

TITLE

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ABSTRACT. The sum-product conjecture, posed in 1983 by Erdős and Szemerédi [ES83], posits that any sufficiently large set must have a ‘relatively large’ number of distinct sums or products between its elements. A newer conjecture extends this idea, positing that for convex sets there must be a ‘relatively large’ number of distinct sums between its elements. Both conjectures remain open and far from being solved. This report provides an entirely self-contained overview of some known results and the methods used to achieve them.

issue of "for any set A" in statements, but only as A gets sufficiently large
GO THROUGH WHOLE THING AND USE TILDE NOTATION
introduce $f(A)$ notation

1. INTRODUCTION AND MOTIVATION

For any sets A, B and binary operation \cdot which acts on elements of A and B , we define

$$A \cdot B = \{a \cdot b : a \in A, b \in B\}.$$

Observe the following example which motivates the study of the sum-product problem.

Let $A = \{1, 2, 3, \dots, n\}$ and $G = \{2, 2^2, 2^3, \dots, 2^n\}$.

Notice that A is given by an arithmetic sequence and G by a geometric sequence. We are going to calculate $|A + A|$, $|AA|$, $|G + G|$, and $|GG|$.

Observe that

$$\begin{aligned} |A + A| &= |\{2, 3, \dots, 2n\}| \\ &= 2n - 1 \\ &= 2|A| - 1, \end{aligned}$$

and by the same argument,

$$\begin{aligned} |GG| &= |\{2^2, 2^3, \dots, 2^{2n}\}| \\ &= 2|G| - 1. \end{aligned}$$

We have

$$|G + G| = |\{2^i + 2^j : i, j \in \{1, \dots, n\}\}|$$

For all i, j , $2^i + 2^j$ is a number written in base 2. By the uniqueness of representations in different bases (see appendix), we have that $2^i + 2^j$ is distinct for every choice of $\{i, j\}$.

Therefore,

$$\begin{aligned} |G + G| &\geq |\{\{i, j\} : i, j \in \{1, \dots, n\}\}| \\ &= \binom{n}{2} + n \\ &= \binom{|G| + 1}{2} \end{aligned}$$

Finally, I prove in the appendix that there exists $c \in \mathbb{R}^+$ such that

$$|AA| \geq \frac{c |A|^2}{\log(|A|)}.$$

Observing the trivial bounds $2|S| - 1 \leq |S \cdot S| \leq \binom{|S|}{2}$ for any set S and any commutative operation \cdot , it is clear that both AA and $G + G$ are, as $|A|, |G| \rightarrow \infty$, almost as large as they can be, and $A + A, GG$ are as small as they can be.

This is the phenomenon which motivates this problem. One question is: “does there exist a set for which neither the sum nor the product set is large?” The sum-product conjecture states that such a set does not exist. Another question we study in this report is: “what determines if the sum set is large or the product set is large?” A conjecture which partially attacks this question states that the sum set is large when the set itself is convex.

The rest of the report will proceed in the following way : **finish this**

2. PRELIMINARIES

As a useful shorthand, for any natural number n , let

$$[n] = \{1, 2, \dots, n\}.$$

The study of these problems requires the notion of orders of magnitude. For any functions $f, g : \mathbb{R} \rightarrow \mathbb{R}$, write

$$f(x) \gg g(x) \text{ as } x \rightarrow \infty$$

if

$$\exists x_0, c \in \mathbb{R}^+ \text{ s.t. } x > x_0 \implies |f(x)| \geq c |g(x)|,$$

write

$$f(x) \ll g(x) \text{ as } x \rightarrow \infty$$

if

$$\exists x_0, c \in \mathbb{R}^+ \text{ s.t. } x > x_0 \implies |f(x)| \leq c |g(x)|,$$

and write

$$f(x) \asymp g(x) \text{ as } x \rightarrow \infty$$

if

$$f(x) \ll g(x) \text{ and } f(x) \gg g(x) \text{ as } x \rightarrow \infty.$$

We write $\ll_\epsilon, \gg_\epsilon$ if the constant depends on ϵ . For example,

$$f(x) \gg_\epsilon g(x)^\epsilon$$

means

$$\forall \epsilon > 0, f(x) \gg g(x)^\epsilon.$$

More precisely, there is some function $c : \mathbb{R}^+ \rightarrow \mathbb{R}^+$ so

$$\forall \epsilon > 0, \exists x_0 \in \mathbb{N} \text{ s.t. } x > x_0 \implies |f(x)| \geq c(\epsilon) |g(x)^\epsilon|.$$

We write \lesssim, \gtrsim if, along with a constant factor, there is also a logarithmic factor. That is

$$f(x) \lesssim g(x)$$

if there is some $c \in \mathbb{R}$ such that

$$f(x) \ll (\log(x))^c g(x).$$

Throughout this report, the asymptotic parameter (in this case x) will always tend to ∞ , so it will no longer be mentioned. Oftentimes the parameter will not even be in the expression. For example, if for some set A we write

$$|A + A| \gg |A|,$$

it is taken to mean that A is defined implicitly by $|A|$, and $|A|$ is the parameter which tends to ∞ .

For any sets A, B and any binary operation \cdot acting on elements of A and B , define the representation function $r_{A \cdot B} : A \cdot B \rightarrow \mathbb{N}$ by

$$r_{A \cdot B}(x) = |\{(a, b) \in A \times B : x = a \cdot b\}|.$$

Throughout this report the shorthand

$$\delta_{A,B}(x) = r_{A-B}(x), \quad \sigma_{A,B}(x) = r_{A+B}(x), \quad \delta_A(x) = \delta_{A,A}(x), \quad \sigma_A(x) = \sigma_{A,A}(x)$$

will be used.

