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On ultrafriable integers

Gérald Tenenbaum

Abstract. Say that an integer n is y -ultrafriable if its canonical decomposition is free of prime powers exceeding y . We investigate the asymptotic behaviour of the distribution function $\Upsilon(x, y)$, equal to the number of y -ultrafriable integers not exceeding x . The study being restricted to the range $\psi(y) > 2 \log x$ (where ψ denotes Chebyshev's function) by a symmetry argument, and writing $\Psi(x, y)$ for the number of y -friable integers not exceeding x , $\beta = \beta(x, y)$ for the saddle-point associated to the Dirichlet series $Z(s, y) := \prod_{p \leq y} (1 - p^{-(\nu_p+1)s}) / (1 - p^{-s})$ where $\nu_p := \lfloor (\log y) / \log p \rfloor$, we obtain full description via a number of effective estimates, qualitative versions of which may be stated as follows:

- (i) $\Upsilon(x, y) \sim \Psi(x, y) \Leftrightarrow y / (\log x)^2 \rightarrow \infty \quad (x \rightarrow \infty)$;
- (ii) $\Upsilon(x, y) = o(\Psi(x, y)) \Leftrightarrow y / (\log x)^2 \rightarrow 0 \quad (x \rightarrow \infty)$;
- (iii) $\Upsilon(x, y) \sim x^\beta Z(\beta, y) G(\beta \sqrt{\sigma_2}) \quad (x \rightarrow \infty, 2 \log x < \psi(y) \leq (\log x)^3)$.

Here $\sigma_2 := d^2 \{ \log Z(\beta, y) \} / ds^2$ and $G(z) := e^{z^2/2} \int_z^\infty e^{-t^2/2} dt / \sqrt{2\pi} \quad (z > 0)$.

Moreover, the phase transition between domains (i) and (ii) is quantitatively described in terms of the Dickman function, the Gaussian behaviour as $\log x$ approaches $\frac{1}{2}\psi(y)$ is made explicit, and a quantitative estimate for the local behaviour with respect to the variable x is derived.

Keywords: Friable integers, integers free of large prime factors, saddle-point method, local behaviour, large deviations in the central limit theorem.

1. Introduction and statement of results

A positive integer n is said to be y -friable if its largest prime factor $P^+(n)$ (with the convention that $P^+(1) = 1$) does not exceed y . In the last twenty years, friable integers, and in particular the counting function

$$\Psi(x, y) := \sum_{\substack{n \leq x \\ P^+(n) \leq y}} 1,$$

received considerable attention in the literature. In this paper, we investigate a related structure.

Let us say that an integer n is y -ultrafriable if no prime power dividing n exceeds y . Intrinsically a sieve problem but also relevant to other fields such as irreducibility of polynomials [1], graph theory (see, e.g., [9]), or the study of so-called economical integers [4],⁽¹⁾ the distribution of ultrafriable integers raises interesting methodological questions which hopefully can be fairly satisfactorily answered using available techniques previously developed in the context of friable integers.

Denote by $\Upsilon(x, y)$ the number of y -ultrafriable integers not exceeding x and put $\nu_p = \nu_p(y) := \lfloor (\log y) / \log p \rfloor$ for each prime $p \leq y$, so that p^{ν_p} is the largest power of p not exceeding y . Writing $N_y := e^{\psi(y)}$ where $\psi(y) := \sum_{p \leq y} \nu_p \log p$ is Chebyshev's function, we note that all integers counted in $\Upsilon(x, y)$ are divisors of N_y , and hence

$$(1.1) \quad \Upsilon(x, y) = \tau(N_y) = \prod_{p \leq y} (1 + \nu_p) = 2^{\pi(y) + O(\sqrt{y}/\log y)} \quad (x \geq N_y),$$

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1. An integer n is said to be economical in base q if its prime factorisation can be written with no more digits than n itself in base q . Thus $14 = 2 \cdot 7$, $15 = 3 \cdot 5$ and $16 = 2^4$ are economical in base 10 but $18 = 2 \cdot 3^2$ is not.

where $\pi(y)$ denotes the number of primes not exceeding y .

Next, we observe that

$$\Upsilon(x, y) = \tau(N_y) - \Upsilon\left((N_y/x)-, y\right) \quad (\sqrt{N_y} \leq x \leq N_y).$$

Therefore, we may restrict the study of $\Upsilon(x, y)$ to the case

$$(1.2) \quad x < \sqrt{N_y}, \text{ i.e. } \psi(y) > 2 \log x.$$

The Dirichlet series associated to the counting function $\Upsilon(x, y)$ is

$$(1.3) \quad Z(s, y) := \prod_{p \leq y} \frac{1 - p^{-(\nu_p+1)s}}{1 - p^{-s}} \quad (\Re s > 0).$$

While, for large values of y , we readily obtain satisfactory estimates for $\Upsilon(x, y)$ from results on the local behaviour of $\Psi(x, y)$ (see [3]), we need to perform a direct evaluation of the Perron integral when y is small. The details are then similar to those appearing in the study of squarefree friable integers provided in [2], with however some significant discrepancies. When it will seem appropriate, we shall skip certain calculations and refer to the corresponding details in [2].

The saddle-point, say $\beta = \beta(x, y)$, relevant to the Perron integral for $\Upsilon(x, y)$ is defined by the equation

$$(1.4) \quad \varphi_1(\beta, y) = \log x$$

where

$$(1.5) \quad \varphi_1(s, y) := \frac{-Z'(s, y)}{Z(s, y)} = \sum_{p \leq y} \left\{ \frac{\log p}{p^s - 1} - \frac{(\nu_p + 1) \log p}{p^{(\nu_p+1)s} - 1} \right\} \quad (\Re s > 0, y \geq 2).$$

Then $\varphi_1(\sigma, y)$ is a decreasing function of σ such that $\varphi_1(0+, y) = \frac{1}{2}\psi(y)$, $\varphi_1(\infty, y) = 0$. Thus, under assumption (1.2), equation (1.4) has a unique solution and β is well defined. For convenience, we put

$$(1.6) \quad \beta := 0 \quad (\psi(y) \leq 2 \log x).$$

We write

$$(1.7) \quad \varphi_j(s, y) := (-1)^{j-1} \frac{d^{j-1} \varphi_1}{d\sigma^{j-1}}(s, y), \quad \sigma_j := \varphi_j(\beta, y) \quad (j \geq 1),$$

let

$$\Phi(z) := \frac{1}{\sqrt{2\pi}} \int_z^\infty e^{-t^2/2} dt \quad (z \in \mathbb{R})$$

denote the decreasing distribution function of the Gaussian law, and put

$$(1.8) \quad G(z) := e^{z^2/2} \Phi(z) \quad (z \in \mathbb{R}).$$

Thus

$$(1.9) \quad G(z) = \frac{1}{2} + O(z) \quad (z \rightarrow 0), \quad G(z) = \frac{1}{\sqrt{2\pi}z} \left\{ 1 - \frac{1}{z^2} + O\left(\frac{1}{z^4}\right) \right\} \quad (z \rightarrow +\infty).$$

We also define

$$\mathcal{H}_y(\sigma) := \prod_{p \leq y} \left(1 - \frac{1}{p^{(\nu_p+1)\sigma}}\right) \quad (\sigma > 0),$$

and make systematic use of the notation

$$u := \frac{\log x}{\log y} \quad (x \geq y \geq 2).$$

Theorem 1.1. *Let $\varepsilon > 0$. For $x \geq y \geq 2$, we have*

$$(1.10) \quad \Upsilon(x, y) = \Psi(x, y) \left\{ 1 + O\left(\frac{u \log 2u}{\sqrt{y} \log y}\right) \right\} \quad (x \geq y > (\log x)^{2+\varepsilon}),$$

$$(1.11) \quad \Upsilon(x, y) = x^\beta Z(\beta, y) G(\beta \sqrt{\sigma_2}) \left\{ 1 + O\left(\frac{1}{u}\right) \right\} \quad (2 \log x < \psi(y) \ll (\log x)^3).$$

From this result, a number of more explicit estimates may be easily deduced.

