Polynomial-time Matrix-based Method of Determining Subset Sum Solutions

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Abstract

Reducing the conditions under which a given set satisfies the stipulations of the subset sum proposition to a set of linear relationships, the question of whether a set satisfies subset sum may be answered in a polynomial number of steps by forming, solving, and constraining a series of linear systems whose dimensions and number are both polynomial with respect to the length of the set. Following implementation of this algorithm, initial testing over 20,000,000 trials using random inputs of length 1 to 20, 1,000,000 trials each, reveal 100% accuracy with exactly 0 failures.

1 INTRODUCTION

The subset sum problem is a well-known member of the NP-complete complexity class: given a set of integers A and some constant c, is there some subset of A which sums to c?

Previously, there have been no known algorithms or methods to solve the value-unbounded, general-case subset sum problem in polynomial time. The naive algorithm performs in exponential time by cycling through the possible subsets of A until it has either seen all subsets or found one which sums to c. A common pseudo-polynomial method employs a dynamic programming algorithm whose complexity is polynomial with respect to the length of the set and the range of the inputs, O(n(M-N)), where n is the length of the set and B-A is the range of inputs; however, this solution is not truly polynomial, as it is polynomial with respect to M-N, which is exponential in its number of bits. Approximate algorithms exist which can be modified to find exact solutions; however, they too degrade to being exponential in the number of bits required to represent elements in the set.

In contrast with pre-existing algorithms, the method described here does not concern itself with the various subsets that exist within the input set, but rather searches the solution space of a set of linear constraints when applied to an input set to deduce if a solution can exist; the method to be discussed is a strategy which may be employed to find solutions satisfying the constraints of the subset sum problem in time polynomial with respect only to the length of the input, having general-case applicability on the basis of universally occuring properties in sets satisfying the problem.

2 Method

2.1 Preliminary Conventions, Definitions, and Properties

Given a set A of n greater than four elements and an instance of subset sum for a constant c, satisfied by a subset S of length greater than 2 (the algorithm first catches trivial cases for S of length 1 or 2), index A as follows:

$$A = \{a_1, a_2, ..., a_n\} \tag{1}$$

The strategy outlined will conform A to a set of linear constraints which will reveal a subset sum-satisfying subset if one exists. To this end, define a subset membership vector m specific to A such that the ith value in m is 1 if the ith element of A is an element of a given subset S of A summing to c, 0 otherwise. If such a subset S exists, m exists and encodes S within A.

$$m = \begin{pmatrix} m_1 \\ \vdots \\ \vdots \\ m_n \end{pmatrix} \tag{2}$$

If S exists, the four following linear constraints surrounding A, S, and m will be satisfied:

1. S sums to c.

$$a_1 m_1 + a_2 m_2 + \dots + a_n m_n = c (3)$$

2. S has finite length t.

$$m_1 + m_2 + \dots + m_n = |S| = t \tag{4}$$

3. There exists an index r for which the rth element of A is or is not in S.

$$m_r = v \in \{0, 1\} \tag{5}$$

4. There exists an index s for which the sth element of A is or is not in S.

$$m_s = v \in \{0, 1\} \tag{6}$$

Using the above four constraints, an underdetermined system Z can be constructed in the parameters of the constraints listed:

$$Z(A, t, r, v_r, s, v_s)$$
:
$$a_1m_1 + a_2m_2 + \dots + a_nm_n = c$$

$$m_1 + m_2 + \dots + m_n = t$$

$$m_r = v_r$$

$$m_s = v_s$$
(7)

The algorithm given below explores the solution spaces of a polynomial number of forms of Z to construct the characteristic membership vector m for some subset S of A summing to c if one exists. The convention for the determining solution space of $Z(A, t, r, v_r, s, v_s)$ is to first form an equivalent set representation A' by interchanging index 3 of a with index r, interchanging index 4 with index r in r in r in r in r index r in r in r index r in r in r index r in r in

$$\begin{pmatrix}
a_1 & a_2 & a_r & a_s & \cdots & a_n \\
1 & 1 & 1 & 1 & \cdots & 1 \\
0 & 0 & 1 & 0 & \cdots & 0 \\
0 & 0 & 0 & 1 & \cdots & 0
\end{pmatrix}
\begin{pmatrix}
m_1 \\
m_2 \\
m_r \\
m_s \\
\vdots \\
m_n
\end{pmatrix} = \begin{pmatrix}
c \\
t \\
v_r \\
v_s
\end{pmatrix}$$
(8)

