

in each list L_i is at most

$$\begin{aligned}
 \log_{1+\epsilon/2n} t + 2 &= \frac{\ln t}{\ln(1 + \epsilon/2n)} + 2 \\
 &\leq \frac{2n(1 + \epsilon/2n) \ln t}{\epsilon} + 2 \quad (\text{by inequality (3.16)}) \\
 &\leq \frac{4n \ln t}{\epsilon} + 2 \quad (\text{by inequality (35.23)}) .
 \end{aligned}$$

This bound is polynomial in the size of the input—which is the number of bits $\lg t$ needed to represent t plus the number of bits needed to represent the set S , which is in turn polynomial in n —and in $1/\epsilon$. Since the running time of APPROX-SUBSET-SUM is polynomial in the lengths of the L_i , APPROX-SUBSET-SUM is a fully polynomial-time approximation scheme. ■

Exercises

35.5-1

Prove equation (35.21). Then show that after executing line 5 of EXACT-SUBSET-SUM, L_i is a sorted list containing every element of P_i whose value is not more than t .

35.5-2

Prove inequality (35.24).

35.5-3

Prove inequality (35.27).

35.5-4

How would you modify the approximation scheme presented in this section to find a good approximation to the smallest value not less than t that is a sum of some subset of the given input list?

Problems

35-1 Bin packing

Suppose that we are given a set of n objects, where the size s_i of the i th object satisfies $0 < s_i < 1$. We wish to pack all the objects into the minimum number of unit-size bins. Each bin can hold any subset of the objects whose total size does not exceed 1.

- a.** Prove that the problem of determining the minimum number of bins required is NP-hard. (*Hint*: Reduce from the subset-sum problem.)

The **first-fit** heuristic takes each object in turn and places it into the first bin that can accommodate it. Let $S = \sum_{i=1}^n s_i$.

- b.** Argue that the optimal number of bins required is at least $\lceil S \rceil$.
- c.** Argue that the first-fit heuristic leaves at most one bin less than half full.
- d.** Prove that the number of bins used by the first-fit heuristic is never more than $\lceil 2S \rceil$.
- e.** Prove an approximation ratio of 2 for the first-fit heuristic.
- f.** Give an efficient implementation of the first-fit heuristic, and analyze its running time.

35-2 Approximating the size of a maximum clique

Let $G = (V, E)$ be an undirected graph. For any $k \geq 1$, define $G^{(k)}$ to be the undirected graph $(V^{(k)}, E^{(k)})$, where $V^{(k)}$ is the set of all ordered k -tuples of vertices from V and $E^{(k)}$ is defined so that (v_1, v_2, \dots, v_k) is adjacent to (w_1, w_2, \dots, w_k) if and only if for each i , $1 \leq i \leq k$ either vertex v_i is adjacent to w_i in G , or else $v_i = w_i$.

- a.** Prove that the size of the maximum clique in $G^{(k)}$ is equal to the k th power of the size of the maximum clique in G .
- b.** Argue that if there is an approximation algorithm that has a constant approximation ratio for finding a maximum-size clique, then there is a fully polynomial-time approximation scheme for the problem.

35-3 Weighted set-covering problem

Suppose that we generalize the set-covering problem so that each set S_i in the family \mathcal{F} has an associated weight w_i and the weight of a cover \mathcal{C} is $\sum_{S_i \in \mathcal{C}} w_i$. We wish to determine a minimum-weight cover. (Section 35.3 handles the case in which $w_i = 1$ for all i .)

Show that the greedy set-covering heuristic can be generalized in a natural manner to provide an approximate solution for any instance of the weighted set-covering problem. Show that your heuristic has an approximation ratio of $H(d)$, where d is the maximum size of any set S_i .