For any sets A, B , define the Additive Energy $E(A, B)$ and Multiplicative Energy $M(A, B)$ by

$$E(A, B) = |\{(a_1, a_2, b_1, b_2) \in A^2 \times B^2 : a_1 - b_1 = a_2 - b_2\}|$$

and

$$M(A, B) = \left| \left\{ (a_1, a_2, b_1, b_2) \in A^2 \times B^2 : \frac{a_1}{b_1} = \frac{a_2}{b_2} \right\} \right|.$$

Observe that

$$E(A, B) = \sum_{x \in A-B} \delta_{A,B}(x)^2$$

and

$$M(A, B) = \sum_{x \in \frac{A}{B}} r_{\frac{A}{B}}(x)^2.$$

This definition is symmetric in the sense that a 4-tuple $(a_1, a_2, b_1, b_2) \in A^2 \times B^2$ is a solution to

$$a_1 - b_1 = a_2 - b_2$$

if and only if it is a solution to

$$a_1 + b_2 = a_2 + b_1,$$

and therefore

$$E(A, B) = \sum_{x \in A-B} \delta_{A,B}(x)^2 = \sum_{x \in A+B} \sigma_{A,B}(x)^2.$$

There is a similar argument for multiplicative energy. Any 4-tuple $(a_1, a_2, b_1, b_2) \in A^2 \times B^2$ with nonzero entries is a solution to

$$\frac{a_1}{b_1} = \frac{a_2}{b_2}$$

if and only if it is a solution to

$$a_1 b_2 = a_2 b_1.$$

There are at most

$$\sum_{i=1}^4 \binom{4}{i} = 15$$

4-tuples with zero entries, so

$$M(A, B) = \sum_{x \in \frac{A}{B}} r_{\frac{A}{B}}(x)^2 \asymp \sum_{x \in AB} r_{AB}(x)^2.$$

We also define higher energies

$$E_n(A, B) = \sum_{x \in A-B} \delta_{A,B}(x)^n,$$

so

$$E(A) = E_2(A),$$

and as a shorthand use

$$E_n(A) = E_n(A, A).$$

Similar definitions and shorthand are used for multiplicative energy.

We can relate the energies to the sizes of the sum and product sets by the Cauchy-Schwarz Inequality.

$$|A| |B| = \sum_{x \in A+B} \sigma_{A,B}(x) \leq |A+B|^{\frac{1}{2}} E(A, B)^{\frac{1}{2}},$$

and

$$|A| |B| = \sum_{x \in AB} r_{AB}(x) \leq |AB|^{\frac{1}{2}} M(A, B)^{\frac{1}{2}}.$$

Similar inequalities can be derived for $|A - B|$ and $|\frac{A}{B}|$.

Let $I \subset \mathbb{R}$ be an interval. We call a function $f : I \rightarrow \mathbb{R}$ convex if for any 2 points $x_1, x_2 \in I$, and any $\lambda \in [0, 1]$,

$$f((x_1 - x_2)\lambda + x_2) \leq (f(x_1) - f(x_2))\lambda + f(x_2).$$

A finite set $A \subset \mathbb{R}$ is convex if there is a function $f : [1, |A|] \rightarrow \mathbb{R}$ such that

$$A = \{f(i) : i \in \{1, \dots, |A|\}\}.$$

A property of convex functions which will come up later, and is worth proving now, is

Lemma 2.1. *Let $I \subset \mathbb{R}$ be an interval. Let $f : I \rightarrow \mathbb{R}$ be convex. Let $T \subset \mathbb{R}^2$.*

Take $\ell = \{(x, f(x)) : x \in I\}$ to be the graph of f . We have that

$$\forall t_1, t_2 \in T, \quad |(\ell + t_1) \cap (\ell + t_2)| \leq 1.$$

That is that translations of the graph of a convex function intersect in at most one point.

Proof. **PROVE THIS** □

Dyadic partitioning introduction

3. STATEMENT OF PROBLEMS

We'll now move on to precise statements of the mentioned problems, and statements of the most modern results.

The idea that there does not exist a set with a small sum and product set is stated precisely as

Conjecture 3.1 (Sum-Product Conjecture). *For every finite set $A \subset \mathbb{R}$,*

$$\max(|A + A|, |A \cdot A|) \gtrsim |A|^2$$

The idea that convexity opposes additive structure is stated precisely as

Conjecture 3.2 (Convex Sumset Conjecture). *For finite and convex set A ,*

$$|A + A| \gtrsim |A|^2$$

To date, the best results for both of these conjectures are proven in [RS21].

Theorem 3.3. *For every sufficiently large finite set $A \subset \mathbb{R}$,*

$$\max(|A + A|, |A \cdot A|) \gtrsim |A|^{\frac{4}{3} + \frac{2}{1167}}$$

and

Theorem 3.4. *For every sufficiently large, finite, and convex set A ,*

$$|A + A| \gtrsim |A|^{\frac{30}{19}}$$

These results will not be covered in this report. In this report, the strongest results proven are

Theorem 3.5. *For finite sets $A \subset \mathbb{R}$,*

$$\max(|A + A|, |AA|) \gtrsim |A|^{\frac{4}{3}}.$$

Theorem 3.6. *For finite and convex sets $A \subset \mathbb{R}$,*

$$|A + A| \gtrsim |A|^{\frac{14}{9}}.$$

and

Theorem 3.7. *For finite and convex sets $A \subset \mathbb{R}$,*

$$|A - A| \gtrsim |A|^{\frac{8}{5}}.$$

Theorem 3.5 is the weakest amongst these results, although it is still near the best known result. Theorem 3.7 gives the best known exponent on $|A|$, and Theorem 3.6 is obtained by a slight variation on the same methods.

4. GRAPHS AND THE CROSSING NUMBER INEQUALITY

simple, connected

cite szekely and tao

Tao's article reread to see what I missed

First proven in ...

Theres something going on with connectedness here

rewrite this paragraph

A profoundly useful theorem in the study of sum-product conjectures is the Szemerédi-Trotter theorem, a statement about systems of points and lines.

The easiest way to prove this theorem is through the use of the crossing number inequality, a statement about how planar a given graph can be. The purpose of this section is to give a proof of the crossing number inequality.

Due to the brevity of this report, the definitions and proofs given will not be as rigorous as they should be. There are also pictures to help understand.

define connected

A graph G is a pair $G = (V, E)$ where each $e \in E$ is of the form $e \subset V$ with $|e| = 2$. We call the set V the vertices, and the set E the edges. A drawing is a representation of a graph with vertices as points in the plane and edges as curves between their respective vertices. For example:

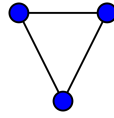


FIGURE 1. A drawing of a graph with $V = \{A, B, C\}$ and $E = \{\{A, B\}, \{B, C\}, \{A, C\}\}$

A graph is connected if for any 2 vertices, there exists a sequence of edges which join them. Note that connectedness is a property of a graph and not of the drawing of a graph.

There are infinitely many ways to draw any given graph. A crossing in a drawing of a graph is an intersection between 2 curves which represent edges. The crossing number of a graph is the minimum number of crossings a drawing of the graph can have. Denote this by $\text{cr}(G)$. A graph G is called planar if its crossing number is 0.

