Bounding the Sum of Square Roots via Lattice Reduction *

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Abstract. Let k and n be positive integers. Define R(n,k) to be the minimum positive value of

$$|e_i\sqrt{s_1}+e_2\sqrt{s_2}+\cdots+e_k\sqrt{s_k}-t|$$

where s_1, s_2, \dots, s_k are positive integers no larger than n, t is an integer and $e_i \in \{1, 0, -1\}$ for all $1 \le i \le k$. It is important in computational geometry to determine a good lower and upper bound of R(n, k). In this paper we show that this problem is closely related to the shortest vector problem in certain integral lattices and present an algorithm to find lower bounds based on lattice reduction algorithms. Although we can only prove an exponential time upper bound for the algorithm, it is efficient for large k when an exhaustive search for the minimum value is clearly infeasible. It produces lower bounds much better than the root separation technique does. Based on numerical data, we formulate a conjecture on the length of the shortest nonzero vector in the lattice, whose validation implies that our algorithm runs in polynomial time and the problem of comparing two sums of square roots of small integers can be solved in polynomial time. As a side result, we obtain constructive upper bounds for R(n, k) when n is much smaller than 2^{2k} .

1 Introduction

Comparing sums of square roots of integers is a famous open problem in computational geometry and numerical analysis. It arises when we need to compare the length of two polygonal paths in a Euclidean space. The problem takes another form when one compares a sum of square roots with an integer. This problem is not known to be in NP. In fact, PSPACE is the smallest well studied complexity class that provably contains this problem [11]. In practice, however, it can usually be solved quickly.

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Definition 1. Define $r_1(n,k)$ to be the minimum positive value of

$$\left|\sqrt{s_1} + \sqrt{s_2} + \dots + \sqrt{s_{\lfloor k/2 \rfloor}} - \sqrt{s_{\lfloor k/2 \rfloor+1}} - \dots - \sqrt{s_k}\right|$$

where s_1, s_2, \dots, s_k are positive integers less than or equal to n. Define $r_2(n, k)$ to be the minimum positive value of

$$|\sqrt{s_1} + \sqrt{s_2} + \dots + \sqrt{s_k} - t|$$

where s_1, s_2, \dots, s_k are positive integers less than or equal to n and t is an integer. Define R(n,k) to be the minimum positive value of

$$|e_1\sqrt{s_1} + e_2\sqrt{s_2} + \cdots + e_k\sqrt{s_k} - t|$$

where s_1, s_2, \dots, s_k are positive integers no larger than n, t is an integer and $e_i \in \{1, 0, -1\}$ for all $1 \le i \le k$.

For example, we have that

$$r_1(3,3) = \left| \sqrt{3} - \sqrt{1} - \sqrt{1} \right| \approx 0.268,$$

$$r_2(3,3) = \left| \sqrt{3} + \sqrt{2} + \sqrt{1} - 4 \right| \approx 0.146$$

and

$$R(3,3) = \left| \sqrt{2} + \sqrt{2} - \sqrt{3} - 1 \right| \approx 0.096.$$

It is easy to see that $R(n,k) \leq r_1(n,k)$ and $R(n,k) \leq r_2(n,k)$. Since we are mainly interested in the lower bounds, we shall be concentrating on R(n,k). If one can show that $R(n,k) \geq 1/2^{\operatorname{poly}(k \log n)}$, then comparing the sum of k square roots of integers no larger than n can be done in time polynomial in k and $\log n$.

The problem of sum of square roots has recently attracted attention. First it is the main barrier to accurately classify some of the most fundamental computational problems in Euclidean space, such as the shortest path problem, the minimum spanning tree problem and the traveling salesman problem [5]. Secondly it is the simplest among the problems of the sign determination of algebraic numbers of high degree. Thirdly it has been used to show hardness of problems in other area such as approximation of 3-player Nash equilibrium [4].

1.1 Previous work

The zeroes of

$$f(s_1, \dots, s_k) = \sqrt{s_1} + \sqrt{s_2} + \dots + \sqrt{s_{\lfloor k/2 \rfloor}} - \sqrt{s_{\lfloor k/2 \rfloor + 1}} - \dots - \sqrt{s_k}$$

form a surface in \mathbf{R}_{+}^{k} , where s_{i} is nonnegative for $1 \leq i \leq k$. To bound $r_{1}(n, k)$ we consider an equivalent problem: how near to the surface can an integral point of absolute height no larger than n get and still miss? In general finding a near-miss integral point to a surface is a very hard problem. Elkies [3] presented algorithms

for these kind of problems with time complexity better than an exhaustive search. As an example, he showed how to find integral points near the curve $x^3 - y^2 = 0$. It seems hard to generalize his algorithm to the sum of square roots problem as the dimension is much higher.

