

Moments of Gamma type and the Brownian supremum process area*

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Abstract: We study positive random variables whose moments can be expressed by products and quotients of Gamma functions; this includes many standard distributions. General results are given on existence, series expansion and asymptotics of density functions. It is shown that the integral of the supremum process of Brownian motion has moments of this type, as well as a related random variable occurring in the study of hashing with linear displacement, and the general results are applied to these variables.

AMS 2000 subject classifications: Primary 60E10; secondary 60J15.

Keywords and phrases: Moments, Gamma function, Brownian motion, supremum process, generalized Pólya urns.

Received March 2010.

1. Introduction

We say that a positive random variable X has *moments of Gamma type* if, for s in some interval,

$$\mathbb{E} X^s = CD^s \frac{\prod_{j=1}^J \Gamma(a_j s + b_j)}{\prod_{k=1}^K \Gamma(a'_k s + b'_k)} \quad (1.1)$$

for some integers $J, K \geq 0$ and some real constants $C, D > 0, a_j, b_j, a'_k, b'_k$. We may and will assume that $a_j \neq 0$ and $a'_k \neq 0$ for all j and k . We often denote the right hand side of (1.1) by $F(s)$; this is a meromorphic function defined for all complex s (except at its poles).

Similarly we say that a real random variable Y has *moment generating function of Gamma type* if, for s in some interval,

$$\mathbb{E} e^{sY} = C e^{ds} \frac{\prod_{j=1}^J \Gamma(a_j s + b_j)}{\prod_{k=1}^K \Gamma(a'_k s + b'_k)} \quad (1.2)$$

for some integers $J, K \geq 0$ and some real constants $C, d, a_j \neq 0, b_j, a'_k \neq 0, b'_k$, and that Y has *characteristic function of Gamma type* if, for all real t ,

$$\mathbb{E} e^{itY} = C e^{itd} \frac{\prod_{j=1}^J \Gamma(b_j + ia_j t)}{\prod_{k=1}^K \Gamma(b'_k + ia'_k t)} \quad (1.3)$$

for some such constants.

*This is an original survey paper.

Of course, (1.1) and (1.2) are the same if $X = e^Y$ and $D = e^d$; further, it will be shown that (1.2) and (1.3) are equivalent by analytic continuation. Moreover, we shall see that the range of validity of (1.1), or (1.2) is always the largest possible. We amplify these simple but useful observations in Theorem 2.1 below.

Remark 1.1. The representations in (1.1)–(1.3) are far from unique. Using the duplication formula (A.3), or more generally the multiplication formula (A.5), and other relations such as the functional equation $\Gamma(z+1) = z\Gamma(z)$, a function $F(s)$ of this form may be rewritten in many different ways. (The rewriting can be more or less transparent; some equivalent versions may look quite different to the unaided eye.) See for example Theorems 1.6 and 1.12.

Remark 1.2. The constant C is determined by the relation $F(0) = \mathbb{E} X^0 = 1$, which shows that

$$C = \frac{\prod_{k=1}^K \Gamma(b'_k)}{\prod_{j=1}^J \Gamma(b_j)} \quad (1.4)$$

provided no b_j or b'_k is a non-positive integer. In general, C can be found by taking limits as $s \rightarrow 0$.

Remark 1.3. The constant D is just a scale factor: X satisfies (1.1) if and only if X/D satisfies the same equation with D replaced by 1 (i.e., without the factor D^s). Similarly, Y satisfies (1.2) or (1.3) if and only if $Y - d$ satisfies the same equation with d replaced by 0. Hence we might assume $D = 1$ or $d = 0$ if convenient (but we will not do so in general).

Remark 1.4. If $r \in \mathbb{R}$, then $x - r = \Gamma(x - r + 1)/\Gamma(x - r)$. Hence, any rational function $Q(x)$ with all poles and zeros real can be written as a finite product $\prod_{\ell} \Gamma(x + c_{\ell})/\Gamma(x + c'_{\ell})$ with $c_{\ell}, c'_{\ell} \in \mathbb{R}$. Consequently we may allow such a rational factor $Q(s)$ in (1.1) and (1.2), or $Q(it)$ in (1.3), without changing the class of distributions.

Remark 1.5. If X has moments of Gamma type and α is a real number, then X^{α} has moments of Gamma type. (Just substitute αs for s in (1.1).) Similarly, if X_1 and X_2 are independent and both have moments of Gamma type, then $X_1 X_2$ has too. (Just use $\mathbb{E}(X_1 X_2)^s = \mathbb{E} X_1^s \mathbb{E} X_2^s$.)

Several well-known distributions have moments or moment generating functions of Gamma type. We give a number of examples in Section 3.

The main motivation for the present paper is that also several less well-known distributions have moments of Gamma type. It is then straightforward to use Mellin transform techniques to obtain expansions or asymptotics of the density function, and it seems advantageous to do so, and to study other properties, in general for this class of distributions. We give some basic results in Section 2, and further results on poles and zeros in Section 4. Asymptotics of the moments are studied in Section 5, and asymptotics and series expansions of the density function are given in Section 6.

This paper thus was inspired by the realization that some recently studied random variables have moments of Gamma type; we treat these in Sections 7–9. The first example is the integral of the supremum process of a Brownian motion, i.e., the area under the supremum process (up to some fixed time T). Let $B(t)$, $t \geq 0$, be a standard Brownian motion. Consider the supremum process $S(t) := \max_{0 \leq s \leq t} B(s)$, and its integral

$$\mathcal{A}(T) := \int_0^T S(t) dt. \quad (1.5)$$

We further let $\mathcal{A} := \mathcal{A}(1)$. For any given $T > 0$, the usual Brownian scaling $\{B(Tt)\}_{t \geq 0} \stackrel{d}{=} \{T^{1/2}B(t)\}_{t \geq 0}$ implies the corresponding scaling for the supremum process $\{S(Tt)\}_{t \geq 0} \stackrel{d}{=} \{T^{1/2}S(t)\}_{t \geq 0}$, and thus

$$\mathcal{A}(T) = T \int_0^1 S(Tt) dt \stackrel{d}{=} T^{3/2} \mathcal{A}. \quad (1.6)$$

In particular, $\mathbb{E} \mathcal{A}(T)^s = T^{3s/2} \mathbb{E} \mathcal{A}^s$ and it is enough to study \mathcal{A} .

The random area \mathcal{A} was studied by Janson and Petersson [19], and using their results we will in Section 7 prove the following formula, showing that \mathcal{A} has moments of Gamma type. (The result for the integer moments $\mathbb{E} \mathcal{A}^n$, $n \in \mathbb{N}$, was given in [19].) We give several different, but equivalent, formulas of the type (1.1) for $\mathbb{E} \mathcal{A}^s$, which exemplifies Remark 1.1. The third version, with only two non-constant Gamma factors is perhaps the simplest. The last, where all Gamma factors are of the type $\Gamma(s/2 + b)$ with $0 < b \leq 1$ is of a canonical type where there are no cancellations of poles, and it is thus easy to see the poles and zeros, cf. Remark 4.5.

Theorem 1.6. *The moments of \mathcal{A} are given by, for $\operatorname{Re} s > -1$,*

$$\begin{aligned} \mathbb{E} \mathcal{A}^s &= \frac{\Gamma(s+1) \Gamma(s+2/3)}{\Gamma(2/3) \Gamma(3s/2+1)} \cdot \left(\frac{3}{\sqrt{8}}\right)^s \\ &= \frac{\Gamma(s+1) \Gamma(s+5/3)}{\Gamma(5/3) \Gamma(3s/2+2)} \cdot \left(\frac{3}{\sqrt{8}}\right)^s \\ &= \frac{2\Gamma(1/3)}{3\sqrt{\pi}} \cdot \frac{\Gamma(3s/2+3/2)}{\Gamma(s+4/3)} \cdot \left(\frac{\sqrt{8}}{9}\right)^s \\ &= \frac{\Gamma(1/3)}{2^{1/3}\pi} \cdot \frac{\Gamma(s/2+1/2) \Gamma(s/2+5/6)}{\Gamma(s/2+2/3)} \cdot \left(\frac{2}{3}\right)^{s/2}. \end{aligned}$$

Further, $\mathbb{E} \mathcal{A}^s = \infty$ for real $s \leq -1$.

Remark 1.7. Several related Brownian areas are studied in Janson [17], for example the integral of $|B(t)|$ or the integral of a normalized Brownian excursion. These areas do not have moments of Gamma type. In fact, most of the Brownian areas studied there have entire functions $\mathbb{E} X^s$ [17, §29], which is impossible for moments of Gamma type, see Theorem 4.1(iv). (For the remaining two areas in

[17], we have no formal proof that they do not have moments of Gamma type, but it seems very unlikely since the integer moments satisfy more complicated recursion formulas [17].)

As a consequence of Theorem 1.6 and our general results in Section 6, we can express the density function of \mathcal{A} using the confluent hypergeometric function ${}_1F_1$ (denoted M in [1] and Φ in [20]) or the confluent hypergeometric function of the second kind U [1] (also denoted Ψ [20]). (Also the proofs of the next two theorems are given in Section 7.)

Theorem 1.8. \mathcal{A} has a density function $f_{\mathcal{A}}(x)$ given by, for $x > 0$,

$$\begin{aligned} f_{\mathcal{A}}(x) &= \frac{2^{2/3}\Gamma(1/3)}{3^{1/2}\pi} \sum_{n=0}^{\infty} (-1)^n \frac{\Gamma(n+5/6)}{n! \Gamma(n+2/3)} \cdot \left(\frac{3}{2}\right)^{n+1/2} x^{2n} \\ &\quad + \frac{2^{2/3}\Gamma(1/3)}{3^{1/2}\pi} \sum_{n=0}^{\infty} (-1)^n \frac{\Gamma(n+7/6)}{n! \Gamma(n+4/3)} \cdot \left(\frac{3}{2}\right)^{n+5/6} x^{2n+2/3} \\ &= \frac{2^{1/2}}{\pi^{1/2}} {}_1F_1\left(\frac{5}{6}; \frac{2}{3}; -\frac{3}{2}x^2\right) + \frac{2^{-1/6}3^{1/3}}{\Gamma(5/6)} x^{2/3} {}_1F_1\left(\frac{7}{6}; \frac{4}{3}; -\frac{3}{2}x^2\right) \\ &= e^{-\frac{3}{2}x^2} \left(\frac{2^{1/2}}{\pi^{1/2}} {}_1F_1\left(-\frac{1}{6}; \frac{2}{3}; \frac{3}{2}x^2\right) + \frac{2^{-1/6}3^{1/3}}{\Gamma(5/6)} x^{2/3} {}_1F_1\left(\frac{1}{6}; \frac{4}{3}; \frac{3}{2}x^2\right) \right) \\ &= \frac{2^{7/6}}{\Gamma(2/3)} e^{-\frac{3}{2}x^2} U\left(-\frac{1}{6}; \frac{2}{3}; \frac{3}{2}x^2\right) \\ &= \frac{2^{5/6}3^{1/3}}{\Gamma(2/3)} x^{2/3} e^{-\frac{3}{2}x^2} U\left(\frac{1}{6}; \frac{4}{3}; \frac{3}{2}x^2\right). \end{aligned}$$

It follows (most easily from the second formula above) that $f_{\mathcal{A}}$ has a finite, positive limit $f_{\mathcal{A}}(0+) = \sqrt{2/\pi}$ as $x \searrow 0$. More precisely, $f_{\mathcal{A}}(x) = \sqrt{2/\pi} + O(x^{2/3})$.

As $x \rightarrow \infty$, we obtain from Theorem 1.6 and our general theorems in Section 6 the following asymptotic result. Note that the two terms in the first or second formula for $f_{\mathcal{A}}$ in Theorem 1.8 are each much larger, of the order $x^{-5/3}$ by the asymptotics of ${}_1F_1$ in [1, (13.5.1)], but they cancel each other almost completely for large x .

Theorem 1.9.

$$f_{\mathcal{A}}(x) \sim \frac{3^{2/3}\Gamma(1/3)}{\pi} x^{1/3} e^{-3x^2/2} = \frac{2 \cdot 3^{1/6}}{\Gamma(2/3)} x^{1/3} e^{-3x^2/2}, \quad x \rightarrow \infty.$$

This result was conjectured in [19], where the weaker result $\mathbb{P}(\mathcal{A} > x) = \exp\{-3x^2/2 + o(x^2)\}$ was shown from the moment asymptotic

$$\mathbb{E} \mathcal{A}^s \sim \frac{\Gamma(1/3)}{\pi^{1/2}} s^{1/6} \left(\frac{s}{3e}\right)^{s/2}, \quad s \rightarrow \infty, \quad (1.7)$$

for integer s and a Tauberian theorem. (Only integer moments were considered in [19]. Note that (1.7) for arbitrary real $s \rightarrow \infty$ follows easily from Theorem 1.6 and Stirling's formula; see Theorem 5.7 and (7.8).)

Remark 1.10. Theorem 1.9 also follows from any of the last two formulas in Theorem 1.8 and the asymptotic formula for U in [1, (13.5.2)]. Indeed, this gives an asymptotic expansion with further terms, cf. Remark 6.3; in this case, by [1, (13.5.2)], the complete asymptotic expansion can be written

$$f_{\mathcal{A}}(x) \sim \frac{2 \cdot 3^{1/6}}{\Gamma(2/3)} x^{1/3} e^{-3x^2/2} {}_2F_0\left(\frac{1}{6}, -\frac{1}{6}; ; -\frac{2}{3}x^{-2}\right), \quad x \rightarrow \infty,$$

where the hypergeometric series ${}_2F_0$ is divergent and the asymptotic expansion is interpreted in the usual way: if we truncate the series after any fixed number of terms, the error is of the order of the first omitted term. (For the general definition of the (generalized) hypergeometric series ${}_pF_q$, see e.g. [13, Section 5.5].)

Theorem 1.9 may be compared with similar results for several other Brownian areas in Janson and Louchard [18], see also Janson [17] and Remark 1.7. In these results for other Brownian areas, the exponent of x is always an integer (0, 1 or 2), while here the exponent is $1/3$, which is related to the power $s^{1/6}$ in (1.7).

Another example with moments of Gamma type comes from Petersson [24]. He studied the maximum displacement in hashing with linear probing, and found for dense tables, after suitable normalization, convergence to a limit distribution given by a random variable \mathcal{M} with the distribution [24, Theorem 5.1]

$$\mathbb{P}(\mathcal{M} \leq x) = 1 - \psi(x^{3/2}), \quad x > 0, \quad (1.8)$$

where $\psi(s) := \mathbb{E} e^{-s\mathcal{A}}$ is the Laplace transform of \mathcal{A} . Equivalently,

$$\mathbb{P}(\mathcal{M} > x) = \psi(x^{3/2}) = \mathbb{E} e^{-x^{3/2}\mathcal{A}}, \quad x > 0. \quad (1.9)$$

This type of relation preserves moments of Gamma type; we give a general result.

Lemma 1.11. *Suppose that V and Z are two positive random variables and $\alpha > 0$. Then*

$$\mathbb{P}(V > x) = \mathbb{E} e^{-x^\alpha Z}, \quad x > 0, \quad (1.10)$$

if and only if

$$V \stackrel{d}{=} T^{1/\alpha} / Z^{1/\alpha}, \quad (1.11)$$

where $T \sim \text{Exp}(1)$ is independent of Z .

If (1.10) or (1.11) holds, then

$$\mathbb{E} V^s = \Gamma(s/\alpha + 1) \mathbb{E} Z^{-s/\alpha}, \quad s > -\alpha. \quad (1.12)$$

In particular, if one of Z and V has moments of Gamma type, then so has the other.

We postpone the simple proof until Section 8. By (1.9), Lemma 1.11 applies to \mathcal{M} and \mathcal{A} , and thus \mathcal{M} has moments of Gamma type. More precisely, Theorem 1.6 implies the following, see Section 8 for details.

Theorem 1.12. For $-3/2 < \operatorname{Re} s < 3/2$,

$$\begin{aligned} \mathbb{E} \mathcal{M}^s &= \frac{\Gamma(1+2s/3)\Gamma(2/3-2s/3)\Gamma(1-2s/3)}{\Gamma(2/3)\Gamma(1-s)} \cdot \left(\frac{2}{3^{2/3}}\right)^s \\ &= \frac{2\Gamma(1/3)}{3\sqrt{\pi}} \cdot \frac{\Gamma(1+2s/3)\Gamma(3/2-s)}{\Gamma(4/3-2s/3)} \cdot \left(\frac{3^{4/3}}{2}\right)^s \\ &= \frac{\Gamma(1/3)}{2^{1/3}\pi} \cdot \frac{\Gamma(1+2s/3)\Gamma(1/2-s/3)\Gamma(5/6-s/3)}{\Gamma(2/3-s/3)} \cdot \left(\frac{3}{2}\right)^{s/3} \\ &= \frac{\Gamma(1/3)}{2^{1/3}\pi^{3/2}} \cdot \frac{\Gamma(1/2+s/3)\Gamma(1+s/3)\Gamma(1/2-s/3)\Gamma(5/6-s/3)}{\Gamma(2/3-s/3)} \cdot 6^{s/3}. \end{aligned}$$

Further, $\mathbb{E} \mathcal{M}^s = \infty$ for real $s \leq -3/2$ or $s \geq 3/2$.