A precise statement of the Crossing Number Inequality is

Theorem 4.1. *Let $G = (V, E)$ be a connected graph. If $|E| \geq 4|V|$ then*

$$\text{cr}(G) \gg \frac{|E|^3}{|V|^2}.$$

For a drawing of a planar graph, we call any region of the plane which is bounded by edges a face. We also call the unbounded region of the plane a face. Here is an example of a drawing of a planar graph with labelled faces:

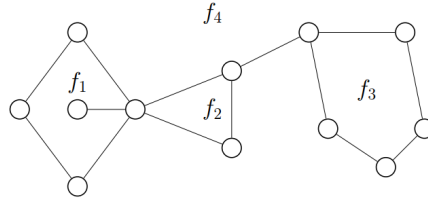


FIGURE 2. Drawing of Planar Graph with Labeled Faces f_i

cite tao

Observe that any non-planar graph $G = (V, E)$ can be turned into a planar graph by removing at most $\text{cr}(G)$ edges from E . Therefore, a bound on the number of edges a graph

can have and remain planar yields a bound on the crossing number for any graph. A famous theorem relating the vertices and edges of planar graphs is

Theorem 4.2 (Euler's Formula for Planar Graphs). *Let $G = (V, E)$ be a connected planar graph, with $|V| \geq 1$, and consider some drawing with 0 crossings. Let F be the set of all faces of this drawing.*

$$|V| - |E| + |F| = 2.$$

Proof. We may construct our drawing of G by first drawing a vertex, and then doing combination of the following steps:

PROOF □

The dependence on $|F|$ in Euler's formula can be removed by using its obvious dependence on $|E|$.

We call an edge incident to a face if the edge is one of the bounding edges which define the face. Define $\chi : F \times E \rightarrow \{0, 1\}$ as the incidence function, so $\chi(f, e) = 1$ if f and e are incident, and $\chi(f, e) = 0$ otherwise. The total number of face edge incidences is

$$I(F, E) = \sum_{f \in F} \sum_{e \in E} \chi(f, e).$$

JUSTIFY THIS ASSUMPTION SOMEWHERE We assume $|E| \geq 3$, so that every face is incident to at least 3 edges. It follows that

$$I \geq \sum_{f \in F} 3 = 3|F|.$$

Every edge is incident to at most 2 faces, so it follows that

$$I \leq \sum_{e \in E} 2 = 2|E|.$$

Therefore

$$3|F| \leq 2|E|$$

or

$$|F| \leq \frac{2}{3}|E|.$$

Applying this to Euler's formula,

$$|V| - |E| + \frac{2}{3}|E| \geq 2$$

or

$$|E| \leq 3|V| - 6$$

when $|E| \geq 3$. Now suppose that $G = (V, E)$ is non-planar and connected. As mentioned before, G may be turned planar by removing at most $\text{cr}(G)$ edges.

READ TAO AND JUSTIFY ABOVE BETTER Therefore, for any graph G with $|E| \geq 3$,

$$|E| - \text{cr}(G) \leq 3|V| - 6$$

or

$$\text{cr}(G) > |E| - 3|V|.$$

To further improve this inequality, we apply the probabilistic method to the deletion of vertices of G .

Let each $v \in V$ be removed with a probability $1 - p$, $p \in (0, 1)$. Let the remaining set of vertices be V' .

An edge is removed whenever either of the corresponding vertices are removed. Let the remaining set of edges be E' . Let the remaining graph be $G' = (V', E')$. We have that if $|E'| \geq 3$,

$$\text{cr}(G') \geq |E'| - 3|V'|,$$

and so

$$\mathbb{E}(\text{cr}(G')) \geq \mathbb{E}(|E'| - 3|V'|),$$

or, by the linearity of the expected value,

$$\mathbb{E}(\text{cr}(G')) \geq \mathbb{E}(|E'|) - 3\mathbb{E}(|V'|).$$

Each $v \in V$ is removed with probability $1 - p$, so

$$\mathbb{E}(|V'|) = p|V|.$$

Each edge remains only when both corresponding vertices remain. Each vertex remains independently with a probability p , so

$$\mathbb{E}(|E'|) = p^2|E|.$$

We can bound $\mathbb{E}(\text{cr}(G'))$ by considering a drawing of G with the minimum number of crossings. Each crossing remains and only when both corresponding edges remain, each of which occurs independently with probability p^2 .

The expected value of the number of crossings remaining in the drawing is $p^4 \text{cr}(G)$. There is no guarantee that this drawing is optimal to minimize the crossings of G' , but we may conclude that

$$\mathbb{E}(\text{cr}(G')) \leq p^4 \text{cr}(G),$$

and therefore that

$$p^4 \text{cr}(G) \geq \mathbb{E}(\text{cr}(G')) \geq p^2|E| - 3p|V|$$

for any $p \in (0, 1)$.

Because we are only concerned with an order of magnitude bound, we assume $|E| \geq 4|V|$, and take $p = \frac{4|V|}{|E|}$. This yields the result

$$\text{cr}(G) \geq \frac{|E|}{\left(\frac{4|V|}{|E|}\right)^2} - \frac{3|V|}{\left(\frac{4|V|}{|E|}\right)^3} = \frac{1}{16} \left(\frac{|E|^3}{|V|^2} - \frac{3|E|^3}{4|V|^2} \right) \gg \frac{|E|^3}{|V|^2}.$$

5. THE SZEMEREDI-TROTTER THEOREM

LOOK OVER WHOLE SECTION CONCAVE OR CONVEX

With this, we can employ the argument of [SZE97] to prove the Szemerédi-Trotter Theorem. A precise statement of the Szemerédi-Trotter Theorem, first proven in [ST83] is

Theorem 5.1 (Szemerédi-Trotter Theorem). *Let $P \subset \mathbb{R}^2$ be a finite set of points. Let \mathcal{L} be a finite set of curves in \mathbb{R}^2 .*

Let $\chi : P \times \mathcal{L} \rightarrow \{0, 1\}$ be the incidence function between a point and a line, so

$$\chi(p, l) = \begin{cases} 1 & \text{if } p \in l \\ 0 & \text{otherwise} \end{cases}$$

If any two $l \in \mathcal{L}$ intersect in at most one point, then the total number of point-curve incidences,

$$I(P, \mathcal{L}) = \sum_{(p,l) \in P \times \mathcal{L}} \chi(p, l)$$

satisfies

$$I(P, \mathcal{L}) \ll |P|^{\frac{2}{3}} |\mathcal{L}|^{\frac{2}{3}} + |P| + |\mathcal{L}|.$$

Proof. This is proven by turning a system of curves and points into a graph. We first omit all points and curves which contribute to one or fewer incidences.

For each remaining curve, if there are n incidences along it, we partition it into $n - 1$ curves, each with 2 incidences. These become the edges of the drawing of some graph. Let the set of edges be E .

If $I_0(P, \mathcal{L})$ is the remaining number of incidences,

$$|E| \geq I_0(P, \mathcal{L}) - |\mathcal{L}|.$$

Let the vertices of the graph, V , be the remaining points in the system. Obviously, $|V| \leq |P|$.