The known lower bound comes from the root separation technique (for instance see [2] and [1]), which shows that

$$r_1(n,k) \ge \max\left((k\sqrt{n})^{-2^{k-1}}, (k\sqrt{n})^{-2^{\pi(n)-1}}\right)$$

where $\pi(n)$ is the number of primes no larger than n, and

$$R(n,k) \ge \max\left((2k\sqrt{n})^{-2^{k-1}}, (2k\sqrt{n})^{-2^{\pi(n)-1}}\right).$$

For example, it gives

$$R(165, 100) \ge \left(200\sqrt{165}\right)^{-2^{37}} \approx 10^{-468635490828}.$$
 (1)

The lower bound is too small when k and n are large. However no significantly better lower bound has been reported as far as we are aware.

Qian and Wang [9] presented an upper bound for $r_1(n,k)$ based on the inequality:

$$\left| \sum_{i=0}^{k} {k \choose i} (-1)^i \sqrt{t+i} \right| \le \frac{1 * 3 * 5 * \dots * (2k-3)}{2^k t^{k-\frac{1}{2}}}.$$

Note that $\binom{k}{i}$ can be as large as $\binom{k}{k/2} \ge 2^k/k$. For any fixed positive integer k, taking

$$n = 2^{2k}t \ge \max_{0 \le i \le k} {k \choose i}^2 (t+i), \tag{2}$$

we have

$$r_1(n,k) \le \left| \sum_{i=0}^k (-1)^i \sqrt{\binom{k}{i}^2 (t+i)} \right|$$

$$\le \frac{1 * 3 * 5 * \dots * (2k-3) * 2^{2k^2 - 2k}}{(2^{2k}t)^{k - \frac{1}{2}}}$$

$$= \frac{C_k}{n^{k - \frac{1}{2}}},$$

where $C_k = 1*3*5*\cdots*(2k-3)*2^{2k^2-2k}$ is a constant depending only on k. By (2), we have that Qian and Wang's result only applies when n is much greater than 2^{2k} . In particular it does not give a meaningful bound when k = 100 and $n \le 2^{200} \approx 10^{60}$.

1.2 Our results

We present a method to numerically bound R(n,k) from below based on lattice reduction. Our method is efficient for large k and n such as k=100 and n=165, where an exhaustive search is clearly infeasible. The lower bounds we obtain are much better than the root separation bound. See Table 1 that compares our lower bounds with that of the root separation technique.

Table 1. Comparing	our lower	bounds wi	th those of	f the root	separation	technique

R(n,k)	Root Separation Technique	Lattice Reduction Technique
	Lower Bound	Lower Bound
R(15, 10)	10^{-60}	10^{-20}
R(33, 20)	10^{-2418}	10^{-50}
R(47,30)	10^{-42832}	
R(66, 40)	$10^{-368688}$	10^{-120}
R(82,50)	$10^{-6201084}$	
R(97,60)	$10^{-51549123}$	
R(113,70)	$10^{-1703312763}$	10^{-240}
R(131, 80)	$10^{-7006714363}$	10^{-290}
R(146, 90)	$10^{-28668468036}$	10^{-335}
R(165, 100)	$10^{-468635490828}$	10^{-390}

Define $[x] = \lfloor x+1/2 \rfloor$ and $\{x\} = x-[x]$. We call an integer b square-free if there does not exist an integer a>1 such that $a^2|b$. We denote the i-th square free integer starting from 2 by $\sigma(i)$. It is known that the square roots of distinct square-free integers are linearly independent over \mathbf{Q} and $\sigma(i)$ satisfies (see [8])

$$\sigma(i) = \pi^2 i/6 + O(\sqrt{i}). \tag{3}$$

Let s_1, s_2, \dots, s_k be the distinct square-free integers no smaller than 2. Let N be a positive integer. Our method is based on studying the integral lattice generated by the following k+1 vectors in \mathbf{R}^{k+1} ,

$$\mathbf{v}_0 = (N, 0, 0, 0, \cdots, 0)$$

$$\mathbf{v}_1 = ([N\sqrt{s_1}], 1, 0, 0, \cdots, 0)$$

$$\mathbf{v}_2 = ([N\sqrt{s_2}], 0, 1, 0, \cdots, 0,)$$

$$\mathbf{v}_3 = ([N\sqrt{s_3}], 0, 0, 1, \cdots, 0,)$$

$$\vdots$$

$$\mathbf{v}_k = ([N\sqrt{s_k}], 0, 0, 0, \cdots, 1).$$

We denote the lattice by $L_{s_1,s_2,\dots,s_k}(N)$. If $s_1=2,s_2=3,\dots,s_k=\sigma(k)$ are the consecutive square free integers, we will simply use L(k,N) to denote the lattice.

In this paper, we are mainly concerned with $R(\sigma(k), k)$, since a good lower bound on $R(\sigma(k), k)$ can imply a good lower bound on R(n, k) whenever $n = k^{O(1)}$.

Lemma 1. If $R(\sigma(k), k) \ge 1/2^{\text{poly}(k)}$, then $R(n, k) \ge 1/2^{\text{poly}(nk)}$.