Some further notation is necessary to describe the results. Let $\alpha = \alpha(x, y)$ denote the saddle-point of the Perron integral for $\Psi(x, y)$. So α is defined by the equation

$$(1.12) \quad \sum_{p \leq y} \frac{\log p}{p^\alpha - 1} = \log x.$$

Explicit approximations to α are widely available in the literature. In particular, it is proved in [5] that

$$(1.13) \quad \alpha = \frac{1}{\log y} \log \left(1 + \frac{y}{\log x}\right) \left\{ 1 + O\left(\frac{\log_2 2y}{\log y}\right) \right\} \quad (x \geq y \geq 2),$$

and, more precisely, that, for any $\varepsilon > 0$,

$$(1.14) \quad \alpha = \begin{cases} 1 - \frac{\xi(u)}{\log y} + O\left(\frac{1}{L_\varepsilon(y)} + \frac{1}{u(\log y)^2}\right) & ((\log x)^{1+\varepsilon} \leq y \leq x), \\ \frac{\log(1 + y/\log x)}{\log y} \left\{ 1 + O\left(\frac{1}{\log y}\right) \right\} & (2 \leq y \leq (\log x)^3), \end{cases}$$

where $\xi(v)$ is defined as the unique solution of the equation $e^{\xi(v)} = 1 + v\xi(v)$ for $v > 1$ and $\xi(1) := 0$. Here and in the sequel, we let $\log_k$ denote the k -fold iterated logarithm and put

$$(1.15) \quad L_\varepsilon(y) := e^{(\log y)^{3/5-\varepsilon}} \quad (\varepsilon > 0, y \geq 2).$$

Available studies of the local behaviour of the counting function $\Psi(x, y)$ —see in particular th. 2.4 of [3]—provide various effective versions of the approximation

$$(1.16) \quad \Psi\left(\frac{x}{h}, y\right) \approx \frac{\Psi(x, y)}{h^\alpha} \quad (x \geq y \geq 2, 1 \leq h \leq x).$$

For $P^+(d) \leq y$, put $h_d := \prod_{p|d} p^{\nu_p+1}$. Then the characteristic function of y -ultrafriable integers may be written as

$$\sum_{\substack{P^+(d) \leq y \\ h_d | n}} \mu(d)$$

and we have the sieve formula

$$\Upsilon(x, y) = \sum_{P^+(d) \leq y} \mu(d) \Psi\left(\frac{x}{h_d}, y\right) \quad (x \geq y \geq 2).$$

Considering (1.16), this leads to the expectation that, in suitable ranges, we should have

$$(1.17) \quad \Upsilon(x, y) \sim \Psi(x, y) \sum_{P^+(d) \leq y} \frac{\mu(d)}{h_d^\alpha} = \Psi(x, y) \mathcal{H}_y(\alpha).$$

Our first corollary below shows that the right-hand side of (1.17) always constitutes an upper bound for the order of magnitude of $\Upsilon(x, y)$ and that this turns into a lower bound when α is replaced by β . The second part of the statement provides the exact domain of validity of (1.17).

We introduce the remainder term

$$E = E(x, y) := \begin{cases} \frac{u \log 2u}{\sqrt{y}} & \text{if } y > \frac{(\log x)^3}{\log_2 2x}, \\ \frac{1}{\sqrt{u}} + \frac{(\log x)^3}{y^2 \log_2 2x} & \text{if } 2 \leq y \leq \frac{(\log x)^3}{\log_2 2x}, \end{cases}$$

and note that

$$E \ll \sqrt{\frac{\log_2 2x}{\log x}} + \frac{(\log x)^3}{y^2 \log_2 2x} \quad (x \geq y \geq 2).$$

Corollary 1.2. (i) *We have*

$$(1.18) \quad \mathcal{H}_y(\beta) \Psi(x, y) \ll \Upsilon(x, y) \ll \mathcal{H}_y(\alpha) \Psi(x, y) \quad (x \geq y \geq 2).$$

(ii) *As $x \rightarrow \infty$, we have*

$$(1.19) \quad \Upsilon(x, y) \sim \mathcal{H}_y(\alpha) \Psi(x, y) \Leftrightarrow y \sqrt{\log_2 x} / (\log x)^{3/2} \rightarrow \infty,$$

$$(1.20) \quad \Upsilon(x, y) \sim \mathcal{H}_y(\beta) \Psi(x, y) \Leftrightarrow y \sqrt{\log_2 x} / (\log x)^{3/2} \rightarrow \infty.$$

Moreover,

$$(1.21) \quad \Upsilon(x, y) = \{1 + O(E)\} \mathcal{H}_y(\alpha) \Psi(x, y) \quad (y \gg (\log x)^{3/2} / \sqrt{\log_2 x}),$$

$$(1.22) \quad \Upsilon(x, y) = \{1 + O(E)\} \mathcal{H}_y(\beta) \Psi(x, y) \quad (y \gg (\log x)^{3/2} / \sqrt{\log_2 x}).$$

In view of (1.6), the lower bound in (1.18) vanishes when $\psi(y) \leq 2 \log x$. However, as observed earlier, we have $\frac{1}{2} \tau(N_y) \leq \Upsilon(x, y) \leq \tau(N_y)$ in this circumstance.

As is already apparent in the statement of Theorem 1.1, the asymptotic fluctuations of $\Upsilon(x, y)$ present a threshold around $y \approx (\log x)^2$. Our next corollary exhibits the behaviours on either side of this threshold and describes the phase transition.

