The Crouton Algorithm for Optimal Addition Chains

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Abstract

We describe a new algorithm for calculating optimal addition chains that has very good performance for single values. The algorithm performs well with or without a database of previously created optimal addition chain lengths. Improved performance comes from the realization that the end of an addition chain can be viewed as a vector addition chain. The values generated by the start of the addition chain together with the values from the trailing vector addition chain satisfy a linear Diophantine equation. By broadening the search tree by differentiating cases based on which constructed chain elements will be used in the future, a collection of new pruning techniques are possible. Estimates of the minimum Hamming weight of solutions to a linear Diophantine equation along with bounds propagation techniques are used to prove that partially constructed chains will not be successful and hence can be pruned.

Introduction

An addition chain is a finite sequence of integers (we call elements) $1 = a_0 < a_1 < \ldots < a_r = n$ with $a_i = a_j + a_k$, $i > j \geqslant k \ge 0$ for a target n of length r. We will refer to an addition chain as formal if we explicitly specify how each element is constructed. For example the chain 1, 2, 3, 4 has two ways of constructing 4 since 4 = 2 + 2 = 1 + 3. We denote with l(n) the length of the optimal (shortest) addition chain for n. We will call each a_i , i > 0 a step. We define $\lambda(n) = \lfloor log_2(n) \rfloor$ and use it to partition the steps into two cases. If $\lambda(a_i) = \lambda(a_{i-1})$ we say it is a small step. Otherwise we must have $\lambda(a_i) = \lambda(a_{i-1}) + 1$ and call it a large step. A step of the form $a_i = a_{i-1} + a_j$, i > j is called a star step. If we denote with s(n) the number of small steps in an optimal chain for n we must have $s(n) = l(n) - \lambda(n)$. We will denote with s(n) the binary digit sum of s(n). The smallest target with an optimal addition chain of length s(n) is denoted with s(n). Knuth [13] notes that proving

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via computer searches that l(c(r)) > r - 1 is particularly time consuming. On the other hand finding an addition chain of length r for c(r) appears quite fast in comparison. Scholz and Brauer [16, 3] conjectured (Scholz-Brauer conjecture) and partially proved that $l(2^n - 1) \leq l(n) + n - 1$. Stolarsky [17] noted that he saw only equality in all values he could prove $(l(2^n - 1) = l(n) + n - 1)$. Knuth and Stolarsky [12, 17] conjectured (Knuth-Stolarsky conjecture) that $v(n) \leq 2^{s(n)}$.

We say an addition sequence is an addition chain that contains a set of specified values. We will extend the notion of an optimal addition sequence using $l(\{a_1,...,a_z\})$ where $a_i \geq 1$. Similarly we say a vector addition chain is a chain of vectors produced via addition from a set of q basis vectors $(v_0, v_{-1}, ..., v_{-q+1})$:

$$v_{-q+1}, v_{-q+2}, ..., v_0, v_1, ..., v_d, v_i = v_j + v_k, i > j \ge k \ge -q + 1$$

We define the length of the shortest vector addition chain as $l([a_1,...,a_z])$ treating the basis vectors as all being the first element for calculating the length. We will use the 2-adic valuation $v_2(n) = k, 2^k | n, 2^{k+1} \nmid n$ which is value of the least significant bit in n.

History of Computer Calculations

Knuth appears to have written the first computer program for calculating optimal addition chains in December 1963 (private communication). He used a program written in Algol that ran on the Burroughs B5000 computer and calculated all optimal addition chains for $n \leq 1024$. He did further calculations on the CDC 6600 in assembler [14]. He worked backward from the target splitting n = p + q, $p \ge q$ and used memoization of prior calculated l(n) values. The properties of the function l(n) are very interesting and often counter intuitive. As each new technique is developed to calculate optimal chains for larger targets new properties often emerge. The first computer calculations for example found l(n) = l(2n) with n = 191. Kato [11] found minimum length merges of addition chains for n-k and k with $1 \le k \le \frac{n}{2}$ as well as enumerations of partitions for n-1 to find addition chains for n. A breadth first search (BFS) was used by Chin and Tsai [5] to find optimal addition chains. They attempted to prune by having all early steps being doubles but discovered that there are only two optimal chains for 2617 and both include the element 5. Tsai [21] found optimal chains for batches of numbers and forced the construction of the final element in a chain to be a star step. Brlek and Mallette [4] calculated batches of optimal addition chains and used bounds for addition chain elements. Using the nomenclature developed by Thurber [18] these would be the most simple class 1 (sequence A) bounds so that an element a_i for an optimal addition chain for n of length r is bounded by $a_i \geq \left\lceil \frac{n}{2^{r-i}} \right\rceil$. Bleichenbacher and Flammenkamp [9] made huge strides by calculating all l(n) for $n \leq 2^{22}$ by using graph reductions and memoization of computed l(m), m < n values for computing l(n). Thurber [18] developed a set of bounds for the elements of an addition chain and the sum of two consecutive elements. These can be used in an addition chain algorithm when elements are selected to cut down the search space. They can also be used to work out the minimum number of steps between two given values in an addition chain. Bahig [1] outlined conditions where steps must be star and limits on the types of non-star steps and hence cut down the time needed to generate branches in the search tree. Clift [6] calculated batches of addition chains using graph enumeration avoiding sub-graphs proven to not be present in optimal chains.

