# CLUSTERING OF LINEAR COMBINATIONS OF MULTIPLICATIVE FUNCTIONS

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ABSTRACT. A real-valued arithmetic function F is said to cluster about the point  $u \in \mathbb{R}$  if the upper density of n with  $u - \delta < F(n) < u + \delta$  is bounded away from 0, uniformly for all  $\delta > 0$ . We establish a simple-to-check sufficient condition for a linear combination of multiplicative functions to be nonclustering, meaning not clustering anywhere. This provides a means of generating new families of arithmetic functions possessing continuous distribution functions. As a specific application, we resolve a problem posed recently by Luca and Pomerance.

## 1. Introduction

Let F be a real-valued arithmetic function. We say that F clusters around the real number u if there is some  $\epsilon > 0$  such that, for every  $\delta > 0$ , the solutions n to

$$u - \delta < F(n) < u + \delta$$

form a set of upper density at least  $\epsilon$ . If F does not cluster around any u, we say that F is nonclustering. The main result of this note is the following criterion for a linear combination of multiplicative functions to be nonclustering.

**Theorem 1.** Let  $f_1, \ldots, f_k$  be multiplicative arithmetic functions taking values in the nonzero real numbers and satisfying the following conditions:

- (i)  $f_1$  is nonclustering,
- (ii) none of  $f_1, \ldots, f_k$  cluster around 0,
- (iii) for all i < j with  $i, j \in \{1, 2, ..., k\}$ , the function  $f_i/f_j$  is nonclustering.

Then for all nonzero  $c_1, \ldots, c_k \in \mathbb{R}$ , the arithmetic function  $F := c_1 f_1 + \cdots + c_k f_k$  is nonclustering.

Theorem 1 has consequences for the study of limit laws of arithmetic functions (for background, see, e.g., [14, Chapters III.2 and III.4] and [12, Chapter 4]). It is easy to see that for an arithmetic function F possessing a limit law (i.e., possessing a distribution function), the distribution function is continuous precisely when F is nonclustering. Now it is often the case that one can prove a distribution function exists by some general principle, but that the proof does not offer any insight into whether that function is continuous. Theorem 1 sometimes provides a convenient way of establishing continuity.

We illustrate by proving a recent conjecture of Luca and Pomerance. Let s(n) be the sum-of-proper-divisors function, so that  $s(n) = \sigma(n) - n$ . Let  $s_{\phi}(n) = n - \phi(n)$  denote the

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cototient function. In [10], Luca and Pomerance noted that  $s(n)/s_{\phi}(n) \geq 1$  for all  $n \geq 2$  and showed that the sequence  $\{s(n)/s_{\phi}(n)\}_{n=2}^{\infty}$  is dense in  $[1, \infty)$ . We prove:

**Theorem 2.** The arithmetic function  $s(n)/s_{\phi}(n)$  possesses a continuous distribution function  $D_{s/s_{\phi}}$ . Moreover,  $D_{s/s_{\phi}}(u)$  is strictly increasing for  $u \geq 1$ .

Theorem 2 was conjectured at the end of [10, §1].

## 2. Nonclustering of $c_1f_1 + \cdots + c_kf_k$ : Proof of Theorem 1

Our argument is modeled on work of Galambos and Kátai [6] concerning pairs of additive functions (generalizing an earlier result of Fein and Shapiro [5]).

2.1. **Setup.** Since  $f_1$  is nonclustering and  $c_1$  is nonzero, the theorem is obvious when k = 1. Proceeding inductively, we may assume that  $k \geq 2$  and that the theorem is already known to hold for all smaller values of k.

Let  $u \in \mathbb{R}$ . We will show that by making a judicious choice of  $\delta$ , the upper density of the set of n satisfying

$$(1) u - \delta < F(n) < u + \delta$$

can be made arbitrarily small.

Let  $\epsilon > 0$ . We let Y and Z be large, fixed real numbers (independent of n); their values will be specified more precisely in the course of the proof. To begin with, we assume that Y, Z > 2.

