

A Fully Polynomial Time Approximation Scheme for Subset Sum

CLRS – Chapter 35

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Approximation Algorithms

For any given optimization (minimization) problem and approximation algorithm \mathcal{A} to solve it:

- Let $\Pi =$ Set of all *instances* of the problem.
- For all instances $I \in \Pi$ define $size(I)$.
- For all instances $I \in \Pi$ define

$OPT(I) =$ cost of optimal solution for I

$A(I) =$ cost of solution produced by A on I .

Let $\rho(n)$ be a function such that

$$\forall I \in \Pi, \text{ with } size(I) = n, \quad \frac{A(I)}{OPT(I)} \leq \rho(n).$$

Then A is a

factor $\rho(n)$ approximation algorithm
($\rho(n)$ -approximation algorithm) .

(there is a similar definition for *maximization* problems)

The Subset-Sum Problem

Definition: An instance of the *subset-sum decision problem* is (S, t) where: $S = \{x_1, x_2, \dots, x_n\}$ a set of positive integers; t a positive integer.

The problem is whether some subset of S adds up exactly to t . This problem is NP-complete

The *subset-sum optimization problem* is to find a subset of S whose sum is as large as possible but no greater than t .

We will define a class of algorithms A_ϵ , such that, $\forall \epsilon > 0$,

- A_ϵ is an ϵ -approximation algorithm for subset-sum.
- A_ϵ runs in time polynomial in n , $\log t$ and $\frac{1}{\epsilon}$.

Such a class of algorithms is known as a

A fully polynomial-time approximation scheme.

An Exponential Time Algorithm

If $S = \{x_1, x_2, \dots, x_n\}$ is a set or list and x a real number then define

$$S + x = \{x_1, x_2, \dots, x_n\} = \{x_1 + x, x_2 + x, \dots, x_n + x\}.$$

If $L = \{x_1, x_2, \dots, x_n\}$ and $L' = \{u_1, u_2, \dots, u_m\}$ are both sorted lists then define **Merge-Lists**(L, L') to be the procedure that returns the sorted union of the two lists. This procedure runs in time $O(|L'| + |L|)$.

Exact-Subset Sums

$n \leftarrow |S|$

$L_0 \leftarrow \langle 0 \rangle$

for $i = 1$ **to** n

$L_i = \text{Merge-Lists}(L_{i-1}, L_{i-1} + x_i)$

remove from L_i **all elements bigger than** t .

return largest element in L_n .

Let P_i be the set of all values that can be obtained by selecting some subset of $\{x_1, x_2, \dots, x_i\}$ and summing its members. Then L_i is a sorted list containing all elements in P_i of size no greater than t .

The algorithm therefore returns the correct answer.

Since L_i can have as many as 2^i items this algorithm can take $\Theta(2^n)$ time!

Trimming

Let $L\{x_1, x_2, \dots, x_m\}$ be a list. To **trim** the list by parameter δ means to remove as many elements from L as possible in such a way that the list L' of remaining elements has the following property:

For every $y \in L$ there exists a $z \in L'$ such that
 $(1 - \delta)y \leq z \leq y$.

Example:

$L = \langle 10, 11, 12, 15, 20, 21, 22, 23, 24, 29 \rangle$ **and** $\delta = 0.1$.

A trimmed list would be $L' = \langle 10, 12, 15, 20, 23, 29 \rangle$.

```
Trim(L,  $\delta$ )  
 $L' = \langle x_1 \rangle$   
 $last = x_1$   
for  $i = 2$  to  $m$   
    if  $last < (1 - \delta)x_i$   
        then append  $x_i$  onto end of  $L'$ .  
             $last = x_i$   
return  $L'$ 
```

This algorithm returns a trimmed list in $O(m)$ time.
(It assumes that input list is sorted in non-decreasing order.)

The Actual Approximation Algorithm

```
Approx-Subset-Sum( $S, t, \epsilon$ )  
 $n \leftarrow |S|$ .  
 $L_0 = \langle 0 \rangle$ .  
for  $i = 1$  to  $n$   
     $L_i \leftarrow$  Merge-Lists( $L_i, L_{i-1} + x_i$ )  
     $L_i \leftarrow$  Trim( $L_i, \epsilon/n$ )  
    remove from  $L_i$  all elements bigger than  $t$   
return largest element in  $L_n$ 
```

Note that when list L_i is trimmed we introduce a relative error of at most ϵ/n between the representative values remaining and the elements of the list. By induction can show that, $\forall y \in P_i$ there exists some $z \in L_i$ such that

$$\left(1 - \frac{\epsilon}{n}\right)^i y \leq z \leq y.$$

Let \bar{z} be the largest element in L_n . If y^* is a solution to the exact subset-sum problem then there exists a $z^* \in L_n$ such that

$$\left(1 - \frac{\epsilon}{n}\right)^n y^* \leq z^* \leq \bar{z} \leq y^*.$$

But $\forall n > 1$,

$$1 - \epsilon \leq \left(1 - \frac{\epsilon}{n}\right)^n \Rightarrow (1 - \epsilon)y^* \leq \bar{z},$$

and A_ϵ is an ϵ -approximation algorithm.

Running Time

The running time of the i th stage of the algorithm is $O(|L_i|)$.

After trimming, successive elements $z', z \in L_i$ have the property

$$z' < z \left(1 - \frac{\epsilon}{n}\right).$$

Therefore the total number of elements in L_i is at most

$$\begin{aligned} \log_{\frac{1}{1-\frac{\epsilon}{n}}} t &= \frac{\ln t}{-\ln \left(1 - \frac{\epsilon}{n}\right)} \\ &\leq \Theta \left(\frac{n \ln t}{\epsilon}\right) \end{aligned}$$

The running time of A_ϵ is proportional to

$$\frac{n^2 \ln t}{\epsilon}$$

and the A_ϵ form a

Fully Polynomial Time Approximation Scheme

for subset-sum.