

EE608, Homework #1 Solution

(1) CLR 1.2-2

INSERTION-SORT beats MERGE-SORT when $8n^2 < 64n \lg n$, $\Rightarrow n < 8 \lg n$,
 $\Rightarrow \frac{n}{8} < \lg n$, $\Rightarrow 2^{n/8} < n$, which is true when $2 \leq n \leq 43$.

(You can solve for n by trial and error using a calculator. Observe that $5 \cdot 8 < n < 6 \cdot 8 = 48$, since $2^5 = 32 < 40$ and $2^6 = 64 > 48$, then narrow it down with the calculator.)

(2) CLR 1.2-3

ALGORITHM 1 will be faster when $100n^2 < 2^n$, $\Rightarrow n \geq 15$.

(You can solve for n by trial and error using a calculator.)

(3) CLR 1.1

	1 second	1 minute	1 hour	1 day
$\lg n$	2^{10^6}	$2^{6 \times 10^7}$	$2^{3.6 \times 10^9}$	$2^{8.64 \times 10^{10}}$
\sqrt{n}	10^{12}	3.6×10^{15}	1.269×10^{19}	7.4649×10^{21}
n	10^6	6×10^7	3.6×10^9	8.64×10^{10}
$n \lg n$	6.2746×10^4	2.801417×10^6	1.33378058×10^8	2.755147513×10^9
n^2	10^3	7.745×10^3	6×10^4	2.93938×10^5
n^3	100	391	1532	4420
2^n	19	25	31	36
$n!$	9	11	12	13

	1 month	1 year	1 century
$\lg n$	$2^{2.59 \times 10^{12}}$	$2^{3.1536 \times 10^{13}}$	$2^{3.1556736 \times 10^{15}}$
\sqrt{n}	6.7081×10^{24}	$9.94519296 \times 10^{26}$	$9.95827587 \times 10^{30}$
n	2.59×10^{12}	3.1536×10^{13}	3.1556736×10^{15}
$n \lg n$	$7.1870856404 \times 10^{10}$	$7.97633893349 \times 10^{11}$	$6.8654697441062 \times 10^{13}$
n^2	1.609347×10^6	5.615692×10^6	5.6175382×10^7
n^3	13733	31593	146677
2^n	41	44	51
$n!$	15	16	17

(4) CLR 2.1-1

The operation of INSERTION-SORT can be illustrated by Figure 1

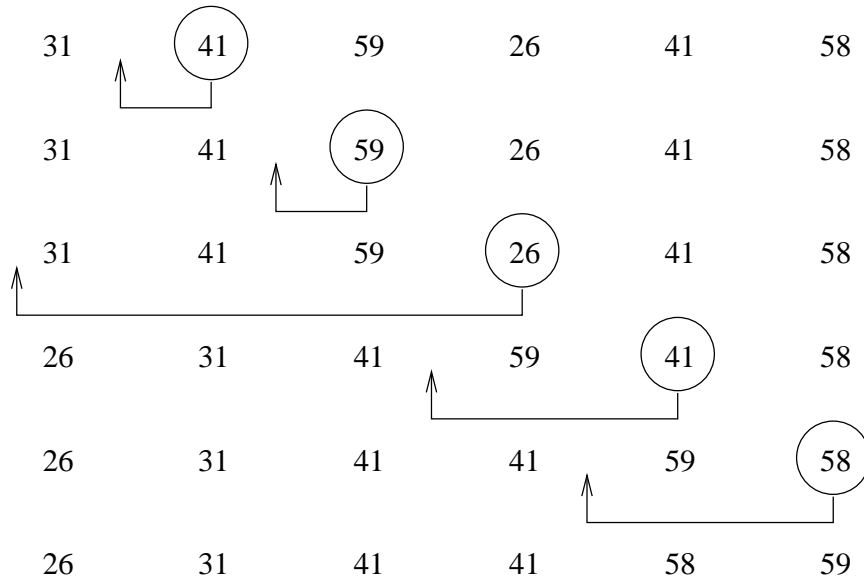


Figure 1: The operation of insertion sort for Problem CLR 2.1-1.