For each solution n to (1), we split off the Y-smooth part of n, writing

$$n = st$$
, where  $p \mid s \implies p \le Y$ , and  $p \mid t \implies p > Y$ .

(Here and below, p always refers to a prime.) We refer to this way of writing n as the 'basic decomposition', and we reserve the letters s and t for this purpose. We sometimes make use of obvious modifications of this notation, e.g., using s' and t' for the components in the decomposition of n'.

For a set S of positive integers, we write dS for its upper density.

2.2. Those n with large smooth part. It is known that the upper density of n with Y-smooth part larger than  $Y^Z$  is

$$\ll \exp(-cZ),$$

where c > 0 is an absolute constant, and the implied constant is also absolute (see [8, Theorem 07, p. 4]). Hence, this same expression bounds the upper density of solutions n to (1) with  $s > Y^Z$ .

2.3. Splitting the set of remaining n. Let S be the set of n satisfying (1) with  $s \leq Y^Z$ . We split S into two pieces,  $S_1$  and  $S_2$ , where

$$S_1 = \{ n \in S : \text{there is an } n' \in S \text{ with } t = t' \text{ and with } f_i(s) \neq f_i(s') \text{ for some } i \},$$
  
$$S_2 = S \setminus S_1.$$

We proceed to bound the upper densities of  $S_1$  and  $S_2$ .

2.4. Bounding  $\bar{\mathbf{d}}\mathcal{S}_1$ . Let  $n \in \mathcal{S}_1$ , and choose n' as in the definition of  $\mathcal{S}_1$ . Since n and n' both satisfy (1),

$$|F(n) - F(n')| = \left| \sum_{i=1}^{k} c_i f_i(n) - \sum_{i=1}^{k} c_i f_i(n') \right| < 2\delta.$$

Writing  $f_i(n) = f_i(s)f_i(t)$ ,  $f_i(n') = f_i(s')f_i(t')$  and keeping in mind that t = t', the preceding inequality becomes

$$\left| \sum_{i=1}^{k} c_i (f_i(s) - f_i(s')) f_i(t) \right| < 2\delta.$$

Let r = r(n) be the largest index in  $\{1, 2, ..., k\}$  with  $f_r(s) \neq f_r(s')$ . Then

$$\left| \sum_{i=1}^{r-1} c_i (f_i(s) - f_i(s')) \frac{f_i}{f_r}(t) + c_r (f_r(s) - f_r(s')) \right| < \frac{2}{|f_r(t)|} \delta.$$

Since none of  $f_1, \ldots, f_k$  cluster around 0, we may select  $\rho > 0$  (depending on the  $f_i$ ,  $\epsilon$ , Y, and Z) in such a way that the set  $\mathcal{T}$  of positive integers m satisfying  $|f_i(m)| < \rho$  for some i has upper density less than  $\epsilon Y^{-Z}$ . If  $|f_r(t)| < \rho$ , then  $t = n/s \in \mathcal{T}$ , and so  $n \in s\mathcal{T}$ . For each s,

$$\bar{\mathbf{d}}(s\mathcal{T}) = \frac{1}{s}\bar{\mathbf{d}}(\mathcal{T}) \le \bar{\mathbf{d}}(\mathcal{T}) < \epsilon Y^{-Z}.$$

But the number of possibilities for s is at most  $Y^Z$ . Thus, the set of  $n \in \mathcal{S}_1$  with  $|f_r(t)| < \rho$  has upper density at most  $\epsilon$ .