Proof. If $n < \sigma(k)$, then $R(n,k) \ge R(\sigma(k),k)$. If $n > \sigma(k)$, by (3), there exists $k' = 6n/\pi^2 + O(\sqrt{n})$ such that $\sigma(k') \le n < \sigma(k'+1)$, then

$$R(n,k) \ge R(\sigma(k'), k') = 1/2^{\text{poly}(nk)}.$$

The following theorem relates the shortest vector of L(k, N) to a lower bound of $R(\sigma(k), k)$.

Theorem 1. If there is a positive integer N such that the shortest nonzero vector in L(k, N) has length greater than $\sqrt{(1 + k\sqrt{\sigma(k)}/2)^2 + k^2\sigma(k)}$, then

$$R(\sigma(k), k) \ge 1/N$$
.

We can also obtain constructive upper bounds from the following theorem.

Theorem 2. Let $(s, a_1, a_2, \dots, a_k)$ be a vector in L(k, N). Then there exists an integer b such that

$$\left| \sum_{i=1}^{k} a_i \sqrt{\sigma(i)} - b \right| \le \left(|s| + \sum_{i=1}^{k} |a_i|/2 \right) \frac{1}{N}.$$

We set N to be large and use a lattice reduction algorithm to find a short vector $(s, a_1, a_2, \dots, a_k)$ in the lattice L(k, N). It gives us a constructive upper bound. For example, we have found that integers a_1, a_2, \dots, a_{100} and t such that $\max_{1 \le i \le 100} a_i^2 \sigma(i) = 19796$ and

$$\left| \sum_{i=1}^{100} a_i \sqrt{\sigma(i)} - t \right| \approx 10^{-115},$$

which implies a constructive upper bound $R(19796, 100) \le 10^{-115}$.

1.3 Organization

In Section 2, we review some relevant facts about lattice and present our algorithm to find a lower bound for $R(\sigma(k), k)$. In Section 3, we prove a rigorous exponential time upper bound $\exp(O(k))$ for the algorithm and present some numerical data. Based on the data we formulate a conjecture which implies that our algorithm runs in time $O(\operatorname{poly}(k))$. In Section 4, we prove Theorem 1. In Section 5, we prove Theorem 2 and another theorem on a provable upper bound for some R(n,k) where n is much smaller than 2^{2k} . Throughout this paper, we use lattice functions in Victor Shoup's NTL package to produce numerical data. The block size of the BKZ reduction is set to be 10.

2 Lattices and Our Algorithm

In the m-dimensional Euclidean space ${\bf R}^m,$ a (full rank) integral lattice is the set

$$\left\{ \sum_{i=1}^{m} x_i \mathbf{b}_i | x_i \in \mathbf{Z} \right\},\,$$

where $\mathbf{b}_1, \mathbf{b}_2, \cdots \mathbf{b}_m$ are linearly independent vectors over \mathbf{R} and $\mathbf{b}_i \in \mathbf{Z}^m$ for all $1 \leq i \leq m$. The determinant of a lattice is defined to be the absolute value of the determinant of the matrix (b_{ij}) where b_{ij} is the j-th coordinate of \mathbf{b}_i . Assume that a lattice has determinant D and the shortest nonzero vector has length λ . Minkowski's first theorem (see page 12 in [7]) asserts that $\lambda \leq \sqrt{m}D^{1/m}$.

Finding the shortest nonzero vector in a lattice is a well studied problem. The Block-Korkine-Zolotarev (BKZ) lattice reduction algorithm, which is based on the famous LLL lattice reduction algorithm, can find a nonzero vector whose length is at most $2^{O(m(\ln \ln m)^2/\ln m)}\lambda$ in polynomial time [10]. Although the algorithm usually performs better than the worst case approximation ratio, it is not believed that a polynomial time algorithm can find nonzero vectors of length $2^{o(\sqrt{\log m})}\lambda$ for general lattices [6]. See [7] for a survey on computational lattice problems.

To use Theorem 1, we need a good lower bound on the length of the shortest nonzero vector in L(k, N). We first apply the BKZ reduction algorithm on L(k, N) to obtain a reduced base. We then apply the Gram-Schmidt orthogonalization on the reduced base. Let $\lambda^*(k, N)$ denote the length of the shortest Gram-Schmidt vector (we will omit k and N if they are clear from the context). Then $\lambda^*(k, N)$ is a lower bound for the length of the shortest nonzero vector in L(k, N). The main process of our method can be illustrated as follows:

$$L(k,N) \overset{BKZ}{\Longrightarrow} (\mathbf{v}_0',\mathbf{v}_1',\cdots,\mathbf{v}_k') \overset{Gram-Schmidt}{\Longrightarrow} (\mathbf{v}_0^*,\mathbf{v}_1^*,\cdots,\mathbf{v}_k^*).$$

Note that one should not apply the Gram-Schmidt orthogonalization directly on L(k, N). Otherwise the shortest Gram-Schmidt vector will always have length 1. The algorithm is described as follows.