(5) CLR 2.1-2

NON-INCREASING-INSERTION-SORT(A)

1. **for** $j \leftarrow 2$ to $\text{length}[A]$
2. **do** $\text{key} \leftarrow A[j]$
3. \triangleright Insert $A[j]$ into the sorted sequence $A[1..j-1]$
4. $i \leftarrow j-1$
5. **while** $i > 0$ and $A[i] < \text{key}$
6. **do** $A[i+1] \leftarrow A[i]$
7. $i \leftarrow i-1$
8. $A[i+1] \leftarrow \text{key}$

(6) CLR 2.1-3

The pseudocode is as follows:

LINEAR-SEARCH(A, v)

1. $j \leftarrow 1$
2. **while** $j \leq \text{length}[A]$
3. **if** $A[j] = v$
4. **then return** j
5. **else** $j \leftarrow j+1$
6. **return** NIL

We need to show the algorithm is correct by using a loop invariant in the algorithm. The loop invariant is as follows:

At the start of the j th loop iteration of the while loop in lines 2-5, the subarray $A[1..j - 1]$ consists of elements not equal to v .

Now we will show that the loop invariant satisfies the three properties mentioned in the book:

Initialization: When $j = 1$, $A[1..j - 1]$ contains no elements; hence, there can be no elements in $A[1..j - 1]$ that are equal to v . Thus, the loop invariant holds.

Maintenance: If at the start of the $j = i$ iteration, the subarray $A[1..i - 1]$ contains no elements equal to v , then there are two conditions to check. If $A[j] = v$, then we exit the loop and j is not incremented; hence, the loop invariant still holds. If $A[j] \neq v$, then j is incremented to $i + 1$ and $A[1..i]$ contains no elements equal to v at the start of the $j = i + 1$ iteration; hence, the loop invariant still holds.

Termination: If the loop terminates with $j = \text{length}[A] + 1$, then $A[1..\text{length}[A]]$ contains no elements equal to v , and the loop invariant holds. The algorithm will then return NIL at line 6, which is the correct response for not finding an element in A equal to v . If the loop terminates with $j < \text{length}[A] + 1$, then the algorithm has found an element $A[j]$ which is equal to v , returning key j at line 4 without incrementing j . Note that it is still the case that no elements in $A[1..j - 1]$ are equal to v , and so the loop invariant holds.

Hence, the algorithm is correct when there is an element in A which is equal to v and when there is no such element.

(7) CLR 2.2-2

SELECTION-SORT(A)

1. $n \leftarrow \text{length}[A]$
2. **for** $i \leftarrow 1$ to $n - 1$
3. **do** $\text{min} \leftarrow \infty$
4. \triangleright Find smallest element of current portion of $A[i..n]$
5. **for** $j \leftarrow i$ to n
6. **if** $A[j] < \text{min}$
7. **do** $\text{min} \leftarrow A[j]$
8. $\text{key} \leftarrow j$
9. \triangleright Swap the smallest $A[\text{key}]$ with the element $A[i]$ and work with $A[i + 1..n]$
10. **exchange** $A[i] \leftrightarrow A[\text{key}]$
11. **return** A

The worst case running time for SELECTION-SORT is $\Theta(n^2)$, because the algorithm must find the next smallest element for each iteration. The best case could

be $O(n)$ if the array is sorted by adding a linear check to see if the array is already sorted; however, without this check the best case is $\Theta(n^2)$.

The loop invariant is that at the start of the i th iteration $A[1..i - 1]$ contains the smallest $i - 1$ elements of A in sorted order.

At the end of the $(n - 1)$ th iteration, the n th element of A is no smaller than the first $n - 1$ elements (by the loop invariant). And since $A[1..n - 1]$ is sorted at this point, the whole array is sorted. Thus, there is no need for the n th iteration.

(8) CLR 2.2-4

Just modify the algorithm to store a precomputed answer for some of the inputs. Check for those inputs at the outset, and return the corresponding answer if it is detected.

(9) CLR 2.3-3

Base Case: For $n = 2$, $T(2) = 2 \lg 2 = 2$

Induction Hypothesis: Assume for $n = 2^k$, $T(2^k) = 2^k \lg 2^k$.