Because any two curves intersect in at most one point, $\text{cr}(G)$ is at most $|\mathcal{L}|^2$.

Supposing that $|E| \geq 4|V|$, we have that

$$|\mathcal{L}|^2 \geq \text{cr}(G) \gg \frac{(I_0(P, \mathcal{L}) - |\mathcal{L}|)^3}{|P|^2}$$

or

$$I_0(P, \mathcal{L}) \ll |\mathcal{L}|^{\frac{2}{3}} |P|^{\frac{2}{3}} + |\mathcal{L}|.$$

If $|V| \geq \frac{1}{4}|E|$, then $|V| \gg |E|$, or

$$|P| \gg I_0(P, \mathcal{L}) - |\mathcal{L}| \implies I_0(P, \mathcal{L}) \ll |P| + |\mathcal{L}|.$$

Finally, the remaining incidences not yet counted is

$$I(P, \mathcal{L}) - I_0(P, \mathcal{L}) \ll |\mathcal{L}| + |P|,$$

so

$$I(P, \mathcal{L}) \ll |P|^{\frac{2}{3}} |\mathcal{L}|^{\frac{2}{3}} + |P| + |\mathcal{L}|.$$

□

The Szemerédi Trotter theorem has a handful of direct applications to the sum product conjecture and the convex sumset conjecture. The main theorem which leads to these results is proven in [ENR00].

Theorem 5.2. Let $A \subset \mathbb{R}$ be finite, with $|A| = n$.

Label the elements of A so that $a_1 < a_2 < \dots < a_n$.

Let $f : [a_1, a_n] \rightarrow \mathbb{R}$ be convex. Let $S = \{(a, f(a)) : a \in A\}$ and $T \subset \mathbb{R}^2$ be finite.

We have

$$|S + T| \gg \max \left(|S|^{\frac{3}{2}} |T|^{\frac{1}{2}}, |S| |T| \right).$$

Proof. Let

$$L_t = \{(x, f(x)) + t : x \in [a_1, a_n], t \in T\},$$

and let

$$\mathcal{L} = \{L_t : t \in T\}.$$

For every $x \in A$, $(x, f(x)) + t \in S + T$. Therefore, there are $|A| = |S|$ incidences between L_t and the point set $S + T$, for all $t \in T$. It follows that there are $|S| |T|$ total incidences. The set \mathcal{L} consists of translations of the graph of a convex function, so the Szemerédi-Trotter theorem is satisfied. Thus

$$|S| |T| \ll |S + T|^{\frac{2}{3}} |T|^{\frac{2}{3}} + |S + T| + |T|.$$

Trivially,

$$|S + T| \geq |T|,$$

so

$$|S| |T| \ll \max \left(|S + T|^{\frac{2}{3}} |T|^{\frac{2}{3}}, |S + T| \right),$$

or

$$|S + T| \gg \max \left(|S|^{\frac{3}{2}} |T|^{\frac{1}{2}}, |S| |T| \right).$$

□

The following corollaries are also proven in [ENR00].

Corollary 5.3. *For convex and finite sets $A \subset \mathbb{R}$, and finite sets $B \subset \mathbb{R}$,*

$$|A + B| \gg |A| |B|^{\frac{1}{2}}.$$

Proof. Let $A \subset \mathbb{R}$ be finite. Let $n = |A|$. Let $f : [1, n] \rightarrow \mathbb{R}$ be the convex function for which

$$A = \{f(i) : i \in [n]\}.$$

Take

$$S = \{(i, f(i)) : i \in [n]\}$$

and

$$T = [n] \times B.$$

We have

$$S + T \subset ([n] + [n]) \times (A + B),$$

so

$$|S + T| \leq |[n] + [n]| |A + B| \ll |A| |A + B|.$$

Apply Theorem 5.2 to get

$$|A| |A + B| \gg |S + T| \gg \max \left(|A|^{\frac{3}{2}} (|A| |B|)^{\frac{1}{2}}, |A| (|A| |B|) \right)$$

or

$$|A| |A + B| \gg |A|^2 |B|^{\frac{1}{2}} \implies |A + B| \gg |A| |B|^{\frac{1}{2}}.$$

□

This will be used to achieve stronger results later, but in particular, this gives the result

Corollary 5.4. *For sufficiently large sets $A \subset \mathbb{R}$ which are finite and convex,*

$$|A + A| \gg |A|^{\frac{3}{2}}.$$

We can also use Theorem 5.2 to get a result on the Sum-Product conjecture.

Corollary 5.5. *For finite sets $A \subset \mathbb{R}^+$ of positive real numbers,*

$$\max(|A + A|, |A \cdot A|) \gg |A|^{\frac{5}{4}}.$$

Proof. Let $A \subset \mathbb{R}$ be finite.

Label the elements of A so that $a_1 < \dots < a_n$. Let $f : [a_1, a_n] \rightarrow \mathbb{R}$ be concave or convex. Take

$$S = \{(a, f(a)) : a \in A\},$$

and

$$T = A \times f(A).$$

Observe that

$$S + T \subset (A + A) \times (f(A) + f(A)),$$

so that

$$|S + T| \ll |A + A| |f(A) + f(A)|.$$

Apply Theorem 5.2 to get

$$|A + A| |f(A) + f(A)| \gg |S + T| \gg \max \left(|A|^{\frac{3}{2}} (|A|^2)^{\frac{1}{2}}, |A| |A|^2 \right)$$

or

$$\max(|A + A|, |f(A) + f(A)|) \gg |A|^{\frac{5}{4}}$$

for any convex or concave function $f : [a_1, a_n] \rightarrow \mathbb{R}$.

If $A \subset \mathbb{R}^+$, we may take $f(x) = \log(x)$, immediately yielding the desired result. \square

The Szemerédi-Trotter theorem also provides more general tools which can be used alongside other results to sharpen the above bounds.

Theorem 5.6. *Let $A \subset \mathbb{R}$ be convex, then for every finite set $B \subset \mathbb{R}$ we have*

$$|\{x \in A - B : \delta_{A,B}(x) \geq \tau\}| \ll \frac{|A| |B|^2}{\tau^3}.$$

Proof. **PROOF** \square

An immediate corollary of this is

Corollary 5.7. *Let $A \subset \mathbb{R}$ be a convex and finite set. Let $B \subset \mathbb{R}$ be finite. Order elements $s_i \in A - B$ such that*

$$\delta_{A,B}(s_1) \geq \delta_{A,B}(s_2) \geq \dots \geq \delta_{A,B}(s_{|A-B|}).$$

For every $1 \leq r \leq |A - B|$ we have

$$\delta_{A,B}(s_r) \ll \frac{|A|^{\frac{1}{3}} |B|^{\frac{2}{3}}}{r^{\frac{1}{3}}}.$$

Proof.

$$r = |\{x \in A - B : \delta_{A,B}(x) \geq \delta_{A,B}(s_r)\}| \ll \frac{|A| |B|^2}{\delta_{A,B}(s_r)^3} \implies \delta_{A,B}(s_r) \ll \frac{|A|^{\frac{1}{3}} |B|^{\frac{2}{3}}}{r^{\frac{1}{3}}}.$$

\square

Later in the next section we'll use these results to find bounds on the additive energies between certain sets, but first we'll introduce some simpler results which use additive and multiplicative energy to show why such bounds are useful.

