

ASSIGNMENT 2

SUB-SET SUM

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1 DYN

This approach has time complexity of $\mathcal{O}(nk)$ and space complexity of $\mathcal{O}(k)$. It is greatly effected by the size of array **A** and/or targeted sum **k**. If we introduce big numbers this problem becomes really hard for this approach. Example of such problem would be:

```
5
10000000
2000000
2000000
2000000
2000000
2000000
```

2 EXH

This approach has time and space complexity of $\mathcal{O}(2^n)$. The worst case scenario is when the input is a long array **A** of some relatively small numbers and some really big targeted sum **k**. If the targeted sum **k** is greater or equal then sum of the whole array **A**, this approach will perform exhaustive search without any kind of pruning. Example of such problem would be:

```
3000000
6000000
1
2
3
1
2
3
...
1
2
3
```

3 GREEDY

This approach has time complexity of $\mathcal{O}(n \log n)$ and space complexity of $\mathcal{O}(n)$. In the worst case this algorithm returns at most 2 times worse result. Example of such problem would be:

```
3
1000000
500001
500000
500000
```

In this case optimal result is 1000000 but the algorithm would return 500001. The result is almost 2 times worse then the optimal ($500001 \approx 1000000/2$).

4 FPTAS

This is ϵ -approximation. Its execution time is greatly effected by the parameter ϵ . We have compared how the ϵ parameter effect the execution time of this algorithm. We have generated random inputs using code listed in the Listing 1. We observed (Figure 1) that with smaller values of ϵ this algorithm performs slower and has higher standard deviation.

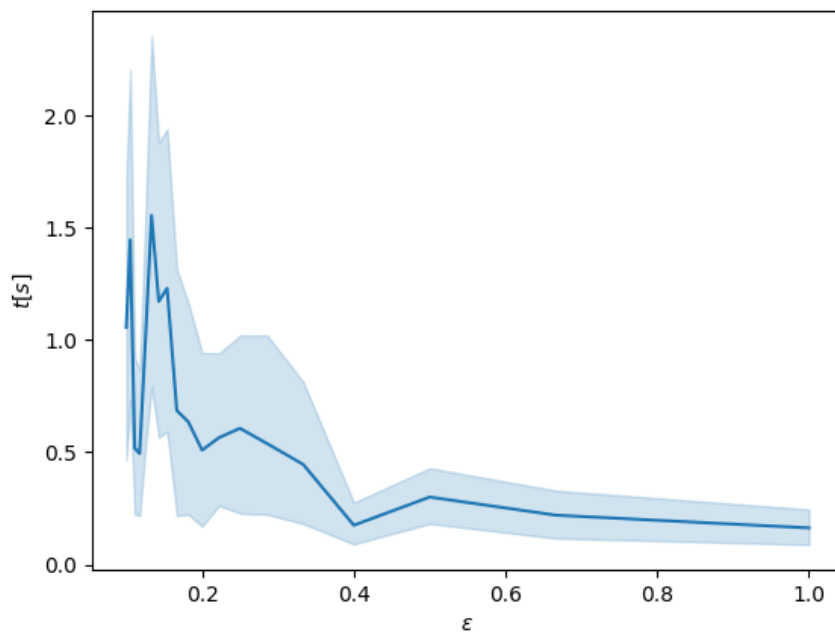


Figure 1: Execution time compared to the parameter ϵ

Listing 1: Random problem generator

```
1 def generate(n: int) -> Tuple[List[int], int]:
2   A = []
3   for _ in range(n):
4     A.append(randint(1, int(n/2)))
5
6   k = sum(sample(A, int(n*random()))))
7   return (A, k)
```