Induction Step: Show this is true for $n = 2^{k+1}$, that is, $T(2^{k+1}) = 2^{k+1} \lg 2^{k+1}$.

$$\begin{aligned} T(2^{k+1}) &= 2T\left(\frac{2^{k+1}}{2}\right) + 2^{k+1} \\ &= 2T(2^k) + 2^{k+1} = 2(2^k \lg 2^k) + 2^{k+1} \\ &= 2^{k+1}(\lg 2^k + 1) = 2^{k+1}(\lg 2^k + \lg 2) \\ &= 2^{k+1}(\lg(2 \times 2^k)) = 2^{k+1} \lg(2^{k+1}) \end{aligned}$$

(10) CLR 2.3-4

The running time for the recursive version of INSERTION-SORT is the following:

$$T(n) = \begin{cases} \Theta(1) & \text{if } n \leq 1, \\ T(n - 1) + C(n) & \text{otherwise.} \end{cases}$$

where $C(n)$ is the time to insert the n^{th} element into $A[1..n - 1]$.

(11) CLR 2.3-7

FindMatch(S,x) ;return true if found, false if not

{

1. Mergesort(S,1,n) ; $\Theta(n \cdot \log n)$

2. if $x > S[n] + S[n - 1]$ or $x < S[1] + S[2]$ return false ; this line is optional, it can save some time by skipping the search for cases where x is too small or too large to be the sum of a pair of values in the sequence.

```

3. for  $i \leftarrow 1$  to  $n$ 
4.   do if (BinarySearch( $S, x - S[i]$ )=true);  $\Theta(\log n)$ 
5.     then return true
6. return false
}

```

(12) CLR 2-2

- (a) The output $A'[1] \dots A'[n]$ should be a permutation of the input $A[1] \dots A[n]$.
- (b) The loop-invariant for the loop in lines 2-4 is that at the end of each iteration $A[j-1] \leq A[k]$ for all $k > j-1$. This can be proven as follows:

Initialization: Before the first iteration either $A[n-1] \leq A[n]$ or $A[n-1] > A[n]$. In the former case, the invariant will be true at the end of the first iteration because the **If** in line 3 evaluates to false and the exchange does not take place. In the latter case the **If** in line 3 evaluates to true, causing the last elements of the array A to be exchanged, and ensuring the invariant to hold true at the end of the iteration.

Maintenance: At the beginning of each iteration, $A[j]$ is known to be no greater than $A[k]$ for all $k > j$. During each iteration the elements $A[j]$ and $A[j-1]$ are compared and possibly exchanged such that the greater of the two is placed in $A[j]$. This ensures that at the end of the iteration $A[j-1]$ is not greater than $A[j]$, and by transitivity also not greater than $A[k]$ for all $k > j-1$.

Termination: At termination $j = i + 1$, so $j - 1 = i$. Because of the loop invariant we have that $A[i] \leq A[k]$ for all $k > i$.

- (c) The loop-invariant for the loop in lines 1-4 is that at the end of each iteration the $A[1] \leq A[2] \leq A[3] \dots \leq A[i]$.

Initialization: At the end of the first iteration of the outer loop $i = 1$ and the loop-invariant is trivially true.

Maintenance: The termination condition of the inner loop as proved in part(b) ensures that at the end of each iteration of the outer loop $A[i] \leq A[k]$ for all $k > i$. In the next iteration the element that is brought into $A[i+1]$ will be chosen from amongst $A[k]$ for $k > i$. We are guaranteed from the termination condition of the inner loop in the previous iteration of the outer loop that this element $A[i+1] \geq A[i]$. Hence on each iteration of the outer loop the element brought into $A[i]$ is such that $A[i-1] \leq A[i]$. The elements $A[1] \dots A[i-1]$ are not disturbed during the i 'th iteration. So at the end of the i 'th iteration we have that $A[1] \leq A[2] \leq A[3] \dots \leq A[i]$.

Termination: At the end of the last iteration of the outer loop $i = \text{length}[A]$ and the loop-invariant ensures that $A[1] \leq A[2] \leq A[3] \dots \leq A[i]$. Thus the algorithm terminates with the array A sorted in non-ascending order.