Broadening the Search Tree

Prior work[19] has shown that the end of an addition chain can be viewed as a vector addition chain that consumes some of the earlier values to get to the target. We can show that the number of elements from the start of the chain that are consumed must be small and related to the small step count of the target. With this in mind we will describe an algorithm that splits the problem such that a small packet of numbers (called a *crouton*) is selected at each stage that must be used by the chain in the future. A crouton can also be thought of as the temporary variables needed to compute the rest of the chain. This will make the search tree broader but enable new pruning techniques. We will then show how to bound the values in the vector addition chain such that many cases can be rejected early.

Recursive Algorithm Outline

We now describe the initial state and recursive transformations the new algorithm uses to traverse the search space. We will be searching for an optimal addition chain $A = b_0 < b_1 < \ldots < b_d = n$ for n of length $d \le l(n)$. Obviously when d < l(n) the search for a chain will fail. We will say we are at depth r with $0 \le r \le d$ and mean we have generated the sequence $1 = b_0 < b_1 < \ldots < b_r$. At depth r though we will not keep or use all $b_i, 0 \le i \le r$ but instead we will have chosen to use a subset of these values $a_1 < a_2 < \ldots < a_z$ with $z \ge 1$. There must be a mapping α such that $a_i = b_{\alpha(i)}$. Each $a_i < n$ must be consumed in at least one later addition chain step $\beta(i) > r$ so $b_{\beta(i)} = a_i + b_j$ with $j < \beta(i)$.

Initial State

We start with a single crouton < 1 >. So $a_1 = b_0 = 1$ and z = 1.

Recursive Transformation

If we start with a crouton $< a_1, \ldots, a_z >$ we form a new element $a_{z+1} = a_i + a_j$, $i \ge j$. From this we potentially can form up to four new croutons to search:

 $\langle a_1, \ldots, a_{z+1} \rangle$ The new crouton has z+1 elements.

 $< a_1, \ldots, a_{i-1}, a_{i+2}, \ldots, a_{z+1} >$ The new crouton has z elements since element a_i is dropped.

 $< a_1, \ldots, a_{j-1}, a_{j+2}, \ldots, a_{z+1} >$ The new crouton has z elements since element a_j is dropped.

 $< a_1, \ldots, a_{j-1}, a_{j+2}, \ldots, a_{i-1}, a_{i+2}, \ldots, a_{z+1} >, i \neq j$ The new crouton has z-1 elements since elements a_i and a_j are dropped.

It should be clear from these transformations that in the crouton $\langle ..., a_z \rangle$ at depth r must have $a_z = b_r$.

Termination State

The final crouton at depth d must be $\langle a_1 = n \rangle$ if the search is successful. The initial and termination state along with the fact that the recursive step can only increase or decrease z by at most 1 leads to the following simple bounds:

$$z \le \min(r+1, d-r+1)$$

Bounding the Crouton's Size

Let $\langle a_1, a_2, ..., a_z \rangle$ be a crouton with $z \geq 1$. Let these a_i be part of an optimal addition chain within the first r+1 values for n with l(n)=d. Hence $\{a_i|1\leq i\leq z\}\subset A$ where A is an addition chain for n. Further we will specify that the members of the crouton are the only elements in the addition chain within the first r+1 values that are used in the rest of the chain. We specify that all elements in the crouton must be used later in the chain. From the requirement that A is an optimal chain we must have $l(\{a_1, a_2, \ldots, a_z\}) = r$. The corresponding vector chain that consumes the crouton must have $l([x_1, x_2, \ldots x_z]) = d - r$ and hence $l(\{x_1, x_2, \ldots x_z\}) = d - r - z + 1$ [15]. We know that