Suppose now that  $n \in \mathcal{S}_1$  and that  $|f_r(t)| \geq \rho$ . Then continuing the above calculation,

(2) 
$$\left| \sum_{i=1}^{r-1} c_i(f_i(s) - f_i(s')) \frac{f_i}{f_r}(t) + c_r(f_r(s) - f_r(s')) \right| < \frac{2}{\rho} \delta.$$

We enforce the condition that  $\delta > 0$  is small enough that

$$\frac{2}{\rho}\delta < \min_{1 \le i \le k} \min_{\substack{S, S' \le Y^Z \\ f_i(S) \ne f_i(S')}} |c_i(f_i(S) - f_i(S'))|.$$

Then (2) implies that there is at least one value of  $i \in \{1, 2, ..., r-1\}$  with  $f_i(s) \neq f_i(s')$ . We now apply the induction hypothesis to the list of functions  $f_i/f_r$ , where i runs over those indices not exceeding r-1 for which  $f_i(s) \neq f_i(s')$ . (It is easy to see that condition (iii) for the original list  $f_1, ..., f_k$  implies all of conditions (i)–(iii) for the new list of functions  $f_i/f_r$ .) This induction hypothesis implies that

$$\sum_{i=1}^{r-1} c_i (f_i(s) - f_i(s')) \frac{f_i}{f_r}$$

does not cluster around  $-c_r(f_r(s) - f_r(s'))$ . We may thus fix  $\delta_{r,s,s'} > 0$  small enough to guarantee that the set  $\mathcal{U}_{r,s,s'}$  of positive integers m satisfying

$$\left| \sum_{i=1}^{r-1} c_i (f_i(s) - f_i(s')) \frac{f_i}{f_r}(m) + c_r (f_r(s) - f_r(s')) \right| < \frac{2}{\rho} \delta_{r,s,s'}$$

has upper density smaller than  $\epsilon Y^{-2Z}k^{-1}$ . We make the further stipulation that our choice of  $\delta > 0$  satisfies

$$\delta < \min \delta_{r,s,s'}$$

where the minimum runs over all of the (finitely many!) possible triples r, s, s' that arise in this way.

With  $\delta$  so restricted, whenever (2) holds,  $n \in s\mathcal{U}_{r,s,s'}$ . Each set  $s\mathcal{U}_{r,s,s'}$  has upper density smaller than  $\epsilon Y^{-2Z}k^{-1}$ , while the number of possibilities for the triple r, s, s' is at most  $kY^{2Z}$ . Hence, the set of  $n \in \mathcal{S}_1$  with  $|f_r(t)| \geq \rho$  has upper density smaller than  $\epsilon$ .

We conclude that  $S_1$  has upper density smaller than  $2\epsilon$ .

2.5. Bounding  $dS_2$ . For each large real number x, we partition  $S_2 \cap [1, x]$  as follows. Given a pair of nonnegative integers U, V, we let  $S_2(U, V)$  be the subset of  $S_2 \cap [1, x]$  consisting of those n with

$$x/2^{U+1} < n \le x/2^U$$
 and  $x/2^{(U+1)+V} < t \le x/2^{U+V}$ .

Thus,

$$\mathcal{S}_2 \cap [1, x] = \bigcup_{U, V > 0} \mathcal{S}_2(U, V).$$

If  $n \in \mathcal{S}_2(U, V)$ , then

$$2^{V-1} < s = n/t < 2^{V+1}.$$

Since each  $n \in \mathcal{S}_2$  has  $s \leq Y^Z$ , the set  $\mathcal{S}_2(U, V)$  is empty unless  $2^{V-1} < Y^Z$ , and so we will assume this condition on V. To bound  $\#\mathcal{S}_2(U, V)$ , we first fix the large-primes component t and count the number of corresponding n. List these as

$$n_1 = s_1 t$$
,  $n_2 = s_2 t$ , ...,  $n_J = s_J t$ .

Then for each  $1 \leq i \leq k$ ,

$$f_i(s_1) = f_i(s_2) = \cdots = f_i(s_J);$$

otherwise, some of  $n_1, \ldots, n_J$  would belong to  $\mathcal{S}_1$ . In particular, every  $n \in \mathcal{S}_2(U, V)$  corresponding to this particular t has

$$f_1(s) = d$$

for a fixed d. By a theorem of Halász, the number of positive integers  $S < 2^{V+1}$  with  $f_1(S) = d$  is

