

# CS 301 – Algorithms

Homework 1 – 04/03/2019

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## **Problem 1.** Asymptotic Growth

Arrangement of the functions  $n2^n$ ,  $n\log(n)$ ,  $n!$ ,  $n$ ,  $n^{100}$ ,  $2^n$ ,  $\log(n)$ ,  $\log(n!)$ ,  $(\log(n))!$ ,  $2^{2^n}$  are provided below;

$$2^{2^n} > n! > n^{100} > n2^n > 2^n > (\log(n))! > n\log(n) > \log(n!) > n > \log(n)$$

## **Problem 2.** Recurrences

For this question, consider the below notations in order to compare and contrast the applied theorem and the questions;

- **Master's Theorem**

$$T(n) = aT(n/b) + f(n) \text{ where } a \geq 1, b > 1 \text{ and } f(n) = \theta(n^{\log_b a})$$

Case 1:  $\log_b a > k$  then  $O(n^{\log_b a})$

Case 2:  $\log_b a = k$

↳ If  $p > -1$ ,  $\theta(n^k \log^{p+1} n)$

↳ If  $p = -1$ ,  $\theta(n^k \log \log n)$

↳ If  $p < -1$ ,  $\theta(n^k)$

Case 3:  $\log_b a < k$

↳ If  $p \geq 0$ ,  $\theta(n^k \log^p n)$

↳ If  $p < 0$ ,  $O(n^k)$

Assuming  $T(n)$  is constant for  $n \leq 2$ .

**(a)**  $T(n) = 2 \cdot T(n/2) + n^3$

- Using the Master's Theorem;  $a = 2$ ,  $b = 2$  and  $f(n) = \theta(n^k \log^p n)$  since  $\log_b a < k$  ( $\log_2 2 < 3$ ) where  $k = 3$  and  $p = 0$  ( $f(n) = n^3$ ).

$$\hookrightarrow p = 0 : \theta(n^3) \rightarrow T(n) = \theta(n^3)$$

**(b)**  $T(n) = 7 \cdot T(n/2) + n^2$

- Using the Master's Theorem;  $a = 7$ ,  $b = 2$  and  $f(n) = O(n^{\log_b a})$  since  $\log_b a > k$  ( $\log_2 7 > 2$ ) where  $k = 2$  and  $p = 0$  ( $f(n) = n^2$ ).

$$\hookrightarrow p = 0 : \log_2 7 > 2 \rightarrow T(n) = O(n^{\log_2 7})$$

**(c)**  $T(n) = 2 \cdot T(n/4) + \sqrt{n}$

- Using the Master's Theorem;  $a = 2$ ,  $b = 4$  and  $f(n) = \theta(n^k \log^{p+1} n)$  since  $\log_b a = k$  ( $\log_4 2 = 0.5$ ) and  $p > -1$  ( $0 > -1$ ) where  $k = 0.5$  and  $p = 0$  ( $f(n) = \sqrt{n}$ ).

$$\hookrightarrow p > -1 : \theta(n^{0.5} \log n)$$

**(d)**  $T(n) = T(n-1) + n$

- Master's Theorem is not suitable for this recurrence. Using Substitution Method;

$$\hookrightarrow \text{Guess } T(n) = O(n)$$

$$\hookrightarrow \text{Now Assume } "T(k) \leq c.k" \text{ and prove } "T(n) \leq c.n"$$

$$\hookrightarrow T(n) = \underbrace{T(n-1) + n}_{\leq c.n - 1 + n}$$

$$\leq \underbrace{c.n - 1 + n}_{= c.n - (c - n)}$$

$$= \underbrace{c.n}_{\text{Desired}} - \underbrace{(c - n)}_{\text{Residual}} \rightarrow (c - n) \geq 0 \text{ if } c \geq n \text{ and thus proving } T(n) = O(n)$$

Desired    Residual

### **Problem 3.** Binary Search – Python

**(a)**

Iterative Binary Search	Recursive Binary Search
<pre>def binarySearch(alist,item):     first = 0     last = len(alist)-1     found = False      while first&lt;=last and not found:         midpoint = (first + last)/2         if alist[midpoint] == item:             found = True         else:             if item &lt; alist[midpoint]:                 last = midpoint-1             else:                 first = midpoint+1     return found</pre>	<pre>def binarySearch(alist,item):     if len(alist) == 0:         return False     else:         midpoint = len(alist)/2         if alist[midpoint] == item:             return True         else:             if item&lt;alist[midpoint]:                 return binarySearch(alist[:midpoint],item)             else:                 return binarySearch(alist[midpoint+1:],item)</pre>

(i) For the Iterative Binary Search, size of the subarray splits (e.g. assuming the size of the array is  $2^n$ , at each iteration, the size will shrink as  $2^{n-1}$ ,  $2^{n-2}$ , ...,  $2^{n-n}$ ) at each iteration. Iterations terminate when the subarray length reaches  $2^{n-n}$ . Thus, for a given array with the size of  $n$ ,  $\log(n) + 1$  operations occur, resulting in  $O(\log(n))$ .

