{22}[i][j] = S_1[i][j] + S_3[i][j] - S_2[i][j] + S_6[i][j]$$

*end Strassen.ijk(A,B)*

Implementation & Results

# **STRASSEN'S ALGORITHM**

```
C:\Windows\system32\cmd.exe

**** MATRIA A ****
  41      6334      19169      11478
26962     5705      23281      9961
2995      4827      32391      3902
292       17421     19718      5447

**** MATRIA B ****
 18467     26500     15724      29358
24464     28145     16827      491
11942      5436     14604      153
12382      18716     19895      21726
```

## Matrix – A and Matrix - B

4 x 4 matrices



```
**** A11 ****
41      6334
26962   5705
```

```
**** B11 ****
18467   26500
24464   28145
```

```
**** A12 ****
19169   11478
23281   9961
```

```
**** B12 ****
15724   29358
16827   491
```

```
**** A21 ****  
2995      4827  
292       17421
```

```
**** B21 ****  
11942     5436  
12382     18716
```

```
**** A22 ****  
32391     3902  
19718     5447
```

```
**** B22 ****  
14604     153  
19895     21726
```

```

**** S11 ****
32432  10236
46680  11152

**** S12 ****
33071  26653
44359  49871
Likewise calculate S21, S32...S71, S72

**** S1 ****
1526617396      1374889652
2038445848      1800323432
Likewise calculate S2, S3...S6, S7

**** C11 ****
526748917      498381862
1038833178     1188045817

**** C12 ****
615525788      256617557
1058118342     1014326230

**** C21 ****
608524279      464330723
734496818      707185145

**** C22 ****
678985763      180027942
694064312      138484623
Total time taken for 4 * 4 ordered matrix multiplication [Strassen's method] is 0.053000 seconds

```

Theory

# **WINOGRAD'S ALGORITHM**

# Theory

- Modification and improvement of S – Method
- Again, consider a 2-by-2 matrix multiplication
- S – method requires 7 multiplications and 18 additions
- W – method required 7 multiplications but only 15 additions.
- The complexity is improved to  $O(n^{2.795})$

Algorithm

# **WINOGRAD'S ALGORITHM**

# Algorithm

- $S1 = (A21 + A22),$
- $S2 = S1 - A11,$
- $S3 = A11 - A21,$
- $S4 = A12 - S2,$
- $S5 = B12 - B11,$
- $S6 = B22 - S5,$
- $S7 = B22 - B12,$
- $S8 = S6 - B21,$

# Algorithm cont.,

- $M1 = S2S6,$
- $M2 = X11Y11,$
- $M3 = X12Y21,$
- $M4 = S3S7,$
- $M5 = S1S5,$
- $M6 = S4Y22,$
- $M7 = X22S8,$



# Algorithm cont.,

- $T1 = M1 + M2,$
- $T2 = T1 + M4,$
- $Z11 = M2 + M3,$
- $Z12 = T1 + M5 + M6,$
- $Z21 = T2 - M7,$
- $Z22 = T2 + M5$

Theory

# **COPPERSMITH–WINOGRAD ALGORITHM**

# Theory

- Asymptotically fastest known algorithm for square matrix multiplication as of today !
- Asymptotic complexity is ( $O^{2.376}$ )
- “Not practical, required so large matrices that the modern hardware can’t handle” Robinson 2005

# Evaluation of Powers

Theory

# **BINARY METHOD**

# Theory

- How to calculate  $x^n$  ?
  1. write  $n$  in binary number system
  2. Replace 1 by SX and 0 by S
  3. Cross off left most SX
  4. S – squaring and X – multiply by  $x$
  5.  $n = 23 \Rightarrow 10111 \Rightarrow SXSSXSXSX \Rightarrow SSXSXSX$