(3) 
$$\ll 2^{V+1}/\sqrt{E(2^{V+1})}$$

with an absolute implied constant, where E(T) is defined for real values of T by

$$E(T) = \sum_{\substack{p \le T \\ f_1(p) \neq \pm 1}} \frac{1}{p}.$$

(To deduce this from the main theorem of [7], apply that result to the additive function  $\log |f_1(n)|$ .) Our hypothesis that  $f_1$  is nonclustering implies that the unrestricted sum

 $\sum_{p:\ f_1(p)\neq\pm 1} \frac{1}{p}$  diverges: Otherwise, the set of squarefree n divisible only by primes p with  $f_1(p)=\pm 1$  has density

$$\prod_{p:\ f_1(p)\neq \pm 1} \left(1-\frac{1}{p}\right) \prod_{p:\ f_1(p)=\pm 1} \left(1-\frac{1}{p^2}\right) > 0,$$

which forces  $f_1$  to cluster around one of  $\pm 1$ . Hence, the denominator in (3) tends to infinity with V. Thus, there is a positive integer  $V_0 = V_0(\epsilon)$  such that whenever  $V \geq V_0$ , the number of  $S < 2^{V+1}$  satisfying  $f_1(S) = d$  is at most  $\epsilon \cdot 2^{V+1}$ . (We could also have reached this conclusion by applying [3, Theorem IV] instead of [7].) We conclude that, for each fixed t, the number of corresponding  $n = st \in \mathcal{S}_2(U, V)$  is

$$\leq \begin{cases}
2^{V+1} & \text{always,} \\
\epsilon \cdot 2^{V+1} & \text{when } V \geq V_0.
\end{cases}$$

On the other hand, since  $t \leq x/2^{U+V}$  and has no prime factors in [2, Y], inclusion-exclusion shows that the number of possibilities for t is

$$\leq \frac{x}{2^{U+V}} \prod_{p \leq Y} \left( 1 - \frac{1}{p} \right) + O(2^Y) \leq \frac{x}{2^{U+V} \log Y} + O(2^Y).$$

Combining these upper bounds, we deduce that

$$\#\mathcal{S}_2(U,V) \le \begin{cases} \frac{2x}{2^U \log Y} + O(2^{V+Y}) & \text{always,} \\ \frac{2\epsilon x}{2^U \log Y} + O(2^{V+Y}) & \text{for } V \ge V_0. \end{cases}$$

Finally we sum over U and V. Let  $S_2(U) = \bigcup_V S_2(U, V)$ . Since we need only consider V with  $2^{V-1} < Y^Z$ , we have

$$\# \mathcal{S}_{2}(U) \leq \sum_{\substack{0 \leq V < V_{0} \\ V < \frac{\log(Y^{Z})}{\log 2} + 1}} \left( \frac{2x}{2^{U} \log Y} + O(2^{V+Y}) \right) + \sum_{\substack{V \geq V_{0} \\ V < \frac{\log(Y^{Z})}{\log 2} + 1}} \left( \frac{2\epsilon x}{2^{U} \log Y} + O(2^{V+Y}) \right) \\
\leq \frac{2V_{0}}{\log Y} \frac{x}{2^{U}} + 4\epsilon Z \frac{x}{2^{U}} + O(2^{Y} \cdot Y^{Z}).$$

Now we sum on all nonnegative U with  $2^U \leq x$  to find that

$$\#\mathcal{S}_2 \cap [1, x] \le \frac{4V_0}{\log Y} \cdot x + 8\epsilon Z \cdot x + O(2^Y \cdot Y^Z \cdot \log x).$$

It follows that  $S_2$  has upper density at most

$$\frac{4V_0}{\log Y} + 8\epsilon Z.$$

2.6. **Denouement.** Putting everything together, we see that the upper density of solutions to (1) is at most

$$C \exp(-cZ) + 2\epsilon + \frac{4V_0}{\log Y} + 8\epsilon Z,$$

where C and c are absolute positive constants. We now fix our choices of parameters  $\epsilon, Y, Z$ . Given any  $\eta > 0$ , we first fix Z large enough to make  $C \exp(-cZ) < \eta/3$ , then fix

 $\epsilon > 0$  small enough to make  $2\epsilon + 8\epsilon Z < \eta/3$ , and then finally fix Y large enough to make  $4V_0/\log Y < \eta/3$ . Our arguments then show that for a suitable of choice of  $\delta > 0$ , the set of n satisfying (1) has upper density  $< \eta$ .

