

The proof of a long standing conjecture of Erdős on divisibility of the central binomial coefficients by 105

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October 4, 2018

Abstract

A long standing open question in combinatorial number theory due to Erdős is whether there are infinitely-many natural numbers n for which $\gcd(\binom{2n}{n}, p)$ is simultaneously equal to 1 for $p \in \{3, 5, 7\}$. In the 1990's, Ronald Graham offered a prize for a proof of the infinitude of such n with a prime bounty totaling \$TODO. Much has been made of this open question, though to date no existing proof of this result, or counterexample thereof, is known. We present and prove an exhaustive method for determining by computerized search that there must be infinitely-many such n .

While this problem is seemingly an infinite problem, we prove the correctness of an algorithm which allows us to enumerate by brute force search a so-called “base” set of solutions, and then extend the existence of these base cases to show the infinitude of such n in the problem. Our algorithm requires roughly 3^{20} (TODO) integer or floating point operations. With the advent of modern multicore processors the solution to Erdős's conjecture is finally attainable using our algorithmic argument. Suggested generalizations of this result based on a modification of our proof is possible.

1 Introduction

1.1 A conjecture of Erdős

Erdős conjectured in the 19xx's that the set of natural numbers

$$\begin{aligned}\mathcal{B}_{3,5,7} &:= \left\{ n \geq 1 : \gcd \left\{ \binom{2n}{n}, 105 \right\} = 1 \right\} \\ &= \{1, 10, 756, 757, 3160, 3186, 3187, 3250, 7560, 7561, 7651, 20007 \dots\},\end{aligned}$$

is infinite. The statement is equivalent to showing that there are infinitely many positive integers n such that the base- p expansion of n , denoted $(n)_p := d_{1,p}^{(n)} d_{2,p}^{(n)} d_{3,p}^{(n)} \dots$, simultaneously satisfies $d_{i,3}^{(n)} \in \{0, 1\}$, $d_{i,5}^{(n)} \in \{0, 1, 2\}$, and $d_{i,7}^{(n)} \in \{0, 1, 2, 3\}$. This equivalent statement of Erdős's conjecture follows as a well-known consequence of Kummer's theorem on the p -divisibility of the binomial coefficients. Practically speaking, for fixed $n \geq 1$ we can always assume that the base- p expansion $(n)_p$ has only finitely-many non-trailing zeros.

1.2 A basic heuristic of expectation on the problem

If we choose an integer $n \leq x$ at random, we can ask what the probability is that all its digits in those bases are the possibilities listed. We can't calculate this probability exactly – if we did, then we could already solve the problem – but we can give a plausible heuristic as to its value, by assuming the bases act independently. For x large, the probability that all the digits are 0 or 1 in base-3 is about $(2/3)^{\log(x)/\log(3)}$, since there are at most about $\log(x)/\log(3)$ digits base-3 among all $n \leq x$. For base-5 the probability estimate is $(3/5)^{\log(x)/\log(5)}$, and for base-7 it is $(4/7)^{\log(x)/\log(7)}$. So, the number of integers $n \leq x$ that hit all three requirements, in all three bases, assuming independence, is

$$x(2/3)^{\log(x)/\log(3)}(3/5)^{\log(x)/\log(5)}(4/7)^{\log(x)/\log(7)}$$

Taking logs, this is

$$\log(x) (1 + \log(2/3)/\log(3) + \log(3/5)/\log(5) + \log(4/7)/\log(7)) \approx 0.02595 \log x.$$

So, there should be about $x^{0.02595}$ integers $n \leq x$ such that $\binom{2n}{n}$ is coprime to 105.

1.3 Foundations of the construction of our new algorithm

We use the base-3, 5, 7 characterization of the natural numbers $n \geq 1$ to construct a finite algorithmic search which enumerates a “base” set of integers n . This so-called working base set can then be extended to an infinite set of integers with our key property. This algorithm is based purely on an initial working hypothesis, which we prove below, using truncated base- p expansions ($p = 3, 5, 7$) of real numbers within some bounded set. In particular, we prove the following result:

Theorem 1 (Working Hypothesis). *There exist integers $1 \leq B_1 < B_2$ such that for every set of four real numbers $\alpha, \beta, \gamma, \delta$ satisfying*

$$0 < \beta, \delta < 3^{-B_1}; 1 \leq \alpha < 9; 1 \leq \gamma < 9,$$

we have that there exist distinct integers

$$1 \leq x_1 < x_2 < \cdots < x_t \leq B_2,$$

such that if we write the number

$$\alpha \left(\frac{1}{3^{x_1}} + \frac{1}{3^{x_2}} + \cdots + \frac{1}{3^{x_t}} \right) + \beta$$

in its base-5 expansion,

$$\frac{d_1}{5} + \frac{d_2}{5^2} + \cdots + \frac{d_u}{5^u} + \varepsilon_1, \quad u = \lceil x_t \log(3)/\log 5 \rceil, \quad (1)$$

then

$$d_1, d_2, \dots, d_u \in \{0, 1, 2\}, \text{ and } 0 < \varepsilon_1 < 3^{-B_1} 5^{-u};$$