(ii) For the recurrence;

$$T(n) = \begin{cases} 1 & , \quad n = 1 \\ T(n/2) + O(n) + 1 & , \quad n > 1 \end{cases}$$

The asymptotic running time of the recursive binary algorithm is as follows:

○ Using Substitution Method;

$$\hookrightarrow T(n) = T(n/2) + n + 1 = T(n/4) + 3n/2 + 2 = T(n/8) + 7n/4 + 3$$

$$= \dots = T(n/2^k) + \frac{(2^k-1) \cdot n}{2^{k-1}} + k$$

$$\hookrightarrow T(n/2^k) = 1 \rightarrow n = 2^k \text{ where } k = \log(n)$$

$$\hookrightarrow \text{By plugging in } k = \log(n) \text{ we get } T(n) = \underbrace{1 + 2n - 2 + \log(n)}_{\quad \quad \quad}$$

$$O(n + \log(n)) = O(\log(n))$$

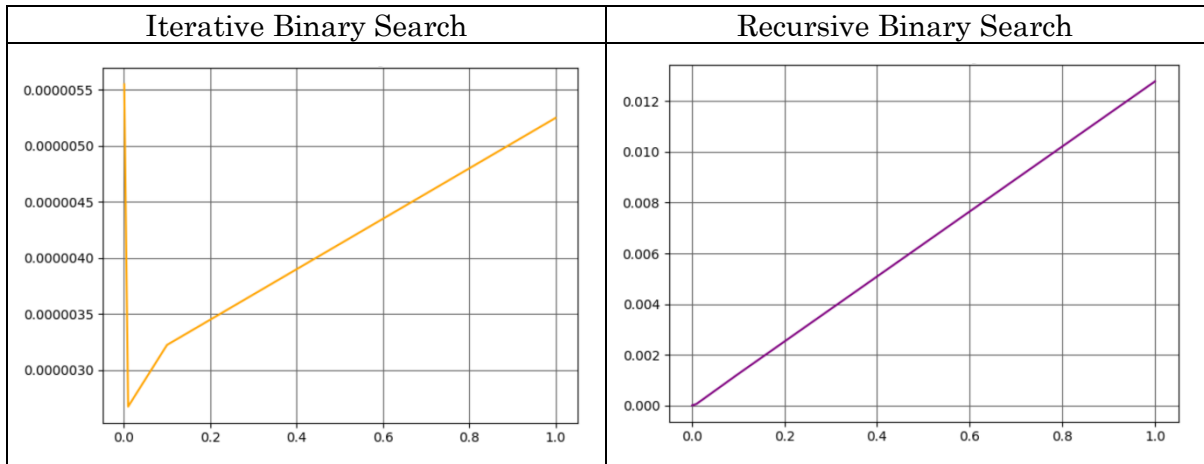
(b)

(i) Computer Specs:

**CPU** – Intel® Core™ i7-6700HQ (6M Cache, up to 3.50 GHz)**Chipset** – Intel® HM170**Memory** – 32 GB DDR4**Storage** – 512 GB SSD + 1 TB SATA HDD**Operating System** – Windows 10 Home

Algorithm	$n = 10^4$	$n = 10^5$	$n = 10^6$	$n = 10^7$
Iterative	2.2200000000 083264e-05	1.070000000025217 4e-05	1.29000000002044 15e-05	2.1000000000270 802e-05
Recursive	4.3999999999 93298e-05	0.000262300000000 15963	0.00498159999999 9808	0.0511002000000 0004

(ii) Graphics for experimental results are provided below;

*x-axis: Input Size, y-axis: Running Time (in seconds)*

(iii) Considering the scalability with respect to these experimental results, the given binary search algorithms are sufficient for large numbers. The given algorithms sustain  $O(n)$  with high efficiency.