The special case $s = 1$ yields $\mathbb{E} \mathcal{M} = 2\Gamma(1/3)/3^{2/3}$, as found by Petersson [24]. Petersson [24] further proved that $\mathbb{E} \mathcal{M}^s = \infty$ for $s \geq 2$; we now see that the sharp threshold is $s = 3/2$.

Our general theorems apply again; they show that \mathcal{M} has a density, and they yield a series expansion and asymptotics for the density. (Proofs are given in Section 8.) Again, the results can be expressed using various hypergeometric functions and series. (Again, see [13] for definitions.)

Theorem 1.13. \mathcal{M} has a continuous density function given by, for $x > 0$,

$$\begin{aligned} f_{\mathcal{M}}(x) &= \frac{3^{1/2}\Gamma(1/3)}{2^{5/6}\pi} \sum_{n=0}^{\infty} (-1)^n \frac{\Gamma(1+n/2)\Gamma(4/3+n/2)}{\Gamma(7/6+n/2)n!} \left(\frac{2}{3}\right)^{n/2} x^{3n/2+1/2} \\ &= \frac{2^{1/2}}{\pi^{1/2}} x^{1/2} {}_2F_2\left(\frac{4}{3}, 1; \frac{7}{6}, \frac{1}{2}; \frac{x^3}{6}\right) - \frac{5}{8} x^2 {}_1F_1\left(\frac{11}{6}; \frac{5}{3}; \frac{x^3}{6}\right). \end{aligned}$$

In particular, for small x we have the asymptotic formula

$$f_{\mathcal{M}}(x) = \frac{2^{1/2}}{\pi^{1/2}} x^{1/2} + O(x^2), \quad x \searrow 0. \quad (1.13)$$

For large x , there is a similar formula, which is the beginning of a divergent asymptotic expansion (interpreted as in Remark 1.10):

Theorem 1.14. As $x \rightarrow \infty$,

$$f_{\mathcal{M}}(x) = \frac{3}{\sqrt{2\pi}} x^{-5/2} + O(x^{-7/2}). \quad (1.14)$$

More precisely, $f_{\mathcal{M}}(x)$ has as $x \rightarrow \infty$ an asymptotic expansion

$$\begin{aligned} f_{\mathcal{M}}(x) &\sim \frac{3}{\sqrt{2\pi}} x^{-5/2} {}_3F_1\left(\frac{3}{2}, \frac{5}{6}, 1; \frac{2}{3}; -\frac{6}{x^3}\right) \\ &\quad + \frac{5\Gamma(1/3)}{2^{1/2}3^{2/3}\pi^{1/2}} x^{-7/2} {}_2F_0\left(\frac{11}{6}, \frac{7}{6}; -\frac{6}{x^3}\right). \end{aligned} \quad (1.15)$$

Yet other recent examples of moments of Gamma type come from the study of generalized Pólya urns [10], [16]; see Section 9 for definitions and results.

As said above, we give some basic results in Section 2, many examples with standard distributions in Section 3, further results on poles and zeros in Section 4, asymptotics of the moments in Section 5, and asymptotics and series expansions of the density function in Section 6. We give proofs of the results above for \mathcal{A} and \mathcal{M} in Sections 7 and 8, and we give some results for generalized Pólya urns in Section 9. We end with a couple of more technical examples (counter examples) in Section 10 and some further remarks in Section 11. Some standard formulas for the Gamma function are for convenience collected in Appendix A.

2. The basic theorem and some notation

Let $F(s)$ denote the right hand side of (1.1) or (1.2). (Thus, the right hand side of (1.3) is $F(it)$.) Evidently, F is a meromorphic function in the complex plane, and all poles are on the real axis. Let ρ_+ and ρ_- be the poles closest to 0:

$$\begin{aligned}\rho_+ &:= \min\{x > 0 : x \text{ is a pole of } F\}, \\ \rho_- &:= \max\{x < 0 : x \text{ is a pole of } F\},\end{aligned}\tag{2.1}$$

with the interpretation that $\rho_+ = \infty$ [$\rho_- = -\infty$] if there is no pole on $(0, \infty)$ [$(-\infty, 0)$]. Thus $-\infty \leq \rho_- < 0 < \rho_+ \leq \infty$. Note that we ignore any pole at 0 in the definitions (2.1); however, it follows from Theorem 2.1 that such a pole cannot exist; $F(s)$ is always analytic at $s = 0$.

Theorem 2.1. *Let $X > 0$ and Y be random variables connected by $X = e^Y$ and thus $Y = \log X$, and let $C, D > 0$, $d = \log D$, $a_j \neq 0$, $b_j, a'_k \neq 0$, b'_k be real constants, for $j = 1, \dots, J \geq 0$ and $k = 1, \dots, K \geq 0$. Let $F(s)$ be the meromorphic function in (1.1) and (1.2) and let $\rho_+ \in (0, \infty)$ and $\rho_- \in (-\infty, 0)$ be defined by (2.1). Then the following are equivalent:*

- (i) (1.1) holds for all real s in some non-empty interval.
- (ii) (1.1) holds for all complex s in the strip $\rho_- < \operatorname{Re} s < \rho_+$.
- (iii) (1.2) holds for all real s in some non-empty interval.
- (iv) (1.2) holds for all complex s in the strip $\rho_- < \operatorname{Re} s < \rho_+$.
- (v) (1.3) holds for all real $t \neq 0$ in some interval $|t| < t_0$ with $t_0 > 0$.
- (vi) (1.3) holds for all real t .

In this case, further $\mathbb{E} X^s = \mathbb{E} e^{sY} = \infty$ if $s \leq \rho_-$ or $s \geq \rho_+$; thus

$$\{s \in \mathbb{R} : \mathbb{E} X^s < \infty\} = \{s \in \mathbb{R} : \mathbb{E} e^{sY} < \infty\} = (\rho_-, \rho_+).$$

Equivalently,

$$\begin{aligned}\rho_+ &= \sup\{s \geq 0 : \mathbb{E} X^s < \infty\}, \\ \rho_- &= \inf\{s \leq 0 : \mathbb{E} X^s < \infty\}.\end{aligned}$$

Furthermore, $F(s) = \mathbb{E} X^s = \mathbb{E} e^{sY} \neq 0$ when $\rho_- < \operatorname{Re} s < \rho_+$.

Proof. (i) \iff (iii) and (ii) \iff (iv) are trivial, since $\mathbb{E} X^s = \mathbb{E} e^{sY}$. Further, trivially (iv) \implies (iii), (vi) \implies (v) and (iv) \implies (vi). Hence, to show the equivalences it suffices to show that (iii) \implies (v) and (v) \implies (iv).

(v) \implies (iv). Note first that $\varphi(t) := \mathbb{E} e^{itY} \rightarrow 1$ as $t \rightarrow 0$. Hence, $F(it) \rightarrow 1$ as $t \searrow 0$, and thus F does not have a pole at 0.

This shows that $F(z)$ is analytic in the strip $\rho_- < \operatorname{Re} z < \rho_+$, and thus $F(iz)$ is analytic in the strip $-\rho_+ < \operatorname{Im} z < -\rho_-$. By continuity, $\varphi(t) = F(it)$ also for $t = 0$ and thus for the entire interval $(-t_0, t_0)$. Hence, on this interval at least, $\varphi(t)$ equals the boundary values of the function $F(it)$ which is analytic for $0 \leq \operatorname{Im} z < -\rho_-$ and by a theorem of Marcinkiewicz [22], $\mathbb{E} e^{-rY} < \infty$ for every $r \in (0, -\rho_-)$; equivalently, $\mathbb{E} e^{rY} < \infty$ if $\rho_- < r < 0$. By considering $-Y$, we find similarly that $\mathbb{E} e^{rY} < \infty$ if $0 < r < \rho_+$. Consequently,

$$\mathbb{E} e^{rY} < \infty \quad \text{if } \rho_- < r < \rho_+. \quad (2.2)$$

It is well-known that (2.2) implies that $\psi(z) := \mathbb{E} e^{zY}$ is defined and finite for $\rho_- < \operatorname{Re} z < \rho_+$ and that $\psi(z)$ is an analytic function of z in this strip. Since $\psi(it) = \varphi(t) = F(it)$ for $|t| < t_0$, analytic continuation yields $\psi(z) = F(z)$ in this strip, i.e. (iv) holds.

(iii) \implies (v). Suppose that $\mathbb{E} e^{sY} = F(s)$ for $s \in (a, b)$, with $-\infty < a < b < \infty$. Let $s_0 \in (a, b)$ with $s_0 \neq 0$ and suppose that $s_0 > 0$. (The case $s_0 < 0$ is similar, or follows by considering $-Y$.) Thus $\mathbb{E} e^{s_0 Y} = F(s_0) < \infty$, and it follows that $z \mapsto \mathbb{E} e^{zY}$ is defined and analytic for $0 < \operatorname{Re} z < s_0$. Since $\mathbb{E} e^{zY} = F(z)$ on an interval in this strip, $\mathbb{E} e^{zY} = F(z)$ for $0 < \operatorname{Re} z < s_0$.

For any real t , we may take $z = it + \varepsilon$ for $0 < \varepsilon < s_0$; letting $\varepsilon \searrow 0$ we have $\mathbb{E} e^{(it+\varepsilon)Y} \rightarrow \mathbb{E} e^{itY}$ by dominated convergence (using $\mathbb{E}(1 + e^{s_0 Y}) < \infty$). If further $t \neq 0$, then also $\mathbb{E} e^{(it+\varepsilon)Y} = F(it + \varepsilon) \rightarrow F(it)$ since F has only real poles, and thus $\mathbb{E} e^{itY} = F(it)$. Hence (v) holds.

This completes the proof of the equivalences. Suppose $\mathbb{E} e^{sY} < \infty$ for some $s \geq \rho_+$ (and thus $\rho_+ < \infty$). Letting $z \nearrow \rho_+$, we then have, by dominated convergence, $F(z) = \mathbb{E} e^{zY} \rightarrow \mathbb{E} e^{\rho_+ Y} < \infty$, while the definition of ρ_+ as a pole yields $F(z) = |F(z)| \rightarrow \infty$. This contradiction shows that $\mathbb{E} e^{sY} = \infty$ for $s \geq \rho_+$. Similarly, or by considering $-Y$, $\mathbb{E} e^{sY} = \infty$ for $s \leq \rho_-$.

Finally observe that $F(s) = 0$ only when some $a'_k s + b'_k$ is a pole of Γ , i.e., a non-positive integer, which implies that s is real. However, if $\rho_- < s < \rho_+$, then $F(s) = \mathbb{E} e^{sY} > 0$. \square

Remark 2.2. The equivalence (iii) \iff (iv) (or, equivalently, (i) \iff (ii)) is an instance of the well-known fact that a (two-sided) Laplace transform of a positive function or measure has singularities where the real axis intersects the boundary of the natural strip of definition, see e.g. [6, §3.4]. The result by Marcinkiewicz [22] used above is a sharper version of this.

We make some simple but useful observations.

Corollary 2.3. *The distribution of X is determined by the function $F(s)$ on the right hand side of (1.1): If $\mathbb{E} X_1^s = F(s)$ for $s \in I_1$ and $\mathbb{E} X_2^s = F(s)$ for $s \in I_2$, for non-empty intervals I_1 and I_2 , then $X_1 \stackrel{d}{=} X_2$.*

Proof. By Theorem 2.1, $Y_1 := \log X_1$ and $Y_2 := \log X_2$ have the same characteristic function $F(it)$. Hence, $Y_1 \stackrel{d}{=} Y_2$ and $X_1 \stackrel{d}{=} X_2$. \square

Remark 2.4. As said above, Theorem 2.1 (e.g., by (vi)) implies that 0 is not a pole of F ; furthermore, $F(0) = \mathbb{E} X^0 = 1$.

Remark 2.5. Every pole or zero of $F(s)$ must be a pole of one of the Γ factors in (1.1). However, the converse does not hold, since poles in the Γ factors may cancel; if s_0 is a pole of some factors in the numerator, but also a pole of at least as many factors in the denominator, then s_0 is a removable singularity of F , and $F(s_0)$ is well-defined by continuity. Note that such s_0 do not count in the definition (2.1) of ρ_{\pm} .

In particular, $s = 0$ may be a pole of some of the Gamma factors in $F(s)$. (This happens when some b_j or b'_k is 0 or a negative integer. This is the reason we exclude $t = 0$ in Theorem 2.1(v).) However, by Remark 2.4, all such poles must cancel; i.e., there must be an equal number of such factors in the numerator and denominator in (1.1).

Remark 2.6. In Theorem 2.1(v), it is important that we consider an interval about 0 (unlike in (i) and (iii)). In fact, for any $\varepsilon > 0$, there exist a random variable Y and $C, d, a_j, b_j, a'_k, b'_k$ such that (1.3) holds for $|t| \geq \varepsilon$ but not for all ε .

For an example, let Y be any random variable with characteristic function of Gamma type, say $\mathbb{E} e^{itY} = F(it)$. Further, let Z be a random variable with characteristic function $(1 - |t|)_+$, see [9, Section XV.2], and let W be the mixture of Z and the constant 0 obtained as $W := VZ$ with $V \sim \text{Be}(1/2)$, i.e. $\mathbb{P}(V = 0) = \mathbb{P}(V = 1) = 1/2$, and V independent of Z ; assume further that Y is independent of Z and V . Then, for $|t| \geq 1$, the characteristic functions $\mathbb{E} e^{itZ} = 0$ and $\mathbb{E} e^{itW} = \frac{1}{2}(\mathbb{E} e^{itZ} + 1) = \frac{1}{2}$, and thus $\tilde{Y} := Y + \varepsilon^{-1}W$ has the characteristic function

$$\mathbb{E} e^{it\tilde{Y}} = \mathbb{E} e^{itY} \mathbb{E} e^{i\varepsilon^{-1}tW} = \frac{1}{2}F(it), \quad \text{when } |t| \geq \varepsilon. \quad (2.3)$$

Here $\tilde{F}(it) := \frac{1}{2}F(it)$ is another function of the type in (1.3); however (2.3) does not hold for all t since $\tilde{F}(0) = 1/2 \neq 1$.

We do not know whether (1.3) for some interval, together with $F(0) = 1$, implies that (1.3) holds for all t .

For future use, in particular for the asymptotic results in Sections 5 and 6, we define the following parameters, given a random variable X or Y or a function $F(s)$ as in (1.1) or (1.2):

$$\gamma := \sum_{j=1}^J |a_j| - \sum_{k=1}^K |a'_k|, \quad (2.4)$$

$$\gamma' := \sum_{j=1}^J a_j - \sum_{k=1}^K a'_k, \quad (2.5)$$

$$\delta := \sum_{j=1}^J b_j - \sum_{k=1}^K b'_k - \frac{J-K}{2}, \quad (2.6)$$

$$\varkappa := \sum_{j=1}^J a_j \log |a_j| - \sum_{k=1}^K a'_k \log |a'_k| + d, \quad (2.7)$$

$$C_1 := |C|(2\pi)^{(J-K)/2} \frac{\prod_{j=1}^J |a_j|^{b_j-1/2}}{\prod_{k=1}^K |a'_k|^{b'_k-1/2}}. \quad (2.8)$$

Proposition 2.7. *The parameters $\gamma, \gamma', \delta, \varkappa, C_1$ depend on F only and not on the particular representation in (1.1) or (1.2).*

The proof is given in Section 5.

Remark 2.8. To replace X by X^{-1} , or equivalently Y by $-Y$, means that $F(s)$ is replaced by $F(-s)$, which has the same form but with d replaced by $-d$ (D by D^{-1}) and similarly the sign of each a_j and a'_k is changed. This does not affect γ, δ , and C_1 , but γ' and \varkappa change signs. (This also follows from (5.2) and (5.3) below.)

Remark 2.9. More generally, X^α , with α real and non-zero, has parameters $|\alpha|\gamma, \alpha\gamma', \delta, \alpha\varkappa + \gamma'\alpha \log |\alpha|, C_1|\alpha|^\delta$.

Remark 2.10. If $X = X_1 X_2$ with X_1, X_2 independent and both having moments of Gamma type, cf. Remark 1.5, then the parameters $\gamma, \gamma', \delta, \varkappa$ for X are the sums of the corresponding parameters for X_1 and X_2 , while C_1 is the corresponding product.