6. ADDITIVE AND MULTIPLICATIVE ENERGY **RENAME SECTION**

Recall that

$$|A + A| \geq \frac{|A|^4}{E(A)}$$

and

$$|AA| \geq \frac{|A|^4}{M(A)}.$$

Observe that finding an upper bound on $E(A)$ or $M(A)$ in terms of $|A + A|$, $|AA|$, and $|A|$ yields a sum product theorem. This section consists of results which employ this general idea.

The simplest of which is a result in [Sol09], which gives a stronger result on the sum-product conjecture than the Szemerédi-Trotter theorem.

Theorem 6.1. *Let $A \subset \mathbb{R}^+$ be finite and sufficiently large.*

$$\max(|A + A|, |AA|) \gg |A|^{\frac{4}{3}-o(1)}.$$

Proof. **FIX THIS BEGINNING PART**

THE WHOLE THING NEEDS REVISION

We begin with a construction. Consider the set A^2 , along with the smallest set of lines through the origin which cover A^2 .

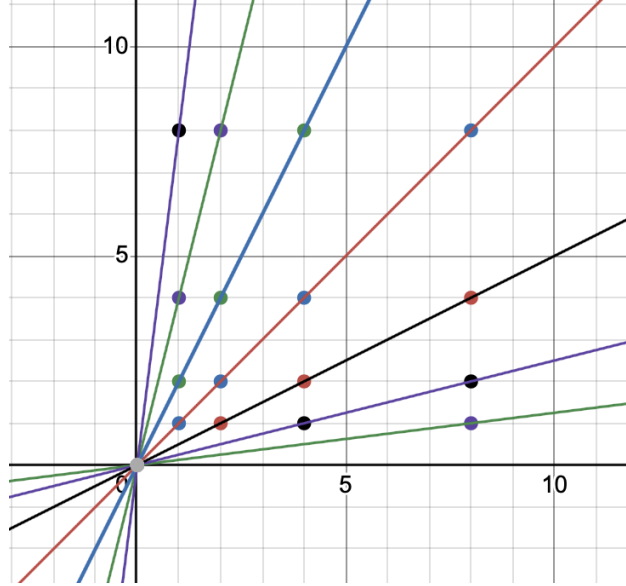


FIGURE 3. Example with $A = \{1, 2, 4, 8\}$.

Two pairs $(a_1, a_2), (b_1, b_2) \in A^2$ give the same representation as a quotient if and only if

$$\frac{a_2}{a_1} = \frac{b_2}{b_1}.$$

Observe that this is the slope of a line through the origin and the points $(a_1, a_2), (b_1, b_2)$.

It follows that $|\frac{A}{A}|$ is the number of lines through the origin necessary to cover the point set A^2 . The slope of the line is the value the line represents in $\frac{A}{A}$, and the number of points on the line is the number of representations of that number in $\frac{A}{A}$.

Consider 2 consecutive lines and the set of all vector sums between a point on each line. If our points are (a_1, a_2) and (b_1, b_2) , with

$$\frac{a_2}{a_1} > \frac{b_2}{b_1},$$

then the slope of their sum is

$$\frac{a_2 + b_2}{a_1 + b_1}$$

which satisfies

$$\frac{b_2}{b_1} < \frac{a_2 + b_2}{a_1 + b_1} < \frac{a_2}{a_1}.$$

That is, the vector sum must “lie between” the two lines which the original vectors are on. More precisely, for any pairs of consecutive lines, the vector sums of all points along the lines are disjoint.

We also have that the sums of any 2 points on each line are distinct. This is because a solution to

$$\lambda_1 v + \lambda_2 w = \lambda_3 v + \lambda_4 w \iff (\lambda_1 - \lambda_3) v + (\lambda_2 - \lambda_4) w$$

where $\lambda_1 \neq \lambda_3$ or $\lambda_2 \neq \lambda_4$ exists only if v and w are linearly dependent.

By dyadic partitioning on $M(A)$, we have

$$M(A) = \sum_{x \in \frac{A}{A}} r_{\frac{A}{A}}(x)^2 \leq \log \left(\left| \frac{A}{A} \right| \right) \tau^2 |S|$$

for some τ , where $S = \left\{ x \in \frac{A}{A} : r_{\frac{A}{A}}(x) \asymp \tau \right\}$.

Consider the reduced system of points and lines, consisting only of the $|S|$ many lines which have $\asymp \tau$ many points on them. Consider the set of all vector sums between points on consecutive lines. Because all pairs of lines give disjoint sets, each with $\asymp \tau^2$ many sums, there are $\tau^2 |S|$ many vector sums. Because this is a subset of $(A + A)^2$,

$$\tau^2 |S| \leq |A + A|^2,$$

so

$$\frac{|A|^4}{|AA|} \leq M(A) \ll \log \left(\left| \frac{A}{A} \right| \right) |A + A|^2 \implies \max(|A + A|, |AA|) \gg |A|^{\frac{4}{3} - o(1)}$$

□

This is nearly the best known result, all others are only slight improvements, ... Only result purely obtaining upper bound on energy

Another instance of this type of result is

Theorem 6.2. *For convex sets $A \subset \mathbb{R}$ which are finite and sufficiently large,*

$$E(A) \ll |A|^{\frac{32}{13} - o(1)},$$

which immediately leads to the result

Corollary 6.3. *For convex sets $A \subset \mathbb{R}$ which are finite and sufficiently large,*

$$|A + A| \gg |A|^{\frac{3}{2} + \frac{1}{26} - o(1)}.$$

FIX BELOW PARAGRAPH This is proven in [?]. I did not read this paper because it includes a lot of unique ideas and complex notation. I plan to read it in the future.

REWRITE BELOW Many arguments involving additive energy are not as straightforward as finding an upper bound. We often explore quantities such as $E(A, S)$, $E(D)$, $E(S)$, $E_3(A)$ etc. The following are examples of those style of arguments.

I'll first prove a handful of theorems which will be useful. The first of which I will restate from before, still leaving out proof.