We recall that the Dickman function $\varrho : \mathbb{R}^+ \rightarrow [0, 1]$ is defined as the continuous solution to the delay differential equation $v \varrho'(v) + \varrho(v-1) = 0$ such that $\varrho(v) = 1$ for $0 \leq v \leq 1$. An explicit expression of its Laplace transform is well-known (see, e.g., [11], th. III.5.10):

$$\widehat{\varrho}(s) = \int_0^\infty e^{-sv} \varrho(v) dv = e^{\gamma + I(-s)}, \quad I(s) := \int_0^s \frac{e^t - 1}{t} dt,$$

where γ denotes Euler's constant.⁽²⁾ On the real axis, $\widehat{\varrho}$ has a simple behaviour:

$$(1.23) \quad \widehat{\varrho}(t) = \frac{1 + O(e^{-t}/t)}{t} \quad (t \rightarrow +\infty), \quad \widehat{\varrho}(t) = \exp \left\{ \frac{e^{|t|}}{|t|} + O\left(\frac{e^{|t|}}{|t|^2}\right) \right\} \quad (t \rightarrow -\infty).$$

Corollary 1.3. *As $x \rightarrow \infty$, we have*

$$(1.24) \quad \Upsilon(x, y) \sim \Psi(x, y) \Leftrightarrow y/(\log x)^2 \rightarrow \infty,$$

$$(1.25) \quad \Upsilon(x, y) = o(\Psi(x, y)) \Leftrightarrow y/(\log x)^2 \rightarrow 0.$$

Moreover, for any $\varepsilon > 0$, and uniformly for $(\log x)^2/L_\varepsilon(\log x) \leq y \leq (\log x)^{5/2}/(\log_2 x)^{3/2}$, we have

$$(1.26) \quad \Upsilon(x, y) = \left\{ 1 + O\left(\frac{1}{\sqrt{u}}\right) \right\} \mathcal{H}_y(\beta) \Psi(x, y) = \left\{ 1 + O\left(\frac{1}{L_\varepsilon(y)}\right) \right\} \frac{\widehat{\varrho}(h)}{2\widehat{\varrho}(2h)} \Psi(x, y),$$

with $h := \frac{1}{2} \log y - \xi(u)$.

To clarify expectations, we note that the parameter h appearing in (1.26) satisfies $e^h \sim 2\sqrt{y}/\log x$ in the critical range $y = (\log x)^{2+o(1)}$. We also observe that when $y > (\log x)^{5/2}/(\log_2 x)^{3/2}$, the estimate (1.10) is more precise than (1.26).

The following result concerns smaller values of y , when the saddle-point estimates for $\Upsilon(x, y)$ and $\Psi(x, y)$ assume different shapes. This corresponds essentially to the case $2\log x < \psi(y) < (2+c)\log x$ for suitable $c \in]0, \frac{1}{2}[$. We obtain a large deviation result which measures the Gaussian distribution of the divisors of $N_y = e^{\psi(y)}$. The proof will be omitted since it is identical, *mutatis mutandis*, to that of Corollary 2.2 of [2].

Corollary 1.4. *Let $y \geq 2$, $N_y = e^{\psi(y)}$, $D_y^2 := \frac{1}{12} \sum_{p \leq y} \nu_p(\nu_p + 2)(\log p)^2$. Uniformly for $0 \leq z \ll (y/\log y)^{1/4}$ and $x = \sqrt{N_y} e^{-zD_y}$, we have*

$$(1.27) \quad \Upsilon(x, y) = \tau(N_y) \Phi(z) \left\{ 1 + O\left(\frac{1+z^4}{u}\right) \right\}.$$

According to a remark developed in [2] and still valid in the present context, we note that Petrov's effective theorem on large deviations in the central limit theorem (see [7], th. VIII.2) provides estimates that are similar in nature to, but less precise than (1.27) in its range of validity.

Finally, as a specific by-product of saddle-point asymptotic formulae, we state a result on the local behaviour of $\Upsilon(x, y)$. We also omit the proof, since it is identical to that of corollary 2.4 of [3],⁽³⁾ and furthermore leave to the reader the possibility of deriving corresponding short interval estimates parallel to theorem 2.5 and corollary 2.6 of [2].

Corollary 1.5. *Uniformly under the conditions $x \geq y \geq 2$, $1 \leq d \leq y$, $\psi(y) > 2\log(dx)$, we have*

$$(1.28) \quad \Upsilon(dx, y) = d^\beta \Upsilon(x, y) \left\{ 1 + O\left(\frac{1}{\sqrt{u}}\right) \right\}.$$

2. We shall take the liberty to use the letter γ for other purposes later in the paper.

3. Apart from the simplification due to the fact that, in view of (1.10), the required result follows directly, for $y > (\log x)^3$, from known results on the local behaviour of $\Psi(x, y)$ —see [3].

2. Lemmas

We start with a useful elementary inequality.

Lemma 2.1. *Let $\nu \in \mathbb{N}^*$, $z > 1$. Then*

$$(2.1) \quad \frac{1}{z+1} \leq \frac{1}{z-1} - \frac{\nu+1}{z^{\nu+1}-1} \leq \frac{\nu}{z+1}.$$

Moreover, the right-hand inequality is strict when $\nu \geq 2$.

Proof. The left-hand inequality is equivalent to

$$\frac{\nu+1}{z^{\nu+1}-1} \leq \frac{1}{z-1} - \frac{1}{z+1} = \frac{2}{z^2-1}.$$

This is clear since $(z^v - 1)/v$ is an increasing function of $v > 0$ when $z > 1$.

To prove the right-hand inequality, we may assume $\nu \geq 2$ since we trivially have equality when $\nu = 1$. Then, the required inequality may be rewritten as

$$(z+1)(z^{\nu+1}-1) < (\nu+1)(z^2-1) + \nu(z-1)(z^{\nu+1}-1),$$

which, after straightforward transformations, amounts to

$$(2.2) \quad f(z) := (\nu-1)(z^\nu+1) - 2 \sum_{1 \leq j < \nu} z^j > 0.$$

However, we have $f(1) = 0$, and $f'(z) = \nu(\nu-1)z^{\nu-1} - 2 \sum_{1 \leq j < \nu} j z^{j-1} > 0$ for $z > 1$. $\square$

Our next lemma provides uniform estimates for the sums

$$(2.3) \quad g_{\pm}(\sigma, y) := \sum_{p \leq y} \frac{\log p}{p^{\sigma} \pm 1};$$

as observed in [2], these may be proved by partial summation from a strong form of the prime number theorem along the lines described in [5], lemma 13. We omit the details.

Lemma 2.2. *Uniformly for $0 < \sigma < 2$, $y \geq 2$, we have*

$$(2.4) \quad g_{\pm}(\sigma, y) = \frac{y^{1-\sigma} - 1}{(1 \pm y^{-\sigma})(1 - \sigma)} \left\{ 1 + O\left(\frac{1}{\log y}\right) \right\} + O(1).$$

Moreover, given any $\sigma_0 > 0$, $\varepsilon > 0$, the remainder term $O(1/\log y)$ may be replaced by $O(1/L_{\varepsilon}(y))$ when $\sigma \geq \sigma_0$.