Algorithm 1:

Input: k, step;

- 1. N = 1;
- 2. $\lambda^* = 0$;
- 3. while $\lambda^* \leq \sqrt{(1+k\sqrt{\sigma(k)}/2)^2 + k^2\sigma(k)}$ do
- 4. $N = N * \dot{step};$
- 5. Apply the BKZ lattice reduction algorithm on L(k, N);
- 6. Apply the Gram-Schmidt orthogonalization on the reduced base;
- 7. Let λ^* be the length of the shortest vector in the Gram-Schmidt base.
- 8. endwhile
- 9. Output 1/N as the lower bound for $R(\sigma(k), k)$.

3 Time Complexity Analysis and Numerical Data

Theorem 3. Algorithm 1 runs in time at most $\exp(O(k))$.

Proof. Denote the length of the shortest vector in L(k, N) by λ . Let l be the length of the shortest vector in the reduced base. From the proof of Lemma 2.8 in [7], we derive that

$$\lambda \le l \le 2^{k+1} \lambda^*.$$

We shall prove that if $N \geq 2^{3k2^k}$, then $\lambda \geq 2^{2k}$, which implies that for k > 7

$$\lambda^* \ge \frac{\lambda}{2^{k+1}} \ge 2^{k-1} > 3k^{1.5}.$$

On the other hand, we know from formulae (3) that for k big enough, $\sigma(k) < 2k$. Hence

$$\sqrt{(1+k\sqrt{\sigma(k)}/2)^2+k^2\sigma(k)}<\sqrt{(1+k\sqrt{2k}/2)^2+2k^3}<\sqrt{2k^3+2k^3}<3k^{1.5}.$$

This shows that the algorithm will terminate before N exceeds 2^{3k2^k} . The time complexity is thus at most $O\left(k2^k \operatorname{poly}\left(k \log 2^{3k2^k}\right)\right)$, which is $\exp(O(k))$.

Assume that $N \geq 2^{3k2^k}$. Any nonzero vector in the lattice has form

$$\left(\sum_{i=1}^{k} a_i \left[N \sqrt{\sigma(i)} \right] - bN, a_1, a_2, \cdots, a_k \right)$$

for some integers a_1, a_2, \dots, a_k and b. It is enough to show that the length of the vector is greater than 2^{2k} . If for some a_i , $|a_i| > 2^{2k}$, then the length of the vector is greater than 2^{2k} . So we may assume that $|a_i| \le 2^{2k}$ for all $1 \le i \le k$.

$$\left| \sum_{i=1}^{k} a_i \left[N \sqrt{\sigma(i)} \right] - b N \right| = \left| \sum_{i=1}^{k} a_i \left(N \sqrt{\sigma(i)} - \left\{ N \sqrt{\sigma(i)} \right\} \right) - b N \right|$$

$$\geq \left| \sum_{i=1}^{k} a_i N \sqrt{\sigma(i)} - b N \right| - \left| \sum_{i=1}^{k} a_i \left\{ N \sqrt{\sigma(i)} \right\} \right|$$

$$\geq N \left| \sum_{i=1}^{k} a_i \sqrt{\sigma(i)} - b \right| - \sum_{i=1}^{k} \frac{|a_i|}{2}$$

$$\geq N \left| \sum_{i=1}^{k} a_i \sqrt{\sigma(i)} - b \right| - \frac{k2^{2k}}{2}$$

By the root separation bound,

$$N\left|\sum_{i=1}^{k} a_i \sqrt{\sigma(i)} - b\right| - \frac{k2^{2k}}{2} \ge N\left(2k\sqrt{2^{4k}\sigma(k)}\right)^{-2^{k-1}} - \frac{k2^{2k}}{2}$$

which is greater than 2^{2k} as $N \ge 2^{3k2^k}$.

The above theorem gives us an exponential upper bound of the time complexity. However from numerical experiments, we can see that the algorithm terminates quickly and enables us to find a lower bound of $R(\sigma(k), k)$ much better than the root separation bound.

We list in Table 2 the values of l^2 and $(\lambda^*)^2$ for L(100,N) where N starts from 10^{50} and keeps increasing by a factor of 10^5 . From Table 2, we learn that the square length of the shortest nonzero vector in the lattice $L\left(100,10^{390}\right)$ is greater than 3102794. Since $\left(1+100\sqrt{165}/2\right)^2+100^2*165=2063785.52...$, we obtain that $R(165,100)\geq 10^{-390}$. Similarly we can get the other data on the right-hand side in Table 1.