Theorem 6.4. *For convex sets $A \subset \mathbb{R}$ which are finite and sufficiently large,*

$$E_3(A) \ll |A|^{3-o(1)}.$$

Proof. Recall that upon ordering a_i such that $\delta_A(a_1) \geq \delta_A(a_2) \geq \cdots \geq \delta_A(a_{|A+A|})$, we have that

$$\delta_A(a_r) \ll \frac{|A|}{r^{\frac{1}{3}}}.$$

With this,

$$\begin{aligned} E_3(A) &= \sum_{x \in A-A} \delta_A(x)^3 \\ &\ll |A|^3 \sum_{r=1}^{|A+A|} \frac{1}{r} \\ &\asymp |A|^3 \int_1^{|A+A|} \frac{1}{r} \, dr \\ &= |A|^{3-o(1)} \end{aligned}$$

□

Theorem 6.5. *For convex sets $A \subset \mathbb{R}$ which are finite and sufficiently large, and any finite set $B \subset \mathbb{R}$*

$$E(A, B) \ll |A| |B|^{\frac{3}{2}}.$$

I NEED TO REDO PROOF WITH MORE LOGIC BEHIND WHY I CHOSE B 1/2

Proof. Denote the elements of $A - B$ by s_i where $\delta_{A,B}(s_1) \geq \cdots \geq \delta_{A,B}(s_{|A-B|})$

Let $P = \left\{x \in A - B : \delta_{A,B}(x) \geq |B|^{\frac{1}{2}}\right\}$, and let $P^* = (A - B) \setminus P$.

$$\begin{aligned}
\sum_{x \in P} \delta_{A,B}(x)^2 &= \sum_{i=1}^{|P|} \delta_{A,B}(s_i)^2 \\
&\ll |A|^{\frac{2}{3}} |B|^{\frac{4}{3}} \sum_{i=1}^{|P|} \frac{1}{r_i^{\frac{2}{3}}} \\
&\asymp |A|^{\frac{2}{3}} |B|^{\frac{4}{3}} |P|^{\frac{1}{3}} \\
&\ll |A|^{\frac{2}{3}} |B|^{\frac{4}{3}} \left(\frac{|A| |B|^2}{|B|^{\frac{1}{2}}} \right)^{\frac{1}{3}} \\
&= |A| |B|^{\frac{3}{2}},
\end{aligned}$$

and

$$\sum_{x \in P^*} \delta_{A,B}(x)^2 < |B|^{\frac{1}{2}} \sum_{x \in P^*} \delta_{A,B}(x) = |A| |B|^{\frac{3}{2}}.$$

Therefore,

$$E(A, B) = \sum_{x \in P} \delta_{A,B}(x)^2 + \sum_{x \in P^*} \delta_{A,B}(x)^2 \ll |A| |B|^{\frac{3}{2}}.$$

□

7. IDK THE TITLE

In this section we'll prove the ... **more exposition** [SS11]

Throughout this section, in the same convention as **CITED**, we'll use the notation

$$A_x = A \cap (A + x) = \{a \in A : a - x \in A\} = \{a \in A : \exists a_0 \in A \text{ s.t. } a - a_0 = x\}.$$

Observe that

$$|A_x| = \delta_A(x).$$

Theorem 7.1. *Let $A \subset \mathbb{R}$ be a convex set. Then*

$$\max(|A + A|, |A - A|) \gg_{\epsilon} |A|^{\frac{8}{5} - \epsilon},$$

$$|A - A| \gg_{\epsilon} |A|^{\frac{8}{5} - \epsilon},$$

$$|A + A| \gg_{\epsilon} |A|^{\frac{14}{9} - \epsilon}.$$

We require the following lemmata

Lemma 7.2. *For every set $A \subset \mathbb{R}$ we have*

$$\sum_x E(A, A_x) = E_3(A).$$

Proof. **PROOF**

□

Lemma 7.3. *For every set $A \subset \mathbb{R}$, and $P \subset A - A$, if η is the number for which*

$$\sum_{x \in P} |A_x| = \eta |A|^2,$$

then

$$\sum_{x \in P} |A \pm A_x| \geq \frac{\eta^2 |A|^6}{E_3(A)}.$$

Proof. **PROOF** □

We're now able to prove Theorem 7.1.

Proof of Theorem 7.1. Define the difference and sum sets with $D = |A - A|$ and $S = |A + A|$. Observe that

$$A - A_x \subset D \cap D_x$$

and

$$A + A_x \subset S \cap S_x.$$

We use this observation along with Lemma 7.3 to find bounds on the additive energies between D and S and A .

Consider the popular sets of differences P, P' defined by

$$P = \left\{ x \in A - A : \delta_A(x) \geq \frac{|A|^2}{2|A - A|} \right\}$$

and

$$P' = \left\{ x \in A - A : \delta_A(x) \geq \frac{|A|^2}{2|A + A|} \right\}.$$

Let P^* and P'^* be their respective compliments. We have

$$\sum_{x \in P^*} |A_x| < \frac{|A|^2}{2|A - A|} \cdot |P^*| \leq \frac{|A|^2}{2},$$

so

$$\sum_{x \in P} |A_x| \gg |A|^2. \tag{1}$$

We also have

$$\sum_{x \in P'^*} |A_x|^2 < \frac{|A|^2}{2|A + A|} \cdot |A|^2,$$

so

$$\sum_{x \in P'} |A_x|^2 \gg \frac{|A|^4}{|A + A|}.$$

□

We can directly apply (1) to Lemma 7.3. We have

$$\eta |A|^2 = \sum_{x \in P} |A_x| \gg |A|^2 \implies \eta \gg 1$$

so

$$\sum_{x \in P} |A \pm A_x| \gg \frac{|A|^6}{E_3(A)} \gtrsim |A|^3.$$

Recall that $A - A_x \subset D \cap D_x$, or $|A - A_x| \leq |D_x| = \delta_D(x)$. It follows that

$$\sum_{x \in P} \delta_D(x) \gtrsim |A|^3.$$

By the definition of P it follows that

$$E(A, D) \geq \sum_{x \in P} \delta_A(x) \delta_D(x) \geq \frac{|A|^2}{2|A - A|} \sum_{x \in P} \delta_D(x) \gtrsim \frac{|A|^5}{|A - A|}.$$

By applying Theorem 6.5,

$$\frac{|A|^5}{|A - A|} \lesssim |A| |A - A|^{\frac{3}{2}} \implies |A - A| \gtrsim |A|^{\frac{8}{5}}.$$

Recalling that $A + A_x \subset S_x$, we get that

$$\sum_{x \in P} \delta_S(x) \gtrsim |A|^3,$$

which by an identical argument yields

$$\frac{|A|^5}{|A - A|} \lesssim E(A, S) \ll |A| |S|^{\frac{3}{2}},$$

or

$$|A - A|^2 |A + A|^3 \gtrsim |A|^8 \implies \max(|A - A|, |A + A|) \gtrsim |A|^{\frac{8}{5}}.$$

REFERENCE EARLIER STATEMENT OF THEOREM

Acknowledgements. This report is based on work supported by NSF grant DMS-2051032, which we gratefully acknowledge. I would also like to express my thanks to the Mathematics department of Indiana University for hosting the program, my mentor XXX and....

