



Design and Analysis of Algorithms

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Strassen's Matrix Multiplication

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- Divide and Conquer:

$$\begin{array}{|c|c|} \hline C_{11} & C_{12} \\ \hline C_{21} & C_{22} \\ \hline \end{array} = \begin{array}{|c|c|} \hline A_{11} & A_{12} \\ \hline A_{21} & A_{22} \\ \hline \end{array} \times \begin{array}{|c|c|} \hline B_{11} & B_{12} \\ \hline B_{21} & B_{22} \\ \hline \end{array}$$

$\begin{array}{c} \updownarrow n/2 \\ \leftarrow n/2 \rightarrow \end{array}$

Now we have:

- $C_{11} = A_{11} \times B_{11} + A_{12} \times B_{21}$
- $C_{12} = A_{11} \times B_{12} + A_{12} \times B_{22}$
- $C_{21} = A_{21} \times B_{11} + A_{22} \times B_{21}$
- $C_{22} = A_{21} \times B_{12} + A_{22} \times B_{22}$

In this case we have $T(n) = 8T(n/2) + O(n^2) = \Theta(n^3)$! **How we can reduce the time complexity?**

Strassen's Matrix Multiplication

In order to improve the algorithm we have to reduce the number of multiplication by introducing the new variables as follows:

- $P = (A_{11} + A_{22}) \times (B_{11} + B_{22})$
- $Q = (A_{21} + A_{22}) \times B_{11}$
- $R = A_{11} \times (B_{12} - B_{22})$
- $S = A_{22} \times (B_{21} - B_{11})$
- $T = (A_{11} + A_{12}) \times B_{22}$
- $U = (A_{21} - A_{11}) \times (B_{11} + B_{12})$
- $V = (A_{12} - A_{22}) \times (B_{21} + B_{22})$

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Now, we have:

- $C_{11} = P + S - T + V$
- $C_{12} = R + T$
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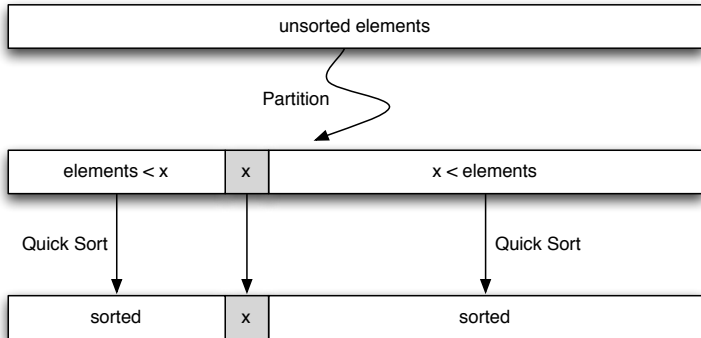
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and so the $T(n)$ can be expressed as:

$$T(n) = 7T(n/2) + O(n^2) = \Theta(n^{\log_2(7)}).$$

Quick Sort



Quick Sort: Algorithm

```
Quick Sort( $A, s, e$ ){  
    if( $s < e$ ){  
        Partition( $A, s, e, m$ );  
        Quick Sort( $A, s, m - 1$ );  
        Quick Sort( $A, m + 1, e$ );  
    }  
}
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```
Partition( $A, s, e, m$ ){  
     $x \leftarrow A[s]$ ;  $i \leftarrow s + 1$ ;  $j \leftarrow e$ ;  
    do{  
        while( $A[i] < x$ )  $i++$ ;  
        while( $A[j] > x$ )  $j--$ ;  
        if( $i < j$ ) swap( $A[i], A[j]$ );  
    } while( $i < j$ );  
    swap( $A[s], A[j]$ );  
    return( $j$ );  
}
```

Quick Sort: Analysis

- **Best case:** when Partition divides the input array into two subarrays with almost equal length.

$$T(n) = 2T(n/2) + O(n) \implies T(n) = O(n \cdot \log(n)).$$

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- **Average case:** Average over all possible length for subarrays...

$$\begin{array}{rcccccl} T(n) & = & T(0) & + & T(n-1) & + & O(n) \\ T(n) & = & T(1) & + & T(n-2) & + & O(n) \\ \vdots & & \vdots & & \vdots & & \\ T(n) & = & T(n-2) & + & T(1) & + & O(n) \\ T(n) & = & T(n-1) & + & T(0) & + & O(n) \\ \hline nT(n) & = & \sum_{i=0}^{n-1} T(i) & + & \sum_{i=0}^{n-1} T(i) & + & nO(n) \end{array}$$

Quick Sort: Average case analysis

$$T(n) = \frac{2}{n} \sum_{i=0}^{n-1} T(i) + cn$$

Now we prove that $T(n) = O(n \cdot \log(n))$ (This is just a guess!). The proof is based on induction:

- **Initiation:** $n = 2 \implies T(2) = T(1) + 2c = O(1)$. ✓

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- **Hypothesis:** $\forall i < n \implies T(i) = O(i \cdot \text{Log}(i)) \leq c' i \cdot \text{Log}(i)$.✓

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Now we prove that $T(n) = O(n \cdot \text{Log}(n))$ (This is just a guess!). The proof is based on induction:

- **Initiation:** $n = 2 \implies T(2) = T(1) + 2c = O(1) \cdot \sqrt{}$
- **Hypothesis:** $\forall i < n \implies T(i) = O(i \cdot \text{Log}(i)) \leq c' i \cdot \text{Log}(i) \cdot \sqrt{}$
- **Induction step:** prove the statement for n :

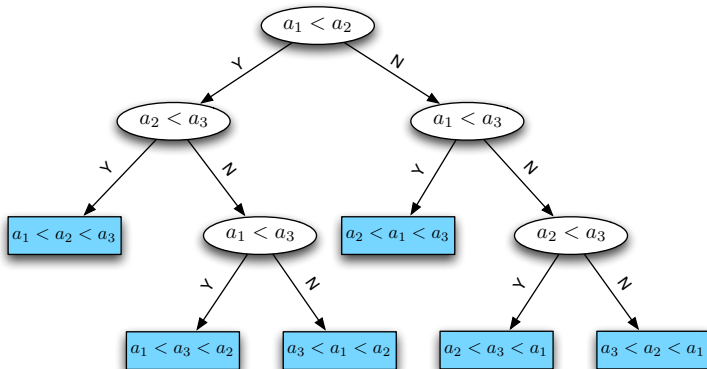
$$\begin{aligned} T(n) &\leq \frac{2}{n} \sum_{i=0}^{n-1} c' i \cdot \text{Log}(i) + cn \\ &\leq \frac{2c'}{n} \left(\sum_{i=0}^{n/2} i \cdot \text{Log}(n/2) + \sum_{i=n/2+1}^{n-1} i \cdot \text{Log}(n) \right) + cn \\ &\leq \frac{2c'}{n} \left(\sum_{i=0}^{n/2} i \cdot \text{Log}(n) - \sum_{i=0}^{n/2} i + \sum_{i=n/2+1}^{n-1} i \cdot \text{Log}(n) \right) + cn \\ &\leq \frac{2c'}{n} \left(\text{Log}(n) \frac{n(n-1)}{2} - \frac{(n/2)(n/2-1)}{2} \right) + cn \\ &= O(n \cdot \text{Log}(n)) \cdot \sqrt{} \end{aligned}$$

Sorting Lower Bound

At least how many comparisons we need to sort three items?

Sorting Lower Bound

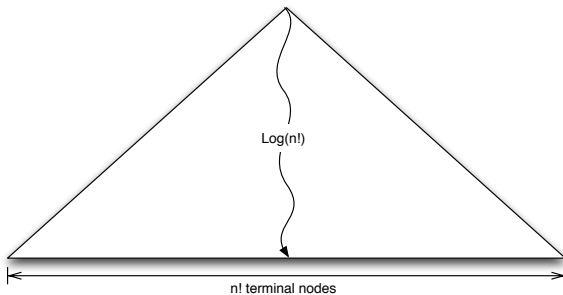
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We need at least $3 = \lceil \log_2(3!) \rceil$ to sort three items.

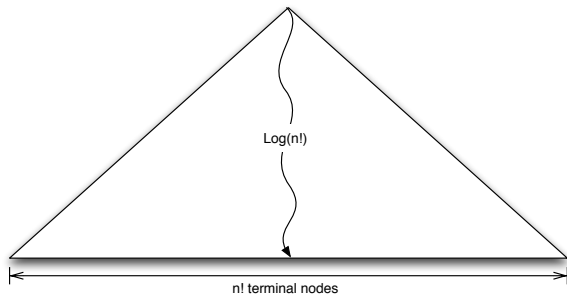
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In general we have:



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$$n! \simeq \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \left(1 + \Theta\left(\frac{1}{n}\right)\right) \quad (\text{Stirling Formula})$$

$$\Rightarrow \log_2 n! \simeq \log_2(\sqrt{2\pi n}) + \log_2 \left(\frac{n}{e}\right)^n + \log_2 \left(1 + \Theta\left(\frac{1}{n}\right)\right)$$

$$\Rightarrow \log_2 n! = \Omega(n \log_2(n)).$$

Exercises

1. Show the details of Matrix Multiplication in which each matrix is divided into nine blocks (each of size $n/3 \times n/3$).
2. Draw a comparison tree for five elements and then show that at most six comparisons are enough to find the median of five elements.