Table 2 and Table 3 illustrate that the ratio between λ^* and $N^{\frac{1}{k+1}}$ remains about the same when N increases. Note that $\lambda^*(k,N)$ is the lower bound of the length of the shortest nonzero vector in L(k,N). Based on this observation, we formulate the following conjecture on the length of the shortest nonzero vector in L(k,N):

Conjecture 1. The shortest nonzero vector in the lattice L(k, N) has length greater than $N^{\frac{1}{k+1}}/k$.

Corollary 1. If Conjecture 1 is true then

- 1. $R(\sigma(k), k) \ge 1/(2\sigma(k)k^3)^k$,
- 2. Algorithm 1 runs in time O(poly(k)).

Proof. Set

$$N = \left\lceil \left(k \left(\left(1 + k \sqrt{\sigma(k)} / 2 \right)^2 + k^2 \sigma(k) \right) \right)^k \right\rceil.$$

The first item follows from Theorem 1 and Conjecture 1. The time complexity of Algorithm 1 is $O(k \cdot \text{poly}(k \log N))$, which is at most O(poly(k)). Thus the second item holds.

It is interesting to contrast L(k, N) with a similar lattice generated by

$$(N,0,0,0,\cdots,0)$$

$$(N[\sqrt{s_1}],1,0,0,\cdots,0)$$

$$(N[\sqrt{s_2}],0,1,0,\cdots,0,)$$

$$(N[\sqrt{s_3}],0,0,1,\cdots,0,)$$

$$\vdots$$

$$(N[\sqrt{s_k}],0,0,0,\cdots,1).$$

It is easy to see that the shortest vector in the lattice always has length 1 no matter how large N is.

Table 2. The data for L(100, N) ($\sigma(100) = 165$)

$\log_{10} N$	l^2	$(\lambda^*)^2$	$\lambda^*/N^{\frac{1}{k+1}}$	$\log_{10} N$	l^2	$(\lambda^*)^2$	$\lambda^*/N^{\frac{1}{k+1}}$
50	189	0.920102	0.31	55	187	1	0.29
60	267	0.848578	0.23	65	318	1.075123	0.24
70	379	1.251328	0.23	75	531	1.496321	0.22
80	711	1.905386	0.22	85	824	2.645646	0.23
90	1039	3.06967	0.23	95	1466	3.489959	0.21
100	1959	5.242274	0.23	105	2339	6.414824	0.23
110	2726	7.750273	0.23	115	3639	10.30219	0.23
120	4370	11.97402	0.22	125	5512	14.85322	0.22
130	6936	17.7999	0.22	135	9345	26.00225	0.23
140	10479	32.10648	0.23	145	11789	38.72952	0.23
150	18949	45.87119	0.22	155	20579	71.66641	0.25
160	23457	81.06815	0.23	165	33572	108.6672	0.24
170	40148	122.6395	0.23	175	52839	157.6955	0.23
180	72509	185.265	0.22	185	81229	271.003	0.24
190	96242	279.2017	0.22	195	116002	416.3477	0.24
200	165201	421.7492	0.21	205	182891	662.0805	0.24
210	267509	786.044	0.23	215	307450	1103.5	0.25
220	411290	1241.363	0.23	225	530717	1566.732	0.23
230	484931	1948.158	0.23	235	761232	2508.116	0.24
240	1010500	2877.924	0.23	245	1273090	3674.155	0.23
250	1628420	4691.54	0.23	255	1699623	5815.91	0.23
260	2345069	8097.233	0.24	265	2735544	10196.72	0.24
270	3830216	12159.34	0.23	275	4483731	14841.72	0.23
280	6448489	20310.26	0.23	285	7963507	21892.18	0.23
290	9576142	29173.78	0.23	295	11065095	37113.27	0.23
300	14625831	47887.96	0.23	305	20017801	54521.73	0.23
310	22107891	70218.28	0.23	315	30329692	92013.94	0.23
320	42326718	133759.6	0.25	325	50110184	160807.8	0.23
330	58226453	163842	0.20	335	73620063	209813	0.24
340	101230523	289148.6	0.23	345	116341856	367148.5	0.22
350	134263638	446526.2	0.23	355	176973638	606588.9	0.23
360	195623258	577848.5	0.23 0.21	365	295497369	953276.7	0.24
370	254486097	1108836	0.21 0.23	375	365813532	1383326	0.24
380	620569774	1500616	0.23 0.21	385	857210733	1906762	0.23
390	936892309	3102794	0.21 0.24	395	1214701229	3716443	0.21
400	1512222196	4440063	0.24 0.23	405	1815150428	4762491	0.24
410	2479097817	6683891	0.23	415	2825781352	8365474	0.21
420	3315764769	1	0.23 0.23	425	4759873617	13620881	0.23
430	5567641140		0.23 0.24	435		20738831	0.23
440			0.24 0.24	445	7323262414 10529606845		0.22
450			0.24 0.23	445			0.22
460			0.24	465	42772055636		0.23
470			0.21				0.24
480			0.22	485			0.24
490			0.24		1.18668E+11		0.23
I .	1.37029E+11	1	0.24		1.79821E+11		0.23
1	2.02652E+11		0.24	515			0.22
	4.04579E+11		0.24		4.64658E+11	1.4E+09	0.24
1	5.00824E+11		0.23	535			0.25
1	9.40142E+11		0.22		1.03272E+12		0.22
1	1.23394E+12		0.23	555			0.23
	2.04264E+12		0.23		2.53027E+12		0.23
1	3.14243E+12		0.24		4.00544E+12		0.23
1	5.82342E+12		0.21		6.91689E+12		0.23
590	8.50935E+12	2.13E+10	0.21	595	1.11703E+13	2.95E+10	0.22

