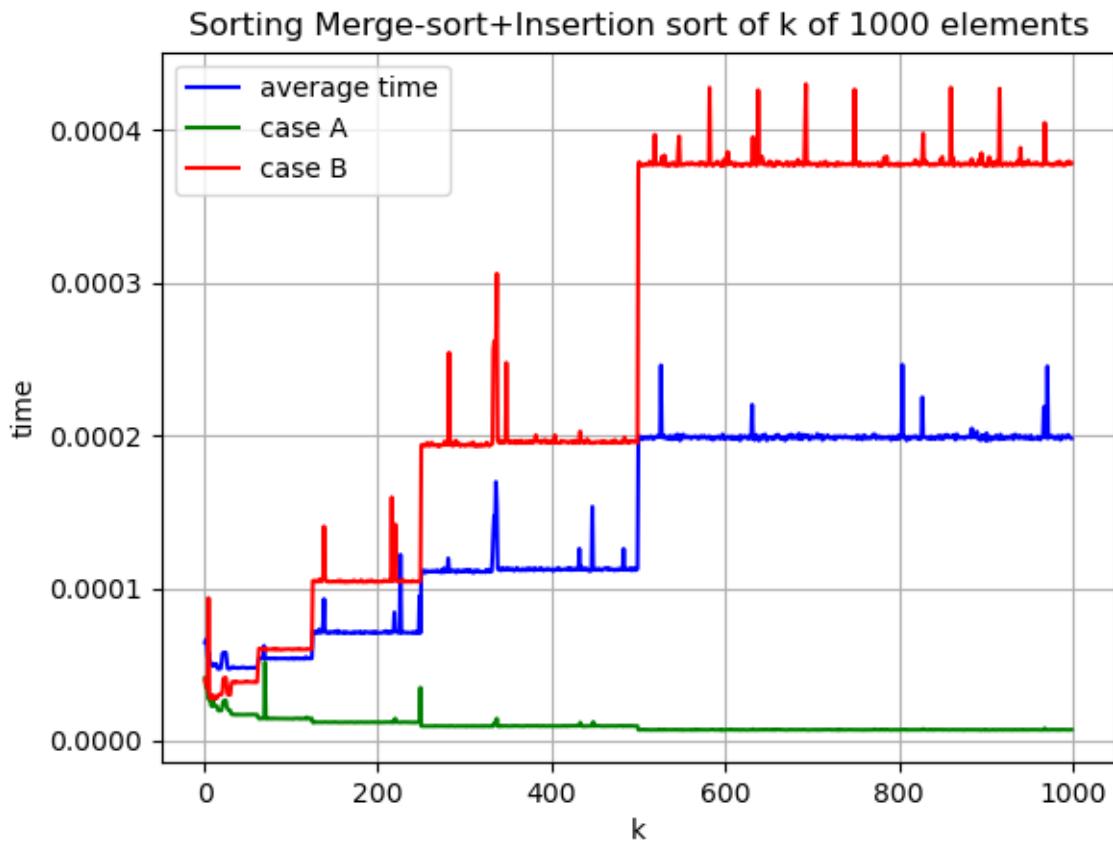


Homework 4 by Timofei Podlorytov

4.1



the graph above was created for sorting of 1000 elements and 30 sample size.

(a-b)

The algorithm for the task is in a p1.cpp file and the graph was created using the graph.py file which I also provide.

The best, worst and average cases are presented on the graph. I tested for different values of k which are shown on the x axis.

(c)

As we see here the graph is kinda like stairs and the jumps appear on $k=2^n$ for the average and worst case. While the best case declines in time as we turn to using insertion sort faster and thus need to merge less lists together → perform less operations and go to n complexity in insertion sort. When we have $k=2^n$ we equally split the array and can do the merging for efficiently with the list of closer size.

The jumps happen between $k=2^n$ and $k=2^{n-1}$ as I have proven by checking the values near those k in the python script. Currently they are commented out, but can be easily run, if needed.

(d)

We can clearly see the time increase for larger k as well as values of $k=2^{n-1}$ being good compared to the ones that come after. So, the best choice would be to choose a small k that equals 2^{n-1} . For instance, $k=1$ would work well.

4.2

Time complexity:

- (a) (2 points) $T(n) = 36T(n/6) + 2n$,
 - (b) (2 points) $T(n) = 5T(n/3) + 17n^{1.2}$,
 - (c) (2 points) $T(n) = 12T(n/2) + n^2 \lg n$,
 - (d) (2 points) $T(n) = 3T(n/5) + T(n/2) + 2^n$,
 - (e) (2 points) $T(n) = T(2n/5) + T(3n/5) + \Theta(n)$.

a)

$$T(n) = 36T(n/6) + 2n,$$

In this case the master theorem can be used as we have only one $T(n)$ and all the values of a and b satisfy.

$$\log_b a = \log_6 36 = 2 \rightarrow f(n) = O(n^{(2-1)}) = O(n) \text{ as } f=2n \rightarrow \text{first case and } T(n) = \Theta(n^{(\log_6 36)}) = \Theta(n^2)$$

b)

