

# ON THE COMPLEXITY OF SOLVING THE GENERALIZED SET PACKING PROBLEM APPROXIMATELY

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**Abstract.** The generalized set packing problem ( $GSP$ ) is as follows. Given a family  $F$  of subsets of  $M = \{1, \dots, m\}$  and a vector  $b \in R^m$ , find a subfamily  $F' \subset F$  of maximum cardinality such that for every  $i \in M$ ,  $i$  does not belong to more than  $b_i$  members of  $F'$ . The subproblem ( $GSP_k$ ) consists of those instances of ( $GSP$ ) where each member of  $M$  belongs to at most  $k$  members of  $F$ . It is shown that for every  $k$ , if there is a polynomial-time approximation algorithm for ( $GSP_k$ ) with a positive performance ratio then  $GSP_k$  has a polynomial approximation scheme. This generalizes a result of Garey and Johnson with regard to the maximum independent set problem in a graph.

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The generalized set packing problem is the following integer linear programming problem:

$$\begin{aligned}
 & \text{Maximize } e^T x \\
 (GSP) \quad & \text{subject to } Ax \leq b \\
 & x \in \{0, 1\}^n
 \end{aligned}$$

where  $A \in \{0, 1\}^{m \times n}$  is a zero-one matrix,  $b \in \mathbb{Z}^m$  is an integral vector, and  $e$  denotes a vector of 1's. For an application of the problem, consider a scheduling problem where jobs require processing during certain time intervals. Thus,  $A_{ij}$  equals 1 if and only if job  $j$  requires processing during the  $i$ -th period. The value of  $b_i$  gives the number of jobs that can run in parallel during the  $i$ -th period. The problem is to select a set of jobs of maximum cardinality which can be processed subject to the constraints.

The definitions in the present paragraph are taken from [GJ]. Given an instance  $I$  of  $(GSP)$ , denote by  $opt(I)$  the value of its optimal solution. If  $F$  is an approximation algorithm for  $(GSP)$ , let  $F(I)$  denote the value of the solution delivered by  $F$ , and let  $r(F) = \inf_I \{F(I)/opt(I)\}$ . The ratio  $r(F)$  is the *worst-case performance ratio* of the algorithm  $F$ . An *approximation scheme* is an algorithm  $\overline{F}$  that receives together with the problem another parameter  $\epsilon > 0$ . Thus, for a fixed  $\epsilon$  there is an implicit algorithm  $F_\epsilon$  that is derived from  $\overline{F}$ . If for every  $\epsilon > 0$  the derived algorithm  $F_\epsilon$  runs in polynomial time and has a performance ratio  $r(F_\epsilon) \geq 1 - \epsilon$  then  $\overline{F}$  is said to be a *polynomial approximation scheme* [GJ].

The maximum independent set problem is a special case of  $(GSP)$  where  $A$  has exactly two 1's in each row (i.e.,  $A$  is the node-arc incidence matrix of some graph), and  $b = e$ . Garey and Johnson [GJ] proved the following:

**Proposition 1.** *If the maximum independent set problem has a polynomial-time approximation algorithm with a positive performance ratio then it also has a polynomial approximation scheme.*

We show that Proposition 1 can be extended to a parameterized family of subproblems of the generalized set packing problem. For every integer  $k \geq 2$ , let  $(GSP_k)$  denote the set of instances of  $(GSP)$  where the matrix  $A$  has at most  $k$  1's per row. A generalization of Proposition 1 is as follows.

**Proposition 2.** *For every  $k \geq 2$ , if there is a polynomial-time approximation algorithm for  $(GSP_k)$  with a positive performance ratio then there is a polynomial approximation scheme for  $(GSP_k)$ .*