We shall also need an estimate for the order of magnitude of the quantity

$$(2.5) \quad V_y(\sigma) := \sum_{p \leq y} \frac{(\nu_p + 1) \log p}{p^{(\nu_p + 1)\sigma} - 1} \quad (\sigma > 0).$$

Lemma 2.3. *For $y \geq 2$, $1/\log y < \sigma \leq 2$, we have*

$$(2.6) \quad V_y(\sigma) \asymp \frac{y^{1/2-\sigma}(1 + y^{1/2-\sigma}) \log y}{1 + |1 - 2\sigma| \log y}.$$

Proof. We have $V_y(\sigma) \asymp (S + T) \log y$, with

$$S := \sum_{p \leq \sqrt{y}} \frac{1}{p^{(\nu_p + 1)\sigma}}, \quad T := \sum_{\sqrt{y} < p \leq y} \frac{1}{p^{2\sigma}}.$$

By the inequality $p^{\nu_p+1} > y$ and the prime number theorem, we may write

$$S \ll \frac{y^{1/2-\sigma}}{\log y}, \quad T \asymp \frac{y^{1-2\sigma} - y^{1/2-\sigma}}{(1-2\sigma)\log y} \asymp \frac{y^{1/2-\sigma}(1+y^{1/2-\sigma})}{1+|1-2\sigma|\log y},$$

in view of the uniform estimate

$$(2.7) \quad \frac{e^t - 1}{t} \asymp \frac{1 + e^t}{1 + |t|} \quad (t \in \mathbb{R}).$$

Thus $S \ll T$, whence $V_y(\sigma) \asymp T \log y$, as required. $\square$

The following result provides explicit estimates for β in terms of x and y . The asymptotic behaviour of the function ξ appearing in (1.14) has been described in [6]. In particular, we have

$$(2.8) \quad \xi(v) = \log v + \log_2 v + \frac{\log_2 v}{\log v} + O\left(\left(\frac{\log_2 v}{\log v}\right)^2\right) \quad (v \geq 3).$$

Lemma 2.4. *Let $\varepsilon > 0$.*

(i) *For $x \geq x_0(\varepsilon)$, $(\log x)^{1+\varepsilon} < y \leq x$, we have*

$$(2.9) \quad \beta = 1 - \frac{\xi(u)}{\log y} + O\left(\frac{1}{u(\log y)^2} + \frac{1}{L_\varepsilon(y)}\right).$$

(ii) *For $x \geq 2$, $2 \log x < \psi(y) \ll (\log x)^3$, we have*

$$(2.10) \quad \beta = \frac{1 + O(1/\log y)}{\log y} \log\left(\frac{\psi(y)}{\log x} - 1\right).$$

(iii) *For $x \geq y \geq 2$, $\psi(y) > 2 \log x$, $r := y/(\log x)^2$, we have,*

$$(2.11) \quad \alpha - \beta \asymp \frac{(\log 2u)(\sqrt{y} + u \log 2u)}{y(\log y)(1 + |\log r|)}.$$

Proof. We note that, in its range of validity, (2.9) follows from (2.11) in view of (1.14). However, it will be convenient to derive (2.9) as a preliminary step. By (2.1), we have, for all $\sigma > 0$,

$$(2.12) \quad g_+(\sigma, y) \leq \varphi_1(\sigma, y) \leq g_+(\sigma, y) + r(\sigma, y),$$

say, with

$$(2.13) \quad r(\sigma, y) := \sum_{p \leq \sqrt{y}} \frac{(\nu_p - 1) \log p}{p^\sigma + 1}.$$

Let us first assume $y > (\log x)^3$. Then we deduce from (2.4) and the left-hand inequality above that $\beta > 3/5$ provided x is sufficiently large. Inserting this back into (1.5) and taking (2.6) into account, we obtain

$$\varphi_1(\beta, y) = g_-(\beta, y) + O(y^{-1/10}).$$

This is sufficient to deduce (2.9) by computations identical to those leading to estimate (7.8) of [5].

Next, we consider the case $(\log x)^{1+\varepsilon} < y \leq (\log x)^3$. Then it follows from (2.8) and (2.9) that $\beta \leq \beta(x, 2(\log x)^3) \leq \frac{2}{3} + O(1/\log y)$. Therefore

$$r(\beta, y) \ll (\log y) \sum_{p \leq \sqrt{y}} \frac{1}{p^\beta} \ll y^{(1-\beta)/2}$$

where the last bound readily follows by partial summation—see lemma 3.6 of [3] for a general estimate. We hence deduce from (2.4) that

$$(2.14) \quad r(\beta, y) \ll g_+(\beta, y)y^{-1/6}.$$

Thus we obtain that the estimate (2.4) for $g_+(\beta, y)$ equally holds for $\varphi_1(\beta, y)$. We may now again deduce (2.9) by computations parallel to those leading to estimate (7.8) of [5]. We refer the reader to [5] for the details.

Let us next evaluate β when $2\log x < \psi(y) \leq (\log x)^3$. In this range, it follows from (2.14) that

$$(2.15) \quad g_+(\beta, y) = \left\{1 + O\left(\frac{1}{y^{1/6}}\right)\right\} \log x.$$

By (2.4), and since we have $\beta < 3/4$, this plainly implies

$$(2.16) \quad y^{1-\beta} \asymp \log x.$$

Now, by (2.1) with $z = p^\beta$ and $\nu = \nu_p = \nu_p(y)$,

$$- \sum_{p \leq \sqrt{y}} \frac{(\nu_p - 1) \log p}{1 + p^\beta} \leq \varphi_1(\beta, y) - \frac{\psi(y)}{1 + y^\beta} \leq \sum_{p \leq y} \frac{\nu_p(y)(y^\beta - p^\beta) \log p}{(1 + p^\beta)(1 + y^\beta)}.$$

By partial summation, we obtain that the above upper bound is $\ll \beta y^{1-\beta} \ll \beta \log x$. Moreover, by (2.14) and (2.15), the lower bound is

$$-r(\beta, y) \ll \frac{\log x}{y^{1/6}}.$$

Let γ denote the solution to the equation $\psi(y)/(1 + y^\gamma) = \log x$. We deduce from the above that

$$|\beta - \gamma|(\log x) \log y \ll \frac{\psi(y)|y^\beta - y^\gamma|}{(1 + y^\beta)(1 + y^\gamma)} \ll \left(\beta + \frac{1}{y^{1/6}}\right) \log x.$$

This yields (2.10) if, say, $\psi(y) > (2 + y^{-1/7}) \log x$. In the complementary case, we appeal to the estimate

$$\varphi_1(\sigma, y) = \frac{1}{2}\psi(y) - \sigma D_y^2 + O(\sigma^2 y (\log y)^2) \quad (0 \leq \sigma \log y < \frac{1}{2})$$

where D_y^2 is defined in Corollary 1.4 and satisfies $D_y^2 = \frac{1}{4}y \log y + O(y)$. This readily follows from the classical expansion of $z/\{e^z - 1\}$ involving Bernoulli numbers, up to the third order. Substituting $\sigma = \beta$ yields (2.10).

We are now in a position to prove the more precise estimate (2.11). From (1.14), (2.9) and (2.10), we see that, for any $\vartheta \in [0, 1]$ and $\gamma = \alpha + \vartheta(\beta - \alpha)$, we have in the considered range

$$(2.17) \quad g'_-(\gamma, y) = e^{O(R)} \frac{(y^{1-\gamma} - 1) \log y}{(1 - y^{-\gamma})^2 (1 - \gamma)} \asymp u(\log y)^2,$$

with $R := 1/\log 2u + 1/\log y$, where the first estimate is proved in lemmas 4 and 13 of [5], using a strong form of the prime number theorem. Moreover, recalling the notations (2.5) and $r := y/(\log x)^2$, we have

$$(2.18) \quad \begin{aligned} g_-(\beta, y) - g_-(\alpha, y) &= g_-(\beta, y) - \varphi_1(\beta, y) = \sum_{p \leq y} \frac{(\nu_p + 1) \log p}{p^{(\nu_p + 1)\beta} - 1} \\ &= V_y(\beta) \asymp \frac{u(\log 2u)(\sqrt{y} + u \log 2u) \log y}{y(1 + |\log r|)}, \end{aligned}$$

where the last estimate follows from (2.6), (2.9) and (2.10), since

$$(2.19) \quad y^{1/2-\beta} \asymp \frac{u \log 2u}{\sqrt{y}} \quad (\psi(y) > 2 \log x).$$

Estimate (2.11) readily follows from this and (2.17), by the mean-value theorem. $\square$

We next evaluate the derivatives σ_j ($j \geq 1$) defined in (1.7) for comparatively small values of y .