To represent the solution space of $Z(A', t, 3, v_r, 4, v_s)$, the outlined algorithm follows the convention of expressing solution space with respect to a particular solution of the system and any linear combination of the null space of the multiplier of $Z(A', t, 3, v_r, 4, v_s)$.

$$m' = m'_p + dN\begin{pmatrix} a_1 & a_2 & a_r & a_s & \cdots & a_n \\ 1 & 1 & 1 & 1 & \cdots & 1 \\ 0 & 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 0 & 1 & \cdots & 0 \end{pmatrix}$$
(9)

The followed solution convention is simple forward elimination, followed by back-substitution. The particular solution is chosen such that all free variables (the rank of the multiplier of the system is 4; given its representation, the free variables are a'_5 , . . . a'_n) are assumed to be zero. The null space is then composed of n-4 special solutions each respectively assuming one unique m_i , $i = 5, \ldots, n$ to be 1, all other m_j to be 0. This convention then allows m' for any A' to be expressed as follows:

$$m' = \begin{pmatrix} b_1 \\ b_2 \\ v_r \\ v_s \\ 0 \\ \vdots \\ 0 \end{pmatrix} + d_1 \begin{pmatrix} k_{1,1} \\ k_{2,1} \\ 0 \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} + d_2 \begin{pmatrix} k_{1,2} \\ k_{2,2} \\ 0 \\ 0 \\ 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} + \dots + d_{n-4} \begin{pmatrix} k_{1,n-4} \\ k_{2,n-4} \\ 0 \\ 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix}$$
(10)

The prescribed convention thus provides a representation of the solution space that may be explored to determine if a valid m' satisfying subset sum exists for a given $Z(A', t, 3, v_r, 4, v_s)$: each vector of the null space can be contributed to m' one or zero times, and, together, the sum of the particular

solutions and applicable vectors of the null space will take b_1 and b_2 (the non-zero and non-one values in the particular solution unique to the values of the first two elements and the chosen v_r and v_s) to 0 or 1. For any solution space for which this applies, the resulting m' will be a vector of 0s and 1s encoding the membership of S. Due to the form found as a result of convention of this method, the first four elements of A' are referred to as the current window, the first two elements are the balance elements, and the second two elements are the pivot elements.

The algorithm below attempts to reveal and exploit solution space properties which may exist universally among all sets for **some** window configuration if the set satisfies subset sum to form m'. If the found properties are, in fact, universal, the algorithm outlined is an exact, general-case algorithm for the subset sum problem.

In order to determine, view, and exploit these properties, the algorithm utilizes a construct which will be called a directional contribution table. A directional contribution table is a tabulation of the contributions of the elements of the null space towards bringing the balance values of the particular solution towards 0 or 1. A directional contribution table D tabulates the contribution of a given vector in the null space of Z towards paired balance targets and is defined with respect to the solution space (as expressed in the previously given convention) of a system S(Z) and given target values of the balance points within the particular solution.

$$D(S(Z), t_{1}, t_{2}):$$

$$\begin{bmatrix} \frac{k_{1,1}}{t_{1}-b_{1}} & \frac{k_{1,2}}{t_{1}-b_{1}} & \cdots & \frac{k_{1,n-4}}{t_{1}-b_{1}} \\ \frac{k_{2,1}}{t_{2}-b_{2}} & \frac{k_{2,2}}{t_{2}-b_{2}} & \cdots & \frac{k_{2,n-4}}{t_{2}-b_{2}} \\ \frac{k_{1,1}}{t_{1}-b_{1}} - \frac{k_{2,1}}{t_{2}-b_{2}} & \frac{k_{1,2}}{t_{1}-b_{1}} - \frac{k_{2,2}}{t_{2}-b_{2}} & \cdots & \frac{k_{1,n-4}}{t_{1}-b_{1}} - \frac{k_{2,n-4}}{t_{2}-b_{2}} \end{bmatrix}$$

$$(11)$$

For a given set A satisfying subset sum, there appears to exit a window configuration for which in $Z(A', t, 3, v_r, 4, v_s)$ and $D(S(Z), t_1, t_2), t, v_r, v_s, t_1$, and t_2 apply to an extant S, characterized by specific properties within D. The following (possibly non-exhaustive) property has been determined and is employed by the algorithm to determine m' encoding S within A':