3. 
$$s \vee s_{\phi}$$
: Proof of Theorem 2

We begin with a result of independent interest.

**Proposition 3.** Fix a nonzero real number R. Then  $F(n) = \frac{\sigma(n)}{n} + R \frac{\phi(n)}{n}$  possesses a continuous distribution function.

Proof that a (possibly discontinuous) distribution function exists. We argue via the method of moments. The argument is very similar to one described in detail in [11, §4], and so we only sketch the proof. For each positive integer k, define

$$\mu_k = \lim_{x \to \infty} \frac{1}{x} \sum_{n < x} \left( \frac{\sigma(n)}{n} + R \frac{\phi(n)}{n} \right)^k.$$

To see that  $\mu_k$  exists, it suffices to note that

$$(\sigma(n)/n + R\phi(n)/n)^k = \sum_{j=0}^k \binom{k}{j} R^{k-j} \sigma(n)^j \phi(n)^{k-j}/n^k$$

and that each of the functions  $\sigma(n)^j \phi(n)^{k-j}/n^k$  possesses a finite mean value, by a straightforward application of Wintner's mean value theorem [12, Theorem 1, p. 138]. Since

$$\binom{k}{j} \le 2^k$$
 and  $\sigma(n)^j \phi(n)^{k-j} / n^k \le (\sigma(n)/n)^k \le (n/\phi(n))^k$ ,

we can use the estimation of the moments of  $n/\phi(n)$  appearing in the proof of [11, Proposition 4.3] to deduce that

$$\mu_k \ll \exp(O(k \log \log(3k))).$$

(Here we allow implied constants to depend on R.) In particular, the condition

$$\limsup_{k\to\infty}\mu_{2k}^{1/2k}/k<\infty$$

that is required for application of [2, Theorem 3.3.12, p. 123] is satisfied, and so F(n) possesses a distribution function.

Proof of continuity. We apply Theorem 1 with  $f_1(n) = \sigma(n)/n$  and  $f_2(n) = \phi(n)/n$ . The Erdős-Wintner theorem [4] (see also [12, §4.7]), applied to  $\log f_1$ ,  $\log f_2$ , and  $\log(f_1/f_2)$  shows that all of  $f_1, f_2, f_1/f_2$  have continuous distribution functions, which immediately implies conditions (i)-(iii).

Remark 4. Results closely related to Proposition 3 can already be found in the literature. For example, [9] contains a proof of the continuity of the distribution function of  $\frac{\sigma(n)}{n} + \frac{\phi(n)}{n}$  in a strong form (a sharp estimate for the modulus of continuity). The strength of Theorem

1 is its ease of applicability and wide generality. To illustrate with a random example, an argument analogous to the above will prove that

$$c_1 \frac{\phi(n)}{\sigma(n)} + c_2 \exp\left(\sum_{p|n} \frac{1}{\log p}\right) + c_3 \frac{\sigma(n)\lambda(n)}{n}$$

has a continuous distribution function for any nonzero  $c_1, c_2, c_3$ . Here  $\sigma, \phi$  are as usual, and  $\lambda$  is the Liouville function, the completely multiplicative function with  $\lambda(p) = -1$  for every prime p. (To estimate the moments in this case one should appeal to [15, Sätze I, II] in place of Wintner's theorem.)