$$n \le 2^{d-r-z+1} \sum_{i=1}^{z} a_i$$

$$\le 2^{d-r-z+1} \sum_{i=1}^{z} 2^{r+1-i}$$

$$\le 2^{d-2z+2} (2^z - 1)$$

From $s(n) = l(n) - \lambda(n)$ we obtain

$$s(n) = d - \lambda(n)$$

$$\geq d - \lambda(2^{d-2z+2}(2^z - 1))$$

$$\geq 2z - 2 - \lambda(2^z - 1)$$

$$\geq z - 1$$

From this we can see that the crouton remains small since s(n) is slow growing.

The Frobenius Equation

Our two addition sequence problems are linked because the vector addition chain consumes the elements of the crouton to reach the target. So from $l(\{a_1,a_2,\ldots,a_z\})=r$ and $l(\{x_1,x_2,\ldots x_z\})=d-r-z+1$ we know that:

$$\sum_{i=1}^{z} a_i x_i = n \tag{1}$$

We can prune the search tree by proving there are no solutions to the Frobenius equation when limited by the complementary addition sequence problem. The most obvious way of eliminating cases uses the GCD:

$$GCD(a_1, a_2, \ldots, a_z)|n$$

The many divisions needed to calculate the GCD make this test better done after the following bounds on x_i . For example we must have

$$1 \le x_i \le 2^{d-r-z+1}$$

We can do much better than this though if we can estimate a lower bound for $s(\{x_1, x_2, \dots, x_z\})$. We can do this by getting a lower bound for

$$\max_{i=1}^{z} v(x_i)$$

We start with

$$v(n) = v\left(\sum_{i=1}^{z} a_i x_i\right)$$

$$\leq \sum_{i=1}^{z} v(a_i x_i)$$

$$\leq \sum_{i=1}^{z} v(a_i) v(x_i)$$

$$\leq \max_{i=1}^{z} v(x_i) \sum_{i=1}^{z} v(a_i)$$

So we have

$$\max_{i=1}^{z} v(x_i) \ge \left\lceil \frac{v(n)}{\sum_{i=1}^{z} v(a_i)} \right\rceil \tag{2}$$

We have a number of upper bounds on v(n) based on s(n) which we will denote by $s_l(n)$. If s(n)=0,1 then simple bounds on n yield v(n)=1,2 respectively. See [13] pp 468. For the more adventurous see [7] were the result is stated but without much detail. The case v(n)=3 was proved to have s(n)=2 in [10]. Knuth in [13] pp 468-469 outlines the v(n)=3,4 cases in Theorem B and C showing $s(n)\geq 2$. Thurber in [20] did a detailed analysis of the case s(n)=3 and showed $v(n)\leq 8$. Flammenkamp in [8] wrote a computer program to enumerate the s(n)=3 cases. The author wrote a computer program similar to that of Flammenkamp but using a different underlying finite set to represent addition chains of fixed small step count. With this the detailed enumeration of s(n)=4,5 was possible yielding $v(n)\leq 16,32$ Small Step Enumeration. The number of cases explodes as the small step count increases. By limiting the enumeration to those cases that could violate the Knuth-Stolarsky conjecture the author also covered the case s(n)=6 and found $v(n)\leq 64$. This gives us overall:

$$s(x) \ge s_l(x) = \begin{cases} \lceil log_2(v(x)) \rceil & v(x) \le 64 \\ 7 & v(x) > 64 \end{cases}$$

Immediately we can see that

$$1 \le x_i < 2^{d-r-z+2-\max_{i=1}^z s_l(x_i)}$$

For $\max_{i=1}^z s_l(x_i) \leq 6$ we have $\max_{i=1}^z v(x_i) \leq 2^{\max_{i=1}^z s_l(x_i)}$. We can therefor refine the limit. The maximum value in a chain of length l with at least $s = \max_{i=1}^z s_l(x_i)$ small steps must be:

$$x_i \le 2^{l-s} + 2^{l-s-1} + \dots + 2^{l-s-2^s+1}$$

 $\le 2^{l+1-s-2^s} (2^{2^s} - 1)$

$$1 \le x_i \le 2^{d-r-z+2-\max_{i=1}^z s_l(x_i) - 2^{\max_{i=1}^z s_l(x_i)}} (2^{2^{\max_{i=1}^z s_l(x_i)}} - 1)$$
 (3)