Lemma 2.5. *Let $j \in \mathbb{N}^*$ be fixed. Uniformly for $2 \log x < \psi(y) \ll (\log x)^3$, we have*

$$(2.20) \quad \sigma_j \ll u(\log y)^j.$$

For $j = 3$, the right-hand side may be multiplied by $\min(1, \beta \log y) + 1/\sqrt{u}$. Moreover, when $j = 2$, we may replace the $\ll$ -sign by $\asymp$. More precisely,

$$(2.21) \quad \sigma_2 = \left\{ 1 + O\left(\frac{1}{\log y}\right) \right\} \frac{w-1}{w} u(\log y)^2$$

with $w := \psi(y)/\log x$.

Proof. Put $R_p(z) := \sum_{0 \leq h \leq \nu_p} z^h$. A simple induction provides the formula

$$\sigma_j = \sum_{p \leq y} \frac{Q_{j,p}(p^\beta)(\log p)^j}{R_p(p^\beta)^j}$$

where $Q_{j,p}$ is a polynomial of degree $\nu_p j - 1$ with coefficients $\ll \nu_p^j$. This immediately implies (2.20) in view of the first inequality in (2.12).

Recall that our hypotheses imply $\beta \leq \frac{2}{3} + O(1/\log y)$. To prove the complementary assertions, we observe that, for any fixed $j \geq 1$,

$$(2.22) \quad \begin{aligned} \sigma_j - \frac{d^{j-1} g_+}{d\sigma^{j-1}}(\beta, y) &= -\frac{d^{j-1} g_+}{d\sigma^{j-1}}(\beta, \sqrt{y}) + \sum_{p \leq \sqrt{y}} \frac{Q_{j,p}(p^\beta)(\log p)^j}{R_p(p^\beta)^j} \\ &\ll (\log y)^j \sum_{p \leq \sqrt{y}} \frac{1}{p^\beta} \ll (\log y)^{j-1} y^{(1-\beta)/2} \ll \sqrt{u}(\log y)^j. \end{aligned}$$

Since

$$\frac{d^2 g_+}{d\sigma^2}(\beta, y) = \sum_{p \leq y} \frac{(\log p)^3 p^\beta (p^\beta - 1)}{(1 + p^\beta)^3} \ll \min(1, \beta \log y) u(\log y)^3,$$

we obtain the statement regarding the case $j = 3$.

As for the case $j = 2$, we first note, on applying the prime number theorem as for the proof of (2.4), that the estimate

$$(2.23) \quad g'_+(\sigma, y) = \left\{ 1 + O\left(\frac{1}{\log y}\right) \right\} \frac{1}{(1 + y^{-\sigma})^2} \int_1^y \frac{\log t}{t^\sigma} dt$$

holds uniformly for $\sigma > 0$. Taking (2.4) and (2.15) into account and evaluating $1 + y^{-\beta}$ by (2.10), we get

$$g'_+(\beta, y) = e^{O(R)} \frac{w-1}{w} u(\log y)^2,$$

with $R := 1/\log 2u + 1/\log y \ll 1/\log y$. By (2.22), this estimate is equally valid for σ_2 : indeed $\sqrt{u} > y^{1/7}$ in the range under study. $\square$

We shall need the following estimate to control the decay of $|Z(s, y)|$ along the line $\sigma = \beta$. We write $s = \beta + i\tau$ with $\tau \in \mathbb{R}$ and set

$$Y_\varepsilon := e^{(\log y)^{3/2-\varepsilon}} \quad (y \geq 2).$$

Lemma 2.6. *Let $\varepsilon > 0$. For a suitable absolute constant $c > 0$, we have*

$$(2.24) \quad \left| \frac{Z(\beta + i\tau, y)}{Z(\beta, y)} \right| \leq \begin{cases} e^{-cu(\tau \log y)^4} & \text{if } |\tau| \leq 1/\log y, \\ e^{-cu\tau^4/(1+\tau^4)} & \text{if } 1/\log y < |\tau| \leq Y_\varepsilon. \end{cases}$$

Proof. For $s = \beta + i\tau$ ($\tau \in \mathbb{R}$), a standard computation yields

$$(2.25) \quad \left| \frac{Z(s, y)}{Z(\beta, y)} \right|^2 = \prod_{p \leq y} \frac{1 + 4 \sin^2(\frac{1}{2}\tau(\nu_p + 1) \log p) / \{p^{\beta(\nu_p + 1)}(1 - p^{-\beta(\nu_p + 1)})^2\}}{1 + 4 \sin^2(\frac{1}{2}\tau \log p) / \{p^\beta(1 - p^{-\beta})^2\}}.$$

Now, observe that

$$\nu^2 \leq \left(\sum_{0 \leq j < \nu} p^{j\beta} \right) \left(\sum_{0 \leq j < \nu} p^{-j\beta} \right) = p^{(\nu-1)\beta} \left(\frac{1 - p^{-\nu\beta}}{1 - p^{-\beta}} \right)^2 \quad (\nu \geq 1)$$

and, by lemma 1 of [10],

$$\left| \frac{\sin \nu \vartheta}{\nu \sin \vartheta} \right| \leq 1 - \frac{2}{3} \min(1, \nu^2 \|\vartheta/\pi\|^2) \leq 1 - \frac{2}{3} \|\vartheta/\pi\|^2 \quad (\vartheta \in \mathbb{R}, \nu \geq 1)$$

where $\|z\|$ denotes the distance from the real number z to the set of integers. Therefore, writing $\vartheta_p := \|(\tau/2\pi) \log p\|$ and $B_p := p^\beta(1 - p^{-\beta})^2$, we obtain that the generic factor in (2.25) does not exceed

$$\begin{aligned} \frac{1 + 4 \sin^2(\frac{1}{2}\tau(\nu_p + 1) \log p) / \{(\nu_p + 1)^2 B_p\}}{1 + 4 \sin^2(\pi \vartheta_p) / B_p} &\leq \frac{B_p + 4(1 - 2\vartheta_p^2/3)^2 \sin^2(\pi \vartheta_p)}{B_p + 4 \sin^2(\pi \vartheta_p)} \\ &\leq \frac{B_p + 4(1 - 2\vartheta_p^2/3) \sin^2(\pi \vartheta_p)}{B_p + 4 \sin^2(\pi \vartheta_p)} = 1 - \frac{8\vartheta_p^2 \sin^2(\pi \vartheta_p)}{3B_p + 12 \sin^2(\pi \vartheta_p)}, \end{aligned}$$

and so, appealing to the lower bound $|\sin(\pi \vartheta_p)| \geq 2\vartheta_p$, we arrive at

$$(2.26) \quad \left| \frac{Z(s, y)}{Z(\beta, y)} \right|^2 \leq e^{-W},$$

say, with

$$W := \frac{8}{3} \sum_{p \leq y} \frac{4\vartheta_p^4}{16\vartheta_p^2 + p^\beta(1 - p^{-\beta})^2} \gg \sum_{p \leq y} \frac{\|(\tau/2\pi) \log p\|^4}{p^\beta}.$$

If $|\tau| \leq 1/\log y$, we have

$$(2.27) \quad W \gg \tau^4 \sum_{p \leq y} \frac{(\log p)^4}{p^\beta} \gg \frac{\tau^4 (\log y)^3 (y^{1-\beta} - 1)}{1 - \beta} \gg (\tau \log y)^4 u,$$

where, for the last estimate, we used (2.9) if $y > (\log x)^3$ and (2.16) in the complementary case.