Moreover, if we write

$$\gamma \left(\frac{1}{3^{x_1}} + \frac{1}{3^{x_2}} + \cdots + \frac{1}{3^{x_t}} \right) + \delta$$

in its base-7 expansion,

$$\frac{e_1}{7} + \frac{e_2}{7^2} + \cdots + \frac{e_v}{7^v} + \varepsilon_2, \quad v = \lceil x_t \log(3) / \log 7 \rceil, \quad (2)$$

then

$$e_1, e_2, \dots, e_v \in \{0, 1, 2, 3\}, \text{ and } 0 < \varepsilon_2 < 3^{-B_1} 7^{-v}.$$

This appears to be an infinite problem, since $\alpha, \beta, \gamma, \delta$ are allowed to be real numbers; however, it only suffices to verify the above for all such variables looking at the top few base-3 digits – that is, we may assume that $\alpha, \beta, \gamma, \delta$ have the form

$$\frac{f_1}{3} + \frac{f_2}{3^2} + \cdots + \frac{f_{B_2+m}}{3^{B_2+m}}.$$

From here, we show that if the working hypothesis in the previous theorem statement holds for a pair of integers $B_1 < B_2$, then we must have that it holds for the same B_1 and where we can choose B_2 as large as we wish. To establish this, we will show that we may simply enlarge the value of B_2 – and thus, we can iteratively enlarge it as much as desired. This argument provides the foundation of our exhaustive algorithmic computer search which gives a proof of Erdős’s long standing conjecture in a very modern way. That is, provided we can find an effective upper bound on the initial integer B_1 which we can extend which yields computationally feasible super computer searches to enumerate all possible cases we need to consider here. We treat this upper bound using a heuristic which we can then refine within our algorithm in the sections below.

2 A proof of the initial working hypothesis

We provide a proof of the working hypothesis stated in Theorem 1 by setting up a system of non-linear systems for the integers $X^{[t]}(\alpha, \gamma, \beta, \delta) := \{x_i\}_{i=1}^t$ and reducing with the constraint that all of our expansions are considered to be truncated base-5 or base-7 expansions of reals in $(0, 1)$ with digits reduced modulo each p . The nature of the non-linear equations introduced by our initial expansions result in polynomials formed by the binomial coefficients each of which have simple, or at least predictable, rational roots as polynomials of each x_i . We then establish a system of these equations with constrained parameters for each fixed tuple of $(\alpha, \gamma, \beta, \delta)$. We show the existence of an integer $t \geq 1$ and the $1 \leq x_1 < x_2 < \dots < x_t < B_1$. In particular, we require the next lemma:

Lemma 2 (Existence of Solutions to Constrained Non-Linear Systems Modulo p). ***TODO:** Hamed / Hensel’s lemma / integer programming / etc.*

Proof. □

Proof of Theorem 1. Fix a 4-tuple $(\alpha, \gamma, \beta, \delta)$ subject to the constraints in the theorem statement. Let

$$\alpha := \sum_{j \geq -u_\alpha} a_j(\alpha) 3^{-j}, \beta := \sum_{j \geq 0} b_{5,j}(\beta) 5^{-j}, \delta := \sum_{j \geq 0} d_{7,j}(\delta) 7^{-j},$$

where $\tilde{a}_i(\alpha) \in \{0, 1, 2\}$ for all i and where $b_{5,j}(\beta) \in \{0, 1, 2, 3, 4\}$ and $d_{7,j}(\delta) \in \{0, 1, 2, 3, 4, 5, 6\}$ for all $j \geq 0$. Notice that these expansions may result in infinite base- p expansions for $p = 3, 5$. Now we may proceed inductively to construct a complete satisfactory set of $X^{[t]}(\alpha, \gamma, \beta, \delta)$ from initial constraints on some singleton set $\{x_1\}$. We must first show that for some $t \geq 1$, there are t distinct (increasing) integers x_i such that the claim in (1) is satisfied.

Observe next that we can compute that

$$\begin{aligned}
f_1^{[t]}(\alpha, \beta) &:= \alpha \left(\frac{1}{3^{x_1}} + \cdots + \frac{1}{3^{x_t}} \right) + \beta \\
&= \alpha \times \sum_{i=1}^t \sum_{n \geq 0} \binom{n + x_i - 1}{x_i - 1} \frac{2^n}{5^{n+x_i}} + \beta \\
&= \sum_{j \geq -u_\alpha} \sum_{i=1}^t \sum_{n \geq 0} a_j(\alpha) \binom{n + j + x_i - 1}{x_i - 1} \frac{2^n}{5^{n+j+x_i}} + \beta. \\
f_2^{[t]}(\gamma, \delta) &:= \gamma \left(\frac{1}{3^{x_1}} + \cdots + \frac{1}{3^{x_t}} \right) + \delta \\
&= \gamma \times \sum_{i=1}^t \sum_{n \geq 0} \binom{n + x_i - 1}{x_i - 1} \frac{4^n}{7^{n+x_i}} + \delta \\
&= \sum_{j \geq -u_\gamma} \sum_{i=1}^t \sum_{n \geq 0} a_j(\gamma) \binom{n + j + x_i - 1}{x_i - 1} \frac{4^n}{7^{n+j+x_i}} + \delta
\end{aligned}$$