(A) If the length of S is 4, m' is the exact solution of S(Z) when the elements of the window are the elements of S and membership variables are set appropriately. If the length of S is 5, m' is the exact solution of S(Z) plus the vector representing the column of D for which $D_{1,i}$ and $D_{1,i}$ are both 1 when membership variables are set appropriately. In all other cases for a set A of length greater than four, m' may be formed by taking the vectors represented by each column i for which the absolute value of $D_{3,i}$ is less than one.

2.2 Justification of Properties

Within the parameters of the convention listed above, manual algebraic reduction in the general case yields the following closed forms for the variables of

the particular solution of S(Z), the null space of S(Z), and values within the directional contribution table:

$$\beta = a_1 - a_2 \tag{12}$$

$$b_1 = \frac{c - v_s a_s - v_r a_r - a_2 (t - v_r - v_s)}{\beta}$$
 (13)

$$b_2 = \frac{a_1(t - v_r - v_s) + v_s a_s + v_r a_r - c}{\beta}$$
 (14)

$$k_{1,i} = \frac{a_2 - a_{i+4}}{\beta} \tag{15}$$

$$k_{2,i} = \frac{a_{i+4} - a_1}{\beta} \tag{16}$$

$$\delta_1 = t_1(a_1 - a_2) + v_s(a_s - a_2) + v_r(a_r - a_2) + a_2t - c \tag{17}$$

$$\delta_1 = a_2(t - t_1 - v_r - v_s) - (c - t_1 a_1 - v_r a_r - v_s a_s) \tag{18}$$

$$\delta_2 = t_2(a_1 - a_2) - v_s(a_s - a_1) - v_r(a_r - a_1) - a_1 t + c \tag{19}$$

$$\delta_2 = (c - t_2 a_2 - v_r a_r - v_s a_s) - a_1 (t - t_2 - v_r - v_s) \tag{20}$$

$$D_{1,i} = \frac{a_2 - a_{i+4}}{\delta_1} \tag{21}$$

$$D_{2,i} = \frac{a_{i+4} - a_1}{\delta_2} \tag{22}$$

$$D_{3,i} = \left| \frac{\delta_1 a_1 + \delta_2 a_2 - a_{i+4} (\delta_1 + \delta_2)}{\delta_1 \delta_2} \right| \tag{23}$$

Property (A) is to say that an instance of a subset S of A summing to c exists under the following constraints:

(1) The length of S is four, and when the elements of S are set as the window of A, m is found encoding S within A.

Assume all members of S are the current window of A', and set membership to $t_1=1,\,t_2=1,\,v_r=1,\,v_s=1.$

$$c = a_1 + a_2 + a_r + a_s$$

$$b_1 = \frac{c - a_s - v_r - 2a_2}{a_1 - a_2}$$

$$b_1 = \frac{a_1 + a_2 - 2a_2}{a_1 - a_2}$$
$$b_1 = 1$$

$$b_2 = \frac{2a_1 + v_r + v_s - a_1 - a_2 - v_r - v_s}{a_1 - a_2}$$
$$b_2 = \frac{a_1 - a_2}{a_1 - a_2}$$
$$b_2 = 1$$

Thus, the particular solution of S(Z) for this configuration is

$$\begin{pmatrix} b_1 \\ b_2 \\ m_r \\ m_s \\ \vdots \\ m_n \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \\ \vdots \\ 0 \end{pmatrix}$$

encoding S and showing (1) to be true. The above also extends to show that (1) is true for any instance of S having length less than four for which all elements of S are within the window of A' and membership is set appropriately.

(2) The length of S is 5, and when all but one element of S is set as the window of A', m' encoding S within A' is found by adding to the particular solution the vector of the null space corresponding to $D_{1,r} = D_{2,r} = 1$.