Table 3. The data for $\lambda^*(k,N)/N^{\frac{1}{k+1}}$

						I				
$\log_{10} N$	10	20	30	40	50	60	70	80	90	100
50	0.83	0.74	0.62	0.56	0.48	0.42	0.36	0.29	0.27	0.31
60	0.91	0.78	0.64	0.57	0.47	0.41	0.34	0.32	0.27	0.23
70	0.90	0.78	0.65	0.55	0.47	0.39	0.35	0.31	0.29	0.23
80	0.91	0.77	0.66	0.58	0.50	0.42	0.36	0.33	0.28	0.22
90	0.90	0.69	0.65	0.58	0.49	0.41	0.37	0.30	0.24	0.23
100	0.90	0.76	0.65	0.57	0.49	0.43	0.35	0.30	0.28	0.23
110	0.94	0.73	0.64	0.56	0.46	0.44	0.35	0.32	0.27	0.23
120	0.75	0.73	0.62	0.59	0.47	0.41	0.34	0.30	0.27	0.22
130	0.91	0.80	0.63	0.56	0.49	0.42	0.36	0.29	0.27	0.22
140	0.89	0.78	0.67	0.56	0.51	0.39	0.35	0.31	0.29	0.23
150	0.89	0.77	0.66	0.53	0.48	0.43	0.35	0.32	0.27	0.22
160	0.92	0.79	0.64	0.59	0.50	0.39	0.36	0.32	0.25	0.23
170	0.89	0.75	0.64	0.51	0.47	0.41	0.36	0.31	0.28	0.23
180	0.92	0.74	0.65	0.58	0.48	0.43	0.36	0.30	0.25	0.22
190	0.93	0.84	0.65	0.56	0.45	0.38	0.39	0.29	0.26	0.22
200	0.90	0.78	0.66	0.52	0.47	0.42	0.34	0.32	0.28	0.21
210	0.83	0.80	0.64	0.56	0.48	0.42	0.40	0.33	0.25	0.23
220	0.79	0.77	0.67	0.57	0.51	0.44	0.35	0.30	0.28	0.23
230	0.94	0.73	0.62	0.58	0.44	0.43	0.33	0.30	0.27	0.23
240	0.91	0.79	0.67	0.58	0.48	0.42	0.36	0.31	0.25	0.23
250	0.91	0.80	0.63	0.53	0.47	0.41	0.36	0.31	0.26	0.23
260	0.93	0.77	0.63	0.55	0.47	0.43	0.36	0.33	0.28	0.24
270	0.91	0.74	0.66	0.54	0.50	0.43	0.39	0.30	0.27	0.23
280	0.85	0.81	0.63	0.55	0.47	0.42	0.36	0.31	0.26	0.24
290	0.92	0.79	0.64	0.55	0.49	0.40	0.37	0.32	0.26	0.23
300	0.94	0.78	0.68	0.57	0.47	0.42	0.34	0.32	0.25	0.23

4 The Proof of Theorem 1

To prove Theorem 1, we need the following lemma.

Lemma 2. Let s_1, s_2, \dots, s_k be k distinct positive square free integers. Let λ be the length of the shortest nonzero vector in $L_{s_1, s_2, \dots, s_k}(N)$. For any integers a_1, a_2, \dots, a_k, b , if $(b, a_1, a_2, \dots, a_k) \neq (0, 0, 0, \dots, 0)$, and $\lambda^2 \geq \left(1 + \frac{\sum_{i=1}^k |a_i|}{2}\right)^2 + \sum_{i=1}^k a_i^2$, then

$$\left| \sum_{i=1}^{k} a_i \sqrt{s_i} - b \right| \ge \frac{1}{N}.$$

Proof. The vector

$$\left(\sum_{i=1}^{k} a_i [N\sqrt{s_i}] - bN, a_1, a_2, \cdots, a_k\right)$$

is nonzero and in the lattice, hence its length

$$\sqrt{\sum_{i=1}^{k} a_i^2 + \left(\sum_{i=1}^{k} a_i [N\sqrt{s_i}] - bN\right)^2}$$

is no smaller than λ . We have

$$\sum_{i=1}^{k} a_i^2 + \left(\sum_{i=1}^{k} a_i [N\sqrt{s_i}] - bN\right)^2 \ge \lambda^2 \ge \left(1 + \frac{\sum_{i=1}^{k} |a_i|}{2}\right)^2 + \sum_{i=1}^{k} a_i^2.$$