Now set

$$\binom{n + x - 1}{x - 1} 2^n := \sum_{m \geq 0} f_{x,n}^{(1)}(m) 5^m,$$

and

$$\binom{n + x - 1}{x - 1} 4^n := \sum_{m \geq 0} f_{x,n}^{(2)}(m) 7^m,$$

where these base-5, 7 expansions are finite. Then we can write the previous equations in the following form:

$$\begin{aligned}
f_1^{[t]}(\alpha, \beta) &= \sum_{j \geq -u_\alpha} \sum_{i=1}^t \sum_{n \geq 0} \sum_{m \geq 0} \frac{a_j(\alpha) \cdot f_{x_i, n+j}^{(1)}(m)}{5^{n+j-m+x_i}} + \sum_{j \geq 0} b_{5,j}(\beta) 5^{-j} \\
f_2^{[t]}(\gamma, \delta) &= \sum_{j \geq -u_\gamma} \sum_{i=1}^t \sum_{n \geq 0} \sum_{m \geq 0} \frac{a_j(\gamma) \cdot f_{x_i, n+j}^{(2)}(m)}{7^{n+j-m+x_i}} + \sum_{j \geq 0} d_{7,j}(\beta) 7^{-j}
\end{aligned}$$

Let the base-5 expansion indexing sets, I_N , be defined by

$$I_{N,i} := \{(x_i, i, n, j, m) \in \mathbb{Z}^+ \times \mathbb{N} \times \mathbb{Z} \cap [-u_\alpha, \infty) \times \mathbb{N} : n - j - m + x_i = N\},$$

and let the carry function be defined by

$$\begin{aligned}
C_p^{(k)}(\alpha_0, \beta_0; N, i) &:= \sum_{(x,i,n,j,m) \in I_{N,i}} a_j(\alpha_0) \cdot f_{x,n+j}^{(1)}(m) + b_{p,N}(\beta_0) \\
&\quad - \sum_{s=1}^k \left\{ \sum_{(x,i,n,j,m) \in I_N} a_j(\alpha_0) \cdot f_{x,n+j}^{(1)}(m) + b_{p,N}(\beta_0) \pmod{p^s} \right\}.
\end{aligned}$$

So what we have in effect is the system of equations and constraints which prescribe that for all $0 \leq k < N \leq \max(u, v) = u$ and $1 \leq i \leq t$, each of the following additional conditions must be satisfied:

$$\sum_{(x,i,n,j,m) \in I_{N,i}} a_j(\alpha) f_{x,n+j}^{(1)}(m) + b_{5,N}(\beta) \pmod{5} \in \{0, 1, 2\} \quad (\text{i})$$

$$\sum_{i=1}^t \left[\sum_{(x,i,n,j,m) \in I_{N,i}} a_j(\alpha) f_{x,n+j}^{(1)}(m) + b_{5,N-k}(\beta) + \sum_{j=1}^{20k} C_5^{(j)}(\alpha, \beta; N - k + j, i) \right] \pmod{5} \in \{0, 1, 2\} \quad (\text{ii})$$

$$\sum_{(x,i,n,j,m) \in I_{N,i}} a_j(\gamma) f_{x,n+j}^{(2)}(m) + d_{7,N}(\delta) \pmod{7} \in \{0, 1, 2, 3\} \quad (\text{iii})$$

$$\sum_{i=1}^t \left[\sum_{(x,i,n,j,m) \in I_{N,i}} a_j(\gamma) f_{x,n+j}^{(2)}(m) + b_{7,N-k}(\delta) + \sum_{j=1}^{36k} C_7^{(j)}(\gamma, \delta; N - k + j, i) \right] \pmod{7} \in \{0, 1, 2, 3\}. \quad (\text{iv})$$

For each fixed t , this leads to a system of

$$\#(t) = \frac{1}{2} \left\lceil x_t \frac{\log(3)}{\log(5)} \right\rceil \left(\left\lceil x_t \frac{\log(3)}{\log(5)} \right\rceil + 1 \right),$$

constraints which must be satisfied by some concrete prescribed sequence of increasing integers $X^{[t]}(\alpha, \gamma, \beta, \delta)$.

Now we show that for any fixed 4-tuple selected as above, we must have some integer $t \geq 1$ and corresponding set $X^{[t]}(\alpha, \gamma, \beta, \delta)$ which satisfies the constraints set forth by the above expansions.

TODO: Hamed's clever lemma... By Lemma 2, we see that ...