Remark 2.11. If X has moments of Gamma type, then so has a suitably conjugated (a.k.a. tilted) distribution: If, for simplicity, X has a density function $f(x)$, $x > 0$, let \tilde{X} have the density function $x^r f(x) / \mathbb{E} X^r$ for a real r such that $\mathbb{E} X^r < \infty$. Then $\mathbb{E} \tilde{X}^s = \mathbb{E} X^{s+r} / \mathbb{E} X^r$ and thus \tilde{X} has moments of Gamma type, obtained by a simple substitution in (1.1). It follows that γ, γ' and \varkappa are the same for \tilde{X} as for X , while δ is increased by $r\gamma'$ and C_1 is multiplied by $e^{r\varkappa} / \mathbb{E} X^r$. (Cf. (5.3) below.) Clearly, ρ_+ and ρ_- are both decreased by r .

For a random variable Y with moment generating function of Gamma type, the same applies to \tilde{Y} with density function $e^{ry} f(y) / \mathbb{E} e^{rY}$, if Y has density function f .

We note also the following relations, for a function $F(s)$ as above:

Lemma 2.12. *We have*

$$\begin{aligned} \sum_{j:a_j>0} a_j - \sum_{k:a'_k>0} a'_k &= \frac{1}{2}(\gamma + \gamma'), \\ \sum_{j:a_j<0} |a_j| - \sum_{k:a'_k<0} |a'_k| &= \frac{1}{2}(\gamma - \gamma'). \end{aligned}$$

Proof.

$$\begin{aligned}\gamma + \gamma' &= \sum_j (|a_j| + a_j) - \sum_k (|a'_k| + a'_k) = 2 \left(\sum_{j:a_j > 0} a_j - \sum_{k:a'_k > 0} a'_k \right), \\ \gamma - \gamma' &= \sum_j (|a_j| - a_j) - \sum_k (|a'_k| - a'_k) = 2 \left(\sum_{j:a_j < 0} |a_j| - \sum_{k:a'_k < 0} |a'_k| \right). \quad \square\end{aligned}$$

3. Some examples

There are several well-known examples of distributions with moments of Gamma type. We collect some of them here. The results below are all well-known. We usually omit scale parameters that may be added to the definitions of the distributions.

Example 3.1 (Gamma distribution). Let Γ_α have the Gamma distribution $\Gamma(\alpha) = \Gamma(\alpha, 1)$ with density function $f(x) = x^{\alpha-1}e^{-x}/\Gamma(\alpha)$, $x > 0$, for a parameter $\alpha > 0$. Then, for $\operatorname{Re} s > -\alpha$,

$$\mathbb{E} \Gamma_\alpha^s = \int_0^\infty x^s f(x) dx = \frac{1}{\Gamma(\alpha)} \int_0^\infty x^{s+\alpha-1} e^{-x} dx = \frac{\Gamma(s+\alpha)}{\Gamma(\alpha)}, \quad (3.1)$$

a simple example of moments of Gamma type. The right hand side of (3.1) is an analytic function in $\operatorname{Re} s > -\alpha$, with a pole at $-\alpha$; thus $\rho_+ = \infty$, $\rho_- = -\alpha$, and $\mathbb{E} \Gamma_\alpha^s = \infty$ when $s \leq -\alpha$.

We have $\gamma = \gamma' = 1$, $\delta = \alpha - 1/2$, $\varkappa = 0$, $C_1 = \sqrt{2\pi}/\Gamma(\alpha)$.

Note that the different Gamma distributions can be obtained by conjugation from each other, cf. Remark 2.11.

Example 3.2 (Exponential distribution). The exponential distribution $\operatorname{Exp}(1)$ with density function e^{-x} , $x > 0$, is the special case $\alpha = 1$ of the Gamma distribution in Example 3.1. We thus obtain from (3.1) (or directly from (A.1)), for $T \sim \operatorname{Exp}(1)$,

$$\mathbb{E} T^s = \Gamma(s+1), \quad \operatorname{Re} s > -1, \quad (3.2)$$

while $\mathbb{E} T^s = \infty$ when $s \leq -1$.

More generally, if T_μ has an exponential distribution $\operatorname{Exp}(\mu) = \Gamma(1, \mu)$ with mean μ , which has the density function is $\mu^{-1}e^{-x/\mu}$, $x > 0$, then $T_\mu \stackrel{d}{=} \mu T$ and

$$\mathbb{E} T_\mu^s = \Gamma(s+1)\mu^s, \quad \operatorname{Re} s > -1. \quad (3.3)$$

Thus T_μ has moments of Gamma type, with $\rho_+ = \infty$, $\rho_- = -1$, $\gamma = \gamma' = 1$, $\delta = 1/2$, $\varkappa = \log \mu$, $C_1 = \sqrt{2\pi}$.

Example 3.3 (Uniform distribution). Let U have a uniform distribution $U(0, 1)$ on $[0, 1]$. Then, obviously, for $\operatorname{Re} s > -1$,

$$\mathbb{E} U^s = \int_0^1 x^s dx = \frac{1}{s+1}. \quad (3.4)$$

This can be rewritten as a Gamma type formula by the functional equation (A.2), which yields

$$\mathbb{E} U^s = \frac{\Gamma(s+1)}{\Gamma(s+2)}, \quad \operatorname{Re} s > -1, \quad (3.5)$$

while $\mathbb{E} U^s = \infty$ when $s \leq -1$.

Of course, it would be silly to claim that (3.5) is a simplification of (3.4), but it shows that the uniform distribution has moments of Gamma type and thus belongs to the class studied here.

We have $\rho_+ = \infty$, $\rho_- = -1$, $\gamma = \gamma' = 0$, $\delta = -1$, $\varkappa = 0$, $C_1 = 1$.

Example 3.4 (Beta distribution). Let $B_{\alpha,\beta}$ have the Beta distribution $B(\alpha, \beta)$, where $\alpha, \beta > 0$; then $B_{\alpha,\beta}$ has a density function $f(x) = cx^{\alpha-1}(1-x)^{\beta-1}$, $0 < x < 1$, where $c = \Gamma(\alpha)^{-1}\Gamma(\beta)^{-1}\Gamma(\alpha+\beta)$, cf. (A.8). (Note that $\alpha = \beta = 1$ yields the uniform distribution in Example 3.3 as a special case.) Consequently, by (A.8), for $\operatorname{Re} s > -\alpha$,

$$\mathbb{E} B_{\alpha,\beta}^s = \int_0^1 x^s f(x) dx = \left(\frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)} \right)^{-1} \frac{\Gamma(s+\alpha)\Gamma(\beta)}{\Gamma(s+\alpha+\beta)} = \frac{\Gamma(\alpha+\beta)\Gamma(s+\alpha)}{\Gamma(\alpha)\Gamma(s+\alpha+\beta)}, \quad (3.6)$$

while $\mathbb{E} B_{\alpha,\beta}^s = \infty$ if $s \leq -\alpha$. We have $\rho_+ = \infty$, $\rho_- = -\alpha$, $\gamma = \gamma' = 0$, $\delta = -\beta$, $\varkappa = 0$, $C_1 = \Gamma(\alpha+\beta)/\Gamma(\alpha)$.

Example 3.5 (Chi-square distribution). The chi-square distribution $\chi^2(n)$ is the distribution of $Q_n := \sum_{i=1}^n N_i^2$, where N_1, N_2, \dots are i.i.d. standard normal variables. It is well-known, see e.g. [9, Section II.3], that the chi-square distribution is a Gamma distribution, differing from the normalized version in Example 3.1 by a scale factor; more precisely, $Q_n \stackrel{d}{=} 2\Gamma_{n/2}$. Consequently, the chi-square distribution has moments of Gamma type, with, by (3.1),

$$\mathbb{E} Q_n^s = 2^s \frac{\Gamma(s+n/2)}{\Gamma(n/2)}, \quad \operatorname{Re} s > -n/2, \quad (3.7)$$

while $\mathbb{E} Q_n^s = \infty$ for $s \leq -n/2$. We have $\rho_+ = \infty$, $\rho_- = -n/2$, $\gamma = \gamma' = 1$, $\delta = (n-1)/2$, $\varkappa = \log 2$, $C_1 = \sqrt{2\pi}/\Gamma(n/2)$.

If $n = 2$, then $Q_2 \stackrel{d}{=} 2\Gamma_1 \stackrel{d}{=} 2T \stackrel{d}{=} T_2$, which also follows from (3.7) and (3.3).

Example 3.6 (Chi distribution). The chi distribution $\chi(n)$ is the distribution of $R_n := (\sum_{i=1}^n N_i^2)^{1/2} = Q_n^{1/2}$, with notations as in Example 3.5. Hence, using (3.7), the chi distribution has moments of Gamma type, with

$$\mathbb{E} R_n^s = \mathbb{E} Q_n^{s/2} = 2^{s/2} \frac{\Gamma(s/2+n/2)}{\Gamma(n/2)}, \quad \operatorname{Re} s > -n, \quad (3.8)$$

while $\mathbb{E} R_n^s = \infty$ for $s \leq -n$. We have $\rho_+ = \infty$, $\rho_- = -n$, $\gamma = \gamma' = 1/2$, $\delta = (n-1)/2$, $\varkappa = 0$, $C_1 = 2^{1-n/2}\pi^{1/2}/\Gamma(n/2)$, cf. Remark 2.9.

In particular, the special case $n = 1$ shows that if $N \sim N(0, 1)$, then $|N| \sim \chi(1)$ has moments of Gamma type with

$$\mathbb{E} |N|^s = \pi^{-1/2} 2^{s/2} \Gamma(s/2 + 1/2). \quad (3.9)$$

Example 3.7 (F-distribution). The F -distribution is the distribution of

$$F_{m,n} := \frac{Q_m/m}{Q'_n/n} = \frac{n}{m} \frac{Q_m}{Q'_n}, \quad (3.10)$$

where $Q_m \sim \chi^2(m)$ and $Q'_n \sim \chi^2(n)$ are independent. By (3.7), this has moments of Gamma type with

$$\mathbb{E} F_{m,n}^s = \left(\frac{n}{m}\right)^s \mathbb{E} Q_m^s \mathbb{E} (Q'_n)^{-s} = \left(\frac{n}{m}\right)^s \frac{\Gamma(s+m/2)\Gamma(n/2-s)}{\Gamma(m/2)\Gamma(n/2)}, \quad (3.11)$$

for $-m/2 < \operatorname{Re} s < n/2$, while $\mathbb{E} F_{m,n}^s = \infty$ for $s \leq -m/2$ and $s \geq n/2$. We have $\rho_+ = n/2$, $\rho_- = -m/2$, $\gamma = 2$, $\gamma' = 0$, $\delta = (n+m-2)/2$, $\varkappa = \log(n/m)$, $C_1 = 2\pi(\Gamma(m/2)\Gamma(n/2))^{-1}$. (Cf. Remarks 2.8 and 2.10.)

Example 3.8 (t-distribution). The t -distribution is the distribution of

$$\mathcal{T}_n := \frac{N}{R_n/\sqrt{n}}, \quad (3.12)$$

where $N \sim \mathcal{N}(0,1)$ and $R_n \sim \chi(n)$ are independent. This random variable is not positive; in fact the distribution is symmetric. However, by (3.10) and (3.12), $\mathcal{T}_n^2 \stackrel{d}{=} F_{1,n}$, so if we consider $|\mathcal{T}_n| \stackrel{d}{=} F_{1,n}^{1/2}$, we see from Example 3.7 that $|\mathcal{T}_n|$ has moments of Gamma type, with

$$\mathbb{E} |\mathcal{T}_n|^s = \mathbb{E} F_{1,n}^{s/2} = n^{s/2} \frac{\Gamma(s/2+1/2)\Gamma(n/2-s/2)}{\sqrt{\pi}\Gamma(n/2)}, \quad (3.13)$$

for $-1 < \operatorname{Re} s < n$, while $\mathbb{E} |\mathcal{T}_n|^s = \infty$ for $s \leq -1$ and $s \geq n$. We have $\rho_+ = n$, $\rho_- = -1$, $\gamma = 1$, $\gamma' = 0$, $\delta = (n-1)/2$, $\varkappa = \frac{1}{2} \log n$, $C_1 = 2^{3/2-n/2}\pi^{1/2}/\Gamma(n/2)$. (Cf. Remark 2.9.)

Example 3.9 (Weibull distribution). The standard Weibull distribution has the distribution function

$$\mathbb{P}(W_\alpha \leq x) = 1 - e^{-x^\alpha}, \quad x > 0, \quad (3.14)$$

for a parameter $\alpha > 0$, and thus density function $\alpha x^{\alpha-1} e^{-x^\alpha}$, $x > 0$.

Note that $\alpha = 1$ yields the exponential distribution in Example 3.2. Moreover, for any α and $y > 0$,

$$\mathbb{P}(W_\alpha^\alpha \leq y) = \mathbb{P}(W_\alpha \leq y^{1/\alpha}) = 1 - e^{-y} = \mathbb{P}(T \leq y),$$

where $T \sim \operatorname{Exp}(1)$; hence $W_\alpha^\alpha \stackrel{d}{=} T$ and $W_\alpha \stackrel{d}{=} T^{1/\alpha}$. It follows from (3.2) that W_α too has moments of Gamma type, see Remark 1.5, with

$$\mathbb{E} W_\alpha^s = \mathbb{E} T^{s/\alpha} = \Gamma(s/\alpha + 1), \quad \operatorname{Re} s > -\alpha, \quad (3.15)$$

while $\mathbb{E} W_\alpha^s = \infty$ if $s \leq -\alpha$. We have $\rho_+ = \infty$, $\rho_- = -\alpha$, $\gamma = \gamma' = 1/\alpha$, $\delta = 1/2$, $\varkappa = \alpha^{-1} \log \alpha^{-1}$, $C_1 = \sqrt{2\pi/\alpha}$, cf. Remark 2.9.

If $\alpha = 1$, then obviously $W_1 \stackrel{d}{=} T \sim \operatorname{Exp}(1)$. If $\alpha = 2$, then $W_2 \stackrel{d}{=} T^{1/2} \stackrel{d}{=} 2^{-1/2} Q_2^{1/2} \stackrel{d}{=} 2^{-1/2} R_2$, which also follows by comparing (3.15) and (3.8).

Example 3.10 (Stable distribution). Let S_α be a positive stable random variable with the Laplace transform $\mathbb{E} e^{-tS_\alpha} = e^{-t^\alpha}$, with $0 < \alpha < 1$. Recall that any positive stable distribution is of this type, for some $\alpha \in (0, 1)$, up to a scale factor, see Feller [9, Section XIII.6]. (We may also allow $\alpha = 1$, but in this exceptional case S_α is degenerate with $S_1 = 1$ a.s.)

For any $s > 0$, by (A.10),

$$\Gamma(s) \mathbb{E} S_\alpha^{-s} = \int_0^\infty t^{s-1} \mathbb{E} e^{-tS_\alpha} dt = \int_0^\infty t^{s-1} e^{-t^\alpha} dt,$$

while the change of variables $u = t^\alpha$ yields

$$\int_0^\infty t^{s-1} e^{-t^\alpha} dt = \frac{1}{\alpha} \int_0^\infty u^{s/\alpha} e^{-u} \frac{du}{u} = \alpha^{-1} \Gamma(s/\alpha).$$

Hence,

$$\mathbb{E} S_\alpha^{-s} = \frac{\Gamma(s/\alpha)}{\alpha \Gamma(s)} = \frac{\Gamma(1 + s/\alpha)}{\Gamma(1 + s)}, \quad s > 0.$$

(In particular, this moment is finite.) Thus, for $s < 0$,

$$\mathbb{E} S_\alpha^s = \frac{\Gamma(1 - s/\alpha)}{\Gamma(1 - s)}. \quad (3.16)$$

We have shown (3.16) for $s < 0$, but Theorem 2.1 (with $\rho_+ = \alpha$ and $\rho_- = -\infty$) shows that (3.16) holds whenever $\operatorname{Re} s < \alpha$, while $\mathbb{E} S_\alpha^s = \infty$ for $s \geq \alpha$. (The case $\alpha = 1$ is exceptional; $\mathbb{E} S_1^s = 1$ for every real s , so (3.16) holds but $\rho_+ = \infty$.) Thus S_α has moments of Gamma type.

We have $\gamma = \alpha^{-1} - 1$, $\gamma' = -(\alpha^{-1} - 1)$, $\delta = 0$, $\varkappa = \alpha^{-1} \log \alpha$, $C_1 = \alpha^{-1/2}$.