The following table shows a hypothetical limit and how it changes as the small step count of the x_i values increases:

s_l	$x_i \leq$
0	1000000000000000000000000000000000000
1	1100000000000000000000000000000000000
2	1111000000000000000000000000000000000
3	11111111100000000000_2
4	111111111111111111111111111111111111

Many cases are rejected in practice by applying this bound to the original Frobenius equation:

$$n \le \left(\max_{i=1}^{z} x_i\right) \left(\sum_{i=1}^{z} a_i\right) \tag{4}$$

Practically this inequality is not satisfied in a significant proportion of cases and search times are reduced considerably. This pruning will be made clearer with an example.

When trying to find an optimal chain for 42833 of length 19 (which must succeed since computer calculations show l(42833) = 19) we get to a depth r = 8 with the crouton < 5, 32, 67 >. We could have gotten to this point in the search with the partial addition chain 1, 2, 3, 5, 8, 16, 32, 64, 67.

From 1 we have the following Frobenius equation to solve:

$$5x_1 + 32x_2 + 67x_3 = 42833$$
$$101_2x_1 + 100000_2x_2 + 1000011_2x_3 = 1010011101010001_2$$

From $2 \max_{i=1}^z v(x_i) \ge \left\lceil \frac{v(n)}{\sum_{i=1}^z v(a_i)} \right\rceil = \left\lceil \frac{8}{6} \right\rceil = 2$. So at least one x_i must have $v(x_i) \ge 2$ so that $\max_{i=1}^z s_l(x_i) \ge 1$ and hence that $s(\{x_1, x_2, x_3\}) \ge 1$ and so from $3 x_i \le 384$. We can immediately abandon this search since from 4 we have the contradiction $42833 \le 384(5+32+67) = 39936$.

Better Estimates for $\max_{i=1}^{z} v(x_i)$

Bounding the size of the $\max_{i=1}^{z} v(x_i)$ has such a large effect on performance that trying to get a better estimate of $\max_{i=1}^{z} v(x_i)$ can be worthwhile even it it takes a non-trivial amount of computation.

If we say that $n \mod m$ represents the least non-negative element from the residue class modulo m. It is curios that:

$$v(n \bmod (2^i - 2^j)) \le v(n), i > j$$
 (5)

To prove this we will assume a smallest counter example v(r) > v(n) with $r = n \mod (2^i - 2^j)$. We must have n > r. We will first assume that the binary representation of n contains a power of two $(2^{i+k}, k \ge 0)$. Since $2^{i+k} \equiv 2^{i+k} - (2^{i+k} - 2^{j+k}) \pmod{(2^i - 2^j)}$. We have a contradiction since $v(n - 2^{i+k} + 2^{j+k})$ would be a smaller counter example from the residue class:

$$v(n-2^{i+k}+2^{j+k}) \le v(n-2^{i+k}) + v(2^{j+k})$$

 $\le v(n) - 1 + 1$
 $\le v(n)$
 $< v(r)$

We may therefore assume that $2^i-2^j < n \le 2^{i-1}-1$. The binary representation of n must be $2^{i-1}+2^{i-2}+\cdots+2^j+\cdots$ and we obtain the following contradiction: $v(n-(2^i-2^j)) < v(n) < v(r)$ since $n-(2^i-2^j) < n$. From this we obtain the following inequality:

$$\max_{i=1}^{z} v(x_i) \ge \max_{i=1}^{z} v(x_i \bmod m) \ge \left\lceil \frac{v(n \bmod m)}{\sum_{i=1}^{z} v(a_i \bmod m)} \right\rceil, m = 2^j - 2^k, j > k$$
(6)

Proof:

$$n \bmod m = (\sum_{i=1}^{z} a_i x_i) \bmod m$$

$$= (\sum_{i=1}^{z} (a_i \bmod m)(x_i \bmod m)) \bmod m$$

$$v(n \bmod m) = v((\sum_{i=1}^{z} (a_i \bmod m)(x_i \bmod m)) \bmod m)$$

$$\leq v(\sum_{i=1}^{z} (a_i \bmod m)(x_i \bmod m))$$

$$\leq \sum_{i=1}^{z} v(a_i \bmod m)v(x_i \bmod m)$$

$$\leq \sum_{i=1}^{z} v(a_i \bmod m)v(x_i \bmod m)$$

$$\leq \sum_{i=1}^{z} v(a_i \bmod m)v(x_i \bmod m)$$

This is a generalization of the realization that the bottom k $(2^k = 2^{k+1} - 2^k)$ bits of n can be determined only by the bottom k bits of a_i and x_i . This will be clear in the following example. With the crouton < 5, 8 > at depth r = 4 while trying to find an addition chain for n = 35. From 2 we would have $\max_{i=1}^z v(x_i) \ge \left\lceil \frac{v(n)}{\sum_{i=1}^z v(a_i)} \right\rceil = 1$. If we instead look at the bottom two bits (k = 2) then $\max_{i=1}^z v(x_i) \ge \left\lceil \frac{v(n \bmod 2^k)}{\sum_{i=1}^z v(a_i \bmod 2^k)} \right\rceil = \left\lceil \frac{2}{1} \right\rceil = 2$. This window method will also detect in-feasibility when $2^k |GCD(a_1, a_2, \ldots, a_z)|$ but $2^k \nmid n$. An example from the more general case has a crouton of < 25, 84 >at depth r = 8 while searching for an addition chain for n = 42163 of length d = 19 < l(42163) = 20. This search will fail. We could have encountered this situation via the partial chain 1, 2, 4, 8, 16, 17, 25, 42, 84. With m = 12 we get:

$$\max_{i=1}^{z} v(x_i) \ge \left\lceil \frac{v(n \bmod 12)}{\sum_{i=1}^{z} v(a_i \bmod 12)} \right\rceil = \left\lceil \frac{3}{1} \right\rceil = 3$$

So we must have $s({x_1, x_2}) \ge 2$ and so from $3x_i \le 480$. This is not enough to prune this path completely since (25 + 84)480 > 42163 but it limits the possibilities for the x_i .

We may obtain a similar bound by multiplying both sides of 1 by a constant m:

$$\max_{i=1}^{z} v(x_i) \ge \left\lceil \frac{v(mn)}{\sum_{i=1}^{z} v(ma_i)} \right\rceil \tag{7}$$

Practically only m=3 seems to allow performance improvements. For example with the crouton <34,43,86> at depth $r=9,\ n=58951$ and d=20 we obtain:

$$\max_{i=1}^{z} v(x_i) \ge \left[\frac{v(3n)}{\sum_{i=1}^{z} v(3a_i)} \right] = \left[\frac{10}{8} \right] = 2$$

So we must have $s(\lbrace x_1, x_2 \rbrace) \geq 1$ and so from $3x_i \leq 384$.

Enumerating Solutions

For problems that appear small it helps the performance of the algorithm to attempt to enumerate all the solutions to the Frobenius equation and attempt to eliminate them by finding a lower bound for $l(\{x_1,...,x_z\})$. Knowing the bounds of x_i and also the bounds of $s(\{x_1,...,x_z\})$ enumerating problems with values for each $l_i \leq x_i \leq u_i$ limiting sub-problems to ~2000 works well. Trying to solve the Frobenius equation directly using the extended GCD algorithm is too expensive. Presumably because of the many divides needed. Improving l_i, u_i via the standard techniques of constraint satisfaction appears to be too slow as well. It's well known in the field of constraint satisfaction that we select the most constrained x_i and that would correspond to the largest a_i . It turns out we can do better than this and instead select the a_i with smallest 2-adic valuation $v_2(a_i)$. The idea being that some of the low order bits of x_i can be

calculated using multiplicative inverses.

We will follow an example to see the sort of pruning done. We have the crouton < 16, 17, 67 > at depth r = 8 for an addition chain for n = 152557 of length d = 21 < l(152557) = 22. So this search will eventually fail. We have $a_1 = 16$, $a_2 = 17$ and $a_3 = 67$. Simple bit counting has $\left\lceil \frac{v(152557)}{v(16)+v(17)+v(67)} \right\rceil = 2$. So $u_i \leq 1536$ and $l_i \geq 1$. Using 4 we know that $s(\{x_1, x_2, x_3\}) = 1$ and hence $v(x_i) \leq 2$. With these bounds we can compute the initial bounds for x_3 :

$$1521 \le x_3 \le 1536$$

 $1536 \le x_3 \le 1536$ (using $v(x_3) \le 2$)