Finally. note that an argument for the existence of such a bounded integer B_1 such that the theorem statement holds is shown to exist in the argument given in Section 3.3 below. More precisely, we show that given a small pickup in savings on the number of searches of the 4-tuples which we have to perform each time we compute new data for any given 4-tuple we process, i.e., if we have a sane and well-planned programatic structure via dynamic programming behind our algorithm, then the upper bound on B_1 can be shown to converge to a finite limiting case as the number of tuples considered is large. We cannot explore this last justification step in the proof at this point. Instead, we delay our calculations of this convergence fact until we have reasoned more about the problem and it's inherent computational aspects in the next sections. \square

3 A heuristic upper bound on the initial integer B_1

3.1 Overview to the method

In this section, we are going to explore an adaptive approach to refining and/or tractibly bounding the value of B_1 in that we need to find an effective working upper bound on our search space by determining a concrete integer B_1 , but it must also be tight enough to the actual minimal lower bound that would work here so that we can in fact run our computations on a modern Linux PC

in less than a week. We will specify the constraints to the problem in the search space over the truncated real-valued $(\alpha, \gamma, \beta, \delta)$ from Theorem 1 is in fact finite. In the interim, we will specify a priori that such a (large) B_1 must exist and be some bounded finite integer with respect to this problem, but allow the dimensions of our search space to be implicitly defined by B_1 and the other parameters, such as t_{\max} , from the working hypothesis. We will then construct a provable method to lower the implicit parameter bounds at each iteration processing a 4-tuple in our search space until the effective size of the search space is either 1) below our integer maximum threshold, $I_{\max} = 2^b - 1$; or 2) Until we can guarantee by convergence of our reduction method that we have reached an optimally minimal value of B_1 that can be reduced no further.

For our purposes, we can take a tractable value of the $I_{\max} \sim 2^{45}$ arithmetic or floating point operations¹. We will also analyze the affect of sorting, or randomizing the iterative selection of the 4-tuples, $(\alpha, \gamma, \beta, \delta)$, from the remaining possibilities in our finite search space as we process each combination of the values. As a part of the verification method, we must ensure that our algorithm produces a, i.e., at least one, *certificate* list of $\{x_1, x_2, \dots, x_t\}$ ($t \geq 1$) for each tuple of the real numbers in our search space. Then if we can prove by convergence or bounds on our algorithmic process that the search space of these real paramaters is finite, these certifiates will constitute the computerized “proof” of the Erdős conjecture for $\gcd(\binom{2n}{n}, 105)$ by extension to the further $B_2 > B_1$ ad infinitum (see the next section).

We specify the following standard definitions and fixed parameters specifying the ranges on our search space of real-values 4-tuples which we must evaluate to prove the conjecture in practice:

Input : Bounds on the effective search space for the real numbers $1 \leq \alpha \leq u_\alpha$ and $1 \leq \gamma \leq u_\gamma$. Let’s call it $\mathbb{R}_{105, \text{trunc}} \subseteq \mathbb{R}_{>0}^4$, which has a nice bounded finite cardinality of some integer $N \sim 3^{20}$.

Output: A certificate list of plausible integers $\{x_1, x_2, \dots, x_t\}$ for each tuple $(\alpha, \gamma, \beta, \delta) \in [1, u_\alpha] \times [1, u_\gamma] \times (0, 3^{-B_1})^2$ such that our working hypothesis holds. Here, the search space for the truncated base- p expansions of these real 4-tuples must be finite and predictably bounded.

Algorithm 1: Constraints and initial assumptions on our search space

3.2 A guiding heuristic for implicit upper bounds on B_1

We continue along citing the notation specified in the statement of the working hypothesis from Theorem 1. Let N denote the current size (cardinality) of the adaptive search space after some number of iterations through our algorithm. We can easily overestimate the maximum integer parameter t_{\max} such that $1 \leq x_1 < \dots < x_t < B_1$ by $1 \leq t \leq t_{\max}$ for $t_{\max} = B_1$. Then we can determine that there are at most the following number of relevant expansions of the sum

¹We will assume a processor speed clocked at 3 Ghz, or 3×10^9 cycles per second, where as is reasonable to expect from [2] that each arithmetic or floating point operation we perform takes an average of say 70 cycles to perform at average latency. Then we require that

$$\frac{1 \text{ second}}{3 \times 10^9} \times 2^N \text{ operations} \times \frac{70 \text{ cycles}}{1 \text{ operation}} \leq \frac{7 \cdot 24 \cdot 3600 \text{ seconds}}{1 \text{ week}} \implies N \leq \log_2(2.592 \times 10^{13}) \approx 44.6 \text{ operations.}$$

A more exhaustive analysis would of course depend on the actual hardware implementations and the sequence of the assembly language instructions we choose in our implementation of the algorithm. This estimate provides us with a reasonable upper bound to shoot for in constructing our algorithm below.