Proof of Theorem 2. Let u > 0. Writing  $s(n) = \sigma(n) - n$  and  $s_{\phi}(n) = n - \phi(n)$ , the inequality  $s(n)/s_{\phi}(n) \leq u$  can be put in the form

$$\frac{\sigma(n)}{n} + u \frac{\phi(n)}{n} \le 1 + u.$$

By Proposition 3,  $\frac{\sigma(n)}{n} + u \frac{\phi(n)}{n}$  possesses a continuous distribution function, say  $D_{1,u}$ . It follows that, for each u > 0,

$$\lim_{x \to \infty} \frac{1}{x} \# \{ 2 \le n \le x : s(n)/s_{\phi}(n) \le u \}$$

exists and equals  $D_{1,u}(1+u)$ . Since  $s(n)/s_{\phi}(n) \geq 1$ , the same limit also exists for  $u \leq 0$ , where it vanishes. We denote the value of this limit by  $D_{s/s_{\phi}}(u)$ .

We now check the boundary conditions necessary for  $D_{s/s_{\phi}}$  to qualify as a distribution function. It is trivial that  $\lim_{u\to\infty} D_{s/s_{\phi}}(u) = 0$ . To see that  $\lim_{u\to\infty} D_{s/s_{\phi}}(u) = 1$ , suppose that  $s(n)/s_{\phi}(n) > u$ , where u is large and positive. We can write this inequality in the form

$$\frac{\frac{\sigma(n)}{n} - 1}{1 - \frac{\phi(n)}{n}} > u.$$

So either  $\frac{\sigma(n)}{n} > 1 + \sqrt{u}$  or  $\frac{\phi(n)}{n} > 1 - \frac{1}{\sqrt{u}}$ . Each of these inequalities holds on a set of density tending to 0 as  $u \to \infty$ , since  $\frac{\sigma(n)}{n}$  and  $\frac{\phi(n)}{n}$  each have continuous distribution functions (e.g., by the Erdős–Wintner theorem again). It follows that  $1 - D_{s/s_{\phi}}(u) \to 0$  as  $u \to \infty$ , and hence  $D_{s/s_{\phi}}(u) \to 1$  as  $u \to \infty$ , as desired.

Now we show continuity of  $D_{s/s_{\phi}}(u)$ . It is certainly sufficient to consider values of  $u \geq 1$ . Given such a u, we will prove that the set of solutions n to

$$u - \delta < \frac{s(n)}{s_{\phi}(n)} < u + \delta$$

comprise a set of upper density tending to 0 as  $\delta \downarrow 0$ . Therefore  $s/s_{\phi}$  is nonclustering (provided one extends this quotient to be defined at n=1). Rearranging these inequalities for  $s(n)/s_{\phi}(n)$  yields

$$\frac{\sigma(n)}{n} + u \frac{\phi(n)}{n} \le 1 + u + \delta \left(1 - \frac{\phi(n)}{n}\right) \le 1 + u + \delta$$

as well as

$$\frac{\sigma(n)}{n} + u \frac{\phi(n)}{n} \ge 1 + u - \delta \left( 1 - \frac{\phi(n)}{n} \right) \ge 1 + u - \delta.$$

Now the desired result follows from the continuity of the distribution function  $D_{1,u}$ .

So far we have shown that  $s/s_{\phi}$  has a continuous distribution function  $D_{s/s_{\phi}}$ . It remains (only) to prove that  $D_{s/s_{\phi}}(u)$  is strictly increasing for  $u \geq 1$ .

We let  $a, b \ge 1$  with a < b and aim to show that  $D_{s/s_{\phi}}(a) < D_{s/s_{\phi}}(b)$ . By [10], the image of  $s/s_{\phi}$  is dense in  $[1, \infty)$ , and so we may fix an  $n_0$  such that

$$c := s(n_0)/s_{\phi}(n_0) \in (a, b).$$

We now argue that a positive proportion of the multiples n of  $n_0$  also satisfy  $s(n)/s_{\phi}(n) \in (a, b)$ . It is easy to prove (see the start of [10, §3]) that

$$s(n_0 m)/s_{\phi}(n_0 m) \ge s(n_0)/s_{\phi}(n_0) > a$$

for all m, and so it suffices to show that  $s(n_0m)/s_{\phi}(n_0m) < b$  holds a positive proportion of the time.