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APPENDIX

UNIQUENESS OF REPRESENTATIONS OF DIFFERENT BASES
BELOW

$$|A + A| \geq \frac{c|A|^2}{\log(|A|)}.$$

I need to define all things I use like little o and big O.

Theorem (Erdős multiplication table theorem).

$$|[n] \cdot [n]| = o(n^2).$$

This is known as the “multiplication table theorem” because the quantity

$$|[n] \cdot [n]|$$

is the number of distinct numbers in an $n \times n$ multiplication table:

×	1	2	3	4	5	6	7	8	9	10
1	1	2	3	4	5	6	7	8	9	10
2	2	4	6	8	10	12	14	16	18	20
3	3	6	9	12	15	18	21	24	27	30
4	4	8	12	16	20	24	28	32	36	40
5	5	10	15	20	25	30	35	40	45	50
6	6	12	18	24	30	36	42	48	54	60
7	7	14	21	28	35	42	49	56	63	70
8	8	16	24	32	40	48	56	64	72	80
9	9	18	27	36	45	54	63	72	81	90
10	10	20	30	40	50	60	70	80	90	100

FIGURE 4. 10×10 multiplication table with distinct numbers highlighted in red.

The theorem I am proving here is just an “upper bound” on the asymptotic behavior of this quantity. In [For08], Kevin Ford proved that the exact order of this quantity is

$$|[n] \cdot [n]| \asymp \frac{n^2}{\log(n^2)^\delta (\log \log(n^2))^{\frac{3}{2}}},$$

where $\delta = 1 - \frac{1 + \log \log(2)}{2}$ is the Erdős-Tenenbaum-Ford constant.

We require a lemma and a prerequisite theorem.

Lemma. Let $(a_n)_{n=1}^\infty$ be a sequence of real numbers. Let $A : \mathbb{R} \rightarrow \mathbb{R}$ be defined by

$$A(t) = \sum_{n \leq t} a_n.$$

For $x \in \mathbb{R}$ and any differentiable function $\phi : [1, y] \rightarrow \mathbb{R}$,

$$\sum_{n \leq x} a_n \phi(n) = A(x)\phi(x) - A(1)\phi(1) - \int_1^x A(t)\phi'(t) \, dt$$

FINISH THIS LATER

Proof.

$$\begin{aligned}
\sum_{x < n \leq y} a_n \phi(n) &= a_{\lceil x \rceil} \phi(\lceil x \rceil) + \cdots + a_{\lfloor y \rfloor} \phi(\lfloor y \rfloor) \\
&= (A(x+1) - A(x)) \phi(\lceil x \rceil) + \cdots + (A(y) - A(y-1)) \phi(\lfloor y \rfloor) \\
&= \\
&= A(y) \phi(\lfloor y \rfloor) - A(x) \phi(\lceil x \rceil) + \sum_{i=1} (\phi(\lceil x \rceil + i) - \phi(\lceil x \rceil + 1 + i)) A(x+i)
\end{aligned}$$

□

A prerequisite theorem gives the order of magnitude of 3 functions which will prove useful.

Theorem 7.4. *Define the functions $\theta, \psi, \pi : [1, \infty) \rightarrow \mathbb{R}$ by*

$$\theta(x) = \sum_{\substack{p \leq x \\ p \text{ prime}}} \log(p),$$

$$\psi(x) = \sum_{\substack{p^\alpha \leq x \\ p \text{ prime} \\ \alpha \in \mathbb{N}}} \log(p),$$

$$\pi(x) = \sum_{\substack{p \leq x \\ p \text{ prime}}} 1.$$

We have

$$\psi(x) \asymp \theta(x) \asymp x$$

and

$$\pi(x) \asymp \frac{x}{\log(x)}.$$

Proof. The functions θ and ψ are clearly related by

$$\psi(t) = \prod_{\substack{p \leq N^{\frac{1}{\alpha}} \\ p \text{ prime} \\ \alpha \in \mathbb{N}}} \log(p) = \sum_{\alpha \in \mathbb{N}} \theta\left(t^{\frac{1}{\alpha}}\right) = \theta(t) + \sum_{\alpha \geq 2} \theta\left(t^{\frac{1}{\alpha}}\right).$$

Note that the sum over α has only finitely many terms. The sum terminates when

$$2 \geq t^{\frac{1}{\alpha}} \implies \alpha \leq \log_2(t).$$

A trivial upper bound on $\theta(t)$ is

$$\theta(t) = \sum_{\substack{p \leq t \\ p \text{ prime}}} \log(p) \leq t \log(t),$$

so

$$\begin{aligned}\psi(t) &= \theta(t) + \sum_{2 \leq \alpha \leq \log_2(t)} \theta(t^{\frac{1}{\alpha}}) \\ &\leq \theta(t) + \log_2(t) \theta(t^{\frac{1}{2}}) \\ &\leq \theta(t) + \frac{t^{\frac{1}{2}} \log(t)^2}{\log(2)}\end{aligned}$$

or

$$\psi(t) \ll \max\left(\theta(t), t^{\frac{1}{2}} \log(t)^2\right).$$

By showing $\theta(t) \ll t$, we will have shown $\psi(t) \ll \theta(t) \ll t$. Because $\theta(t) \leq \psi(t)$, we will have shown that $\psi(t) \asymp \theta(t) \ll t$.

We have

$$\theta(t) = \sum_{\substack{p \leq t \\ p \text{ prime}}} \log(p) = \log\left(\prod_{\substack{p \leq t \\ p \text{ prime}}} p\right),$$

so it is sufficient to show that

$$\prod_{\substack{p \leq t \\ p \text{ prime}}} p \ll e^t.$$

It is also sufficient to prove it for $t \in \mathbb{N}$ because $\theta(t) = \theta(\lfloor t \rfloor)$.