# Theory

- Known before 200 BC
- Appeared in Pingala's Hindu Classic *chandhasutra*
- Required no temporary storage
- Problem : scanning from left to right
- Algorithm A is devised to scan from right – left
- Al-kashi stated algorithm A in A.D. 1427
- Binary method is not optimal

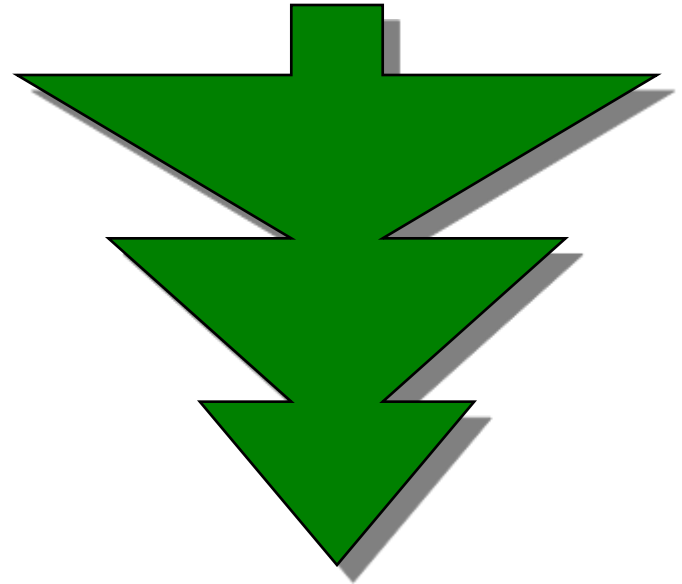
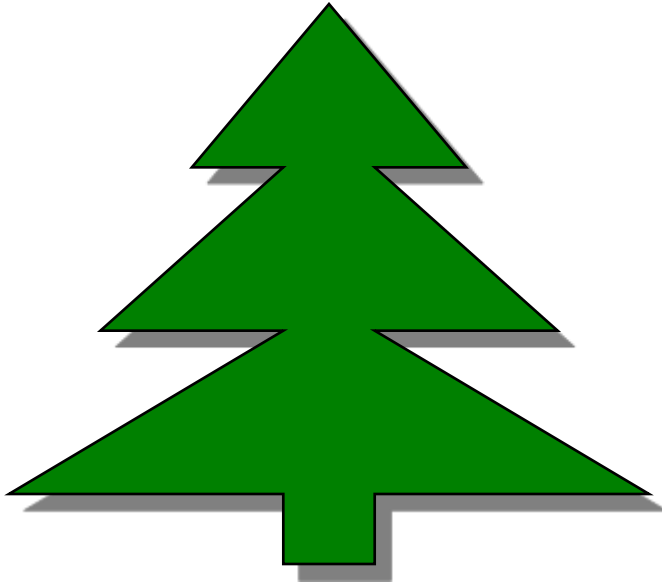
Theory