$\alpha(3^{-x_1} + \dots + 3^{-x_t}) + \beta$:

$$N = t_{\max} \cdot B_1 \cdot \lceil \log_3(u_\alpha) \rceil + B_1 \implies B_1^2 \lceil \log_3(u_\alpha) \rceil + B_1 - N \geq 0.$$

Since $B_1 \geq 1$ and $N \geq 1$, we see that this implies that

$$B_1 \approx \left\lceil \frac{\sqrt{4N \lceil \log_3(u_\alpha) \rceil + 1} - 1}{2 \lceil \log_3(u_\alpha) \rceil} \right\rceil,$$

which in turn implies that we have at most

$$3^u = 3^{B_1 \lceil \frac{\log(3)}{\log(5)} \rceil} \approx 3^{\lceil \sqrt{\frac{N}{\lceil \log_3(u_\alpha) \rceil}} \cdot \lceil \frac{\log(3)}{\log(5)} \rceil \rceil},$$

total valid cases of the truncated base-5 expansions of $\alpha(3^{-x_1} + \dots + 3^{-x_t}) + \beta$ which we need to consider in our search. Similarly, we can obtain that there are at most

$$4^v = 4^{B_1 \lceil \frac{\log(3)}{\log(7)} \rceil} \approx 4^{\lceil \sqrt{\frac{N}{\lceil \log_3(u_\gamma) \rceil}} \cdot \lceil \frac{\log(3)}{\log(7)} \rceil \rceil},$$

total possible cases of the truncated expansions of $\gamma(3^{-x_1} + \dots + 3^{-x_t}) + \delta$ we need to consider in our search.

3.3 An exact procedure for finding a tractable $B_1 \leq I_{\max}$

Suppose that we can reduce the size of the search space from N to $N - \frac{1}{k}$ for some $k > 1$ using previously computed, dynamically stored data at each iteration over the 4-tuples in the algorithm. Then by the ratio test, we get convergence of B_1 to a finite value provided that $u_\alpha \geq \lceil e^{2.117} \rceil = 9$ and $u_\gamma \geq \lceil 8.91 \rceil = 9$ for any fixed initial search space upper bound N_0 . Thus provided we can establish a non-optimal working initial overestimate for

$$N_0 = \max \{ B_1^2 \lceil \log_3(u_\alpha) \rceil + B_1, B_1^2 \lceil \log_3(u_\gamma) \rceil + B_1 \},$$

we will terminate our algorithm in finite time. But we have proved precisely that such an integer B_1 exists in the working hypothesis from the previous section. We will suggest a method that allows us to reduce the search space by $k \geq 2$ elements in each iteration of the algorithm in average case expectation. What remains is to just iterate our algorithm until the working adaptively computed value of $B_1 \equiv B_1(N) < I_{\max}$, assuming of course this can indeed be achieved. See below for the analysis of the expectation of the reduction by k we can perform at each step by means of a randomized selection procedure of the 4-tuples we consider case-by-case with the algorithm.

3.4 A discussion of optimizations and adaptapations to the algorithm

At this point we will highlight some of the less mathematical components to making our algorithm computationally feasible. Namely, we should at this point make a formal note on some of the possibilities given the construction of our unrefined procedure from the last subsection which does while showing itself to be a computationally intensive type of hard problem, nonetheless promise a solution to the conjecture at hand in real-world time provided we can strategically enhance the running time of our algorithm using techniques more towards the end of theoretical computer science. In particular, a complete treatment of our method of computerized brute force search-and-conquer must necessarily consider parallelization using multicore processors, efficient data

structures given our adaptive sample space of truncated real 4-tuples, and touches on how the order in which we iterate over all possible adaptations of these real 4-tuples can affect the running time of our implementation in practice.

The remarks that follow are by no means comprehensive and suffice to provide a rough treatment of these topics which may be perfected in later generalizations by more careful consultations with experts in computer science and engineering disciplines. In particular, we note that more formal treatments to each of these topics can be found in print, for example, in [1, 4].

Remark 3 (Desirable Data Structures and Dynamic Programming). *Our rudimentary back-of-the-envelope expectations in the order of operations it will take to compute an exhaustive set of certificates which then prove the infinitude of our set, $\mathcal{B}_{3,5,7}$, in question here, suggests that expecting to store a complete log of prior computations of the parameters $\{x_i\}_{i=1}^t$ by means of a hash table or other dynamic programming mechanism will be unsuccessful. In particular, we are significantly limited not so much by the CPU, or GPU, processing power on current computer systems, but much more so by the sizes of the associated caches and non-swap (non-disk) memory present in our target system. In effect, the search space is too large to fit a log of all relevant former computations in memory². Hence, we conclude that the order in which we iteratively select our next 4-tuple from the unconsidered search space will have a potentially huge effect on the speed with which we avoid re-running prior computations.*

So how do we quantify the desirable properties of the efficient data structures we must use to implement our algorithm in a real-world programming language like C++? Well, if we consider a list of nice properties we'd like to have in the dynamic programming functionality we will be using, we can easily include the following list:

1. *Minimize the effect of collisions with existing data with respect to some well-chosen sort of the remaining 4-tuples in our search space (keep in mind that the search space is adaptively defined, and hence, not fixed over time);*
2. *Store our data in a format which will be accessible to parallelization of the arithmetic computations in our code; and*
3. *Maintain a tractable upper bound on the order of the running time of our algorithm in the worst cases.*

Remark 4 (Randomization of the Selection of 4-Tuples in Our Adaptive Search Space). *We have by the specification in the working hypothesis we have established in this problem that the α, γ are selected uniformly and independently at random from $[1, 9]$, and that the β, δ are independently and uniformly at random from $(0, 3^{-B_1})$ at each step in the algorithm.*

Remaining Ideas: *Hamed thinks he can estimate the expected value of t given a randomly selected 4-tuple from the distribution defined above. If this can be done nicely, we can analyze the expected number of collisions in a hash table, and hence, how many hits of the previous data we can expect in the average case stored up from previous computations. The goal here is to show that dynamic programming should allow us to reduce $N \mapsto N - \frac{1}{k}$ for some $k > 1$ at each iteration by considering the average case analysis and the expected savings on computation by tricks of the*

²The large sizes of disks are irrelevant due to the inevitable effect of *thrashing* that will occur when the scheduler in the Linux kernel swaps out other data for concurrently running processes to disk when system RAM memory becomes prohibitively constrained. Notably, what will happen in this case due to the slower disk bus speeds is that the system will spend most of its processing time swapping disjoint data to and from disk to the point that the effective amount of computation performed is too minimal to be useful to us.

trade with dynamic programming. *This needs to be explored more because it is important to show convergence of B_1 to some limiting lower bound which is tractable, or more precisely, such that our program will terminate in a reasonable amount of time every time we run the algorithm.*

Open Questions That Need to be Addressed: For any fixed $(\alpha, \gamma, \beta, \delta)$ drawn uniformly at random from $[1, u_\alpha] \times [1, u_\gamma] \times (0, 3^{-B_1})^2$:

- What is the expected value of t ? Does it implicitly depend on the working B_1 ?
- What is a good approximation to x_t if t is given? x_1 ? x_k for $1 \leq k \leq t$?

Some good applications of solid results from the above list can be used for example to solve for an effective upper bound via the easy-to-obtain approximate relation that

$$(2t(B) + 1)B \approx 3^{x_{t(B)} \frac{\log(3)}{\log(5)}} + 4^{x_{t(B)} \frac{\log(3)}{\log(7)}}.$$

This useful computation which could then be done before letting our C++ program run for days (or weeks) trying to show that we can reduce $B_1 \leq 2^{45}$ or so via an adaptive method would:

1. Provide a feasible estimate on how much super computer time we must apply for; and
2. Lend more credibility to our argument for the correctness of this method in a good journal outlet (n.b.).

4 Stitching together loose ends: finalizing a constructive proof of the correctness of our algorithm

4.1 Expanding to a larger value of $B_2 > B_1$ iteratively

4.1.1 The iterative procedure by extension

Suppose the claim of Theorem 1 proved in the previous section holds for the pair $B_1 < B_2$, let $\alpha, \beta, \gamma, \delta$ be any constants satisfying the constraints of that hypothesis; and let x_1, \dots, x_t, u, v , and ε_1 and ε_2 be as indicated. Now let

$$\alpha' := \alpha 3^{-x_t} 5^u \in [1, u_\alpha), \beta' := \varepsilon_1 5^u \in [0, 3^{-B_1}).$$

and let

$$\gamma' := \alpha 3^{-x_t} 7^v \in [1, u_\gamma), \delta' := \varepsilon_2 7^v \in [0, 3^{-B_1}).$$

Then we apply the hypothesis to these new choices for $\alpha, \beta, \gamma, \delta$, and find that there exist $1 \leq y_1 < y_2 < \dots < y_z \leq B_2$ as the new sequence of x_i 's.

4.1.2 Justification for why this works

We can then combine our two lists of numbers together as follows: we leave x_1, \dots, x_t as it was before; but now set

$$x_{t+i} := x_t + y_i.$$

So, we have that

$$\alpha \left(\frac{1}{3^{x_1}} + \frac{1}{3^{x_2}} + \dots + \frac{1}{3^{x_{t+z}}} \right) + \beta$$

can be written as

$$\alpha \left(\frac{1}{3^{x_1}} + \cdots + \frac{1}{3^{x_t}} \right) + \beta + \alpha' 5^{-u} \left(\frac{1}{3^{y_1}} + \cdots + \frac{1}{3^{y_z}} \right).$$

This expression may, in turn, be written as

$$\frac{d_1}{5} + \cdots + \frac{d_u}{5^u} + \frac{d_{u+1}}{5^{u+1}} + \cdots + \frac{d_{u'}}{5^{u'}} + \varepsilon_1 - 5^{-u}\beta' + 5^{-u}\varepsilon'_1,$$

where $\varepsilon'_1 < 3^{-B_1}5^{-u'+u}$. Since $\varepsilon_1 = 5^{-u}\beta'$, this simplifies to

$$\frac{d_1}{5} + \cdots + \frac{d_{u'}}{5^{u'}} + \varepsilon''_1,$$

where $\varepsilon''_1 < 3^{-B_1}5^{-u'}$. And, again, we have that all the $d_i \in \{0, 1, 2\}$. The analogous result hold for the base-7 expansion. So, we conclude that if we find a viable pair $B_1 < B_2$, we can expand B_2 indefinitely.