For some natural number t , and a prime p ,

$$t+1 < p \leq 2t+1 \implies p \mid \binom{2t+1}{t} = \frac{(2t+1)!}{t!(t+1)!}.$$

Therefore, for any t ,

$$\prod_{\substack{t+1 < p \leq 2t+1 \\ p \text{ prime}}} p \mid \binom{2t+1}{t} \implies \prod_{\substack{t+1 < p \leq 2t+1 \\ p \text{ prime}}} p \leq \binom{2t+1}{t},$$

which gives us

$$2 \prod_{\substack{t+1 < p \leq 2t+1 \\ p \text{ prime}}} p \leq 2 \binom{2t+1}{t} \leq (1+1)^{2t+1} \implies \prod_{\substack{t+1 < p \leq 2t+1 \\ p \text{ prime}}} p \leq 4^t.$$

The rest follows by induction on t . Because we are proving a statement about order of magnitude, the base case is trivial. Now suppose that for some $t \in \mathbb{N}$,

$$\prod_{\substack{p \leq m \\ p \text{ prime}}} p \ll e^t.$$

If t is odd, the induction follows trivially. If t is even, let $t = 2m$, so

$$\prod_{\substack{p \leq 2m \\ p \text{ prime}}} p \ll e^{2m}.$$

We have

$$\begin{aligned} \prod_{\substack{p \leq 2m+1 \\ p \text{ prime}}} p &= \prod_{\substack{p \leq m+1 \\ p \text{ prime}}} p \prod_{\substack{m+1 < p \leq 2m+1 \\ p \text{ prime}}} p \\ &\ll e^{m+1} 4^m \\ &\ll e^{m+1} e^m = e^{2m+1}. \end{aligned}$$

Thus, $\psi(t) \asymp \theta(t) \ll t$. To prove $\psi(t) \asymp \theta(t) \asymp t$, it suffices to show that $\psi(t) \gg t$. □

The final lemma needed is

Lemma.

$$\sum_{\substack{p \leq n \\ p \text{ prime}}} \frac{1}{p} = \log \log (n) + O(1).$$

Proof of Lemma. Observe that for some number $N \in \mathbb{N}$, the prime factorization of $N!$ is of the form

$$N! = \prod_{\substack{p \leq N \\ p \text{ prime}}} p^{\alpha(N,p)},$$

where

$$\alpha(N, p) = \sum_{i \in \mathbb{N}} \left\lfloor \frac{N}{p^i} \right\rfloor = \sum_{i \leq \log_p(N)} \left\lfloor \frac{N}{p^i} \right\rfloor = \sum_{i \leq \log_2(N)} \left\lfloor \frac{N}{p^i} \right\rfloor.$$

It follows that

$$\begin{aligned} \log(N!) &= \sum_{\substack{p \leq N \\ p \text{ prime}}} \alpha(N, p) \log(p) \\ &= \sum_{\substack{p \leq N \\ p \text{ prime} \\ i \leq \log_p(N)}} \left\lfloor \frac{N}{p^i} \right\rfloor \log(p) \\ &= \sum_{\substack{p \leq N \\ p \text{ prime} \\ i \leq \log_p(N)}} \left(\frac{N}{p^i} - \delta(p) \right) \log(p) \\ &= N \sum_{\substack{p \leq N \\ p \text{ prime} \\ i \leq \log_p(N)}} \frac{\log(p)}{p^i} - \sum_{\substack{p \leq N \\ p \text{ prime} \\ i \leq \log_p(N)}} \delta(p) \log(p). \end{aligned}$$

We have that

$$i \leq \log_p(N) \iff p^i \leq N,$$

so

$$\sum_{\substack{p \leq N \\ p \text{ prime} \\ i \leq \log_p(N)}} \log(p) = \psi(N),$$

and therefore

$$\sum_{\substack{p \leq N \\ p \text{ prime} \\ i \leq \log_2(N)}} \frac{\log(p)}{p^i} \leq \frac{\log(N!)}{N} + \frac{\psi(N)}{N}.$$

We also have, via a Riemann sum,

$$\begin{aligned} \log(N!) &= \sum_{i=1}^N \log(i) \\ &= \int_2^N \log(i) \, di + O(1) \\ &= [i \log(i) - i]_{i=1}^N + O(1) \\ &= N \log(N) - N + O(1). \end{aligned}$$

This leads to

$$\sum_{\substack{p \leq N \\ p \text{ prime} \\ i \leq \log_2(N)}} \frac{\log(p)}{p^i} = \frac{N \log(N) - N + O(1)}{N} + \frac{O(N)}{N} = \log(N) + O(1).$$

Moreover, we have that

$$\begin{aligned} \sum_{\substack{p \leq N \\ p \text{ prime} \\ 2 \leq i \leq \log_2(N)}} \frac{\log(p)}{p^i} &\leq \sum_{x, i \geq 2} \frac{\log(x)}{x^i} \\ &= \sum_{x \geq 2} \sum_{i \geq 2} \frac{\log(x)}{x^i} \\ &= \sum_{x \geq 2} \frac{1}{x^2} \left(\frac{\log(x)}{1 - \frac{1}{x}} \right) \\ &= \sum_{x \geq 2} \frac{\log(x)}{x^2 - x} \end{aligned}$$

which, by the limit comparison test, converges if

$$\sum_{x \geq 2} \frac{\log(x)}{x^2}$$

converges.

Apply L'Hopital's rule to see that

$$\forall \epsilon > 0, \log(x) = o(x^\epsilon),$$

and therefore that

$$\begin{aligned} \sum_{x \geq 2} \frac{\log(x)}{x^2} &= \sum_{x \geq 2} \frac{o(1)}{x^{2-\epsilon}} \\ &\ll \sum_{x \geq 2} \frac{1}{x^{2-\epsilon}} \\ &\asymp \int_2^\infty \frac{1}{x^{2-\epsilon}} dx \end{aligned}$$

which converges for $\epsilon < 1$. It follows that

$$\sum_{\substack{p \leq N \\ p \text{ prime}}} \frac{\log(p)}{p} = \log(N) + O(1).$$

Applying Abel's summation formula,

$$\begin{aligned} \sum_{\substack{p \leq N \\ p \text{ prime}}} \frac{1}{p} &= \sum_{\substack{p \leq N \\ p \text{ prime}}} \frac{\log(p)}{p} \cdot \frac{1}{\log(p)} \\ &= \left(\sum_{\substack{p \leq N \\ p \text{ prime}}} \frac{\log(p)}{p} \right) \cdot \frac{1}{\log(N)} + O(1) + \int_2^N \frac{\sum_{\substack{p \leq t \\ p \text{ prime}}} \frac{\log(p)}{p}}{t \log^2(t)} dt \\ &= \frac{\log(N) + O(1)}{\log(N)} + O(1) + \int_2^N \frac{\log(t) + O(1)}{t \log^2(t)} dt \\ &= O(1) + \int_2^N \frac{1}{t \log(t)} dt + \int_2^N \frac{O(1)}{t \log^2(t)} dt \end{aligned}$$

□

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