If $1/\log y < |\tau| \leq \sqrt{y}$, we appeal to upper bounds on primes in short intervals and argue as in lemma 5.12 of [8] to show that

$$(2.28) \quad \sum_{z/2 < p \leq z} \|(\tau/2\pi) \log p\|^4 \gg \frac{\tau^4}{1 + \tau^4} \frac{z}{\log z} \quad (y^{3/4} < z \leq y),$$

from which we infer by partial summation that $W \gg \tau^4 u / (1 + \tau^4)$, as required.

When $\sqrt{y} < |\tau| \leq Y_\varepsilon$, we note that the left-hand side of (2.28) is

$$\gg \sum_{z/2 < p \leq z} \sin^4\left(\frac{1}{2}\tau \log p\right) = \frac{3}{8} \sum_{z/2 < p \leq z} 1 - \frac{1}{2} \sum_{z/2 < p \leq z} \left\{ \cos(\tau \log p) - \frac{1}{4} \cos(2\tau \log p) \right\}.$$

However, classical bounds on exponential sums over primes⁽⁴⁾ yield that the last sum is $o(z/\log z)$ for $y^{3/4} < z \leq y$. We may then conclude by partial summation as before. $\square$

By a standard procedure, we deduce from the previous lemma an upper estimate for the number of ultrafriable integers in short intervals.

Lemma 2.7. *Let $\varepsilon > 0$. For a suitable absolute constant $c_0 > 0$, and uniformly for $x \geq y \geq 2$, $1 \leq z \leq \min(Y_\varepsilon, e^{c_0 u})$, we have*

$$(2.29) \quad \Upsilon(x + x/z, y) - \Upsilon(x, y) \ll x^\beta Z(\beta, y)/z.$$

Proof. From (2.24) and a classical bound for short sums of coefficients of Dirichlet series (see e.g. [11], Exercise 171) we infer that the left-hand side of (2.29) is

$$\begin{aligned} &\ll \frac{x^\beta}{z} \int_0^z |Z(\beta + i\tau, y)| d\tau \\ &\ll \frac{x^\beta Z(\beta, y)}{z} \left\{ \int_0^{1/\log y} e^{-cu(\tau \log y)^4} d\tau + \int_{1/\log y}^z e^{-c\tau^4 u/(1+\tau^4)} d\tau \right\} \\ &\ll \frac{x^\beta Z(\beta, y)}{z} \left\{ \frac{1}{u^{1/4}} + ze^{-cu/2} \right\}. \end{aligned}$$

$\square$

4. See, e.g., [11], equation (III.5.72).

3. Proofs

3.1. Proof of Theorem 1.1

Recall the definition (1.12) for the saddle-point $\alpha = \alpha(x, y)$ related to the distribution of friable integers.

First, let us assume $y > (\log x)^{2+\varepsilon}$ with, say, $\varepsilon \in]0, \frac{1}{3}[$. Then, for large x , we have $\alpha > \frac{1}{2} + \frac{1}{5}\varepsilon$ by (1.13). By theorem 2.4(i) of [3] the number of integers $n \leq x$ which are y -friable but not y -ultrafriable does not exceed

$$\sum_{p \leq y} \Psi\left(\frac{x}{p^{\nu_p+1}}, y\right) \ll \Psi(x, y)S$$

with

$$(3.1) \quad S := \sum_{p \leq y} \frac{1}{p^{(\nu_p+1)\alpha}} \ll \frac{y^{1/2-\alpha}}{\log y} \ll \frac{u \log 2u}{\sqrt{y} \log y}$$

where the first estimate follows from (2.6) and the last from (1.14). This proves (1.10).

Thus, it remains to prove the estimate (1.11) when

$$(3.2) \quad 2 \log x < \psi(y) \ll (\log x)^3.$$

In this range, we apply the saddle-point method in a very similar fashion to that of [5]. For purposes of convenience, we note at the outset that the expected main term has order of magnitude

$$(3.3) \quad \frac{x^\beta Z(\beta, y)}{1 + \beta\sqrt{\sigma_2}} \gg \frac{x^\beta Z(\beta, y)}{\sqrt{u} \log u}.$$

This follows from (1.9), (2.10) and (2.21).

First, we apply Perron's formula with remainder (see [11], th. II.2.3) to get

$$(3.4) \quad \Upsilon(x, y) = \frac{1}{2\pi i} \int_{\beta-iT}^{\beta+iT} \frac{x^s Z(s, y)}{s} ds + O\left(\frac{x^\beta Z(\beta, y) \log T}{T}\right)$$

with $T := e^{2c_1(\log u)^{4/3}} < \min(e^{c_0 u}, Y_{1/20})$ where c_1 is absolute, sufficiently small, and c_0 is the constant appearing in the statement of Lemma 2.7. This is proved in a standard way using (2.29) and we omit the details.

Set $T_0 := u^{-1/5}/\log y$. The contribution of the range $T_0 \leq |\tau| \leq T$ to the last integral may be bounded above using (2.24). We obtain that it is

$$\begin{aligned} &\ll x^\beta Z(\beta, y) \left\{ \int_{T_0}^{1/\log y} e^{-cu(\tau \log y)^4} \frac{d\tau}{\beta + \tau} + \int_{1/\log y}^T e^{-cu\tau^4/(1+\tau^4)} \frac{d\tau}{\tau} \right\} \\ &\ll x^\beta Z(\beta, y) \left\{ e^{-cu(T_0 \log y)^4} \log u + e^{-\frac{1}{2}cu/(\log y)^4} \log_2 y + e^{-u/2} \log T \right\} \\ &\ll x^\beta Z(\beta, y) e^{-c_2 u^{1/5}}. \end{aligned}$$

Therefore, under assumption (3.2), we get

$$(3.5) \quad \Upsilon(x, y) = \frac{1}{2\pi i} \int_{\beta-iT_0}^{\beta+iT_0} \frac{x^s Z(s, y)}{s} ds + O\left(x^\beta Z(\beta, y) e^{-c_1(\log u)^{4/3}}\right).$$

The last integral is classically evaluated by expanding the integrand around $s = \beta$. Let $T_1 := u^{-1/3}/\log y$. When $T_1 < |\tau| \leq T_0$, we have

$$Z(s, y)x^s = Z(\beta, y)x^\beta e^{-\tau^2\sigma_2/2 + i\tau^3\sigma_3/6 + \sigma_4\tau^4/24 + O(1)} \ll Z(\beta, y)x^\beta e^{-u^{1/3}/5}$$

by (2.20) and (2.21), since $\tau^5\varphi_5(\beta + i\tau, y) \ll T_0^5\sigma_5 \ll 1$ for $|\tau| \leq T_0$. Therefore, we may replace T_0 by T_1 in (3.5) without altering the error term.