Assume four members of S are the current window of A, and set membership to $t_1=1,\,t_2=1,\,v_r=1,\,v_s=1.$

$$c = a_1 + a_2 + a_r + a_s + a_k$$

Solving for the values in the directional contribution table:

$$D_{1,k-4} = \frac{a_2 - a_k}{a_1 a_4 + a_3 + 2a_2 - c}$$

$$D_{1,k-4} = \frac{a_2 - a_k}{a_2 - a_k} = 1$$

$$D_{2,k-4} = \frac{a_k - a_1}{c - a_2 - a_4 - a_3 - 2a_1}$$

$$D_{2,k-4} = \frac{a_k - a_1}{a_k - a_1} = 1$$

showing (2) to be true. The above may also be generalized for any S having length less than or equal to five for which all but one element of the subset exists within the window and membership is set appropriately.

(3) In all other cases, S of length t > 5 of A exists if and only if there exists a window configuration and membership assignment t_1 , t_2 , v_r , v_s such that there

exists a set I, length t'=t - t_1 - t_2 - v_r - v_s satisfying $D_{3,I_j}<1$ and $\Sigma D_{1,I_j}=\Sigma D_{2,I_j}=1$ for all j in the range of the length of I.

Suppose that such a window exists. It is given that

$$\sum_{j=1}^{t'} D_{1,I_j} = 1$$

and

$$\sum_{i=1}^{t'} D_{2,I_j} = 1$$

Using the closed forms derived earlier in this section:

$$\sum_{j=1}^{t'} D_{1,I_j} = 1 = \sum_{j=1}^{t'} \frac{K_{1,I_j}}{t_1 - b_1}$$

$$t_1 - b_1 = \sum_{j=1}^{t'} K_{1,I_j}$$

$$b_1 + \sum_{j=1}^{t'} K_{1,I_j} = t_1$$

$$\sum_{j=1}^{t'} D_{2,I_j} = 1 = \sum_{j=1}^{t'} \frac{K_{2,I_j}}{t_2 - b_2}$$

$$t_2 - b_2 = \sum_{j=1}^{t'} K_{2,I_j}$$

$$b_2 + \sum_{j=1}^{t'} K_{2,I_j} = t_2$$

Repeating the same process as was used in the justification of (1), m' is derived encoding S within A':

$$m' = x_p + \sum_{j=1}^{t'} N(A)_{I_j} = \begin{pmatrix} t_1 \\ t_2 \\ v_r \\ v_s \\ \vdots \end{pmatrix} \quad m_k, k > 4 = \begin{cases} 1, & \text{if } k - 4 \in I \\ 0, & \text{if } k - 4 \notin I \end{cases}$$
$$S = \begin{cases} A'_k \mid m'_k = 1 \\ \mid S \mid = t_1 + t_2 + v_r + v_s + t' = t \end{cases}$$

Thus, if such a window exists, a S of A of length t exists and is encoded by the derivable m.

Suppose S of length t of A exists. Let

$$S = \{s_1, s_2, ..., s_t\}$$

$$S' = A \setminus S = \{a_l \mid a_l \notin S\}$$

$$I = \{i - 4 \mid A'_i \in S, i > 4\}$$

$$X = \{i - 4 \mid A'_i \notin S, 4 < i < max(I)\}$$

Call window $W=\{\ A'_1,\ A'_2,\ A'_3,\ A'_4\ \}$ and define membership of the window as follows:

$$t_{1} = \begin{cases} 1, & \text{if } w_{1} \in S \\ 0, & \text{if } w_{1} \notin S \end{cases} t_{2} = \begin{cases} 1, & \text{if } w_{2} \in S \\ 0, & \text{if } w_{2} \notin S \end{cases}$$
$$v_{r} = \begin{cases} 1, & \text{if } w_{3} \in S \\ 0, & \text{if } w_{3} \notin S \end{cases} v_{s} = \begin{cases} 1, & \text{if } w_{4} \in S \\ 0, & \text{if } w_{4} \notin S \end{cases}$$

Under this membership assignment,

$$\sum_{j=1}^{t'} D_{1,I_j} = \sum_{j=1}^{t'} \frac{a_2 - a_{I_j+4}}{\delta_1} = \frac{1}{\delta_1} (t'a_2 - \sum_{j=1}^{t'} a_{I_j+4})$$

$$= \frac{1}{\delta_1} (a_2t - a_2t_2 - a_2v_r - a2v_s - (c - a_1t_1 - a_2v_2 - a_rv_r - a_sv_s))$$

$$= \frac{1}{\delta_1} (t_1(a_1 - a_2) + v_s(a_s - a_2) + v_r(a_3 - a_2) + a_2t - c)$$