- (d) The outer loop iterates $n = \text{length}[A]$ times. The inner loop iterates i times where i increases by 1 on each iteration of the outer loop. The running time of the algorithm can be summed as:

$$T(n) = \sum_{i=1}^n \sum_{j=i}^n c_1$$

$$T(n) = \sum_{i=1}^n (n-i)c_1$$

$$T(n) = \sum_{i=1}^n nc_1 - \sum_{i=1}^n ic_1$$

$$T(n) = n^2c_1 - \frac{n(n+1)}{2}c_1$$

$$T(n) = \frac{1}{2}n^2c_1 - \frac{n}{2}c_1$$

$$T(n) = \Theta(n^2)$$

The running time for BUBBLESORT is the same as the worst-case running time for INSERTION-SORT

(13) CLR 2-4

- (a) The input sequence, $\langle 2, 3, 8, 6, 1 \rangle$, has the following inversions: $\{(1, 5), (2, 5), (3, 5), (4, 5), (3, 4)\}$.
- (b) The array with the most inversions contains unique elements and is reverse sorted, for example, $A = \{n, n-1, \dots, 1\}$. In this case, there are the following number of inversions:

$$N_{inv} = (n-1) + (n-2) + \dots + (n-(n-1)) = \sum_{i=1}^{n-1} (n-i) = \sum_{i=1}^{n-1} i = \frac{n(n-1)}{2}$$

- (c) Each element j of the array that is not in the correct position is shifted left until it reaches the element i such that $A[i] > j$. The number of shifts performed by insertion sort corresponds to the number of inversions of every element k , $j \leq k < i$ in the relation to i (i.e., $\text{inversion}(k, i)$). Each shift eliminates an inversion from the set.
- (d) The algorithm to determine the number of inversions in $\Theta(n \lg n)$ uses MERGE-SORT with an enhancement in the MERGE procedure. When an array A_{low} (A with the lower index range) is merged with an array A_{high} (A with the greater index range, further from the index 1), if an element is picked from the top of A_{high} , the number of elements left in A_{low} should be added to form the total number of inversions at the end of the algorithm.

For example, $\langle 5, 2, 4, 6, 1, 3, 2, 6 \rangle$ has $N_{inv} = 13$, using the MERGE-SORT enhancement (see Figure 2).

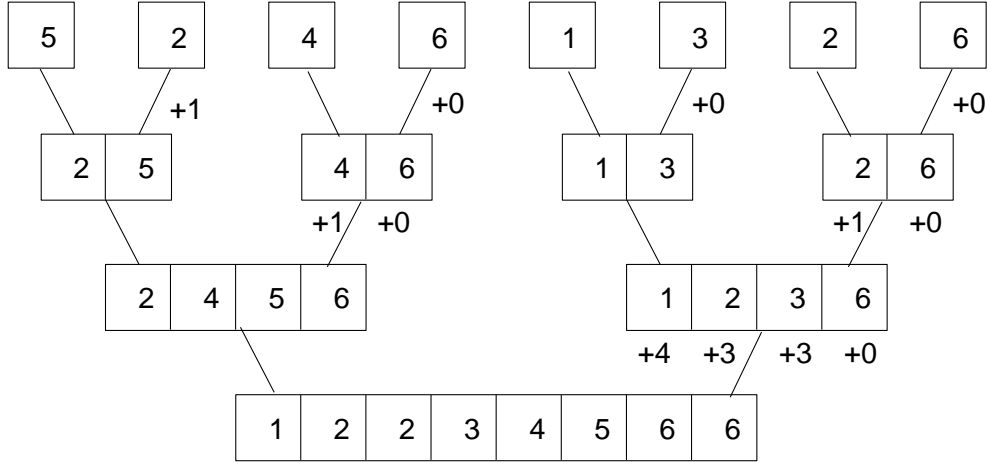


Figure 2: The MERGE-SORT enhancement for the example in Problem CLR 2-4(d).

MERGE-SORT can be modified to count inversions by adding line 18 to MERGE, as shown below. Note that *Num_of_inv* is a global variable initialized to zero that contains the total number of inversions when MERGE-SORT terminates. This will not change the asymptotic behavior of MERGE-SORT.

```

MERGE(A, p, q, r)
1.  $n_1 \leftarrow q - p + 1$ 
2.  $n_2 \leftarrow r - q$ 
3. Create arrays  $L[1..n_1 + 1]$  and  $R[1..n_2 + 1]$ 
4. for  $i \leftarrow 1$  to  $n_1$ 
5.     do  $L[i] \leftarrow A[p + i - 1]$ 
6. for  $j \leftarrow 1$  to  $n_2$ 
7.     do  $R[j] \leftarrow A[q + j]$ 
8.  $L[n_1 + 1] \leftarrow \infty$ 
9.  $R[n_2 + 1] \leftarrow \infty$ 
10.  $i \leftarrow 1$ 
11.  $j \leftarrow 1$ 
12. for  $k \leftarrow p$  to  $r$ 
13.     do if  $L[i] \leq R[j]$ 
14.         then  $A[k] \leftarrow L[i]$ 
15.              $i \leftarrow i + 1$ 
16.         else  $A[k] \leftarrow R[j]$ 
17.              $j \leftarrow j + 1$ 
18.              $Num\_of\_inv \leftarrow Num\_of\_inv + n_1 + 1 - i$ 