Let y be a large, fixed real parameter be specified more precisely below. To begin with, we assume y is so large that  $\prod_{p \leq y} (1 - 1/p) > 1/(2 \log y)$ . (This is true for all large y by Mertens' theorem, since  $e^{\gamma} < 2$ .) Let  $P_y$  be the product of the primes not exceeding y. Then for all sufficiently large x (depending on y),

(4) 
$$\#\{m \le x : \gcd(m, P_y) = 1\} > \frac{1}{2} x \prod_{p \le y} (1 - 1/p) > \frac{1}{4 \log y} x.$$

Moreover, recalling that  $\frac{\sigma(m)}{m} = \sum_{d|m} \frac{1}{d}$ , we have that

$$\sum_{\substack{m \le x \\ \gcd(m, P_y) = 1}} \left( \frac{\sigma(m)}{m} - 1 \right) = \sum_{\substack{m \le x \\ \gcd(m, P_y) = 1}} \sum_{\substack{d \mid m \\ d > 1}} \frac{1}{d} \le \sum_{\substack{d : p \mid d \implies p > y}} \frac{1}{d} \sum_{\substack{m \le x \\ d \mid m}} 1$$

$$\le x \sum_{\substack{d : p \mid d \implies p > y \\ d > 1}} \frac{1}{d^2} = x \left( \prod_{p > y} \left( 1 + \frac{1}{p^2} + \frac{1}{p^4} + \dots \right) - 1 \right).$$

The prime number theorem together with partial summation implies that

$$\prod_{p>y} \left( 1 + \frac{1}{p^2} + \frac{1}{p^4} + \dots \right) < \exp\left( \sum_{p>y} \frac{2}{p^2} \right)$$

$$\leq \exp\left( O\left( \frac{1}{y \log y} \right) \right) = 1 + O\left( \frac{1}{y \log y} \right).$$

Hence,

$$\sum_{\substack{m \le x \\ \gcd(m, P_y) = 1}} \left( \frac{\sigma(m)}{m} - 1 \right) \ll \frac{1}{y \log y} x,$$

so that the number of  $m \leq x$  with  $gcd(m, P_y) = 1$  and  $\frac{\sigma(m)}{m} - 1 \geq \frac{1}{\log y}$  is O(x/y). Comparing with (4), we see that if y is fixed sufficiently large, then for all large x,

(5) 
$$\#\{m \le x : \gcd(m, P_y) = 1, \frac{\sigma(m)}{m} - 1 < \frac{1}{\log y}\} > \frac{1}{8 \log y}x.$$

Increasing y if necessary, we may assume that y exceeds the largest prime factor of  $n_0$ . Then for any m counted on the left-hand side of (5),

$$\frac{\sigma(n_0 m)}{n_0 m} - 1 = \frac{\sigma(n_0)}{n_0} \frac{\sigma(m)}{m} - 1 \le \frac{\sigma(n_0)}{n_0} \left( 1 + \frac{1}{\log y} \right) - 1 = \frac{\sigma(n_0)}{n_0} - 1 + \frac{\sigma(n_0)/n_0}{\log y}.$$

Since also

$$1 - \frac{\phi(n_0 m)}{n_0 m} \ge 1 - \frac{\phi(n_0)}{n_0},$$

we find that

$$\frac{s(n_0 m)}{s_{\phi}(n_0 m)} = \frac{\frac{\sigma(n_0 m)}{n_0 m} - 1}{1 - \frac{\phi(n_0 m)}{n_0 m}} \le \frac{\frac{\sigma(n_0)}{n_0} - 1}{1 - \frac{\phi(n_0)}{n_0}} + \frac{\sigma(n_0)/n_0}{(1 - \frac{\phi(n_0)}{n_0})} \frac{1}{\log y}$$

$$= c + \frac{\sigma(n_0)/n_0}{(1 - \frac{\phi(n_0)}{n_0})} \frac{1}{\log y}.$$

Increasing y if necessary, we can ensure that this last expression is smaller than b.