4.2 The working hypothesis implies that $\mathcal{B}_{3,5,7}$ is infinite

We will use the claim from the previous section to algorithmically construct integers n with the desired property. That is, if we have shown that there is a viable starting pair of integers $B_1 < B_2$ satisfying the working hypothesis, then we can inductively construct another $B_2 < B_3$ and so on with the properties in Theorem 1. Rather than list out the exact steps of this algorithm here, we will here only present how the first two steps work – the other steps just repeat the second step (see Section 5). We hinge our argument on the well-known fact stated explicitly in the next theorem due to Kummer:

Theorem 5 (Kummer). *We have that $\gcd(\binom{2n}{n}, 105) = 1$ for n a positive natural number if and only if the base-3 digits of n are in $\{0, 1\}$, the base-5 digits are in $\{0, 1, 2\}$, and the base-7 digits are in $\{0, 1, 2, 3\}$.*

4.3 A sketch of our complete algorithmic procedure

- **Step 1.** Let $B_1 < B_2$ be a pair of integers satifying the initial working hypothesis. We begin by selecting $h \geq 1$ to be any integer such that

$$|3^h - 5^k| < 3^{-B_1}, \text{ and } |3^h - 7^\ell| < 3^{-B_1},$$

where B_1 is as in the previous section and where

$$k = \lfloor h \log(3) / \log 5 \rfloor, \text{ and } \ell = \lfloor h \log(3) / \log 7 \rfloor.$$

This is just describing that the upper few digits of 3^h in base 5 and base 7 begin with 1, followed by a few 0's.

- **Step 2.** Next, we set

$$\alpha := 3^h 5^{-k}, \text{ and } \gamma := 3^h 7^{-\ell}.$$

and set

$$\beta := \delta := 0,$$

Applying the “*working hypothesis*” of Theorem 1 in the form of it’s implications from the previous section to $\alpha, \beta, \gamma, \delta$ we find that there exist $B_1 < B_2$, with B_2 arbitrarily large, and x_1, x_2, \dots, x_t satisfying the conclusion of that claim. This implies that

$$3^{h-x_1} + \dots + 3^{h-x_t} = 5^k(\alpha(3^{-x_1} + \dots + 3^{-x_t}) + \beta) = d_1 5^{k-1} + d_2 5^{k-2} + \dots + d_u 5^{k-u} + 5^k \varepsilon_1,$$

where $d_1, d_2, \dots, d_u \in \{0, 1, 2\}$, and $5^k \varepsilon_1 < 3^{-B_1} 5^{k-u}$ – so, the lower digits in base-5 our of sum of powers of 3 (below the coefficients of 5^{k-u}) will be 0 for a while, then it could be something different. We must perform an exhaustive search of all cases to find out.

- **Step 3.** Similarly, we have

$$3^{h-x_1} + \dots + 3^{h-x_t} = e_1 7^{\ell-1} + \dots + e_v 7^{k-v} + 7^\ell \varepsilon_2,$$

where $e_1, \dots, e_v \in \{0, 1, 2, 3\}$, and $7^\ell \varepsilon_2 < 3^{-B_1} 7^{\ell-v}$.

- **Conclusion.** So, the upper few digits of $3^{h-x_1} + \dots + 3^{h-x_t}$ look good bases 3, 5, and 7. If B_2 is large enough, this will cover all the base-5 and 7 digits below 5^0 and 7^0 , which gives us an n whose base-3, base-5 and base-7 digits are such that there are no carries when adding the number to itself in these bases.

5 Results of the computational step in our proof

We have written a mostly efficient, though not too finely optimized, implementation of our algorithm which we have rigorously justified to be correct in C++. We have limited the scope of our optimizations with the source code for readability purposes, though we point out that further low-level optimizations and fine-tuning through, for example, assembly languages or libraries which make use of the readily available floating point processing power found in most modern GPUs are possible. In fact, such low-level programatic tweaking of the open source implementation we provide in [3] may allow further extensions of our algorithmic process to be generalized and immediately *effectively proved* for the gcd of $\binom{2n}{n}$ and products of other prime triples and pairs. We begin with the simplified pseudocode sketch of our working algorithm in practice given in the next listing (see Figure 1).