Likewise,

$$\sum_{j=1}^{t'} D_{2,I_j} = \sum_{j=1}^{t'} \frac{a_{I_j+4} - a_1}{\delta_2}$$

$$= \frac{1}{\delta_2} (c - t_1 a_1 - a_2 t_2 - v_r a_r - v_s a_s - a_1 t + a_1 t_1 + a_2 t_2 + a_r v_r + a_s v_s)$$

$$= 1$$

Thus, if S of A exists, $\Sigma D_{2,I_j} = \Sigma D_{1,I_j} = 1$. By (3), all D_{3,I_j} for the window must satisfy

$$-1 < \frac{\delta_1 a_1 + \delta_2 a_2 - a_{I_j + 4}}{\delta_1 \delta_2} < 1$$

Expanding the variables and simplifying, for $\delta_1 \delta_2 > 0$,

$$a_{I_j+4}(\delta_1+\delta_2) - \delta_1\delta_2 < (a_2-a_1)(c-t_1a_1 - t_2a_2 - v_ra_r - v_sa_s) < a_{I_j+4}(\delta_1+\delta_2) + \delta_1\delta_2$$

and for $\delta_1\delta_2 < 0$,

$$a_{I_i+4}(\delta_1+\delta_2)+\delta_1\delta_2<(a_2-a_1)(c-t_1a_1-t_2a_2-v_ra_r-v_sa_s)< a_{I_i+4}(\delta_1+\delta_2)-\delta_1\delta_2$$

It can be shown that the above is satisfied for some window configuration by showing that a tighter contained constraint is also satisfied for some window:

$$-t' < \sum_{j=1}^{t'} D3, I_j < t',$$

$$D_{1,I_i} > 0$$
 and $D_{2,I_i} > 0$

Summing the D_{3,I_j} ,

$$\begin{split} \sum_{j=1}^{t'} D_{3,I_j} &= \frac{1}{\delta_1 \delta_2} \sum_{j=1}^{t'} (\delta_1 a_1 + \delta_2 a_2 - a_{I_j+4} (\delta_1 + \delta_2)) \\ &= t' (\frac{a_1 \delta_1 + a_2 \delta_2}{\delta_1 \delta_2}) - \frac{c - t_1 a_1 - t_2 a_2 - v_r a_3 - v_s a_4}{\delta_1 \delta_2} \\ &= \frac{t' (a_1 - a_2) (t_1 a_1 + t_2 a_2 + v_r a_r + v_s a_s - c) + (t_1 a_1 + t_2 a_2 + v_r a_r + v_s a_s + c)}{\delta_1 \delta_2} \\ &= \frac{(t_1 a_1 + t_2 a_2 + v_r a_r + v_s a_s - c) (t' (a_1 - a_2) + 1)}{\delta_1 \delta_2} \\ &= \frac{(t_1 a_1 + t_2 a_2 + v_r a_r + v_s a_s - c) (t' a_1 - t' a_2 + 1)}{(a_2 t' + a_1 t_1 + a_r v_r + a_s v_s - c) (c - a_1 t' - a_2 t_2 - a_r v_r - a_s v_s)} \\ &= \frac{(c - t_1 a_1 - t_2 a_2 - v_r a_r - v_s a_s) (t' a_1 - t' a_2 + 1)}{(c - a_2 t' - a_1 t_1 - a_r v_r - a_s v_s) (c - a_1 t' - a_2 t_2 - a_r v_r - a_s v_s)} \end{split}$$

The numerator and denominator of this expression are both polynomial with respect to t' (the numerator is of degree 1, and the denominator of degree 2); given that t' > 1, all coefficients of t' in the numerator appear as a product in the denominator, and all non-t' constants in the numerator appear in the denominator, the result is bounded by (-t',t').