# **FACTOR METHOD**



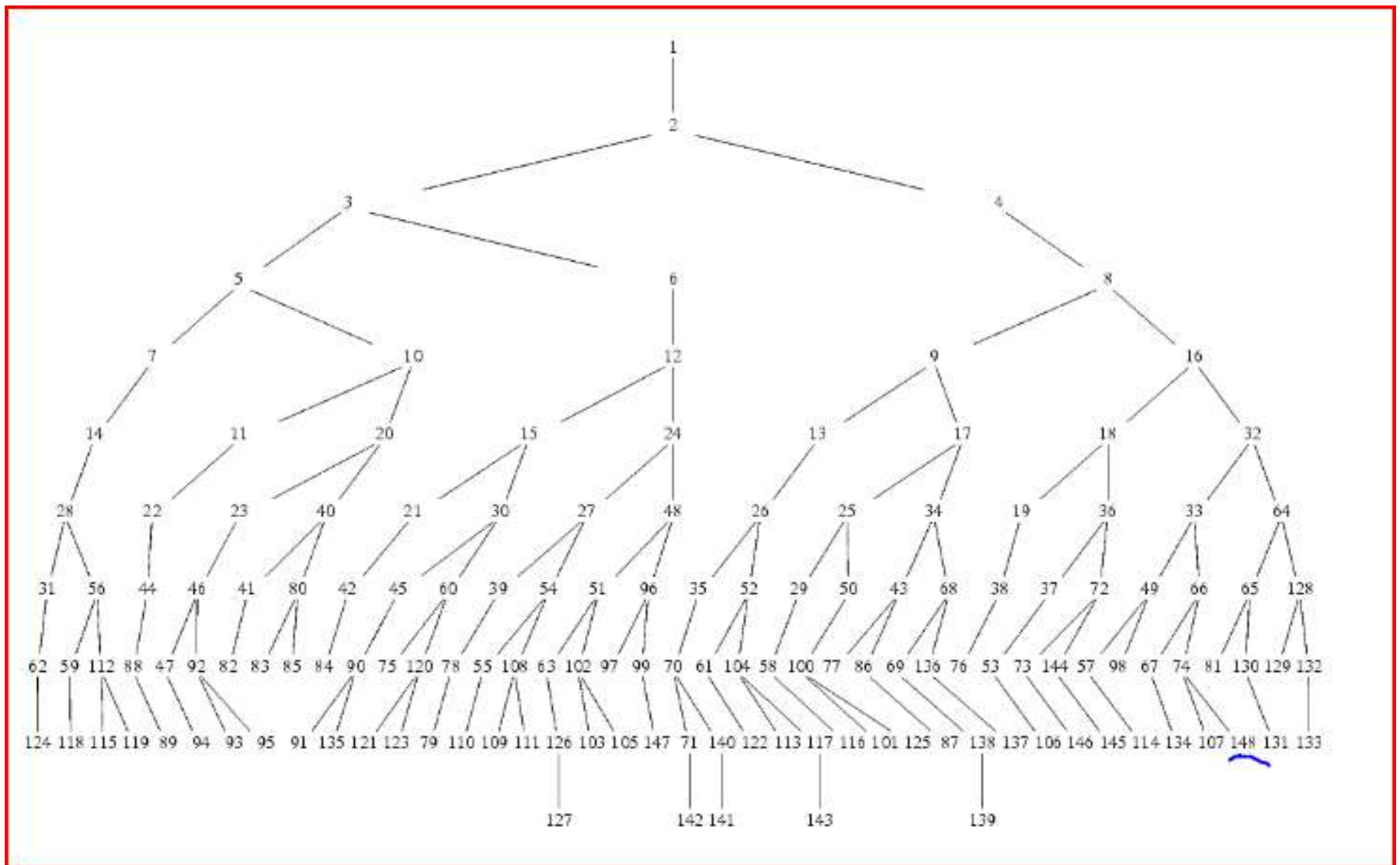
# Theory

- Method based on factorization of  $n$ .
- If  $n = p * q$ , where  $p$  is the smallest prime factor of  $n$  and  $q > 1$ .
- If  $n$  is prime number, calculate  $x^{n-1}$  and multiply by  $x$
- $y^{11} = y^{10} * y = (y^2)^5 * y$
- Required 8 multiplication on contrast 9 by binary method



Theory

# POWER TREE



# Is power tree optimal ?

- Is power tree provides optimal solution for all values of  $n$ ?
- No.
- The smallest example illustrated by knuth was  $n = 77,154,233$  !
- Then what is the economical way to compute  $x^n$  ? Of course, given that the only operation permitted is multiplying two already-computed powers?