Example 3.11 (Mittag-Leffler distribution). The Mittag-Leffler distribution with parameter $\alpha \in (0, 1)$ can be defined as the distribution of the random variable $M_\alpha := S_\alpha^{-\alpha}$, where S_α is a positive stable random variable with $\mathbb{E} e^{-tS_\alpha} = e^{-t^\alpha}$ as in Example 3.4. Since S_α has the moments given by (3.16), the Mittag-Leffler distribution too has moments of Gamma type given by, cf. Remark 1.5,

$$\mathbb{E} M_\alpha^s = \mathbb{E} S_\alpha^{-\alpha s} = \frac{\Gamma(s)}{\alpha \Gamma(\alpha s)} = \frac{\Gamma(s+1)}{\Gamma(\alpha s+1)}, \quad \operatorname{Re} s > -1, \quad (3.17)$$

while $\mathbb{E} M_\alpha^s = \infty$ for $s \leq -1$. In particular, the integer moments are given by

$$\mathbb{E} M_\alpha^n = \frac{n!}{\Gamma(n\alpha + 1)}, \quad n = 0, 1, 2, \dots \quad (3.18)$$

We have $\rho_+ = \infty$, $\rho_- = -1$, $\gamma = \gamma' = 1 - \alpha$, $\delta = 0$, $\varkappa = -\alpha \log \alpha$, $C_1 = \alpha^{-1/2}$, in accordance with Remark 2.9.

The reason for the name ‘‘Mittag-Leffler distribution’’ is that its moment generating function is, by (3.18),

$$\mathbb{E} e^{sM_\alpha} = \sum_{n=0}^{\infty} \mathbb{E} M_\alpha^n \frac{s^n}{n!} = \sum_{n=0}^{\infty} \frac{s^n}{\Gamma(n\alpha + 1)}, \quad (3.19)$$

which converges for any complex s and is known as the Mittag-Leffler function $E_\alpha(s)$ since it was studied by Mittag-Leffler [23]. The formula (3.19), or equivalently (3.18), is often taken as the definition of the Mittag-Leffler distribution.

We may also allow $\alpha = 0$ and $\alpha = 1$, with $M_0 \sim \text{Exp}(1)$ (see Example 3.2) and $M_1 \equiv 1$; in both cases (3.17) and (3.19) hold, although for $\alpha = 0$, (3.19) converges only for $|s| < 1$, and $\delta = 1/2$, $C_1 = \sqrt{2\pi}$, while for $\alpha = 1$, (3.17) trivially holds for all s with $\mathbb{E} M_\alpha^s = 1 < \infty$ and $\rho_- = -\infty$.

The Mittag-Leffler distribution was introduced by Feller [8], see also [9, Sections XI.5 and XIII.8(b)] and Pollard [26]. Blumenfeld and Mandelbrot [4] considered also $\log M_\alpha$, which by (3.17) has moment generating function $\Gamma(s+1)/\Gamma(\alpha s+1)$ of Gamma type, and called its distribution the “logarithmic Mittag-Leffler distribution”. Feller [8; 9] showed that the Mittag-Leffler distribution is the limit distribution as $t \rightarrow \infty$ of the number of renewals up to time t , properly normalized, of an i.i.d. sequence of positive random variables belonging to the domain of attraction of a stable law. It emerges also, for example, as the limit distribution of occupancy times in the Darling–Kac theorem, see Bingham, Goldie, Teugels [3, Section 8.11].

In the special case $\alpha = 1/2$, the duplication formula (A.3) yields

$$\mathbb{E} M_{1/2}^s = \frac{\Gamma(s+1)}{\Gamma(s/2+1)} = \pi^{-1/2} 2^s \Gamma(s/2+1/2), \quad (3.20)$$

which by comparison with (3.9) shows that $M_{1/2} \stackrel{d}{=} \sqrt{2}|N|$ with $N \sim N(0,1)$, which is equivalent to the well-known relation $S_{1/2} \stackrel{d}{=} \frac{1}{2}N^{-2}$.

Example 3.12 (A different Mittag-Leffler distribution). Another distribution related to the Mittag-Leffler function $E_\alpha(s)$ in (3.19), and, rather unfortunately, therefore also called “Mittag-Leffler distribution” was introduced by Pillai [25] as the distribution of a random variable L_α with distribution function $1 - E_\alpha(-x^\alpha)$, where $0 < \alpha \leq 1$; equivalently, by (3.19),

$$\mathbb{P}(L_\alpha > x) = E_\alpha(-x^\alpha) = \mathbb{E} e^{-x^\alpha M_\alpha}. \quad (3.21)$$

This is another instance of the relation (1.10), and Lemma 1.11 shows that L_α has moments of Gamma type with, using (3.17),

$$\mathbb{E} L_\alpha^s = \Gamma(s/\alpha+1) \mathbb{E} M_\alpha^{-s/\alpha} = \frac{\Gamma(1+s/\alpha)\Gamma(1-s/\alpha)}{\Gamma(1-s)}, \quad -\alpha < s < \alpha, \quad (3.22)$$

while $\mathbb{E} L_\alpha^s = \infty$ if $s \leq -\alpha$ or (provided $\alpha < 1$) $s \geq \alpha$. Note also that Lemma 1.11 yields the representation [25]

$$L_\alpha \stackrel{d}{=} T^{1/\alpha} M_\alpha^{-1/\alpha} = T^{1/\alpha} S_\alpha, \quad (3.23)$$

with $T \sim \text{Exp}(1)$ and the stable variable S_α independent. Equivalently, see Example 3.9, $L_\alpha \stackrel{d}{=} W_\alpha S_\alpha$, with the Weibull variable W_α and S_α independent. It follows easily from (3.23) that L_α has the Laplace transform $\mathbb{E} e^{-tL_\alpha} = (1+t^\alpha)^{-1}$, $t > 0$ [25].

For $\alpha = 1$, we have $L_1 \sim \text{Exp}(1)$. For $0 < \alpha < 1$, $\rho_+ = \alpha$, $\rho_- = -\alpha$, $\gamma = 2/\alpha - 1$, $\gamma' = 1$, $\delta = 1/2$, $\varkappa = 0$, $C_1 = \sqrt{2\pi}/\alpha$. (For example by Remark 2.10.)

Example 3.13 (Pareto distribution). The Pareto(α) distribution, where $\alpha > 0$, is the distribution of a random variable P_α with $\mathbb{P}(P_\alpha > x) = x^{-\alpha}$, $x \geq 1$. Hence P_α has density function $\alpha x^{-\alpha-1}$, $x > 1$. Direct integration shows that the moments are of Gamma type and given by

$$\mathbb{E} P_\alpha^s = \int_1^\infty \alpha x^{s-\alpha-1} dx = \frac{\alpha}{\alpha-s} = \frac{\alpha\Gamma(\alpha-s)}{\Gamma(\alpha-s+1)}, \quad \text{Re } s < \alpha. \quad (3.24)$$

Hence $\rho_+ = \alpha$, $\rho_- = -\infty$, $\gamma = \gamma' = 0$, $\delta = -1$, $\varkappa = 0$, $C_1 = \alpha$.

We have $P_\alpha \stackrel{d}{=} U^{-1/\alpha}$ with $U \sim \text{U}(0,1)$, so alternatively these result follow from Example 3.3 and Remarks 1.5 and 2.9.

Example 3.14 (Shifted Pareto distribution). The shifted Pareto variable $\tilde{P}_\alpha := P_\alpha - 1$, where $\alpha > 0$, has support $(0, \infty)$ and density function $\alpha(x+1)^{-\alpha-1}$, $x > 0$. The moments are given by, using (A.9),

$$\mathbb{E} \tilde{P}_\alpha^s = \int_0^\infty \alpha x^s (x+1)^{-\alpha-1} dx = \alpha \frac{\Gamma(s+1)\Gamma(\alpha-s)}{\Gamma(\alpha+1)} = \frac{\Gamma(s+1)\Gamma(\alpha-s)}{\Gamma(\alpha)}, \quad (3.25)$$

for $-1 < \text{Re } s < \alpha$, while $\mathbb{E} \tilde{P}_\alpha^s = \infty$ if $s \leq -1$ or $s \geq \alpha$. Hence also the shifted Pareto distribution has moments of Gamma type, with $\rho_+ = \alpha$, $\rho_- = -1$, $\gamma = 2$, $\gamma' = 0$, $\delta = \alpha$, $\varkappa = 0$, $C_1 = 2\pi/\Gamma(\alpha)$.

For $\alpha = 1$, (3.25) yields $\mathbb{E} \tilde{P}_1^s = \Gamma(1+s)\Gamma(1-s) = \mathbb{E} \tilde{P}_1^{-s}$; hence $\tilde{P}_1 \stackrel{d}{=} \tilde{P}_1^{-1}$ (so $\log \tilde{P}_1$ has a symmetric distribution). Using (A.6), we have

$$\mathbb{E} \tilde{P}_1^s = \Gamma(1+s)\Gamma(1-s) = s\Gamma(s)\Gamma(1-s) = \frac{\pi s}{\sin(\pi s)}, \quad -1 < \text{Re } s < 1.$$

Further, comparing (3.25) and (3.11) we see that $\tilde{P}_1 \stackrel{d}{=} F_{2,2}$.

More generally, (3.25) and (3.1) imply that if $\Gamma_1 \sim \Gamma(1) = \text{Exp}(1)$ and $\Gamma'_\alpha \sim \Gamma(\alpha)$ are independent, then $\mathbb{E}(\Gamma_1/\Gamma'_\alpha)^s = \mathbb{E}\Gamma_1^s \mathbb{E}(\Gamma'_\alpha)^{-s} = \mathbb{E} \tilde{P}_\alpha^s$ (for suitable s); thus $\tilde{P}_\alpha \stackrel{d}{=} \Gamma_1/\Gamma'_\alpha$ by Corollary 2.3. This is also an instance of Lemma 1.11 (with $\alpha = 1$ in (1.10) and (1.11)).

Example 3.15 (Extreme value distributions). There are three types of extreme value distributions, see e.g. Leadbetter, Lindgren and Rootzén [21, Chapter I]. We let $X_I, X_{II,\alpha}, X_{III,\alpha}$ denote corresponding random variables; they have the distribution functions

$$\mathbb{P}(X_I \leq x) = e^{-e^{-x}}, \quad -\infty < x < \infty, \quad (3.26)$$

$$\mathbb{P}(X_{II,\alpha} \leq x) = e^{-x^{-\alpha}}, \quad 0 < x < \infty, \quad (3.27)$$

$$\mathbb{P}(X_{III,\alpha} \leq x) = e^{-(-x)^\alpha}, \quad -\infty < x < 0, \quad (3.28)$$

where (for types II and III) $\alpha > 0$ is a real parameter.

The distribution (3.26) (the Gumbel distribution) has the entire real line as support, and is therefore not qualified to have moments of Gamma type. It has, however, moment generating function of Gamma type, see Example 3.19

The distribution (3.27) (the Fréchet distribution) satisfies, with $T \sim \text{Exp}(1)$,

$$\mathbb{P}(X_{II,\alpha} \leq x) = \mathbb{P}(T \geq x^{-\alpha}) = \mathbb{P}(T^{-1/\alpha} \leq x), \quad x > 0,$$

and thus $X_{II,\alpha} \stackrel{d}{=} T^{-1/\alpha} \stackrel{d}{=} 1/W_\alpha$, see Example 3.9. Hence it has moments of Gamma type with, see (3.2) and (3.15),

$$\mathbb{E} X_{II,\alpha}^s = \Gamma(1 - s/\alpha), \quad \text{Re } s < \alpha. \quad (3.29)$$

We have $\rho_+ = \alpha$, $\rho_- = -\infty$, $\gamma = 1/\alpha$, $\gamma' = -1/\alpha$, $\delta = 1/2$, $\varkappa = \alpha^{-1} \log \alpha$, $C_1 = \sqrt{2\pi/\alpha}$, cf. Remarks 2.8 and 2.9 and Examples 3.2 and 3.9.

The distribution (3.28) satisfies $\mathbb{P}(-X_{III,\alpha} \geq x) = e^{-x^\alpha}$, $x > 0$, so $-X_{III,\alpha} \stackrel{d}{=} W_\alpha$ and $X_{III,\alpha} \stackrel{d}{=} -W_\alpha$, see Example 3.9. By (3.15), $|X_{III,\alpha}| = -X_{III,\alpha}$ has moments of Gamma type with

$$\mathbb{E} |X_{III,\alpha}|^s = \Gamma(1 + s/\alpha), \quad \text{Re } s > -\alpha. \quad (3.30)$$

We have $\rho_+ = \infty$, $\rho_- = -\alpha$, $\gamma = \gamma' = 1/\alpha$, $\delta = 1/2$, $\varkappa = \alpha^{-1} \log \alpha^{-1}$, $C_1 = \sqrt{2\pi/\alpha}$.

Example 3.16 (Fejér distribution). The distribution with density function

$$\frac{1 - \cos x}{\pi x^2} = \frac{1}{2\pi} \left(\frac{\sin(x/2)}{x/2} \right)^2, \quad -\infty < x < \infty, \quad (3.31)$$

has characteristic function $(1 - |t|)_+$, which vanishes outside the interval $[-1, 1]$, see [9, Section XV.2]. The density function (3.31) is in harmonic analysis known as the Fejér kernel on the real line, so we call this distribution the Fejér distribution.

If X has this distribution, then, for $-1 < s < 0$, using integration by parts, dominated convergence, (A.7) and (A.6),

$$\begin{aligned} \mathbb{E} |X|^s &= \frac{2}{\pi} \int_0^\infty (1 - \cos x) x^{s-2} dx = \frac{-2}{\pi(s-1)} \int_0^\infty \sin x \cdot x^{s-1} dx \\ &= \frac{-2}{\pi(s-1)} \lim_{\varepsilon \searrow 0} \text{Im} \int_0^\infty e^{(i-\varepsilon)x} x^{s-1} dx \\ &= \frac{2}{\pi(s-1)} \lim_{\varepsilon \searrow 0} \text{Im} \left(\frac{i-\varepsilon}{s} \int_0^\infty e^{(i-\varepsilon)x} x^s dx \right) \\ &= \frac{2\Gamma(s+1)}{\pi s(s-1)} \lim_{\varepsilon \searrow 0} \text{Im} (-(\varepsilon - i)^{-s}) = -\frac{2\Gamma(s-1)}{\pi} \text{Im}(e^{\pi si/2}) \\ &= -\frac{2\Gamma(s-1)}{\pi} \sin\left(\frac{\pi s}{2}\right) = \frac{-2\Gamma(s-1)}{\Gamma(s/2)\Gamma(1-s/2)}. \end{aligned}$$

The right-hand side has poles at $\rho_{\pm} = \pm 1$ (and a removable singularity at 0), and Theorem 2.1 shows that the formula extends to all s with $-1 < \operatorname{Re} s < 1$. In other words, $|X|$ has moments of Gamma type with, using (A.6) again and (A.3) to obtain several versions, for $-1 < \operatorname{Re} s < 1$,

$$\begin{aligned} \mathbb{E}|X|^s &= -\frac{2\Gamma(s-1)}{\pi} \sin\left(\frac{\pi s}{2}\right) = \frac{-2\Gamma(s-1)}{\Gamma(s/2)\Gamma(1-s/2)} \\ &= -\frac{2^s}{2\sqrt{\pi}} \frac{\Gamma(s/2-1/2)}{\Gamma(1-s/2)} \\ &= -\frac{1}{\pi} \frac{\Gamma(s/2-1/2)\Gamma(3/2-s/2)}{\Gamma(2-s)} \\ &= \frac{1}{\cos(\pi s/2)\Gamma(2-s)} = \frac{\Gamma(1/2-s/2)\Gamma(1/2+s/2)}{\pi\Gamma(2-s)}. \end{aligned} \quad (3.32)$$

We have $\gamma = 0$, $\gamma' = 1$, $\delta = -3/2$, $\varkappa = 0$, $C_1 = (2/\pi)^{1/2}$.

We have so far considered distributions with moments of Gamma type; we now turn to a few examples with moment generating function of Gamma type. Of course, if X is any of the examples above, $\log X$ yields such an example. (One such example was mentioned in Example 3.11.)

Example 3.17 (Exponential distribution again). We noted in Example 3.2 that $T \sim \operatorname{Exp}(1)$ has moments of Gamma type. It has moment generating function of Gamma type too, since

$$\mathbb{E}e^{sT} = \int_0^{\infty} e^{sx-x} dx = \frac{1}{1-s} = \frac{\Gamma(1-s)}{\Gamma(2-s)}, \quad \operatorname{Re} s < 1. \quad (3.33)$$

In fact, $T \stackrel{d}{=} -\log U \stackrel{d}{=} \log P_1$, and equivalently $U \stackrel{d}{=} e^{-T}$ and $P_1 \stackrel{d}{=} e^T$; compare (3.33) to (3.5) and (3.24). We have $\rho_+ = 1$, $\rho_- = -\infty$, $\gamma = \gamma' = 0$, $\delta = -1$, $\varkappa = 0$, $C_1 = 1$.