```

(14) CLR A.1-1

$$\sum_{k=1}^n (2k - 1) = 2 \sum_{k=1}^n k - \sum_{k=1}^n 1 = \frac{2n(n+1)}{2} - n = n(n+1) - n = n^2$$

(15) CLR A.1-6

Prove that $\sum_{k=1}^N O(f_k(n)) = O(\sum_{k=1}^N f_k(n))$.

Proof

Let $A = \sum_{k=1}^N O(f_k(n))$, and $B = O(\sum_{k=1}^N f_k(n))$.

Because A and B are sets of functions, we have to show $A \subseteq B$ and $B \subseteq A$.

(1) $A \subseteq B$

Let $g_k(n) = O(f_k(n))$ for $k \in \{1, 2, \dots, n\}$. Then, there exist $c_k > 0$ and $n_k > 0$ such that

$$0 \leq g_k(n) \leq c_k f_k(n) \quad \forall n \geq n_k$$

Choose $c = \max_{k \in \{1, 2, \dots, n\}} \{c_k\}$ and $n_0 = \max_{k \in \{1, 2, \dots, n\}} \{n_k\}$. Then, we have

$$0 \leq g_k(n) \leq c_k f_k(n) \leq c f_k(n) \quad \forall n \geq n_0$$

By the linear property of the summation,

$$0 \leq \sum_{k=1}^N g_k(n) \leq c \sum_{k=1}^N f_k(n) \quad \forall n \geq n_0$$

Hence, $\sum_{k=1}^N g_k(n) = O(\sum_{k=1}^N f_k(n))$.

Because the above equation holds for an arbitrary $g_k(n)$, we have:

$$\sum_{k=1}^N O(f_k(n)) \subseteq O(\sum_{k=1}^N f_k(n))$$

.

(2) $(B \subseteq A)$

Let $g(n) = O(\sum_{k=1}^N f_k(n))$. Then, there exist $c > 0$ and $n_0 > 0$ such that

$$0 \leq g(n) \leq c \sum_{k=1}^N f_k(n) \quad \forall n \geq n_0$$

By the linear property of summations,

$$0 \leq g(n) \leq \sum_{k=1}^N [c f_k(n)] \quad \forall n \geq n_0$$

If we choose $c_k = c$ and $n_k = n_0$ for $k \in \{1, 2, \dots, n\}$ in the above inequality, we obtain that there exist $c_k > 0$ and $n_k > 0$ for $k \in \{1, 2, \dots, n\}$, such that $0 \leq g(n) \leq c_1 f_1(n) + c_2 f_2(n) + \dots + c_n f_n(n)$ whenever $n \geq n_1, n \geq n_2, \dots, n \geq n_n$. This means that $g(n) = \sum_{k=1}^N O(f_k(n))$.

Therefore, $O(\sum_{k=1}^N f_k(n)) \subseteq \sum_{k=1}^N O(f_k(n))$.

From (1) and (2), we have $A = B$.

(16) CLR A.1-7

One method of solution uses a logarithm to convert the product to a sum:

$$\begin{aligned}\prod_{k=1}^n 2 \cdot 4^k &= 2^n \prod_{k=1}^n 4^k \\ \lg(2^n \prod_{k=1}^n 4^k) &= \lg(2^n) + \lg(\prod_{k=1}^n 4^k) \\ &= \lg(2^n) + \sum_{k=1}^n \lg(4^k) \\ &= \lg(2^n) + \lg(4) \sum_{k=1}^n k \\ &= \lg(2^n) + 2 \lg(2) \frac{n(n+1)}{2} \\ &= n \lg(2) + n(n+1) \lg(2) \\ &= n(n+2) \lg(2) \\ &= \lg(2^{n(n+2)})\end{aligned}$$

$$\prod_{k=1}^n 2 \cdot 4^k = 2^{n(n+2)}.$$

An alternative method of solution uses product and exponent rules:

$$\begin{aligned}\prod_{k=1}^n 2 \cdot 4^k &= 2^n \cdot 4^{\sum_{k=1}^n k} \\ &= 2^n \cdot 4^{\frac{n(n+1)}{2}} \\ &= 2^n \cdot 2^{n(n+1)} \\ &= 2^{n+n(n+1)} \\ &= 2^{n^2+2n} \\ &= 2^{n(n+2)}\end{aligned}$$