We now evaluate the new main term.

Since $T_1^j\sigma_j \ll 1$ for $j = 3, 4$ and $\varphi_4(\beta + i\tau, y) \ll \sigma_4^* := u(\log y)^4$ for $|\tau| \leq 1/\log y$, we may write

$$\begin{aligned} \frac{1}{2\pi i} \int_{\beta - iT_1}^{\beta + iT_1} \frac{x^s Z(s, y)}{s} ds &= \frac{x^\beta Z(\beta, y)}{2\pi} \int_{-T_1}^{T_1} e^{-\tau^2\sigma_2/2 + i\tau^3\sigma_3/6 + O(\tau^4\sigma_4^*)} \frac{d\tau}{\beta + i\tau} \\ &= \frac{x^\beta Z(\beta, y)}{2\pi} \left\{ J_2 + \frac{1}{6}iJ_3 + O(K) \right\} \end{aligned}$$

with

$$\begin{aligned} J_2 &:= \int_{-T_1}^{T_1} e^{-\tau^2\sigma_2/2} \frac{d\tau}{\beta + i\tau}, \quad J_3 := \sigma_3 \int_{-T_1}^{T_1} \tau^3 e^{-\tau^2\sigma_2/2} \frac{d\tau}{\beta + i\tau}, \\ K &:= \int_{-T_1}^{T_1} e^{-\tau^2\sigma_2/2} \left\{ \tau^4\sigma_4^* + \tau^6\sigma_3^2 \right\} \frac{d\tau}{|\beta| + |\tau|}. \end{aligned}$$

The required formula now follows from computations identical to those appearing in the end of the proof of proposition 2.13 of [2], so we do not repeat the details here.

This completes the proof of (1.11).

3.2. Proof of Corollary 1.2

We start with establishing (1.18). When $y > (\log x)^3$, we have, for $\gamma = \alpha$ or $\gamma = \beta$,

$$(3.6) \quad \mathcal{H}_y(\gamma) = 1 + O\left(\frac{V_y(\gamma)}{\log y}\right) = 1 + O\left(\frac{u \log 2u}{\sqrt{y} \log y}\right),$$

by (2.6), (1.14), and (2.9). Thus (1.18) holds in this case—and actually so do (1.21) and (1.22).

We may therefore assume $y \leq (\log x)^3$ henceforth.

By (1.11), (1.14), (2.10), (2.4), (2.21), and the definition of α , we have for $2\log x < \psi(y) \ll (\log x)^3$,

$$(3.7) \quad \Upsilon(x, y) \asymp \frac{x^\beta Z(\beta, y)}{1 + \beta\sqrt{\sigma_2}} \gg \frac{\mathcal{H}_y(\beta)x^\alpha \zeta(\alpha, y)}{\alpha\sqrt{u} \log y} \asymp \mathcal{H}_y(\beta)\Psi(x, y)$$

where the last estimate readily follows from the saddle-point asymptotic formula for $\Psi(x, y)$ —see [5], th. 1—, viz.

$$(3.8) \quad \Psi(x, y) = \frac{x^\alpha \zeta(\alpha, y)}{\alpha\sqrt{2\pi g'_-(\alpha, y)}} \left\{ 1 + O\left(\frac{1}{u} + \frac{\log y}{y}\right) \right\} \quad (x \geq y \geq 2),$$

where

$$(3.9) \quad \zeta(s, y) := \prod_{p \leq y} \left(\frac{1}{1 - p^{-s}} \right) \quad (\Re s > 0).$$

This completes the proof of the lower bound in (1.18).

To prove the upper bound, we let ε be a small positive constant and note that, if $\psi(y) > (2 + \varepsilon) \log x$, we have $\alpha \asymp \beta$ by (1.13) and (2.10). We hence deduce from (1.11), (1.9), (2.4) and (2.21) that

$$\Upsilon(x, y) \asymp \frac{x^\beta Z(\beta, y)}{1 + \beta\sqrt{\sigma_2}} \ll \frac{x^\alpha Z(\alpha, y)}{\alpha\sqrt{u} \log y} \asymp \mathcal{H}_y(\alpha) \Psi(x, y),$$

where the first upper bound is an immediate consequence of the definition of β and the latter follows again from (3.8).

When $\psi(y) \leq (2 + \varepsilon) \log x$, the above argument is not sufficient since $\beta \log y$ may approach 0 while we always have $\alpha \gg 1/\log y$. However, by Taylor's formula to the second order and the definition of β , we have

$$x^\beta Z(\beta, y) \leq x^\alpha Z(\alpha, y) e^{-v(\alpha - \beta)^2}$$

where $v := \frac{1}{2} \varphi_2(\gamma, y)$ for some $\gamma \in]\beta, \alpha[$ —it indeed follows from (1.14) and (2.10) that $\beta < \alpha$ if ε is chosen sufficiently small. Now, we observe that, from lemma 13 of [5] and (1.13), we have $g'_+(\alpha, y) \asymp u(\log y)^2$. Since $g'_+(\sigma, y)$ is a non-increasing function of σ , we infer, taking (2.22) with $j = 2$ and (2.23) into account, that $\varphi_2(\gamma, y) \asymp u(\log y)^2$. It remains to use the fact that $\alpha - \beta \asymp \alpha$ in the range under consideration to obtain, for some suitable absolute constant $c > 0$,

$$(3.10) \quad \Upsilon(x, y) \ll x^\beta Z(\beta, y) \ll x^\alpha Z(\alpha, y) e^{-cu} \ll \mathcal{H}_y(\alpha) \Psi(x, y) e^{-cu/2}.$$

This completes the proof of (1.18).

We next turn our attention to (1.21) and (1.22), which contain the sufficiency part of assertions (1.19) and (1.20) respectively.

In view of (3.6), we may assume that $(\log x)^{3/2}/\sqrt{\log_2 x} < y \leq (\log x)^3$, and note that this is equivalent to $y \gg u^{3/2} \log 2u$ and implies $\log y \asymp \log u \asymp \log_2 x$. Since $\beta\sqrt{\sigma_2} \asymp \sqrt{u} \log y$ in the range under consideration, it follows from (1.9) that

$$(3.11) \quad G(\beta\sqrt{\sigma_2}) = \frac{1 + O(1/u)}{\beta\sqrt{2\pi\sigma_2}}.$$

From (2.22), we see that, in this last expression, we can replace σ_2 by $g'_+(\beta, y)$ at the cost of increasing the error to $\ll 1/\sqrt{u}$. Furthermore, we observe that, in view of (3.13), we have

$$\begin{aligned} g'_-(\beta, y) - g'_+(\beta, y) &\ll \sum_{p \leq y} \frac{(\log p)^2}{p^{2\beta}} \\ &\ll \frac{(y^{1-2\beta} - 1) \log y}{1 - 2\beta} \asymp \frac{\{y + (\log x)^2\} (\log y)^2}{y(1 + |\log r|)} \\ &\ll g'_-(\beta, y) \frac{y + (\log x)^2}{uy(1 + |\log r|)} \ll \frac{g'_-(\beta, y)}{\sqrt{u}}, \end{aligned}$$

where we used (2.19), (2.7) and (2.17). It follows that

$$(3.12) \quad \frac{x^\beta Z(\beta, y)}{\beta\sqrt{2\pi\sigma_2}} = \frac{\mathcal{H}_y(\beta) x^\beta \zeta(\beta, y)}{\beta\sqrt{2\pi g'_-(\beta, y)}} \left\{ 1 + O\left(\frac{1}{\sqrt{u}}\right) \right\}.$$