Example 3.18 (Gamma distribution again). A Gamma distributed random variable $\Gamma_n \sim \Gamma(n)$ with integer $n \geq 1$ can be obtained by taking the sum of n independent copies of $T \sim \operatorname{Exp}(1)$, and thus (3.33) implies that Γ_n has moment generating function of Gamma type with

$$\mathbb{E}e^{s\Gamma_n} = (\mathbb{E}e^{sT})^n = \frac{1}{(1-s)^n} = \frac{\Gamma(1-s)^n}{\Gamma(2-s)^n}, \quad \operatorname{Re} s < 1. \quad (3.34)$$

We have $\rho_+ = 1$, $\rho_- = -\infty$, $\gamma = \gamma' = 0$, $\delta = -n$, $\varkappa = 0$, $C_1 = 1$.

More generally, for any real $\alpha > 0$, Γ_{α} has moment generating function $\mathbb{E}e^{s\Gamma_{\alpha}} = (1-s)^{-\alpha}$. If α is not an integer, then this function has a singularity as $s \nearrow 1$ that is not a pole; hence $\mathbb{E}e^{s\Gamma_{\alpha}}$ cannot be extended to a meromorphic function in the complex plane, and Γ_{α} does *not* have moment generating function of Gamma type. Consequently, the Gamma distribution $\Gamma(\alpha)$ has moment generating function of Gamma type if and only if α is an integer.

Example 3.19 (Gumbel distribution). If X_I has the Gumbel distribution (3.26), and $T \sim \text{Exp}(1)$, then

$$\mathbb{P}(e^{-X_I} \geq x) = \mathbb{P}(X_I \leq -\log x) = e^{-x} = \mathbb{P}(T \geq x), \quad x > 0,$$

so $e^{-X_I} \stackrel{d}{=} T$ and $X_I \stackrel{d}{=} -\log T$. Consequently, the Gumbel distribution has moment generating function of Gamma type with, see (3.2),

$$\mathbb{E} e^{sX_I} = \mathbb{E} T^{-s} = \Gamma(1-s), \quad \text{Re } s < 1. \quad (3.35)$$

We have $\rho_+ = 1$, $\rho_- = -\infty$, $\gamma = 1$, $\gamma' = -1$, $\delta = 1/2$, $\varkappa = 0$, $C_1 = \sqrt{2\pi}$.

Example 3.20 (Lévy area). The Lévy stochastic area is defined by the stochastic integral $A_t := \int_0^t X_u dY_u - \int_0^t Y_u dX_u$, where (X_u, Y_u) , $u \geq 0$, is a two-dimensional Brownian motion starting at 0 (i.e., X_u and Y_u , $u \geq 0$, are two independent standard Brownian motions). By Brownian scaling, $A_t \stackrel{d}{=} tA_1$, so we consider only $A := A_1$. Then, for real t , see e.g. Protter [28, Theorem II.43], using (A.6),

$$\mathbb{E} e^{itA} = \frac{1}{\cosh(t)} = \frac{1}{\cos(it)} = \frac{1}{\sin(\frac{\pi}{2} + it)} = \frac{\Gamma(\frac{1}{2} + \frac{it}{\pi})\Gamma(\frac{1}{2} - \frac{it}{\pi})}{\pi}. \quad (3.36)$$

Consequently, by Theorem 2.1, A has moment generating function of Gamma type, with $\rho_{\pm} = \pm\pi/2$ and

$$\mathbb{E} e^{sA} = \frac{1}{\cos s} = \frac{\Gamma(\frac{1}{2} + \frac{s}{\pi})\Gamma(\frac{1}{2} - \frac{s}{\pi})}{\pi}, \quad |\text{Re } s| < \pi/2. \quad (3.37)$$

We have $\gamma = 2/\pi$, $\gamma' = 0$, $\delta = 0$, $\varkappa = 0$, $C_1 = 2$.

It is known that A has the density function $1/2 \cosh(\pi x/2)$, $-\infty < x < \infty$, see e.g. Protter [28, Corollary to Theorem II.43] and Example 6.18 below.

Of course, e^A has moments of Gamma type, and so has e^{cA} for every real c . A comparison with (3.11) shows that $e^{\pi A} \stackrel{d}{=} F_{1,1}$. Hence, $A \stackrel{d}{=} \pi^{-1} \log F_{1,1}$.

Remark 3.21. We have shown that a large number of classical continuous distributions have moments of Gamma type, but there are exceptions. For example, if $X \sim U(1,2)$, then $\mathbb{E} X^s = (2^{s+1} - 1)/(s+1)$ has complex zeros at $-1 + 2\pi ik/\log 2$, $k = \pm 1, \pm 2, \dots$, which is impossible when (1.1) holds. More generally, if X is non-degenerate and is supported in a finite interval $[a, b]$ with $0 < a < b < \infty$, then $\mathbb{E} X^s$ is an entire function of s , which by Theorem 4.1 below shows that X does not have moments of Gamma type.

4. Poles and zeros

In this section it will be convenient to consider the class \mathcal{F} of all functions of the type in (1.1) and (1.2), regardless of whether they equal $\mathbb{E} X^s$ [$\mathbb{E} e^{sY}$] for some random variable X [Y] or not. Thus, \mathcal{F} is the set of functions

$$F(s) = CD^s \frac{\prod_{j=1}^J \Gamma(a_j s + b_j)}{\prod_{k=1}^K \Gamma(a'_k s + b'_k)}, \quad (4.1)$$

where $J, K \geq 0$ and $C, D, a_j, b_j, a'_k, b'_k$ are real with $D > 0$ and $a_j \neq 0, a'_k \neq 0$ for all j and k . We let $\mathcal{F}_m \subset \mathcal{F}$ be the set of such functions that appear in (1.1) and (1.2), i.e., the set of $F(s) \in \mathcal{F}$ such that $F(s) = \mathbb{E} X^s$ for some positive random variable X and s in some interval.

A function $F(s) \in \mathcal{F}$ is a meromorphic function of s in the complex plane \mathbb{C} . We can easily locate its poles (and zeros) precisely as follows. Define $\nu(s) = \nu_F(s)$ for all $s \in \mathbb{C}$ by

$$\nu_F(s) = \begin{cases} m & \text{if } F \text{ has a pole of order } m \text{ at } s, \\ -m & \text{if } F \text{ has a zero of order } m \text{ at } s, \\ 0 & \text{otherwise (i.e., } F \text{ is regular at } s \text{ with } F(s) \neq 0). \end{cases} \quad (4.2)$$

(For $F \in \mathcal{F}_m$ with $F(s)$ an extension of $\mathbb{E} X^s$, we also write ν_X .)

Since D^s has neither poles nor zeros, while $\Gamma(z)$ has a simple pole at each $z \in \mathbb{Z}_{\leq 0} := \{0, -1, -2, \dots\}$ but no zeros,

$$\begin{aligned} \nu_F(s) &= \sum_{j=1}^J \mathbf{1}[a_j s + b_j \in \mathbb{Z}_{\leq 0}] - \sum_{k=1}^K \mathbf{1}[a'_k s + b'_k \in \mathbb{Z}_{\leq 0}] \\ &= \sum_{j=1}^J \mathbf{1}\left[s \in \left\{-\frac{n+b_j}{a_j} : n \in \mathbb{Z}_{\geq 0}\right\}\right] - \sum_{k=1}^K \mathbf{1}\left[s \in \left\{-\frac{n+b'_k}{a'_k} : n \in \mathbb{Z}_{\geq 0}\right\}\right]. \end{aligned} \quad (4.3)$$

Note that all poles and zeros of F lie on the real axis.

We further define $\mathcal{F}_0 := \{F \in \mathcal{F} : \nu_F(0) = 0\}$, i.e., the set of functions $F \in \mathcal{F}$ that have neither a pole nor a zero at 0. By Remark 2.4, $\mathcal{F}_m \subset \mathcal{F}_0 \subset \mathcal{F}$. Note that \mathcal{F} is a group under multiplication and that \mathcal{F}_0 is a subgroup.

For $F \in \mathcal{F}_m$ we defined ρ_{\pm} in Section 2; the definition (2.1) can be written

$$\begin{aligned} \rho_+ &:= \min\{x > 0 : \nu_F(x) > 0\}, \\ \rho_- &:= \max\{x < 0 : \nu_F(x) > 0\}. \end{aligned} \quad (4.4)$$

For general $F \in \mathcal{F}$ we define

$$\begin{aligned} \rho_+ &:= \min\{x \geq 0 : \nu_F(x) \neq 0\}, \\ \rho_- &:= \max\{x \leq 0 : \nu_F(x) \neq 0\}, \end{aligned} \quad (4.5)$$

and note that this is consistent with (4.4) for $F \in \mathcal{F}_m$ by the fact (Theorem 2.1) that such F has no zeros in (ρ_-, ρ_+) and no pole at 0.

If all $a_j, a'_k, b_j, b'_k > 0$, then $F(s)$ has no poles or zeros in the right halfplane $\operatorname{Re} s \geq 0$, since none of the factors in (4.1) has; thus $\rho_+ = \infty$. Similarly, if all $a_j, a'_k < 0$ and $b_j, b'_k > 0$, then $F(s)$ has no poles or zeros in the left halfplane $\operatorname{Re} s \leq 0$ and $\rho_- = -\infty$. The converses do not hold, because the representation (4.1) is not unique and we may, e.g., add cancelling factors that separately have poles at other places. However, we may always choose a representation of the desired type; this is part of the following theorem.

Theorem 4.1. *Let $F \in \mathcal{F}_0$ and let ρ_{\pm} be as in (4.5).*

- (i) *It is always possible to find a representation (4.1) of $F(s)$ with all $b_j, b'_k > 0$.*
- (ii) *If $\rho_- = -\infty$, then $F(s)$ has a representation (4.1) with all $a_j, a'_k < 0$ and $b_j, b'_k > 0$.*
- (iii) *If $\rho_+ = \infty$, then $F(s)$ has a representation (4.1) with all $a_j, a'_k, b_j, b'_k > 0$.*
- (iv) *If $\rho_+ = \infty$ and $\rho_- = -\infty$ (i.e., $F(s)$ is entire and without zeros), then $F(s) = CD^s$ for some real constants C and $D > 0$.*

In particular, if X is a random variable with moments of Gamma type, this applies to the meromorphic extension $F(s)$ of $\mathbb{E}X^s$. In this case, it is not possible that both $\rho_+ = \infty$ and $\rho_- = -\infty$ (i.e., that $\mathbb{E}X^s$ is entire), except in the trivial case $X = D$ a.s. for some $D > 0$ (and thus $\mathbb{E}X^s = D^s$).

We begin by proving a lemma.

Lemma 4.2. *Suppose that a_j, a'_j, b_k, b'_k are real with $a_j, a'_k > 0$ and that $F(s) := \prod_{j=1}^J \Gamma(a_j s + b_j) / \prod_{k=1}^K \Gamma(a'_k s + b'_k)$ is analytic and non-zero in a half-plane $\operatorname{Re} s < B$ for some $B \in (-\infty, \infty)$. Then $F(s) = e^{\alpha s} Q(s)$ for some rational function Q with real poles and zeros and some $\alpha \in \mathbb{R}$.*

Proof. Say that two non-zero real numbers a and a' are *commensurable* if $a/a' \in \mathbb{Q}$; this is an equivalence relation on $\mathbb{R}^* := \mathbb{R} \setminus \{0\}$. (In algebraic language, the equivalence classes are the cosets of \mathbb{Q}^* in \mathbb{R}^* .) The poles of $\Gamma(as + b)$ are regularly spaced with distances $1/|a|$. Thus, if a and a' are incommensurable, and b, b' are any real numbers, then $\Gamma(as + b)$ and $\Gamma(a's + b')$ have at most one common pole.

We divide the set $\{a_j\}_{j=1}^J \cup \{a'_k\}_{k=1}^K$ into equivalence classes of commensurable numbers. This gives a corresponding factorization $F(s) = \prod_{\ell=1}^L F_{\ell}(s)$ where L is the number of equivalence classes and each F_{ℓ} is of the same form as F but with all a_j and a'_k commensurable. It follows that two different factors $F_{\ell_1}(s)$ and $F_{\ell_2}(s)$ have at most a finite number of common zeros or poles; hence, by decreasing B , we may assume that there are no such common poles or zeros with $\operatorname{Re} s < B$. Hence, a pole or zero of a factor $F_{\ell}(s)$ in $\operatorname{Re} s < B$ cannot be cancelled by another factor, and thus such poles or zeros do not exist. Consequently, each factor $F_{\ell}(s)$ satisfies the assumption of the lemma, and we may thus treat each $F_{\ell}(s)$ separately. This means that we may assume that all a_j and a'_k are commensurable.

In this case, there is a positive real number r such that all a_j and a'_k are (positive) integer multiples of r . Using Gauss's multiplication formula (A.5), we may convert each factor $\Gamma(a_j s + b_j)$ or $\Gamma(a'_k s + b'_k)$ into a product of a constant, an exponential factor $e^{\beta s}$, and a number of Gamma factors $\Gamma(rs + u_i)$ with $u_i \in \mathbb{R}$. Using the functional relation (A.2), we may further assume that each $u_i \in (0, 1]$, provided we also allow factors $(rs + u)^{\pm 1}$ with $u \in \mathbb{R}$. Collecting the

factors, we see that

$$F(s) = e^{\alpha s} Q(s) \frac{\prod_{j=1}^{J'} \Gamma(rs + u_j)}{\prod_{k=1}^{K'} \Gamma(rs + u'_k)} \quad (4.6)$$

for a constant α , a rational function $Q(s)$ with real zeros and poles, and some $u_j, u'_k \in (0, 1]$. We may assume that Q has no poles or zeros with $\operatorname{Re} s < B$ (by again decreasing B if necessary). The factors $\Gamma(rs + u_j)$ and $\Gamma(rs + u'_k)$ have no zeros but each has an infinite number of poles with $\operatorname{Re} s < B$, and two factors $\Gamma(rs + u_j)$ and $\Gamma(rs + u'_k)$ have disjoint sets of poles unless $u_j = u'_k$. Since the poles with $\operatorname{Re} s < B$ must cancel in (4.6), this shows that the Gamma factors must cancel each other completely, and thus (4.6) reduces to $F(s) = e^{\alpha s} Q(s)$. \square

Proof of Theorem 4.1. (i): Using $\Gamma(z) = \Gamma(z+1)/z$ repeatedly on any term with $b_j \leq 0$ or $b'_k \leq 0$, we may write $F(s)$ as

$$CD^s Q(s) \frac{\prod_{j=1}^J \Gamma(a_j s + \tilde{b}_j)}{\prod_{k=1}^K \Gamma(a'_k s + \tilde{b}'_k)}, \quad (4.7)$$

with $\tilde{b}_j, \tilde{b}'_k > 0$, where $Q(s)$ is a rational function with only real poles and zeros; $Q(s) = c \prod_i (s - r_i) / \prod_\ell (s - r'_\ell)$ for some real r_i and r'_ℓ , $i = 1, \dots, I$ and $\ell = 1, \dots, L$, say; we may further assume that $r_i \neq r'_\ell$ for all i and ℓ . Since 0 is not a pole or zero of F , it is by (4.7) not a zero or pole of Q , and thus all $r_i, r'_\ell \neq 0$.

If $r < 0$, then

$$s - r = s + |r| = \frac{\Gamma(s + |r| + 1)}{\Gamma(s + |r|)}, \quad (4.8)$$

and if $r > 0$, then

$$s - r = -(r - s) = -\frac{\Gamma(-s + r + 1)}{\Gamma(-s + r)}, \quad (4.9)$$

so $Q(s)$ may be written as product of quotients of Gamma factors of the desired type (and a constant), and thus the result follows from (4.7).

(ii): We use (i) and may thus assume that (4.1) holds with $b_j, b'_k > 0$. We factorize $F(s)$ as, with $d = \log D$,

$$F(s) = C e^{ds} F_+(s) F_-(s), \quad (4.10)$$

where $F_+(s)$ contains all factors $\Gamma(a_j s + b_j)$ and $\Gamma(a'_k s + b'_k)$ with $a_j, a'_k > 0$, and $F_-(s)$ contains the factors with $a_j, a'_k < 0$, so F_- is of the desired form.