(17) CLR A.2-4

Evaluate $\sum_{k=1}^n k^3$

Since k^3 is a monotonically increasing function,

$$\int_0^n x^3 dx \leq \sum_{k=1}^n k^3 \leq \int_1^{n+1} x^3 dx \leq \int_0^{n+1} x^3 dx$$

Hence:

$$\frac{x^4}{4} \Big|_0^n \leq \sum_{k=1}^n k^3 \leq \frac{x^4}{4} \Big|_0^{n+1}$$

Giving:

$$\frac{n^4}{4} \leq \sum_{k=1}^n k^3 \leq \frac{(n+1)^4}{4}$$

Hence, $\sum_{k=1}^n k^3 = \Theta(n^4)$

(18) CLR A.2-5

If you bound the summation using an integral directly (the integral bounds are from 0 to n) the final solutions becomes $[l n n - l n 0]$.

However, $l n 0$ is negative infinity.

So instead we break up the summation

$\sum_{k=1}^n 1/k$ and rewrite it as $\sum_{k=2}^n 1/k + 1$.

(19) CLR A-1

(a) Use the integral method of bounding the summation:

$$\begin{aligned} \int_0^n x^r dx &\leq \sum_{k=1}^n k^r \leq \int_1^{n+1} x^r dx \\ \frac{1}{r+1} x^{r+1} \Big|_0^n &\leq \sum_{k=1}^n k^r \leq \frac{1}{r+1} x^{r+1} \Big|_1^{n+1} \\ \frac{n^{r+1}}{r+1} &\leq \sum_{k=1}^n k^r \leq \frac{(n+1)^{r+1} - 1}{r+1} \leq \frac{(n+1)^{r+1}}{r+1} \end{aligned}$$

Therefore,

$$\sum_{k=1}^n k^r = \Theta(n^{r+1})$$

(b) On one hand,

$$\sum_{k=1}^n \lg^s k \leq \sum_{k=1}^n \lg^s n = n \lg^s n.$$

Clearly, $\sum_{k=1}^n \lg^s k = O(n \lg^s n)$.

On the other hand,

$$\sum_{k=1}^n \lg^s k = \sum_{k=1}^{\lfloor n/2 \rfloor} \lg^s k + \sum_{k=\lfloor n/2 \rfloor + 1}^n \lg^s k$$

$$\geq \sum_{k=1}^{\lfloor n/2 \rfloor} 0 + \sum_{k=\lfloor n/2 \rfloor + 1}^n \lg^s(n/2)$$

$$\geq (n/2) \lg^s(n/2).$$

We can easily prove that $(n/2) \lg^s(n/2) = \theta(n \lg^s n)$.

So $\sum_{k=1}^n \lg^s k = \theta(n \lg^s n)$.

(c)

On one hand,

$$\sum_{k=1}^n k^r \lg^s k \leq \sum_{k=1}^n k^r \lg^s n = \lg^s n \sum_{k=1}^n k^r$$

and from part (a)

$$\sum_{k=1}^n k^r = \Theta(n^{r+1})$$

so,

$$\sum_{k=1}^n k^r \lg^s k = O(n^{r+1} \lg^s n)$$

On the other hand,

$$\sum_{k=1}^n k^r \lg^s k = \sum_{k=1}^{n/2} k^r \lg^s k + \sum_{k=n/2}^n k^r \lg^s k$$

$$\sum_{k=1}^n k^r \lg^s k \geq \sum_{k=1}^{n/2} 0 + \sum_{k=n/2}^n (n/2)^r \lg^s(n/2)$$

$$\sum_{k=1}^n k^r \lg^s k \geq (n/2)^{r+1} \lg^s(n/2)$$

And then combining the above two results it can be shown that,

$$\sum_{k=1}^n k^r \lg^s k = \Theta(n^{r+1} \lg^s n)$$