1 Introduction

This document provides a runtime analysis of the SLOWSORT algorithm[1]. Despite being presented in a satirical article, its recursion pattern turns out to be rather complex, which in turn makes it harder to analyze. On the other hand, the authors do not provide a thorough derivation of the claimed running time in the aforementioned article. The asymptotic bounds explored here do not match tightly with those, as they differ roughly by a linear factor.

2 Slowsort algorithm

The following figure shows the pseudocode of Slowsort:

```
Slowsort (A[1 \dots n]):

If n = 1: return.

Let m := \frac{n+1}{2}.

Slowsort (A[1 \dots m]).

Slowsort (A[m+1 \dots n]).

If A[n] < A[m]: swap A[n] and A[m].

Slowsort (A[1 \dots n-1]).
```

3 Runtime analysis

3.1 Recurrence relation

Let T(n) be the function that characterizes the worst-case running time of SLOWSORT when given an input array of size n. Being a recursive algorithm, we can define T recurrently like so:

$$T(n) = 2T(n/2) + T(n-1) + \Theta(1)$$

Note that we intentionally leave out base cases as we aim at finding asymptotic bounds for T. From this, we can see that $T(i) - T(i-1) = 2T(i/2) + \Theta(1)$ for a given i. Thus,

$$\sum_{i=1}^{n} T(i) - T(i-1) = \sum_{i=1}^{n} 2T(i/2) + \Theta(1)$$

$$\Rightarrow T(n) - T(0) = \sum_{i=1}^{n} 2T(i/2) + \sum_{i=1}^{n} \Theta(1)$$

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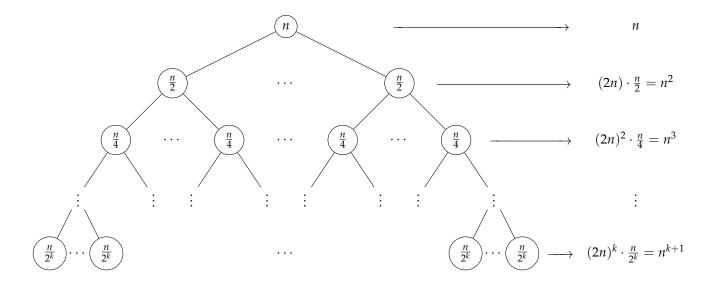
$$\Rightarrow T(n) = \sum_{i=1}^{n} 2T(i/2) + \Theta(n)$$

3.2 Asymptotic bounds

We shall look now for asymptotic bounds for T. Let us consider first the following recurrence U:

$$U(n) = \sum_{i=1}^{n} 2U(n/2) + \Theta(n)$$
$$= 2n U(n/2) + \Theta(n)$$

Clearly, we have that $T(n) \le U(n)$ for all n. In order to find an asymptotic solution for U, let's have a look at its recursion tree:



From this, it follows that

$$U(n) = \sum_{k=0}^{\lfloor \log_2(n) \rfloor} n^{k+1}$$

$$= n \sum_{k=0}^{\lfloor \log_2(n) \rfloor} n^k$$

$$= \frac{n}{n-1} \left(n^{\lfloor \log_2(n) \rfloor + 1} - 1 \right)$$

$$\in \Theta \left(n^{\log_2(n) + 1} \right)$$

This yields an asymptotic upper bound for *T*:

$$T(n) \in \mathcal{O}\left(n^{\log_2(n)+1}\right) \tag{1}$$

In order to find an asymptotic lower bound, let's first rewrite *T* in the following convenient way:

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

Now, define the recurrence L(n) as follows:

$$L(n) = \sum_{i=0}^{\left\lfloor \sqrt{n} \right\rfloor - 1} 2L(n/2 - i/2) + \Theta(n)$$

Such an L nearly satisfies every hypotheses required by the Akra-Bazzi theorem[2]. The only apparent exception is that the summation in the right-hand side goes up to a function of n, and the theorem states that this value should be a constant $k \in \mathbb{N}$. However, the proof outlined in [2] does not seem to rely strongly on this fact¹. In what follows, then, we will use the Akra-Bazzi theorem to obtain an asymptotic solution for L.

Before proceeding, we observe that the perturbation functions $h_i(n)$ are properly bounded:

¹If someone sufficiently qualified to confirm or disprove this assertion ever reads this, please let me know. Thanks in advance!

$$|h_{i}(n)| = |-\frac{i}{2}|$$

$$= \frac{i}{2}$$

$$\leq \frac{\lfloor \sqrt{n} \rfloor - 1}{2}$$

$$\leq \sqrt{n}$$

$$\in O\left(\frac{n}{\log^{2} n}\right)$$
(2)

(2) holds since $\frac{\sqrt{n}}{n/\log^2 n}$ goes to 0 as $n \to \infty$.

References

- [1] A. Broder and J. Stolfi, "Pessimal algorithms and simplexity analysis," ACM SIGACT News, vol. 16, pp. 49–53, 1984.
- [2] T. Leighton, "Notes on better master theorems for divide-and-conquer recurrences," in *Lecture notes*, *MIT*, 1996.