Since $b_j, b'_k > 0$, $F_-(s)$ has no poles or zeros with $\operatorname{Re} s \leq 0$. By assumption, $\rho_- = -\infty$, and thus $F(s)$ is analytic and non-zero in the half-plane $\operatorname{Re} s \leq 0$. Consequently, by (4.10), $F_+(s)$ also has no poles or zeros in $\operatorname{Re} s \leq 0$. By Lemma 4.2, $F_+(s) = Q(s)e^{\alpha s}$ for a rational function $Q(s) = C_1 \prod_j (s -$

$u_j)/\prod_k(s-v_k)$ where u_j and v_k are real. We may assume that $u_j \neq v_k$ for all j, k , and then each u_j or v_k is a zero or pole of $F_+(s)$, and thus all $u_j, v_k > 0$. Using (4.9), we thus can write $Q(s)$ in the desired form; hence $F_+(s)$ and $F(s)$ can be so written.

(iii): Follows from (ii) by replacing $F(s)$ by $F(-s)$.

(iv): In this case, $F(s)$ is an entire function without zeros. We use again the factorization (4.10). By the proof of (ii), $F_+(s) = e^{\alpha s}Q(s)$ for a rational function Q . By symmetry, as in the proof of (iii), similarly $F_-(s) = e^{\alpha-s}Q_-(s)$ for another rational function Q_- . Thus,

$$F(s) = Ce^{(d+\alpha+\alpha_-)s}Q(s)Q_-(s). \quad (4.11)$$

Hence, the rational function $Q(s)Q_-(s)$ is entire and has neither poles nor zeros; consequently, it is constant. Thus $F(s) = C_1e^{d_1s}$ for some C_1 and d_1 . \square

As a consequence we show that the function $\nu_F(s)$ describing the poles and zeros of $F(s)$ essentially determines F , and thus the distribution of X and Y satisfying (1.1) or (1.2).

Theorem 4.3. *If $F_1, F_2 \in \mathcal{F}$ and $\nu_{F_1} = \nu_{F_2}$, then $F_2(s) = cD^sF_1(s)$ for some real constants $C \neq 0$ and $D > 0$.*

Proof. $F := F_2/F_1 \in \mathcal{F}$ and $\nu_F(s) = \nu_{F_1}(s) - \nu_{F_2}(s) = 0$ for every s . Hence, $F \in \mathcal{F}_0$ and $\rho_- = -\infty, \rho_+ = \infty$; thus Theorem 4.1(iv) shows that $F(s) = CD^s$. \square

Corollary 4.4. *If X_1 and X_2 are positive random variables with moments of Gamma type and $\nu_{X_1} = \nu_{X_2}$, then $X_2 \stackrel{d}{=} DX_1$ for some constant $D > 0$. In other words, a distribution with moments of Gamma type is uniquely determined up to a scaling factor by the function $\nu_X(s)$.*

Similarly, if Y_1 and Y_2 have moment generating functions of Gamma type with the same ν , then $Y_2 \stackrel{d}{=} Y_1 + d$ for some real constant d .

Proof. Let $F_j(s)$ be the meromorphic extension of $\mathbb{E}X_j^s, j = 0, 1$. Then $F_2(s) = CD^sF_1(s)$ by Theorem 4.3. Setting $s = 0$ we find $C = 1$, and thus, for s in some interval, $\mathbb{E}X_2^s = F_2(s) = D^sF_1(s) = \mathbb{E}(DX_1)^s$, whence $X_2 \stackrel{d}{=} DX_1$ by Corollary 2.3.

The final statement follows by considering $X_j := e^{Y_j}$. \square

Remark 4.5. The proofs of Theorem 4.1 and Lemma 4.2 yield an almost canonical way of expressing $F(s) \in \mathcal{F}_0$ in the form (4.1). We start by making all $b_j, b'_k > 0$ by Theorem 4.1(i). We then treat positive and negative a_j and a'_k separately; furthermore, if these coefficients are not all commensurable (which they are in most natural examples), we separate them into different equivalence classes of commensurable coefficients. For each class we then rewrite the product of the corresponding factors in the form (4.6) for some real r (different for different classes, and chosen with $|r|$ as large as possible). Note that different factors in (4.6) have no common poles, so it is easy to locate all poles and zeros. It only remains to take care of the rational part in (4.6); in the examples

we know, this is not a problem but we have not studied this in general, and we do not know whether it is possible to use this approach to define a unique canonical representation (4.1) for each F ; we leave this as an open problem. (Alternatively, it might be possible to define a canonical representation including a rational factor.)

See Theorems 1.6 and 1.12 for examples of such 'canonical' versions, but note that they not necessarily are the simplest by other criteria.

Let $N_+(x) := \sum_{0 < s \leq x} \nu_F(s)$, $x > 0$, and $N_-(x) := \sum_{x \leq s < 0} \nu_F(s)$, $x < 0$. Thus $N_+(x)$ is the total number of poles minus the total number of zeros (with multiplicities) in the interval $(0, x]$, and similarly for $N_-(x)$ and the interval $[x, 0)$ on the negative half-axis. The following proposition can be interpreted as giving the density of poles minus zeros on the positive or negative half-axis.

Proposition 4.6. *Let $N_+(x) := \sum_{0 < s \leq x} \nu_F(s)$ for $x > 0$, and $N_-(x) := \sum_{x \leq s < 0} \nu_F(s)$ for $x < 0$. Then $N_+(x)/x \rightarrow \frac{1}{2}(\gamma - \gamma')$ as $x \rightarrow +\infty$ and $N_-(x)/|x| \rightarrow \frac{1}{2}(\gamma + \gamma')$ as $x \rightarrow -\infty$.*

Proof. Use, for simplicity, a representation as in Theorem 4.1(i). Then, using (4.3), the terms with $a_j > 0$ and $a'_k > 0$ give no contributions to $\nu_F(s)$ and $N_+(x)$ for $s > 0$ and $x > 0$, while each $a_j < 0$ gives a contribution $|a_j|x + O(1)$ to $N_+(x)$ (poles regularly spaced at distances $1/|a_j|$), and similarly each $a'_k < 0$ gives a contribution $-|a'_k|x + O(1)$ to $N_+(x)$. Consequently, for $x > 0$, using Lemma 2.12,

$$N_+(x) = x \left(\sum_{j: a_j < 0} |a_j| - \sum_{k: a'_k < 0} |a'_k| \right) + O(1) = x \frac{1}{2}(\gamma - \gamma') + O(1),$$

and the result as $x \rightarrow \infty$ follows. The result as $x \rightarrow -\infty$ follows similarly, or by replacing $F(s)$ by $F(-s)$. \square

5. Asymptotics of moments or moment generating function

In this section we assume that $X > 0$ and $Y = \log X$ are random variables such that (1.1)–(1.3) hold (for $\rho_- < \operatorname{Re} s < \rho_+$ and $t \in \mathbb{R}$), i.e.

$$\mathbb{E} X^s = \mathbb{E} e^{sY} = F(s) = CD^s \frac{\prod_{j=1}^J \Gamma(a_j s + b_j)}{\prod_{k=1}^K \Gamma(a'_k s + b'_k)}, \quad (5.1)$$

and we write as above $D = e^d$. Recall the definitions (2.4)–(2.8).

We begin with asymptotics of F along the imaginary axis and close to it.

Theorem 5.1. *As $t \rightarrow \pm\infty$,*

$$|\mathbb{E} e^{itY}| = |F(it)| \sim C_1 |t|^\delta e^{-\frac{\pi}{2} \gamma |t|}. \quad (5.2)$$

Moreover, for any fixed real σ , and uniformly for σ in any bounded set,

$$|F(\sigma + it)| \sim e^{\varkappa\sigma} |t|^{\gamma'\sigma} |F(it)| \sim C_1 e^{\varkappa\sigma} |t|^{\delta + \gamma'\sigma} e^{-\frac{\pi}{2} \gamma |t|}. \quad (5.3)$$

Proof. It is an easy, and well-known, consequence of Stirling's formula, see e.g. (A.13), that for any complex constant c and all complex z in a sector $|\arg z| < \pi - \varepsilon$ (where $\varepsilon > 0$) with $|z|$ large enough, for example $|z| \geq 2|c|/\sin \varepsilon$,

$$\log \Gamma(z + c) - \log \Gamma(z) = c \log z + O(|z|^{-1}), \quad (5.4)$$

uniformly for c in any bounded set and such $|z|$.

If $a > 0$ and $b \in \mathbb{R}$, we thus have for real $t \rightarrow +\infty$, taking $z = iat$ in (5.4) and in Stirling's formula (A.12),

$$\begin{aligned} \log \Gamma(iat + b) &= \log \Gamma(iat) + b \log(iat) + O(t^{-1}) \\ &= (iat + b - \tfrac{1}{2}) \log(iat) - iat + \log \sqrt{2\pi} + O(t^{-1}) \\ &= (iat + b - \tfrac{1}{2})(\log(at) + i\pi/2) - iat + \log \sqrt{2\pi} + O(t^{-1}). \end{aligned}$$

Taking the real part, we find

$$\begin{aligned} \log |\Gamma(iat + b)| &= \operatorname{Re}(\log \Gamma(iat + b)) \\ &= -\frac{\pi}{2}at + (b - \tfrac{1}{2}) \log(at) + \log \sqrt{2\pi} + O(t^{-1}). \end{aligned}$$

Consequently, for $a > 0$,

$$|\Gamma(iat + b)| \sim \sqrt{2\pi} a^{b-1/2} t^{b-1/2} e^{-\frac{\pi}{2}at}, \quad t \rightarrow +\infty. \quad (5.5)$$

For general real a and t we thus have (by $\Gamma(\bar{z}) = \overline{\Gamma(z)}$)

$$|\Gamma(ait + b)| = |\Gamma(|a|it + b)| \sim \sqrt{2\pi} |a|^{b-1/2} |t|^{b-1/2} e^{-\frac{\pi}{2}|a||t|}, \quad t \rightarrow \pm\infty.$$

The result (5.2) follows by multiplying the various factors in $F(it)$ in (5.1), noting that $|D^{it}| = 1$.

For (5.3) we note that (5.4) implies

$$\begin{aligned} \log |\Gamma(a(\sigma + it) + b)| - \log |\Gamma(iat + b)| &= \operatorname{Re}(a\sigma \log(iat + b) + O(t^{-1})) \\ &= \sigma(a \log |a| + a \log |t| + O(t^{-1})), \end{aligned}$$

and the result follows by multiplying the various factors in $F(\sigma + it)/F(it)$. (Alternatively, at least for fixed σ , we may apply (5.2) with b_j replaced by $b_j + \sigma a_j$, b'_k replaced by $b'_k + \sigma a'_k$ and C replaced by $CD^\sigma = Ce^{d\sigma}$; note that the proof holds for any function F of this type, without assuming the existence of random variables X and Y .) \square

Proof of Proposition 2.7. By (5.2), the values of $F(it)$ determine γ , δ and C_1 . Further, choosing any fixed $\sigma > 0$ in (5.3), we see that F determines γ' and \varkappa too. \square

Corollary 5.2. *We have $\gamma \geq 0$. Further, if $\gamma = 0$, then $\delta \leq 0$.*

Proof. By letting $t \rightarrow \infty$ in (5.2), since $|\mathbb{E} e^{itY}| \leq 1$. \square

Remark 5.3. If $\gamma = 0$ and $\gamma' \neq 0$, then (5.3) implies a better bound for δ , viz. $\delta \leq -\max(\gamma'\rho_+, \gamma'\rho_-)$. (Example 3.16 is one such example, with $\gamma = 0$, $\gamma' = 1$ and strict inequality $\delta = -3/2 < -\gamma'\rho_+ = -1$.)

Theorem 5.4. If $\gamma > 0$, then X and Y are absolutely continuous, with continuous and infinitely differentiable density functions $f_X(x)$ on $(0, \infty)$ and $f_Y(y)$ on $(-\infty, \infty)$ given by

$$f_X(x) = \frac{1}{2\pi i} \int_{\sigma-i\infty}^{\sigma+i\infty} x^{-s-1} F(s) ds, \quad (5.6)$$

$$f_Y(y) = \frac{1}{2\pi i} \int_{\sigma-i\infty}^{\sigma+i\infty} e^{-ys} F(s) ds, \quad (5.7)$$

for any $\sigma \in (\rho_-, \rho_+)$.

Proof. By Theorem 5.1, the characteristic function $\mathbb{E} e^{itY} = F(it)$ is integrable, and thus Y has a continuous density f_Y obtained by Fourier inversion:

$$f_Y(y) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-ity} F(it) dt = \frac{1}{2\pi i} \int_{-i\infty}^{i\infty} e^{-sy} F(s) ds. \quad (5.8)$$

Since $Y = \log X$, X also is absolutely continuous, with the density function

$$f_X(x) = \frac{1}{x} f_Y(\log x) = \frac{1}{2\pi i} \int_{-i\infty}^{i\infty} x^{-s-1} F(s) ds. \quad (5.9)$$

(Alternatively and equivalently, $F(s)$ is the Mellin transform of f_X , and this is the Mellin inversion formula.)

Since Theorem 5.1 further implies that $|t|^N F(it)$ is integrable for every $N \geq 0$, f_Y and f_X are infinitely differentiable and we may differentiate (5.9) and (5.8) under the integral sign an arbitrary number of times.

The integrands in (5.9) and (5.8) are analytic in s for $\rho_- < \operatorname{Re} s < \rho_+$, and thus the estimate (5.3) implies that we can move the line of integration to any line $\operatorname{Re} s = \sigma$ with $\rho_- < \sigma < \rho_+$. \square

Remark 5.5. For X , we consider the density only for $x > 0$, and ‘infinitely differentiable’ here means on $(0, \infty)$. Continuity and differentiability of f_X at 0 will be considered in Theorem 6.11.

Remark 5.6. In the case $\gamma = 0$, the same argument shows that if $\delta < -1$, then X and Y have continuous density functions, which have at least $\lceil |\delta| \rceil - 2$ continuous derivatives. However, Example 3.4, where $\delta = -\beta$, shows that we in general do not have more derivatives. Similarly, Examples 3.3 and 3.4 show that we do not necessarily have continuous density functions for $\gamma = 0$ and $-1 \leq \delta \leq 0$. Example 10.1 gives an example with $\gamma = \delta = 0$ where the distribution is mixed with a point mass besides the absolutely continuous part.

Note that, by (5.2), $\gamma = \delta = 0$ if and only if $|\mathbb{E} X^{it}| = |\mathbb{E} e^{itY}|$ has a non-zero limit as $t \rightarrow \pm\infty$; by the Riemann–Lebesgue lemma, this implies that Y and X do not have absolutely continuous distributions.

We next consider asymptotics of F along the real axis, when possible.

Theorem 5.7. (i) *If $\rho_+ = \infty$, then for real $s \rightarrow +\infty$,*

$$\mathbb{E} X^s = \mathbb{E} e^{sY} = F(s) \sim C_1 s^\delta e^{\gamma s \log s + (\varkappa - \gamma)s}. \quad (5.10)$$

(ii) *If $\rho_- = \infty$, then for real $s \rightarrow -\infty$,*

$$\mathbb{E} X^s = \mathbb{E} e^{sY} = F(s) \sim C_1 |s|^\delta e^{\gamma |s| \log |s| + (\varkappa + \gamma)s}. \quad (5.11)$$

Proof. (i): By Theorem 4.1 and Proposition 2.7, we may assume that all $a_j, a'_k > 0$. We then argue as for Theorem 5.1. If $a > 0$ and $b \in \mathbb{R}$, then for real $s \rightarrow +\infty$,

$$\begin{aligned} \log \Gamma(as + b) &= \log \Gamma(as) + b \log(as) + O(s^{-1}) \\ &= (as + b - \tfrac{1}{2}) \log(as) - as + \log \sqrt{2\pi} + O(s^{-1}) \\ &= as \log s + (a \log a - a)s + (b - \tfrac{1}{2}) \log s \\ &\quad + (b - \tfrac{1}{2}) \log a + \log \sqrt{2\pi} + O(s^{-1}), \end{aligned} \quad (5.12)$$

and the result follows again by multiplying the factors.

(ii): This follows from (i) by replacing Y by $-Y$, see Remark 2.8:

$$\mathbb{E} e^{sY} = \mathbb{E} e^{|s|(-Y)} \sim C_1 |s|^\delta e^{\gamma |s| \log |s| + (\varkappa - \gamma)|s|}. \quad \square$$

If $\rho_+ < \infty$, then F has poles, and possibly zeros, on the positive real axis. Typically, there is an infinite number of such poles (but see Example 3.3 for a counter example), and then we cannot consider asymptotics for all $s \rightarrow +\infty$. However, we can restrict s to a subset of \mathbb{R} and obtain asymptotic results similar to Theorem 5.7 in this case too.