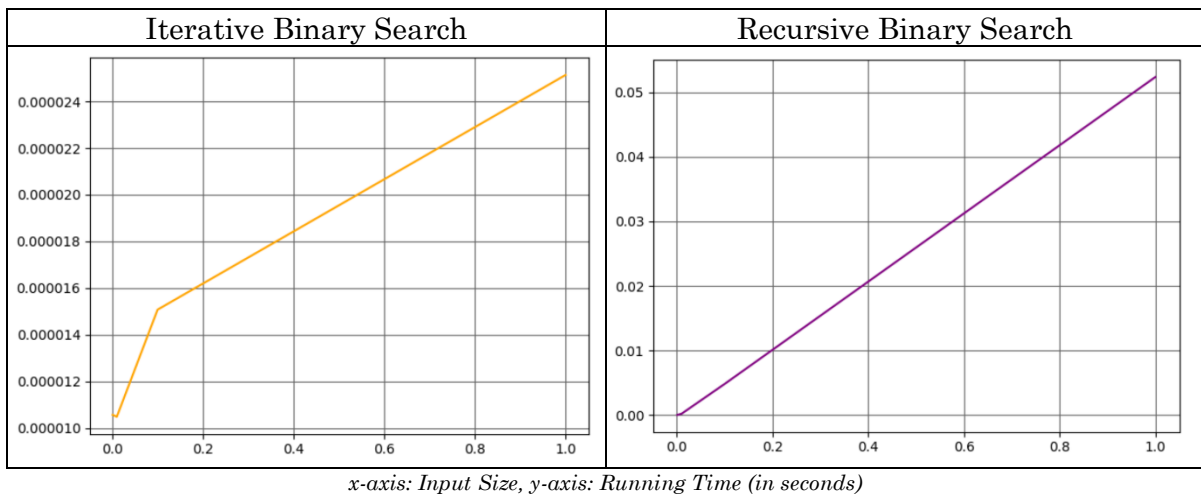
(iv) Considering the theoretical results from part (a), the experimental results are unable to confirm due to the fact that resulting complexity was  $O(n\log(n))$  whereas it is found to be  $O(n)$  for experimental results.

(c)

(i) Average running times in seconds ( $\mu$ ) and the standard deviation ( $\sigma$ ) of the algorithms are provided below:

Algorithm	$n = 10^4$		$n = 10^5$		$n = 10^6$		$n = 10^7$	
	$\mu$	$\sigma$	$\mu$	$\sigma$	$\mu$	$\sigma$	$\mu$	$\sigma$
Iterative	1.056399 9999999 574e-05	0.00015912 519318493 04	1.0480000 00000826 e-05	0.000152 1004930 26132	1.5074000 00001268 9e-05	0.000186 9326024 3311192	2.5134000 00005994 3e-05	0.000291 7690792 085757
Recursive	3.702800 0000005 61e-05	0.00055278 491523783 13	0.0002252 72000000 00747	0.003353 9209563 17864	0.0048399 22000000 008	0.069145 8912967 1335	0.0523535 83999999 974	0.680300 2816117 664

(ii) Graphics for experimental results are provided below;

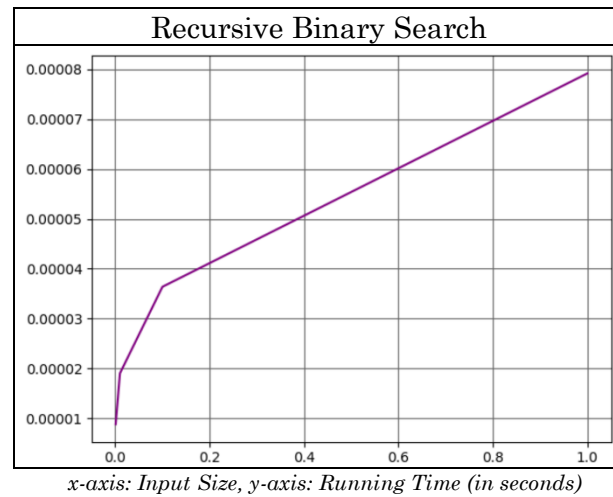


(iii) Considering the average running times, Recursive Binary Search has almost no difference whereas Iterative Binary Search has a drastic increase of rate but not in values.

(d)

In order to improve Recursive Binary Search algorithm so that it can have the same asymptotic running time as the Iterative Binary Search algorithm, start and end points of the indexes can be represented as a parameter in order to avoid using slice function.

- Since aim is to improve the recursion, it can be observed experimentally below:



Ultimately, the modified recursive binary search algorithm serves as expected as it can be observed that the new experimental results are far more familiar with the iterative binary search algorithm than the prior experiment.