It implies that

$$\left| \sum_{i=1}^{k} a_i [N\sqrt{s_i}] - bN \right| \ge 1 + \frac{\sum_{i=1}^{k} |a_i|}{2}.$$

The left hand side is

$$\left| \sum_{i=1}^{k} a_{i} [N\sqrt{s_{i}}] - bN \right| = \left| \sum_{i=1}^{k} a_{i} (N\sqrt{s_{i}} - \{N\sqrt{s_{i}}\}) - bN \right|$$

$$\leq \left| \sum_{i=1}^{k} a_{i} N\sqrt{s_{i}} - bN \right| + \left| \sum_{i=1}^{k} a_{i} \{N\sqrt{s_{i}}\} \right|$$

$$\leq \left| \sum_{i=1}^{k} a_{i} N\sqrt{s_{i}} - bN \right| + \sum_{i=1}^{k} |a_{i} \{N\sqrt{s_{i}}\}|$$

$$\leq \left| \sum_{i=1}^{k} a_{i} N\sqrt{s_{i}} - bN \right| + \frac{\sum_{i=1}^{k} |a_{i}|}{2}$$

So we have

$$\left| \sum_{i=1}^{k} a_i N \sqrt{s_i} - b N \right| \ge 1,$$

therefore $\left|\sum_{i=1}^{k} a_i \sqrt{s_i} - b\right| \ge 1/N$.

Now we are ready to prove Theorem 1.

Proof. Let $n_i, 1 \leq i \leq k$ be positive integers $\leq \sigma(k)$, m be an integer and $e_i \in \{1, 0, -1\}$ for all $1 \leq i \leq k$. We can write $\sum_{i=1}^k e_i \sqrt{n_i} - m$ as $\sum_{i=1}^k a_i \sqrt{\sigma(i)} - b$ where a_1, a_2, \cdots, a_k, b are integers. We have that

$$\sum_{i=1}^{k} |a_i| \le \sum_{i=1}^{k} |a_i| \sqrt{\sigma(i)} \le \sum_{i=1}^{k} \sqrt{n_i} \le k \sqrt{\sigma(k)}$$

and

$$\sum_{i=1}^{k} a_i^2 \le \left(\sum_{i=1}^{k} |a_i|\right)^2 \le k^2 \sigma(k).$$

Assume that $(a_1, a_2, \dots, a_k, b) \neq (0, 0, \dots, 0, 0)$. Since the shortest nonzero vector in the lattice L(k, N) has length at least

$$\left(1 + k\sqrt{\sigma(k)}/2\right)^2 + k^2\sigma(k) \ge \left(1 + \frac{\sum_{i=1}^k |a_i|}{2}\right)^2 + \sum_{i=1}^k a_i^2,$$

we conclude from Lemma 2 that,

$$\left| \sum_{i=1}^{k} e_i \sqrt{n_i} - m \right| = \left| \sum_{i=1}^{k} a_i \sqrt{\sigma(i)} - b \right| \ge \frac{1}{N}.$$

5 Upper Bound

Now we can prove Theorem 2.

Proof. Since $(s, a_1, a_2, \dots, a_k)$ is a vector in the lattice L(k, N), there exists an integer b such that

$$\left| \sum_{i=1}^{k} a_i \left[N \sqrt{\sigma(i)} \right] - bN \right| = |s|.$$

Then

$$\left| \sum_{i=1}^{k} a_i N \sqrt{\sigma(i)} - b N \right| = \left| \sum_{i=1}^{k} a_i \left(\left[N \sqrt{\sigma(i)} \right] + \left\{ N \sqrt{\sigma(i)} \right\} \right) - b N \right|$$

$$\leq \left| \sum_{i=1}^{k} a_i \left[N \sqrt{\sigma(i)} \right] - b N \right| + \left| \sum_{i=1}^{k} a_i \left\{ N \sqrt{\sigma(i)} \right\} \right|$$

$$\leq |s| + \sum_{i=1}^{k} |a_i|/2.$$

Hence the theorem follows.