Lemma 5.8. *Given real a_j, b_j, a'_k, b'_k for $1 \leq j \leq J$ and $1 \leq k \leq K$, with $a_j, a'_k \neq 0$, there exists a closed set $E \subset \mathbb{R}$ and a constant $\xi > 0$ such that $E \cap I$ has measure greater than $1/2$ for every interval I of length 1, and $|\sin(\pi(a_j s + b_j))| \geq \xi$ and $|\sin(\pi(a'_k s + b'_k))| \geq \xi$ for every j and k and all $s \in E$.*

Proof. Let N be the set of all (real) s such that $a_j s + b_j \in \mathbb{Z}$ for some j or $a'_k s + b'_k \in \mathbb{Z}$ for some k . There exists a constant M such that no interval of length 1 contains more than M points of N . (For example, $M = J + K + \sum_j |a_j|^{-1} + \sum_k |a'_k|^{-1}$.) It follows that $E := \{x : |x - s| \geq 1/(2M + 3) \text{ for all } s \in N\}$ satisfies the properties, for some $\xi > 0$. \square

In the sequel we let E denote this set, defined for a given representation (5.1) of $F(s)$. By considering only $s \in E$, we can extend Theorem 5.7 to arbitrary F .

Theorem 5.9. *For real $s \rightarrow \pm\infty$ with $s \in E$,*

$$|F(s)| = |s|^\delta e^{\gamma' s \log |s| + (\varkappa - \gamma')s + O(1)}. \quad (5.13)$$

Proof. If $a < 0$ and $b \in \mathbb{R}$, then for real $s \rightarrow +\infty$, by (A.6) and (5.12),

$$\begin{aligned} \log |\Gamma(as + b)| &= -\log |\Gamma(|a|s - b + 1)| + \log \pi - \log |\sin(\pi(as + b))| \\ &= -|a|s \log s - (|a| \log |a| - |a|)s - \left(\frac{1}{2} - b\right) \log s - \left(\frac{1}{2} - b\right) \log |a| \\ &\quad + \log \sqrt{\pi/2} - \log |\sin(\pi(as + b))| + O(s^{-1}), \\ &= as \log s + (a \log |a| - a)s + \left(b - \frac{1}{2}\right) \log s - \log |\sin(\pi(as + b))| + O(1). \end{aligned} \quad (5.14)$$

If (a, b) is some (a_j, b_j) or (a'_k, b'_k) with $a < 0$, we thus have by Lemma 5.8, for $s \in E$ with $s \rightarrow +\infty$,

$$\log |\Gamma(as + b)| = as \log s + (a \log |a| - a)s + \left(b - \frac{1}{2}\right) \log s + O(1). \quad (5.15)$$

By (5.12), (5.15) holds also for $a > 0$ (and all $s \rightarrow +\infty$).

Further, replacing s by $-s$ and a by $-a$ in (5.15), we see that if (a, b) is some (a_j, b_j) or (a'_k, b'_k) , then for $s \in E$ with $s \rightarrow -\infty$,

$$\log |\Gamma(as + b)| = as \log |s| + (a \log |a| - a)s + \left(b - \frac{1}{2}\right) \log |s| + O(1). \quad (5.16)$$

Thus (5.16) holds for all such (a, b) and $s \rightarrow \pm\infty$ with $s \in E$, and the result follows from (5.1). \square

Note that if $\rho_+ = \infty$, then $\gamma' = \gamma$, while if $\rho_- = \infty$, then $\gamma' = -\gamma$ by (2.4)–(2.5) together with Theorem 4.1 and Proposition 2.7; hence the exponents in Theorems 5.7 and 5.9 agree (as they must).

For complex arguments, we will use the following estimate.

Lemma 5.10. *Let $\Psi(\sigma, t) := \int_0^t \arctan(u/\sigma) du$ for $t \geq 0$. Then, for $\sigma > 0$ with $\sigma \in E$ and all real t ,*

$$\frac{|F(\sigma + it)|}{|F(\sigma)|} = \exp\left(-\frac{\pi}{2}(\gamma - \gamma')|t| - \gamma' \Psi(\sigma, |t|) + O(1 + |t|\sigma^{-1})\right).$$

Proof. We may assume $t > 0$. Consider first a factor $\Gamma(as + b)$ with $a > 0$ and $s = \sigma + it$, $\sigma > 0$. (We may assume that σ is large so that $a\sigma + b > 0$, e.g. by using (5.3) for small σ .) By (A.13),

$$\begin{aligned} \frac{d}{dt} \log |\Gamma(a(\sigma + it) + b)| &= \operatorname{Re} \frac{d}{dt} \log \Gamma(a(\sigma + it) + b) \\ &= \operatorname{Re}(ia \log(a(\sigma + it) + b)) + O(|\sigma|^{-1}) \\ &= \operatorname{Re}(ia \log(a(\sigma + it))) + O(|\sigma|^{-1}) \\ &= -a \operatorname{Im}(\log(\sigma + it)) + O(|\sigma|^{-1}) \\ &= -a \arctan(t/\sigma) + O(|\sigma|^{-1}). \end{aligned} \quad (5.17)$$

Consequently, integrating from 0 to t ,

$$\log |\Gamma(a(\sigma + it) + b)| - \log |\Gamma(a\sigma + b)| = -a\Psi(\sigma, t) + O(t\sigma^{-1}). \quad (5.18)$$

If $a < 0$, we argue as in the proof of Theorem 5.9 and have by (A.6)

$$\begin{aligned} \log |\Gamma(a(\sigma + it) + b)| \\ = -\log |\Gamma(|a|(\sigma + it) + 1 - b)| - \log |\sin(\pi(a(\sigma + it) + b))| + \log \pi. \end{aligned}$$

If further $(a, b) = (a_j, b_j)$ or (a'_k, b'_k) for some j or k , and $\sigma \in E$, then $\log |\sin(\pi(a(\sigma + it) + b))| = \pi|a|t + O(1)$, and it follows, using (5.18) with $|a|$ instead of a , that

$$\begin{aligned} \log |\Gamma(a(\sigma + it) + b)| - \log |\Gamma(a\sigma + b)| &= |a|\Psi(\sigma, t) - \pi|a|t + O(1 + t\sigma^{-1}) \\ &= -a\Psi(\sigma, t) + \pi at + O(1 + t\sigma^{-1}). \end{aligned}$$

The result follows by multiplying the factors in F , using Lemma 2.12. \square

6. Asymptotics of density function

We continue to assume that X and $Y = \log X$ are random variables such that (1.1)–(1.3) hold; as above we write $\mathbb{E} X^s = e^{sY} = F(s)$. We assume $\gamma > 0$, so that density functions of X and Y exist by Theorem 5.4, and consider asymptotics of the density function $f_X(x)$ as $x \rightarrow 0$ or $x \rightarrow \infty$, or equivalently of $f_Y(y)$ as $y \rightarrow -\infty$ or $y \rightarrow \infty$. By symmetry it suffices to consider one side, and we concentrate on $x \rightarrow \infty$, but for convenience in applications we write most results for both sides and for both X and Y .

We consider first $x \rightarrow \infty$ ($y \rightarrow \infty$) and begin with the case $\rho_+ = \infty$, when X has moments of all (positive) orders and f_X decreases rapidly (as we will see in detail soon). We use the saddle point method, see e.g. Flajolet and Sedgewick [12, Chapter VIII], in a standard way.

Theorem 6.1. *Suppose that $\rho_+ = \infty$ and $\gamma > 0$. Then*

$$\begin{aligned} f_X(x) &\sim C_2 x^{c_1-1} e^{-c_2 x^{1/\gamma}}, & x \rightarrow \infty, \\ f_Y(y) &\sim \frac{C_1}{\sqrt{2\pi\gamma}} e^{c_1(y-\varkappa) - \gamma e^{(y-\varkappa)/\gamma}}, & y \rightarrow \infty, \end{aligned}$$

where

$$\begin{aligned} c_1 &:= (\delta + 1/2)/\gamma, \\ c_2 &:= \gamma e^{-\varkappa/\gamma}, \\ C_2 &:= \frac{C_1}{\sqrt{2\pi\gamma}} e^{-c_1 \varkappa}. \end{aligned}$$

Proof. By Theorem 4.1 and Proposition 2.7, we may assume that all $a_j, a'_k, b_j, b'_k > 0$. We will use (5.6), which now is valid for all $x > 0$ and $\sigma > 0$.

By (A.13) and (A.14), for $a, b > 0$ and $\operatorname{Re} s > 0$,

$$(\log \Gamma(as + b))' = a \log(as + b) + O(|s|^{-1}) = a \log(as) + O(|s|^{-1}),$$

$$(\log \Gamma(as + b))'' = \frac{a}{s} + O(|s|^{-2}).$$

Consequently, writing

$$f(s) := \log F(s) = \log C + ds + \sum_{j=1}^J \log \Gamma(a_j s + b_j) - \sum_{k=1}^K \log \Gamma(a'_k s + b'_k),$$

we have for $\operatorname{Re} s > 0$,

$$\begin{aligned} f'(s) &= d + \sum_{j=1}^J (a_j \log a_j + a_j \log s) - \sum_{k=1}^K (a'_k \log a'_k + a'_k \log s) + O(|s|^{-1}) \\ &= \varkappa + \gamma \log s + O(|s|^{-1}), \\ f''(s) &= \frac{\gamma}{s} + O(|s|^{-2}). \end{aligned}$$

Fix $x > 0$ and let $G(s) := x^{-s-1} F(s)$ and $g(s) := \log G(s) = -(s-1) \log x + f(s)$. Then

$$g'(s) = f'(s) - \log x = \gamma \log s + \varkappa - \log x + O(|s|^{-1}). \quad (6.1)$$

We choose (for x large) $\sigma = e^{(\log x - \varkappa)/\gamma}$, so $\gamma \log \sigma = \log x - \varkappa$ and $g'(\sigma) = O(\sigma^{-1})$; thus σ is an approximate saddle point of $G(s)$. Note that $\sigma \rightarrow \infty$ as $x \rightarrow \infty$, so $\sigma^{-1} \rightarrow 0$. Since $g''(s) = f''(s)$, we further have, as $x \rightarrow \infty$,

$$g''(\sigma) = \gamma \sigma^{-1} + O(\sigma^{-2}) \sim \gamma \sigma^{-1}. \quad (6.2)$$

Further, on the line $\operatorname{Re} s = \sigma$,

$$g''(\sigma + it) = f''(\sigma + it) = \frac{\gamma}{\sigma + it} + O(\sigma^{-2}) = \frac{\gamma}{\sigma} + O((1 + |t|)\sigma^{-2}). \quad (6.3)$$

Consequently, Taylor's formula yields

$$g(\sigma + it) = g(\sigma) + O(|t|\sigma^{-1}) - \frac{\gamma}{2\sigma} t^2 + O((|t|^2 + |t|^3)\sigma^{-2}) \quad (6.4)$$

and, uniformly for $|t| \leq \sigma^{0.6}$,

$$G(\sigma + it) = G(\sigma) e^{-\gamma t^2/2\sigma + o(1)}. \quad (6.5)$$

For larger $|t|$, we have a rapid decay, for example by Lemma 5.10 which yields, for large σ and $|t| \geq \sigma^{0.6}$, recalling that now $\gamma' = \gamma$,

$$\begin{aligned} \frac{|G(\sigma + it)|}{|G(\sigma)|} &= \frac{|F(\sigma + it)|}{|F(\sigma)|} = \exp\left(-\gamma \Psi(\sigma, |t|) + O(1 + |t|\sigma^{-1})\right) \\ &\leq \exp\left(-c \min(|t|, |t|^2/\sigma)\right). \end{aligned} \quad (6.6)$$

for some $c > 0$. It follows from (6.5) and (6.6) that (5.6) yields, using Theorem 5.7 and the choice of $\sigma = e^{(\log x - \varkappa)/\gamma} = e^{-\varkappa/\gamma} x^{1/\gamma}$,

$$\begin{aligned} f_X(x) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} G(\sigma + it) dt \sim \frac{G(\sigma)}{2\pi} \int_{-\infty}^{\infty} e^{-\gamma t^2/2\sigma} dt = \frac{G(\sigma)\sqrt{\sigma}}{\sqrt{2\pi\gamma}} \\ &= \frac{x^{-\sigma-1} F(\sigma)\sigma^{1/2}}{\sqrt{2\pi\gamma}} \sim \frac{C_1}{\sqrt{2\pi\gamma}} \sigma^{\delta+1/2} x^{-1} e^{\gamma\sigma \log \sigma + (\varkappa-\gamma)\sigma - \sigma \log x} \\ &= \frac{C_1}{\sqrt{2\pi\gamma}} e^{(\delta+1/2)(\log x - \varkappa)/\gamma} x^{-1} e^{-\gamma\sigma} = C_2 x^{c_1-1} e^{-c_2 x^{1/\gamma}}. \end{aligned}$$

The result for f_Y follows similarly from (5.7), or simpler by $f_Y(y) = e^y f_X(e^y)$. \square

Remark 6.2. The derivative $f'_X(x)$ and higher derivatives $f_X^{(n)}(x)$ can be obtained by repeated differentiation of (5.6) under the integral sign, which multiplies the integrand by a factor $(-s-1)\cdots(-s-n)x^{-n}$. The argument above, including the estimates (6.5) and (6.6), applies to this integral as well and shows that, for any $n \geq 0$,

$$f_X^{(n)}(x) \sim \sigma^n x^{-n} G(\sigma)\sqrt{\sigma}/\sqrt{2\pi\gamma} \sim \sigma^n x^{-n} f_X(x) = (c_2/\gamma)^n x^{n(1/\gamma-1)} f_X(x).$$

In particular, every derivative of f_X tends to 0 rapidly (faster than any power of x) as $x \rightarrow \infty$.

Remark 6.3. The saddle-point method yields also more precise asymptotics including higher-order terms by refining the estimates around $s = \sigma$ in the proof above, see e.g. Flajolet and Sedgewick [12, Section VIII.3]; we leave the details to the reader. This yields an asymptotic expansion in powers of σ^{-1} , with σ as in the proof above, i.e., in powers of $x^{-1/\gamma}$. See Remark 1.10 for an example of such an expansion (there obtained from a known result rather than by performing the calculations).

We continue with the case $\rho_+ < \infty$, when $F(s)$ has a pole at ρ_+ of order $\nu_F(\rho_+) \geq 1$. (Recall the notion ν_F from (4.2).) We denote the coefficients of the singular part of the Laurent expansion of F at a point s_0 by $c_\ell(s_0)$:

$$F(s) = \sum_{\ell=1}^{\nu_F(s_0)} c_\ell(s_0)(s-s_0)^{-\ell} + O(1) \quad \text{as } s \rightarrow s_0. \quad (6.7)$$

In particular, $c_1(s_0)$ is the residue $\text{Res}_{s_0} F$.

We have the following standard result by Mellin inversion, see [11].

Theorem 6.4. *Suppose that $\rho_+ < \infty$ and $\gamma > 0$.*

(i) *As $x \rightarrow \infty$, for some $\eta > 0$,*

$$f_X(x) = x^{-\rho_+-1} \sum_{\ell=0}^{\nu_F(\rho_+)-1} \frac{(-1)^{\ell+1} c_{\ell+1}(\rho_+)}{\ell!} \log^\ell x + O(x^{-\rho_+-1-\eta})$$

In particular, with $\nu := \nu_F(\rho_+) \geq 1$,

$$f_X(x) \sim \frac{(-1)^\nu c_\nu(\rho_+)}{(\nu-1)!} x^{-\rho_+-1} \log^{\nu-1} x.$$

If ρ_+ is a simple pole of F , i.e. $\nu = 1$, this can be written

$$f_X(x) \sim -\operatorname{Res}_{\rho_+}(F) x^{-\rho_+-1}.$$

(ii) More precisely, there is an asymptotic expansion, for any fixed $\sigma > 0$,

$$f_X(x) = \sum_{0 < \rho \leq \sigma} \sum_{\ell=0}^{\nu_F(\rho)-1} \frac{(-1)^{\ell+1} c_{\ell+1}(\rho)}{\ell!} x^{-\rho-1} \log^\ell x + O(x^{-\sigma-1}),$$

summing over all poles ρ of F in $(0, \sigma]$. (The inner sum vanishes unless ρ is a pole, so formally we may sum over all ρ .)

Corresponding asymptotics for $f_Y(y) = e^y f_X(e^y)$ are obtained by replacing each x^{-r-1} by e^{-ry} and $\log^\ell x$ by y^ℓ .

Proof. As said above, this is a standard result, and we refer to [11] for details, but for completeness and later use we give the simple proof.