Theory

# **ADDITION CHAINS**

# Theory

- Addition chain is a finite sequence of positive integers

$$1 = a_0, a_1, \dots, a_r = n$$

With the property that

$$a_i = a_j + a_k \text{ for some } k \leq j < i$$

- Addition chains should be strictly monotonic increasing

# Minimal length

- Let  $l(n)$  be the minimal length of an addition chain for a given number  $n$ .
- The  $l(n)$  is known only for relatively small values
- For  $n$  large,

$$l(n) = \log(n) + (1 + O(1)) (\log(n)/\log(\log(n)))$$

The lower bound was shown by P. Erdos [1960]

# Example: A Shortest Addition Chain for $n = 148$

0	1
1( 0)	2
2( 1)	4
3( 2)	8
4( 3, 0)	9
5( 4)	18
6( 5)	36
7( 6, 0)	37
8( 7)	74
9( 8)	148



# Star Chain

- If  $j$  or  $k$  equals  $i-1$  for all positive indices  $i$  --, then each  $a_i$  in the chain requires the previous  $a_{i-1}$ . Such a chain is a star-chain.
- Its minimal length is denoted by  $l^*(n)$ .
- is  $l(n) = l^*(n)$  ?
- Walter Hansen [1958] proved in his theorem that for large values of  $n$ ,

$$l(n) < l^*(n)$$

Theory & Application

# **MATRIX POWERS USING EIGEN DECOMPOSITION**

# Fibonacci series

- The recurrence relation of the Fibonacci series as matrix-vector product
- $A = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$  and vector  $U_n = \begin{bmatrix} F_{n+1} \\ F_n \end{bmatrix}$
- Recurrence relation is immediate

$$U_n = A U_{n-1} \text{ for all } n \geq 1$$

$$\text{for } n = 0 \text{ that is } U_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

- Therefore,  $U_n = A^n U_0$

# Eigen pairs to Calculate $A^n$

- Eigen values Characteristic equation:

$$AX = \lambda X$$

$$\Rightarrow (A - \lambda I)X = 0; \text{ if there exists inverse}$$

$$\Rightarrow \det((A - \lambda I)) = 0;$$

$$\Rightarrow \lambda_1 = (1 - \sqrt{5})/2$$

$$\text{And } \lambda_2 = (1 + \sqrt{5})/2$$

# Eigen pairs to Calculate $A^n$

$$AX = \lambda X$$

$$\Rightarrow A(AX) = A(\lambda X)$$

$$\Rightarrow A^2 X = \lambda (A X)$$

$$\Rightarrow A^2 X = \lambda^2 X$$

$$\Rightarrow A^n X = \lambda^n X$$

# Eigen pairs to Calculate $A^n$

- In our case, we are interested in  $X = U_0$ .
- Therefore, express the vector  $U_0$  as a linear combination of  $x$  and  $y$
- $U_0 = ax + by$
- Solving the above linear equation results in
- $a = -1/\sqrt{5}$  and  $b = +1/\sqrt{5}$
- $U_0 = (-1/\sqrt{5})x + (1/\sqrt{5})y$

# Eigen pairs to Calculate $A^n$

$$U_n = A^n U_0$$

$$\Rightarrow U_n = (-1/\sqrt{5}) \lambda_1^n x + (1/\sqrt{5}) \lambda_2^n y$$

$$\Rightarrow F_n = (-1/\sqrt{5}) ((1 - \sqrt{5})/2)^n + (1/\sqrt{5}) ((1 + \sqrt{5})/2)^n$$

Putting it together...

# **CONCLUSION**



# Conclusion

- Let us consider a matrix  $A_{n \times n}$ , the asymptotic complexity of calculating the  $A^n$  is as follow:
  1. The naïve method requires  $O(n^4)$ .
  2. The Binary method requires  $O(n^3 \log(n))$ .
  3. Eigen decomposition requires  $O(n * \log(n)) + O(n^3)$ .

Hint:  $(A^{n \times n})^n = (X^{-1} \lambda X)^n = X^{-1} \lambda^n X$

$\lambda^n$  required  $O(n * \log(n))$  and matrix multiplication requires  $O(n^3)$

**QUESTIONS ?**

**THANK YOU.**