The value $x_3 = 1536$ can immediately be dismissed since v(152557-67.1536) = 9 and $v(a_1x_1 + a_2x_2) \le 6$. If this were not the case we would set $a_3 = 1536$ and repeat the bounding process for a_2 with n = 152557 - 67.1536. Pruning techniques like the binary digit count of x_i work well because they can eliminate a large number of possibilities quickly. This examples shows another way to reduce the range via the most significant bit of $u_3 = 1536$. If we assume that $x_3 \ge 2^{\lambda(u_3)}$ then

$$6 \ge v(n - a_3 x_3)$$

$$= v(n - 2^{\lambda(u_3)} a_3 - a_3(x_3 - 2^{\lambda(u_3)}))$$

$$\ge v(n - 2^{\lambda(u_3)} a_3) - v(a_3)(v(x_3) - 1)$$

$$> 9$$

So $x_3 < 1024$. This process can obviously be repeated for the new most significant bit of u_3 if there is still a valid range for x_3 .

Results

To see the potential of this pruning technique in searching for addition chains I took the code used by Ed Thurber to investigate the number of optimal addition chains for n defined as NMC(n). The code had a simple stack of numbers representing the end of a partial addition chain at various depths. New elements are pushed onto the stack by summing all previous elements and pushing them on the stack if they satisfy the addition chain bounds. Duplicates are removed. The code was changed to push croutons on a stack and prune not only with the addition chain bounds but also the bounds described by 4, 2, 6 with modulo 2^k and 7 with the multiplier of 3. The new code searches for formal addition chains were chain construction is differentiated by how elements are constructed as well as their value. So we must filter out duplicates. Despite this we see a big improvement in performance as show in table .

Bahig [2] conjectured that there is always an optimal addition chain for n with the last $\left\lfloor \frac{l(n)}{2} \right\rfloor$ steps star. The first counter example is 5979345 whose 16 optimal addition chains of length 27 each have two small steps:

Table 1: Times to Calculate NMC values

		DIC I. IIIICO CO
$n \leq$	Old Seconds	New Seconds
2^{8}	0.3	0.2
2^{9}	2.8	1.1
2^{10}	26.8	6.0
2^{11}	251.1	32.8
2^{12}	2676.0	199.4
2^{13}	27248.6	1226.5

Table 2: Effects of different Pruning Techniques

n	STN	CR	CR+G	CR+GB	CR+GBD
5979345 All chains	1.1	1.7	0.7	0.1	< 0.1
s(375494703) > 6	554.4	886.1	310.5	29.3	5.2
s(c(27) = 2211837) > 5	6.3	10.2	4.1	0.6	0.1
s(c(30) = 14143037) > 6	49.8	85.3	30.3	4.6	0.8
s(c(32) = 46444543) > 6	292.3	483.8	168.5	16.0	2.5
s(6580895885) > 6	549.4	945.0	286.2	33.1	6.2
s(6442718499) > 6	714.3	1158.7	383.8	38.4	7.3
s(6442517597) > 6	652.0	1125.3	336.1	27.4	5.3
$s(2^{31} - 1) > 6$	53.9	100.5	29.7	0.3	0.1
$s(2^{35} - 1) > 6$	100.5	188.5	52.6	0.5	0.1
$s(2^{37} - 1) > 6$	135.2	257.2	68.6	0.2	0.1
$s(2^{39} - 1) > 6$	179.4	340.6	112.0	0.1	< 0.1
$s(2^{41} - 1) > 6$	234.7	449.7	113.1	< 0.1	< 0.1

STN - Single value algorithm [6]

CR - Crouton, +G with GCD pruning, +B bounds 4 +D memoized l(n) data.

 $1\ 2\ 4\ 8\ 16\ 17\ {\color{red}32}\ 49\ 81\ 162\ 324\ 648\ 1296\ 2592\ 5184\ 5201\ {\color{red}10368}\ 20736\ 25937$ $46673\ 93346\ 186692\ 373384\ 746768\ 1493536\ 2987072\ 2992273\ 5979345$

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Table 3: Time vs. Binary Digit Sum

n	CR+GE	CR+GED
$s(2^{47} - 1) > 7$	104.4	36.8
$s(2^{53} - 1) > 7$	61.9	20.6
$s(2^{55}-1) > 7$	60.2	21.7
$s(2^{57} - 1) > 7$	41.2	20.6
$s(2^{58}-1) > 7$	48.0	16.7
$s(2^{59}-1) > 7$	39.9	35.8
$s(2^{61} - 1) > 7$	31.8	30.5
$s(2^{62} - 1) > 7$	37.6	14.5
$s(2^{63}-1) > 7$	33.4	16.1

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