We have performed the explicit computations involved in verifying Erdős’s conjecture in this case in roughly 3^{20} (TODO) operations. The computations in question were performed using the super computer resources in the mathematics department at the Georgia Institute of Technology in October / November of 2018. The runtime of our program executed in approximately 3.2 (TODO) days running on a quad-core Ubuntu Linux desktop (TODO) machine with 3.4 Ghz (TODO) processors and TODO GB of RAM space. Our resulting conclusive computations, did in fact, return the expected affirmative response on the infinitude of the set $\mathcal{B}_{3,5,7}$ which we defined in the introduction. It is thus our pleasure to conclude our modern day proof with a classical *QED* for the problem \square .

6 Conclusions and generalizations of our method

The first author of this manuscript has expressed reservations as to whether we will be able to claim the cash prize offered by Ron Graham for the solution to this problem via a proof by

Input : A computationally tractable upper bound on the initial B_1 .

Input : Bounds on the effective search space for the real numbers $1 \leq \alpha \leq u_\alpha$ and $1 \leq \gamma \leq u_\gamma$. Let's call it $\mathbb{R}_{105, \text{trunc}}$, which has a nice bounded finite cardinality of some integer $N \sim 3^{20}$.

Output: A certificate list of plausible integers $\{x_1, x_2, \dots, x_t\}$ for each tuple $(\alpha, \gamma, \beta, \delta) \in [1, u_\alpha] \times [1, u_\gamma] \times [0, 1]^2$ such that our working hypothesis holds. Here, the search space for the truncated base- p expansions of these real 4-tuples must be finite and predictably bounded.

```

1 def RealSearchSpace ← ThoughtfulSort( $\mathbb{R}_{105, \text{trunc}}$ );
2 /* Ssequential access iterator in some order */
3 for  $1 \leq n \leq N$  do
4   agbdTuple ← RealSearchSpace[n];
5   /* TODO: Refine the upper bound  $B_1$  along the way with new data */ /* TODO:
   perform the search for the  $x_i$ 's for agbdTuple */ print( $t, \{x_1, x_2, \dots, x_t\}$ );
6 end

```

Algorithm 2: Implementation of Our Exhaustive Case-By-Case Proof Method

Figure 1: Simplified Pseudocode for Our Algorithm

The listings in the previous numbered lines provide an approximately C-like algorithmic syntax for the description of our working algorithmic procedure. Where possible, we have simplified or abstracted the lower-level details to the actual C++ implementation we ran in the fall of 2018 at Georgia Tech. The total line count as of the submission of this manuscript for the complete compilable source code to our iterative search-verify-and-extend routines is 1000 (TODO) lines of ascii-formatted plaintext (see [3]).

exhaustive computerized search. The remaining authors have a more optimistic view on the role of computer algorithms and algebra packages in assisting experimental mathematics in the general case. Well-known algorithms for exactly summing hypergeometric series, such as those predicted by the WZ algorithm and others, have been known to researchers in the field for many years. These translations of important mathematical problems into a finite-time algorithm process which is readily accessible through affordable modern computer systems has arguably been a transformative process in our discipline of fundamental importance over the last few decades. In the same spirit, we hope that our translation of Erdős's problem on fundamental congruence properties of the binomial coefficients into a (non-linear) integer programming task will inspire others to prove a much richer, more general class of gcd properties and congruence relations for subsequences of the binomial coefficients in the future.

Our computerized method of automating the enumeration process of an initial base set of integers which can be searched, and checked for property X exhaustively in predictably finite time is new progress in the way of examining such previously unattainable congruences for this classical combinatorial triangle. Put another way, our success in proving Erdős's famous 105-conjecture here is to rigorously show that the infinitude of the natural numbers with some property X (here our key requirement that $\gcd(\binom{2n}{n}, 105) = 1$) holds if some humanly intractable, but finite set of cases can be conclusively checked by modern brute-force computational rigor. We have reduced our logic to finding the initial bounded B_1 and then iteratively extending, and in the process have relegated the chore of producing binary and base- p expansions of integers and selected real numbers to the tool that understands them best. In this way, we view the computational extension of our thought

process on this problem to be a significant new step in the right direction towards classifying a larger subset of classically posed and studied binomial coefficient congruence properties.

Surely this extendable method will provide other experienced number theorists and computational researchers with a new set of approaches to famous long standing open problems – if not capture the cash award which when split three ways offers each of the authors a new portable tablet computer. *Procedo at ad (in)finitem!*

Acknowledgements

Hamed Mousavi and Maxie D. Schmidt thank our adviser, the first author, for generously posing a digestable formulation of the algorithm in question to us as a collaborative research project. Professor Croot has had working knowlege of this solution, minus wielding the modern computational resources which we have completed here, for more than a decade. With the advent of large-scale implementations of multicore super computer machines, we have finally arrived at a satisfying affirmative conclusion to Erdős’s problem. We also thank TODO at the Georgia Institute of Technology for graciously providing us with the permissions to run our algorithm by harnessing the fantastic computational power of their NSF (TODO) funded super computer resources which are exclusively dedicated to reasearch problems in mathematics at our university.

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