We may apply the BKZ reduction algorithm on the lattice and obtain a nonzero vector $(s, a_1, a_2, \cdots, a_k)$ of length at most $2^{O\left(k(\ln \ln k)^2/\ln k\right)}N^{\frac{1}{k+1}}$. The data is listed in Table 4. More generally, we have

Theorem 4. Let s_1, s_2, \dots, s_k be k distinct square free integers no smaller than 2. For any integer N, we can find integers a_1, a_2, \dots, a_k and b in polynomial time satisfying that $|a_i| \leq 2^{O\left(k(\ln \ln k)^2/\ln k\right)} N^{\frac{1}{k+1}}$ for all $1 \leq i \leq k$ and

$$\left| \sum_{i=1}^k a_i \sqrt{s_i} - b \right| \le 2^{O\left(k(\ln \ln k)^2 / \ln k\right)} N^{\frac{-k}{k+1}}.$$

Proof. The determinant of the lattice L(k, N) is

$$\begin{vmatrix} N & 0 & 0 & 0 & \cdots & 0 \\ [N\sqrt{s_1}] & 1 & 0 & 0 & \cdots & 0 \\ [N\sqrt{s_2}] & 0 & 1 & 0 & \cdots & 0 \\ [N\sqrt{s_3}] & 0 & 0 & 1 & \cdots & 0 \\ \vdots \\ [N\sqrt{s_k}] & 0 & 0 & 0 & \cdots & 1 \end{vmatrix} = N$$

By Minkowski's first theorem, there is a vector of length $\sqrt{k+1}N^{\frac{1}{k+1}}$ or shorter in the lattice. If we apply the BKZ reduction algorithm on the lattice, we obtain a nonzero vector (s,a_1,a_2,\cdots,a_k) of length at most $2^{O(k(\ln \ln k)^2/\ln k)}N^{\frac{1}{k+1}}$. Thus $|a_i| \leq 2^{O\left(k(\ln \ln k)^2/\ln k\right)}N^{\frac{1}{k+1}}$ for all $1 \leq i \leq k$ and $|s| \leq 2^{O\left(k(\ln \ln k)^2/\ln k\right)}N^{\frac{1}{k+1}}$. We have

$$\left| \sum_{i=1}^k a_i \sqrt{s_i} - b \right| \le \left(|s| + \sum_{i=1}^k |a_i|/2 \right) N^{-1} = 2^{O\left(k(\ln \ln k)^2 / \ln k\right)} N^{\frac{-k}{k+1}}.$$

6 Concluding Remarks

In this paper we present a numerical method that finds a much better lower bound for R(n, k) than the previously known methods do. The main open problem is to prove Conjecture 1, which implies that our method runs in polynomial time.

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Table 4. The data for the shortest vector in the BKZ reduced base: $n = \max_i a_i^2 \sigma(i)$

$\log_{10} N$	n	s	$\log_{10} N$	n	s
50	927	1	55	2275	-1
60	1616	-2	65	1680	0
70	3430	2	75	4075	-1
80	7595	-7	85	6016	-1
90	7693	-1	95	10406	-1
100	21744	-1	105	8107	0
110	13310	7	115	19796	-4
120	25650	7	125	39680	4
130	43639	1	135	64375	14
140	88556	6	145	104247	-9
150	144900	4	155	128625	7
160	427228	-15	165	287550	-5
170	236192	12	175	554429	4
180	543600	13	185	1422596	18
190	613965	-7	195	739640	55
200	1056440	31	205	1609650	1
210	1417939	42	215	2003760	-52
220	2321776	136	225	4790753	65
230	3819232	41	235	5270427	18
240	6951744	8	245	10285412	80
250	13564142	-205	255	14948504	232
260	12803364	-125	265	23483741	-187
270	19553816	122	275	27059175	61
280	43818180	480	285	49650120	210
290	90805809	474	295	66427398	-256
300	94303440	42	305	282685488	-71
310	217669834	570	315	208586875	-203
320	282304440	257	325	340940097	288
330	684191361	-451	335	572956706	134
340	380583760	-2671	345	522838807	434
350	2598879987	-1205	355	1209416149	-1467
360	1481771275	743	365	1496653524	1917
370	1564223967	-2843	375	2345301325	3085
380	3944948022	135	385	4095144375	-368
390	8242412940	-2549	395	6185090625	-1475
400	20837120600	-3524	405	16962130500	649
410	35444689191	1235	415	16992489353	-4827
420	52222173525	926	425	29831928284	5987
430	32518315323	-8686	435	31432393528	263
440	97378652871	-20156	445	58208231178	14650
450	1.1644E+11	-4583	455	1.17189E+11	-121
460	1.56626E+11	-24968	465	2.31065E+11	11314
470	2.40379E+11	6759	475	1.88347E+11	-1784
480	3.71914E+11	12199	485	5.79326E+11	25529
490	6.61439E+11	15283	495	7.70885E+11	55710
500	9.82574E+11	39390	505	1.03675E+12	53574
510	1.35333E+12	-74474	515	2.08272E+12	43145
520	2.46576E+12	-18596	525	3.51867E+12	-88646
530	3.53844E+12	-154254	535	7.71874E+12	44316
540	6.34504E+12	12527	545	5.00231E+12	66480
550	2.0875E+13	-50539	555	1.49025E+13	70127
560	1.63683E+13	-10313	565	4.2508E+13	24836
570	4.06111E+13	202903	575	2.54605E+13	63494
580	3.43017E+13	171277	585	7.91203E+13	135561
590	7.65752E + 13	15225	595	8.47789E+13	427965

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