Given that the above is bounded, vewing the second constraint, assert that

$$D_{1,I_j} > 0$$
 and $D_{2,I_j} > 0$

Thus

$$\frac{a_2-a_{I_j+4}}{\delta_1}>0$$
 and $\frac{a_{I_j+4}-a_1}{\delta_2}>0$

meaning

$$\begin{cases} a_2 > a_{I_j+4}, & \text{if } \delta_1 > 0 \\ a_2 < a_{I_j+4}, & \text{if } \delta_1 < 0 \end{cases} and \begin{cases} a_1 < a_{I_j+4}, & \text{if } \delta_2 > 0 \\ a_1 > a_{I_j+4}, & \text{if } \delta_2 < 0 \end{cases}$$

Noting that

$$\begin{cases} \delta_1 > 0, & \text{if } a_2 > \frac{c - t_1 a_1 - v_r a_r - v_s a_s}{t - t_1 - v_r - v_s} \\ \delta_1 < 0, & \text{if } a_2 < \frac{c - t_1 a_1 - v_r a_r - v_s a_s}{t - t_1 - v_r - v_s} \\ \delta_2 > 0, & \text{if } a_1 < \frac{c - t_2 a_2 - v_r a_r - v_s a_s}{t - t_2 - v_r - v_s} \\ \delta_2 < 0, & \text{if } a_1 > \frac{c - t_2 a_2 - v_r a_r - v_s a_s}{t - t_2 - v_r - v_s} \end{cases},$$

it can be seen that a window configuration composed of the minima and/or maxima (and/or points whose values are set according to these) with membership appropriately set of s_i can be made among the set (or possibly created) to satisfy both these (and thus the primary) constraints.

Additionally, note that if an initial ordering of the set is enforced in which the set is ordered by the absolute value of its elements (increasing or decreasing), enumerating and swapping all possible windows of A in a pair-wise order (1 with 2, 1 with 3, . . ., 2 with 3; followed by pair 1 with pair 2 . . . pair 2 with pair 3) creates such a set ordering for which a window may satisfy

$$D_{3,X_k} > 1$$
 or $D_{3,X_k} < -1$

in the same window assignment as the previous constraint.

This shows, then, that if S of length t of A exists, there exists a window configuration and membership assignment t_1 , t_2 , v_r , v_s such that there exists a set I, length $t' = t - t_1 - t_2 - v_r - v_s$ satisfying $D_{3,I_j} < 1$ and $\Sigma D_{1,I_j} = \Sigma D_{2,I_j} = 1$ for all j in the range of the length of I.

2.3 Algorithm

Subset Sum Algorithm. Input: set of constants A and constant c. Output: non-empty subset S of A summing to c if such a subset exists.

```
1: procedure SubsetSum(A,c)
       sort A by descending absolute value
 2:
       S \leftarrow PARTIALSUBSETSUM(A, c)
 3:
       if length(S) > 0 then
 4:
          return S
 5:
       end if
 6:
      sort A by descending signed value
 7:
 8:
       S \leftarrow PARTIALSUBSETSUM(A, c)
       if length(S) > 0 then
 9:
          return S
10:
       end if
11:
       sort A by ascending absolute value
12:
       S \leftarrow PARTIALSUBSETSUM(A, c)
13:
       if length(S) > 0 then
14:
          return S
15:
       end if
16:
       sort A by ascending signed value
17:
       S \leftarrow PARTIALSUBSETSUM(A, c)
18:
```

```
if length(S) > 0 then
19:
          return S
20:
       end if
21:
22:
       return empty list
23: end procedure
24: procedure PartialSubsetSum(A, c)
       Initialize empty list S.
25:
       Initialize empty list pairs.
26:
       Initialize empty list windows.
27:
       runningSum \leftarrow 0
28:
   ▷ Capture case in which S is composed of one or all elements of A.
       for each x in range [1, length(A)] do
29:
          if A_x == c then
30:
              Append A_x to S
31:
              return S
32:
          end if
33:
          runningSum \leftarrow runningSum + A_{x}
34:
          if runningSum == c then
35:
              for each y in range [1, x] do
36:
                  Append A_v to S.
37:
              end for
38:
              return S
39:
40:
          end if
       end for
41:
   ▷ Capture case in which S is composed of all but one element of A.
       for each x in range [1, length(A)] do
42:
          if runningSum - A_x == c then
43:
              for each y in range [1, length(A)] do
44:
                 if y != x then
45:
                     Append A_v to S.
46:
                  end if
47:
              end for
48:
49:
              return S
          end if
50:
       end for
51:
       if length(A) <5 then
52:
          return S
53:
       end if
54:
```