With y fixed as above, (5) implies that the set of m with  $s(n_0m)/s_{\phi}(n_0m) < b$  has positive lower density. It follows that the corresponding values  $n = n_0m$  also comprise a set of positive lower density. Together with our earlier remarks, we conclude that  $D_{s/s_{\phi}}(a) < D_{s/s_{\phi}}(b)$ , as desired. This completes the proof that  $D_{s/s_{\phi}}$  is increasing as well as the proof of Theorem 2.

## 4. Concluding remarks on positive-valued multiplicative functions

Theorem 1 is well-suited to proving the continuity of a distribution function when it exists. It is therefore natural to ask for a general condition guaranteeing that  $F = c_1 f_1 + \cdots + c_k f_k$  possesses a distribution function. We conclude by sketching a proof of the following partial result in this direction. The argument is due essentially to Shapiro [13] (see especially p. 63), but as the case we work in is much simpler than his general set-up, it seems a relatively self-contained discussion is warranted.

**Proposition 5.** Let  $f_1, \ldots, f_k$  be positive-valued multiplicative functions each possessing a distribution function. Then for any  $c_1, \ldots, c_k \in \mathbb{R}$ , the function  $c_1 f_1 + \cdots + c_k f_k$  also has a distribution function.

Note that this result applies, for instance, to the example considered in Proposition 3, but not immediately to the one considered in Remark 4.

Let Y > 0. We keep the notation of §2, where n denotes a positive integer and s denotes the Y-smooth part of n. (There will be no confusion with the sum-of-proper-divisors function.) We say that an arithmetic function F is essentially determined by small primes if for all  $\epsilon > 0$ ,

$$\lim_{Y \to \infty} \bar{\mathbf{d}}\{n : |F(n) - F(s)| > \epsilon\} = 0.$$

If F is an arithmetic function essentially determined by small primes, then F has a distribution function; this is contained in [13, Theorem 2.1], and also follows from [14, Theorem 2.3, p. 427]. Moreover, the converse holds for all additive functions F (see the theorem stretching from pp. 719–720 in [4]).

To relate this back to Proposition 5, we recall that when a positive-valued multiplicative function possesses a limit law, either its distribution function is that of the degenerate distribution at 0, or the additive function  $\log f$  has a distribution function. (See [1, Theorem 4], and note that the convergence of the three series in eq. (3) there is exactly the Erdős-Wintner condition for  $\log f$  to have a distribution function.) Now given  $f_1, \ldots, f_k$  as in Proposition 5, we may reorder the list so that  $f_1, \ldots, f_\ell$  have distributions degenerate at 0, and  $f_{\ell+1}, \ldots, f_k$  do not. It is then easy to see that if  $c_{\ell+1}f_{\ell+1} + \cdots + c_kf_k$  has a distribution function, then  $c_1f_1 + \cdots + c_kf_k$  has the same distribution function. Thus, we can (and do) assume that each of the  $\log f_i$  has a distribution function. As discussed in the previous paragraph, this means that each  $\log f_i$  is essentially determined by small primes. We claim that each  $f_i$  is also so-determined. Indeed, suppose that

$$|f_i(n) - f_i(s)| > \epsilon.$$

Then, with  $\eta > 0$  a parameter at our disposal, either  $f_i(n) > \eta$ , or

$$|f_i(s)/f_i(n)-1|>\epsilon/\eta.$$

This last inequality implies that

$$|\log f_i(n) - \log f_i(s)| \gg_{\epsilon,\eta} 1;$$

since  $\log f_i$  is essentially determined by small primes, this estimate holds on a set of upper density tending to 0 as  $Y \to \infty$ . On the other hand, if  $f_i(n) > \eta$ , then  $\log f_i(n) > \log \eta$ . That occurs on a set of upper density tending to 0 as  $\eta$  tends to infinity, since  $\log f_i$  has a (proper) distribution function. Letting  $Y \to \infty$  and then letting  $\eta \to \infty$  proves our claim.

Since the  $f_i$  are essentially by determined by small primes, so is any  $\mathbb{R}$ -linear combination of the  $f_i$ ; thus, all such combinations possess distribution functions.

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