We follow a similar logic for this case.

$$T(n) = 5T(n/3) + 17n^{1.2}$$

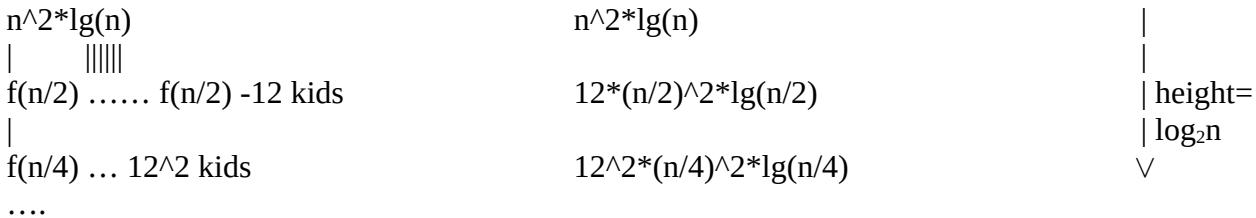
Here we have $\log_b a = \log_3 5 \approx 1.465$. while $f = 17n^{1.2} \in O(n^{1.2}) \in O(n^{(\log_3 5)}) \rightarrow$ first case and $T(n) = \Theta(n^{(\log_3 5)})$

c)

$$T(n) = 12T(n/2) + n^2 \lg n$$

We can't apply the master method here as $n^2 \lg n$ is not a polynomial and we have to use something else.

Let's try using the tree:



$$T(n) = n^2 \lg(n) + 12*(n/2)^2 \lg(n) - 12*(n/2)^2 \lg(2) + 12^2*(n/4)^2 \lg(n) - 12^2*(n/4)^2 \lg(4) \dots = \\ n^2 \lg(n) + 3*(n)^2 \lg(n) - 3*(n)^2 \lg(2) + 3^2*(n)^2 \lg(n) - 3^2*(n)^2 \lg(2)*2 + \\ + 3^3/(2^6)*(n)^2 \lg(n) - 3^3/(2^6)*(n)^2 \lg(2)*3 = \dots$$

We get:

$n^{^2\lg(n)(1+3+9+\dots)} - n^{^2\lg(2)(1+3+9*2+27*3+3^4*4+\dots)}$ in each sum we have $\log_2 n$ elements

we know that $n^2 \lg(n)$ grows faster than n^2 so we can dismiss n^2 for the upper bound. Plus the sum is positive and

Now we get:

$n^{2\lg(n)} * (1 - 3^{\log_2 n}) / (1 - 3)$ which is $\Theta(n^2 * \lg(n) * 3^{\lg(n)})$

d)

$$T(n) = 3T(n/5) + T(n/2) + 2^n$$

here the master theorem won't work as we have two separate function for $T(n)$ inside the main one plus non polynomial as $f(n)$.

Substitution:

let's try 2^n .

Base step:

$$T(1)=2^1=O(2^n) \text{ true}$$

Induction step: assume true for any $k < n$

$$T(n)=3T(n/5)+T(n/2)+2^n=3*2^{n/5}+2^{n/2}+2^n=2*(2^n)-(2^n-3*2^{n/5}-2^{n/2})=\text{desired}>0 \quad \text{residual}>0 \text{ for } n\geq 3$$

$$=c*2^n-\text{residual}>0 \rightarrow O(2^n) \text{ true}$$



Is this the tightest fit though? If not there must exist a tighter $O(f(n))$

$$O(n^k) \subset O(2^n)$$

Base case $n=1$

$T(1)=2^1 \neq 1^k$ we have an issue of an exponential component in the function in its root. Thus getting anything less will be impossible.

$$\Theta(2^n)$$

e)

$$T(n) = T(2n/5) + T(3n/5) + \Theta(n).$$

Master theorem won't work here due to two recursive function calls of $T(n)$ in $T(n)$

Let's try a recursion tree:

$\Theta(n)$	1 child	0
$\Theta(2n/5) \Theta(3n/5)$	2 children	1
$\Theta(4n/25) \Theta(6n/25) \Theta(6n/25) \Theta(9n/25)$	4 children	2
.....
	2^n children	n

but as we know $\Theta(\text{const}*n)=\Theta(n) \rightarrow$ in each step the expression is multiplied by a constant inside Θ and we can simplify it in the following way

$\Theta(n)$	1 child	0	$\Theta(n)$
$\Theta(n) \Theta(n)$	2 children	1	$2\Theta(n)$
$\Theta(n) \Theta(n) \Theta(n) \Theta(n)$	4 children	2	$4\Theta(n)$
.....	

2^n children

n $2^n \Theta(n)$

Overall sum is $\Theta(n) + 2\Theta(n) + 4\Theta(n) + 8\Theta(n) + \dots + 2^n \Theta(n) = \Theta(n)(1+2+4+\dots+2^n) = \Theta(n)*(2^{n+1}-1)/2$

the final n is the height of the tree which is $\log_{0.4}(n)$

$\Theta(n * 2^n * \log_{0.4}(n))$