▷ Construct list of pairs existing within the set; capture case in which S is

```
composed of two elements.
       for each i in range [1, length(A)] do
55:
          for each j in range [i+1, length(A)] do
56:
              if A_i + A_j == c then
57:
                 Append A_i to S.
58:
                 Append A_i to S.
59:
                 return S
60:
              end if
              if A_i != A_i then
62:
                 Append \{A_i, A_i\} to pairs.
63:
              end if
64:
          end for
65:
       end for
66:
   ▶ At this point, all cases for sets of length less than or equal to four have
   been captured. Formulate all possible window configurations of A (order n^4
   in number).
       for each i in range [1, length(pairs)] do
67:
          for each j in range [i+1, length(pairs)] do
68:
69:
              if length(pairs_i intersect pairs_i) == 0 then
                 Append {pairs_{i,1}, pairs_{i,2}, pairs_{i,1}, pairs_{i,2}} to windows.
70:
              end if
71:
          end for
72:
       end for
73:
   ▶ Apply the matrix-based strategy whose convention is given in 2.1.
       for each i in range [1, length(windows)] do
74:
          swap the elements of windows_i into the first four positions of A.
75:
          for two iterations do
76:
              for each t in range[3, length(A) - 1] do
77:
                 CONSTRAIN(A, S, c, t)
78:
                 if length(S) > 0 then
79:
                     return S
80:
                 end if
81:
              end for
82:
              swap the second element with the third
83:
          end for
84:
       end for
85:
       return S
87: end procedure
   ▷ CONSTRAIN constrains a set to the parameterized linear constraints of
   Z, updating the subset list if the characteristic subset membership vector
   can be formed.
88: procedure Constrain(A,S,c,t)
       m \leftarrow BALANCE(A, c, t, 1, 1)
89:
```

```
if m != \text{null then}
90:
            for each index x of m for which m_x == 1 \text{ do}
91:
92:
                Append A_x to S
            end for
93:
94:
            return
        end if
95:
        m \leftarrow BALANCE(A, c, t, 0, 0)
96:
        if m != \text{null then}
97:
            for each index x of m for which m_x == 1 \text{ do}
98:
                Append A_x to S
99:
             end for
100:
             return
101:
         end if
102:
         m \leftarrow BALANCE(A, c, t, 1, 0)
103:
104:
         if m != \text{null then}
             for each index x of m for which m_x == 1 \text{ do}
105:
106:
                 Append A_x to S
107:
             end for
             return
108:
109:
         end if
         m \leftarrow BALANCE(A, c, t, 0, 1)
110:
         if m != \text{null then}
111:
             for each index x of m for which m_x == 1 do
112:
                 Append A_x to S
113:
             end for
114:
115:
             return
         end if
116:
117: end procedure
    \rhd BALANCE attempts to form m from the solution space of Z
118: procedure BALANCE(A, c, t, v_r, v_s)
         m \leftarrow TEST(A, c, t, 1, 1, v_{\rm r}, v_{\rm s})
119:
         if m is not null then
120:
121:
             return m
         end if
122:
         m \leftarrow TEST(A, c, t, 0, 0, v_{\mathrm{r}}, v_{\mathrm{s}})
123:
         if m is not null then
124:
             return m
125:
         end if
126:
         m \leftarrow TEST(A, c, t, 1, 0, v_{\rm r}, v_{\rm s})
127:
         if m is not null then
128:
             \mathbf{return}\ \mathbf{m}
129:
         end if
130:
         m \leftarrow TEST(A, c, t, 0, 1, v_r, v_s)
131:
         if m is not null then
132:
             return m
133:
```

```
end if
134:
         \mathbf{return} \,\, \mathrm{null} \,\,
135:
136: end procedure
137: procedure \text{TEST}(A, c, t, t_1, t_2, v_r, v_s)
         Initialize empty list S.
138:
         Initialize empty list tmp.
139:
         if t_1 equals 1 then
140:
              append A_1 to S
141:
         end if
142:
         if t_2 equals 1 then
143:
              append A<sub>2</sub> to S
144:
         end if
145:
146:
         if v_r equals 1 then
              append A<sub>3</sub> to S
147:
148:
         end if
         if v<sub>s</sub> equals 1 then
149:
              append A<sub>4</sub> to S
150:
151:
         end if
         testSum \leftarrow 0
152:
         for each s in S do
153:
              testSum \leftarrow testSum + s
154:
         end for
155:
         if testSum equals c then
156:
157:
              return S
         end if
158:
         sumTop \leftarrow 0
159:
         sumBottom \leftarrow 0
160:
         for each x in range [1, length(A)-4] do
161:
162:
              calculate D_{1,x}
             calculate D_{2,x}
163:
              calculate D_{3,x}
164:
             if 0 < D_{3,x} < 1 then
165:
                  \mathbf{append} \ \mathbf{x} \ \mathbf{to} \ \mathbf{tmp}
166:
                  sumTop \leftarrow sumTop + D_{1,x}
167:
168:
                  sumBottom \leftarrow sumBottom + D_{2,x}
                  if sumTop equals 1 and sumBottom equals 1 then
169:
                      for each i in tmp do
170:
                          append A<sub>i+4</sub> to S
171:
                      end for
172:
                      return S
173:
                 end if
174:
              end if
175:
         end for
176:
         return null
177:
```