*Proof:* The proof generalizes Garey and Johnson's proof of Proposition 1. Consider an instance of  $(GSP_k)$ , given by a pair  $(A, b)$  ( $A \in \{0, 1\}^{m \times n}$ ,  $b \in Z^m$ ). We produce from  $(A, b)$  another instance  $(\bar{A}, \bar{b})$ ,  $\bar{A} \in \{0, 1\}^{\bar{m} \times \bar{n}}$ ,  $\bar{b} \in Z^{\bar{m}}$  (where  $\bar{m} \leq mn(n^{k-1} + 1)$  and  $\bar{n} = n^2$ ) as follows (see an example below). The rows of  $\bar{A}$  are of two types. The first  $mn$  rows are of the first type and are obtained by placing  $n$  copies of the matrix  $A$  in a diagonal configuration. Also, the first  $mn$  entries of  $\bar{b}$  contain  $n$  copies of  $b$ . The other rows are of the second type and produced as follows. From each row  $i$  of  $A$  with  $k_i$  1's ( $k_i \leq k$ ) we produce  $n^{k_i}$  rows of  $\bar{A}$ . Specifically, a typical row is produced from the  $i$ 'th row of  $A$  in the following way. For each  $j$  such that  $A_{ij} = 1$ , pick a number  $\ell_j$ ,  $n(j-1) + 1 \leq \ell_j \leq nj$ . Given the choices  $\ell_j$ , create a row of  $\bar{A}$  by placing 1's in the  $k_i$  positions of the form  $\ell_j$  and 0's in all the other positions. Obviously, for each  $i$  the number of different ways to choose the  $\ell_j$ 's is  $n^{k_i}$ . The corresponding entry of  $\bar{b}$  is set to  $b_i$ . Consider the generalized set packing problem with  $(\bar{A}, \bar{b})$ . The number of variables is  $n^2$  and it is convenient to use here doubly indexed ones  $z_{jh}$ , defined by  $z_{jh} \equiv x_{n(j-1)+h}$ . Under this equivalence, the symbols  $\bar{A}z$  and  $e^T z$  are well-defined. Consider any zero-one solution  $z = (z_{jh})$  of the system  $\bar{A}z \leq \bar{b}$ . Denote  $x^{(j)} = (z_{j1}, \dots, z_{jn})^T$  ( $j = 1, \dots, n$ ). It is easy to see that  $x^{(j)}$  is a solution of the system  $Ax \leq b$ . Now, denote  $y_j = \prod_h z_{jh}$  ( $j = 1, \dots, n$ ). We claim that  $y = (y_1, \dots, y_n)^T$  is a zero-one solution of  $Ax \leq b$ . For if  $A_i y > b_i$  for some row  $i$  then by choosing  $\ell_j$  (for each  $j$  such that  $A_{ij} = 1$ ) so that  $y_j = z_{j\ell_j}$ , we identify a row  $i'$  of  $\bar{A}$  for which  $\bar{A}_{i'} z > \bar{b}_{i'}$ . Moreover,

$$e^T z = \sum_j e^T x^j = \sum_{j: y_j = 1} e^T x^j \leq (e^T y) \max_j \{e^T x^j\} \leq \max \left\{ (e^T y)^2, \max_j \{(e^T x^j)^2\} \right\}.$$

On the other hand, if  $y = x^{(j)} = x'$  ( $j = 1, \dots, n$ ) then  $e^T z = (e^T x')^2$ . Thus, if the optimal value of the given problem is equal to  $V$  then the optimal value of the problem with  $(\bar{A}, \bar{b})$  is equal to  $V^2$ . Now, suppose there exists a polynomial-time approximation algorithm  $F$  whose performance ratio is  $r > 0$ . Let  $z$  denote the solution given by  $F$  for the system with  $(\bar{A}, \bar{b})$ . Thus,  $(e^T z)/V^2 > r$ . Let  $x^*$  denote one of the vectors  $y, x^{(1)}, \dots, x^{(n)}$  so that  $e^T x^*$  is maximal. It follows from the above that  $(e^T x^*)/V \geq \sqrt{r}$ . This means that there is also a polynomial-time approximation algorithm with a

performance ratio of  $\sqrt{r}$ . Repeating this argument sufficiently many times, we see that for every  $\epsilon > 0$  a ratio of  $1 - \epsilon$  can be guaranteed. ■

**Example 3.** Consider the pair

$$A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 0 \end{bmatrix} \quad b = \begin{bmatrix} 2 \\ 1 \end{bmatrix} .$$

Here the pair  $(\overline{A}, \overline{b})$  is the following:



## Reference

- [GJ] M. R. Garey and D. S. Johnson, *Computers and intractability: A Guide to the theory of NP-completeness*, W. H. Freeman, San Francisco, 1979.