Replacing β by α in all terms except $\mathcal{H}_y(\beta)$ on the right-hand side yields an extra factor

$$F := \exp \left\{ O\left(|\beta - \alpha|D + (\beta - \alpha)^2 u (\log y)^2\right) \right\}$$

with $D := 1/\beta + g''(\beta, y)/g'_-(\beta, y) \ll \log y$. In view of (2.11), we obtain, writing $L := 1 + |\log r|$, that

$$F - 1 \ll \frac{\log 2u}{\sqrt{y}L} + \frac{u(\log u)^2}{yL} + \frac{u^3(\log u)^4}{y^2L^2} \ll \frac{1}{\sqrt{u}} + \frac{u^3(\log u)^2}{y^2} \asymp E.$$

By (3.8), we hence infer that (1.22) holds.

To prove (1.21), we assume $y \leq (\log x)^3$ and note that, for some $\vartheta \in [0, 1]$ and $\gamma = \beta + \vartheta(\alpha - \beta)$, we have

$$\mathcal{H}_y(\alpha) = \mathcal{H}_y(\beta)e^{(\alpha-\beta)V_y(\gamma)}.$$

By (2.11) and (2.18), which is equally valid for γ , we obtain, still using the notation $L := 1 + |\log r|$,

$$\log \left(\frac{\mathcal{H}_y(\alpha)}{\mathcal{H}_y(\beta)} \right) \asymp \frac{u(\log y)^2}{yL^2} + \frac{u^3(\log y)^4}{y^2L^2} \ll \frac{1}{\sqrt{u}} + \frac{(\log x)^3}{y^2 \log_2 x} \asymp E.$$

Thus, we have established (1.21).

It remains to prove that the asymptotic formulae in (1.19) and (1.20) do not hold when $y \ll (\log x)^{3/2}/\sqrt{\log_2 x}$. From (3.7), we have in this range

$$\Upsilon(x, y) \gg \mathcal{H}_y(\beta)\Psi(x, y)e^{(\alpha-\beta)^2 g'_-(\gamma, y)/2}$$

for some $\gamma := \alpha + \vartheta(\beta - \alpha)$ with $0 \leq \vartheta \leq 1$. However, estimates (2.17) and (2.11) yield in this case

$$(\alpha - \beta)^2 g'_-(\gamma, y) \asymp \frac{(\log x)^3}{y^2 \log_2 x} \gg 1.$$

Hence the asymptotic formula in (1.20) cannot hold.

We have already seen in (3.10) that (1.19) fails when $\psi(y) \leq (2 + \varepsilon)$ with ε sufficiently small. In the complementary range, the asymptotic formula (1.11) yields

$$\Upsilon(x, y) \asymp \frac{x^\alpha Z(\alpha, y)}{\alpha \sqrt{u} \log y} e^{-(\alpha-\beta)^2 \varphi_2(\gamma, y)/2} \asymp \mathcal{H}_y(\alpha)\Psi(x, y)e^{-(\alpha-\beta)^2 \varphi_2(\gamma, y)/2}$$

for some γ between α and β . But (2.17) and the last estimate of (2.18), which also holds for γ in place of β , yield $\varphi_2(\gamma, y) \asymp u(\log y)^2$. Hence we get as before

$$(\alpha - \beta)^2 \varphi_2(\gamma, y) \asymp \frac{(\log x)^3}{y^2 \log_2 x} \gg 1.$$

This completes the proof of assertion (1.19).

3.3. Proof of Corollary 1.3

We observe at the outset that the asymptotic formulae (1.24) and (1.25) readily follow, respectively, from (1.10) when $y > (\log x)^{2+\varepsilon}$, and from (1.18) when

$$\log x \ll y \leq (\log x)^2/L_\varepsilon(\log x)$$

with, say $0 < \varepsilon < \frac{1}{10}$. Indeed, for large x , we then have $\alpha \leq \frac{1}{2} - 1/(\log_2 x)^{2/5+2\varepsilon}$, and so

$$\mathcal{H}_y(\alpha) \ll \exp \left\{ - \sum_{\sqrt{y} \leq p \leq y} 1/p^{2\alpha} \right\} \ll \frac{1}{L_{3\varepsilon}(\log x)} = o(1).$$

Moreover, in view of (1.23), it is clear that, in the stated range of validity, (1.26) implies (1.24) and (1.25) since it may be easily checked that h and $\log \{y/(\log x)^2\}$ tend simultaneously to $\pm\infty$.

Thus it only remains to establish (1.26).

We note that the first formula in (1.26) readily follows from (1.22) since, in the considered range, we have $E \ll 1/\sqrt{u}$.

We next need an estimate for $\mathcal{H}_y(\beta)$. We plainly have $y \leq (\log x)^3$. We may hence deduce from (2.10) that, for a suitable constant $c > 0$, we have

$$(3.13) \quad \beta \geq \frac{1}{2} - \frac{c}{(\log y)^{2/5+\varepsilon}}.$$

It follows from (3.13) that

$$\prod_{p \leq \sqrt{y}} (1 - p^{-(\nu_p+1)\beta}) = \exp \left\{ O \left(\frac{y^{(1-3\beta)/2} - y^{(1-3\beta)/3}}{(1-3\beta) \log y} \right) \right\} = 1 + O \left(\frac{u \log u}{y^{2/3} \log y} \right).$$

Since $\log u \asymp \log y$, we get

$$(3.14) \quad \mathcal{H}_y(\beta) = \left\{ 1 + O \left(\frac{u}{y^{2/3}} \right) \right\} \prod_{\sqrt{y} < p \leq y} (1 - p^{-2\beta}).$$

We now apply (2.9) with $\varepsilon/2$ instead of ε and consequently replace β by $1 - \xi(u)/\log y = \frac{1}{2} + h/\log y$ in the latter product. This involves an extra factor

$$\exp \left\{ O \left(\sum_{\sqrt{y} < p \leq y} \frac{\log p}{p^{2\beta} L_{\varepsilon/2}(y)} \right) \right\} = 1 + O \left(\frac{1}{L_{\varepsilon}(y)} \right).$$

Now, we have by lemma III.5.16 of [11]

$$\prod_{\sqrt{y} < p \leq y} (1 - p^{-1-2h/\log y}) = \frac{\widehat{\varrho}(h)}{2\widehat{\varrho}(2h)} \left\{ 1 + O \left(\frac{1}{L_{\varepsilon}(y)} \right) \right\} \quad (y \geq 2).$$

Gathering our estimates so far, we obtain

$$(3.15) \quad \mathcal{H}_y(\beta) = \frac{\widehat{\varrho}(h)}{2\widehat{\varrho}(2h)} \left\{ 1 + O \left(\frac{1}{L_{\varepsilon}(y)} \right) \right\} \quad (y \geq 2).$$

Inserting this into (1.22), we obtain the second formula in (1.26), as required.

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