2.4 Complexity

This algorithm satisfies the classification of strongly polynomial in that (1) the number of operations in the arithmetic model of computation is bounded in a polynomial in the number of integers in the input and (2) the space used by the algorithm is polynomial in the size of the input.¹

- (1) The TEST procedure requires one linear pass through input set A, requiring a constant number of arithmetic computations at each element to calculate the directional contribution table values, followed by at most n appendages to S. The number of arithmetic computations of TEST is O(n). The BAL-ANCE procedure requires four repetitions of the TEST procedure: the number of arithmetic computations required by BALANCE is O(n). The CONSTRAIN procedure requires four repetitions of the BALANCE procedure, followed by at most n appendages to S. The number of arithmetic computations required by CONSTRAIN is O(n). The SUBSETSUM procedure requires n^4 iterations of n iterations of the CONSTRAIN procedure. The number of arithmetic computations required by SUBSETSUM is $O(n^6)$.
- (2) The TEST procedure in one execution requires the space taken by A, the space taken by S which is less than or equal to the length of A, and a finite, constant number of integers of bit length less than or equal to two times the number of bits of the maximum element of A: TEST is O(n) in required space, where n is the number of bits in the input set. BALANCE requires the same space as TEST. CONSTRAIN requires the same space as BALANCE. SUBSETSUM requires storage of all possible length-4 subsets of A, meaning that the total space required by the algorithm is bounded by $O(n^4)$.

3 Discussion

3.1 Accuracy Results

A simple implementation of this algorithm was written (hosted on Github at https://github.com/ad-alston/PolynomialSubsetSum) in Java using 64-bit long integers and tested for accuracy to reveal efficacy of the algorithm.

To test the accuracy of the algorithm, a driver was implemented which generates random sets of integers of a given length n, having range $-2n^2$ to $2n^2$ and tests whether the set satisfies subset sum for c equal to 0 using (a) the conventional exponential algorithm and (b) the SUBSETSUM routine for n permutations of the set. Under the parameters of this test, failure occurs when the output of (b) differs from that of (a).

¹Definition of strongly polynomial algorithm: Grotschel, Martin; Laszlo Lovasz, Alexander Schrijver (1988). "Complexity, Oracles, and Numerical Computation". Geometric Algorithms and Combinatorial Optimization. Springer.

For each n from 1 to 20, 1,000,000 such sets were generated and used to test the implementation of the algorithm. Following all 20,000,000 trials, the success rate was 100%, exhibiting precisely 0 failures.

3.2 Reduction to an Approximation

The method can be reduced to an O(n⁴) approximation method by only using the subset of possible window configurations represented by shifting the entire set to the left (wrapping the element of the first index to the last position in the set) n times and repeating the CONSTRAIN procedure. Performing the same testing as was performed on the exact method as was outlined in section 2.2, yielding a success rate of 99.95% over 20,000,000 trials.

4 Conclusion

An algorithm has been provided which reduces the conditions under which a given set satisfies the stipulations of the subset sum proposition to a set of linear relationships, answering question of whether a set satisfies subset sum may be answered in a polynomial number of steps. Following justification, implementation, and testing of this algorithm, 20,000,000 trials using random inputs of length 1 to 20,1,000,000 trials each, reveal 100% accuracy with exactly 0 failures, in alignment with general-case applicability of this algorithm.