It suffices to prove (ii), since (i) follows by taking $\sigma = \rho_+ + \eta$. We may assume that σ is not a pole of F (otherwise we increase σ a little). We start with (5.6), where we integrate over a line with $\operatorname{Re} s \in (\rho_-, \rho_+)$. We may, using Theorem 5.1, shift the line to $\operatorname{Re} s = \sigma > \rho_+$ too, but then we have to subtract the residues of the traversed poles. Thus

$$f_X(x) = \frac{1}{2\pi i} \int_{\sigma-i\infty}^{\sigma+i\infty} x^{-s-1} F(s) ds - \sum_{0 < \rho \leq \sigma} \operatorname{Res}_{s=\rho}(x^{-s-1} F(s)), \quad (6.8)$$

and the result follows by computing the residues, using (6.7) and $x^{-s-1} = x^{-\rho-1} \sum_{\ell=0}^{\infty} (-\log x)^\ell (s-\rho)^\ell / \ell!$, and noting that, by Theorem 5.1 again,

$$\left| \int_{\sigma-i\infty}^{\sigma+i\infty} x^{-s-1} F(s) ds \right| \leq x^{-\sigma-1} \int_{\sigma-i\infty}^{\sigma+i\infty} |F(s)| ds = O(x^{-\sigma-1}). \quad \square$$

In Theorem 6.4(ii) we have an asymptotic expansion, valid for fixed σ as $x \rightarrow \infty$. It is natural to ask whether this asymptotic expansion actually yields a series representation for $f_X(x)$, i.e., whether we can let $\sigma \rightarrow \infty$ for fixed x (with the error term tending to 0) so that $f_X(x)$ is represented as a convergent series. This is possible sometimes, but not always. In fact, the following theorem shows that this is possible exactly when $\gamma' < 0$, at least provided that there is an infinite number of poles $\rho > 0$ and that these are simple.

Theorem 6.5. *Suppose that $\gamma > 0$.*

(i) *If $\gamma' < 0$, then, for all $x > 0$,*

$$f_X(x) = \sum_{\rho > 0} \sum_{\ell=0}^{\nu_F(\rho)-1} \frac{(-1)^{\ell+1} c_{\ell+1}(\rho)}{\ell!} x^{-\rho-1} \log^\ell x, \quad (6.9)$$

summing over all poles $\rho > 0$ of F . In particular, if F has only simple poles,

$$f_X(x) = \sum_{\rho>0} -\operatorname{Res}_\rho(F)x^{-\rho-1}. \quad (6.10)$$

(ii) If $\gamma' > 0$ and there is an infinite number of poles $\rho > 0$ of F , all simple, then the sum (6.10) diverges for all $x > 0$.

(iii) If $\gamma' = 0$, then (6.9) holds for $x > e^\varkappa$; hence (6.10) holds for $x > e^\varkappa$ provided all poles are simple. However, at least provided that there is an infinite number of poles $\rho > 0$ of F and all such poles are simple, the sum (6.10) diverges for $0 < x < e^\varkappa$.

Corresponding results for $f_Y(y)$ are obtained by replacing $x^{-\rho-1}$ by $e^{-\rho y}$ and $\log^\ell x$ by y^ℓ . The cut-offs in (iii) become $y > \varkappa$ and $y < \varkappa$.

Proof. (i) and (iii) (convergence): We use again (6.8), and have to show that the integral tends to 0 as $\sigma \rightarrow \infty$ for every fixed $x > 0$. We use Lemma 5.10, and note that $0 \leq \Psi(\sigma, t) \leq \frac{\pi}{2}t$ for $t \geq 0$, and thus, because $\gamma' \leq 0$, for $\sigma > 0$ with $\sigma \in E$,

$$\begin{aligned} \frac{|F(\sigma + it)|}{|F(\sigma)|} &\leq \exp\left(-\frac{\pi}{2}(\gamma - \gamma')|t| - \gamma' \frac{\pi}{2}|t| + O(1 + |t|\sigma^{-1})\right) \\ &= \exp\left(-\frac{\pi}{2}\gamma|t| + O(1 + |t|\sigma^{-1})\right). \end{aligned}$$

If $\sigma \in E$ is large enough we thus have $|F(\sigma + it)| = O(e^{-\gamma|t|}|F(\sigma)|)$ for all real t , and hence, because $\gamma > 0$,

$$\int_{\sigma-i\infty}^{\sigma+i\infty} |x^{-s-1}F(s)| |ds| = O(x^{-\sigma-1}|F(\sigma)|) \int_{-\infty}^{\infty} e^{-\gamma|t|} dt = O(x^{-\sigma-1}|F(\sigma)|).$$

If $\gamma' < 0$, then this is by Theorem 5.9 $o(1)$ as $\sigma \rightarrow \infty$ for any fixed $x > 0$, which shows (i).

If $\gamma' = 0$, then Theorem 5.9 yields, for a fixed $x > 0$, $x^{-\sigma}F(\sigma) = O(\sigma^\delta e^{(\varkappa - \log x)\sigma})$, which is $o(1)$ for $x > e^\varkappa$, showing the positive part of (iii).

(ii) and (iii) (divergence): Let $\rho > 0$ be a pole of F that is not too close to a zero or another pole, meaning that the distance to every zero or other pole is at least some small constant $\xi > 0$. (This is true for all poles if all a_j, a'_k are commensurable and ξ is small enough; in general it is true for a large fraction of the poles, and certainly an infinite number of them.) A simple modification of the proof of Theorem 5.9 then yields the same estimate as there for the residue at ρ :

$$|\operatorname{Res}_\rho(F)| = \rho^\delta e^{\gamma' \rho \log \rho + (\varkappa - \gamma')\rho + O(1)},$$

and thus, for every fixed x ,

$$|x^{-\rho-1} \operatorname{Res}_\rho(F)| = \rho^\delta e^{\gamma' \rho \log \rho + (\varkappa - \gamma' - \log x)\rho + O(1)}.$$

Letting $\rho \rightarrow \infty$, we see that the terms of (6.10) are unbounded if $\gamma' > 0$ or $\gamma' = 0$ and $\varkappa > \log x$; hence the sum diverges. \square

Remark 6.6. To show divergence in (ii) and (iii), we assumed for simplicity that F has only simple poles on the positive axis; we conjecture that, more generally, (6.9) diverges also without this restriction.

To show divergence we also assumed that F has an infinite number of positive poles; this is, on the contrary, obviously necessary for divergence, since otherwise the sums (6.9) and (6.10) are finite. However, if F has only a finite number of positive poles, then the sum in (6.9) or (6.10) is not integrable, since it is $\sim cx^{-\rho-1} \log^\ell x$ as $x \rightarrow 0$, where $\rho > 0$ is the largest pole of F and $c \neq 0$, $\ell = \nu_F(\rho) - 1$; hence the sum cannot equal $f_X(x)$ for all $x > 0$. Example 10.2 yields an example where the sum does not equal $f_X(x)$ for any $x > 0$ (although the difference tends to 0 rapidly as $x \rightarrow \infty$ by Theorem 6.4).

In this connection, note that if $\gamma > \gamma'$, then there is an infinite number of poles in $(0, \infty)$ by Proposition 4.6.

We now consider $x \rightarrow 0$ and $y \rightarrow -\infty$. We obtain the following by the same methods as above (now moving the line of integration towards $-\infty$), or more simply by applying the results above to X^{-1} and $-Y$; this replaces $F(s)$ by $F(-s)$ and the Laurent coefficients $c_\ell(s_0)$ by $(-1)^\ell c_\ell(-s_0)$.

Theorem 6.7. *Suppose that $\rho_- = -\infty$ and $\gamma > 0$. Then*

$$\begin{aligned} f_X(x) &\sim C_3 x^{-c_1-1} e^{-c_3 x^{-1/\gamma}}, & x \rightarrow 0, \\ f_Y(y) &\sim \frac{C_1}{\sqrt{2\pi\gamma}} e^{-c_1(y-\varkappa) - \gamma e^{-(y-\varkappa)/\gamma}}, & y \rightarrow -\infty, \end{aligned}$$

where

$$\begin{aligned} c_1 &:= (\delta + 1/2)/\gamma, \\ c_3 &:= \gamma e^{\varkappa/\gamma}, \\ C_3 &:= \frac{C_1}{\sqrt{2\pi\gamma}} e^{c_1 \varkappa}. \end{aligned}$$

Theorem 6.8. *Suppose that $\rho_- > -\infty$ and $\gamma > 0$.*

(i) *As $x \searrow 0$, for some $\eta > 0$,*

$$f_X(x) = x^{|\rho_-|-1} \sum_{\ell=0}^{\nu_F(\rho_-)-1} \frac{c_{\ell+1}(\rho_-)}{\ell!} \log^\ell(1/x) + O(x^{|\rho_-|-1+\eta})$$

In particular, with $\nu := \nu_F(\rho_-) \geq 1$,

$$f_X(x) \sim \frac{c_\nu(\rho_-)}{(\nu-1)!} x^{|\rho_-|-1} \log^{\nu-1}(1/x).$$

If ρ_- is a simple pole of F , i.e. $\nu = 1$, this can be written

$$f_X(x) \sim \operatorname{Res}_{\rho_-}(F) x^{|\rho_-|-1}.$$

(ii) More precisely, there is an asymptotic expansion, for any fixed $\sigma > 0$,

$$f_X(x) = \sum_{0 > \rho \geq -\sigma} \sum_{\ell=0}^{\nu_F(\rho)-1} \frac{c_{\ell+1}(\rho)}{\ell!} x^{|\rho|-1} \log^\ell(1/x) + O(x^{\sigma-1}),$$

summing over all poles ρ of F in $[-\sigma, 0)$.

Corresponding asymptotics for $f_Y(y) = e^y f_X(e^y)$ are obtained by replacing each x^{r-1} by e^{ry} and $\log^\ell(1/x)$ by $(-y)^\ell$.

Theorem 6.9. Suppose that $\gamma > 0$.

(i) If $\gamma' > 0$, then, for all $x > 0$,

$$f_X(x) = \sum_{\rho < 0} \sum_{\ell=0}^{\nu_F(\rho)-1} \frac{c_{\ell+1}(\rho)}{\ell!} x^{|\rho|-1} \log^\ell(1/x), \quad (6.11)$$

summing over all poles $\rho < 0$ of F . In particular, if F has only simple poles,

$$f_X(x) = \sum_{\rho < 0} \text{Res}_\rho(F) x^{|\rho|-1}. \quad (6.12)$$

(ii) If $\gamma' < 0$ and there is an infinite number of poles $\rho < 0$ of F , all simple, then the sum (6.10) diverges for all $x > 0$.

(iii) If $\gamma' = 0$, then (6.11) holds for $0 < x < e^\varkappa$; hence (6.12) holds for $0 < x < e^\varkappa$ provided all poles are simple. However, at least provided that there is an infinite number of poles $\rho < 0$ of F and all such poles are simple, the sum (6.12) diverges for $x > e^\varkappa$.

Corresponding results for $f_Y(y)$ are obtained by replacing $x^{|\rho|-1}$ by $e^{|\rho|y}$ and $\log^\ell(1/x)$ by $(-y)^\ell$. The cut-offs in (iii) become $y < \varkappa$ and $y > \varkappa$.

Theorems 6.5 and 6.9 say that (at least if $\gamma > 0$), $f_X(x)$ has a series expansion in positive (but not necessarily integer) powers of x if $\gamma' > 0$, and a series expansion in negative (but not necessarily integer) powers of x if $\gamma' < 0$, in both cases allowing for terms with logarithmic factors too; if $\gamma' = 0$ one expansion holds for $0 < x < e^\varkappa$ and the other for $x > e^\varkappa$.

Remark 6.10. Suppose that all a_j, a'_k are commensurable; then $F(s)$ may as in Section 4 (see the proof of Lemma 4.2) be rewritten with all $a_j, a'_k = \pm r$, for some real $r > 0$. The poles of F in $(-\infty, 0)$ then form one or several arithmetic series $\{s_j - n/r\}$ with gap $1/r$, possibly apart from a finite number of other poles. If further all poles are simple, then the residue at such a pole $s_l - n/r$ is of the form $(C/r)(-D)^n (n!)^{-1} \prod_{j \neq l} \Gamma(n + c_j) / \prod_k (\Gamma(n + c'_k))$, and the contribution to (6.12) from this series of poles is a (generalized) hypergeometric series with argument $-Dx^{1/r}$, times a constant and a power of x . Consequently, if further $\gamma, \gamma' > 0$, then the density function may be expressed using one or several hypergeometric functions. Typical examples are given in Theorems 1.8 and 1.13.

As a corollary, we get results on continuity and differentiability at 0.

Theorem 6.11. *Suppose that $\gamma > 0$.*

(i) *The density f_X is continuous at 0, and thus everywhere on \mathbb{R} , if and only if $\rho_- < -1$.*

(ii) *The density f_X has a finite jump at 0 if and only if $\rho_- = -1$ and this is a simple pole of F . In this case $f_X(0+) = \text{Res}_{\rho_-}(F)$.*

(iii) *The density f_X is infinitely differentiable on \mathbb{R} if and only if $\rho_- = -\infty$.*

Proof. Note that f_X is infinitely differentiable on $(0, \infty)$ by Theorem 5.4, as well as, trivially, on $(-\infty, 0)$ where it vanishes.

Parts (i) and (ii) follow immediately from Theorems 6.7 and 6.8.

If f_X is infinitely differentiable at 0, then every derivative $f_X^{(n)}(0) = 0$ because f_X vanishes on $(-\infty, 0)$. Hence a Taylor expansion shows that $f_X(x) = O(x^N)$ as $x \rightarrow 0$ for every integer N . If ρ_- were finite, this would contradict Theorem 6.8; hence $\rho_- = -\infty$.

Conversely, if $\rho_- = -\infty$, then Theorem 6.7 shows that $f_X(x)$ tends to 0 rapidly as $x \searrow 0$. Moreover, by Remark 6.2 and the usual change of variables $x \mapsto 1/x$, the same holds for each derivative $f_X^{(n)}(x)$. It follows, by induction, that each derivative $f_X^{(n)}(x)$ exists also at $x = 0$ with $f_X^{(n)}(0) = 0$. Hence f_X is infinitely differentiable. \square

Remark 6.12. More generally, f_X has n continuous derivatives (at 0) if and only if $\rho_- < -n - 1$; we omit the details.

Remark 6.13. We have in this section assumed $\gamma > 0$ in order to have good estimates of $F(s)$ as $|\text{Im } s| \rightarrow \infty$ in the proofs. It seems likely that the results can be extended to the case $\gamma = 0$ too, under suitable conditions, but we have not pursued this beyond noting that the results above hold also for the examples in Section 3 with $\gamma = 0$, with one interesting exception for Example 3.16.

For example, the uniform distribution in Example 3.3 has $\rho_+ = \infty$, $\rho_- = -1$, $\gamma = \gamma' = 0$, $\varkappa = 0$ and a single, simple pole at -1 ; the series in (6.10) is thus 0 and the series in (6.12) is 1, so (6.10) holds for $x > e^{\varkappa} = 1$ and (6.12) holds for $x < e^{\varkappa} = 1$. The asymptotic result in Theorem 6.1 is not directly applicable, since the exponent $1/\gamma = \infty$, but it can be interpreted as $f_X(x) = 0$ for large x , which is correct.

The same holds, *mutatis mutandis*, for the Pareto distribution in Example 3.13, where now there is a single pole at $\alpha > 0$ and the density vanishes on $(0, 1)$.

Similarly, for the Beta distribution $B(\alpha, \beta)$ in Example 3.4, the series in (6.12) is

$$\sum_{n=0}^{\infty} \frac{(-1)^n}{n!} \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta - n)} x^{n+\alpha-1} = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} \sum_{n=0}^{\infty} \binom{\beta-1}{n} (-x)^n,$$

which for $x < 1 = e^{\varkappa}$ converges to the density $f(x) = (\Gamma(\alpha + \beta)/\Gamma(\alpha)\Gamma(\beta)) \cdot x^{\alpha-1}(1-x)^{\beta-1}$; for $x > 1$ this series diverges unless β is an integer (when the series is finite but does not yield $f(x) = 0$ for $x > 1$), while (6.10) holds trivially.

The $\Gamma(n)$ distribution in Example 3.18 is an example with a multiple pole. There is a single pole at 1, with $c_n(1) = (-1)^n$ and $c_\ell(1) = 0$, $\ell \neq n$. Hence the sum in (6.9) is, rewritten for f_Y as stated in Theorem 6.5, $(1/(n-1)!)y^{n-1}e^{-y}$, which is the correct density for $y > \varkappa = 0$.

The Fejér distribution in Example 3.16 has $\gamma = 0$ and $\rho_+ = 1$; further, 1 is a simple pole of $F(s) = \mathbb{E}|X|^s$, with residue $-2/\pi$. Hence, Theorem 6.4, if it applied, would give $f_{|X|}(x) \sim \frac{2}{\pi}x^{-2}$ as $x \rightarrow \infty$, which is false since $f_{|X|}(x) = \frac{2}{\pi}x^{-2}(1 - \cos x)$, see (3.31); it gives the correct average behaviour but misses the oscillations. The reason for this interesting failure of Theorem 6.4 and its proof is the slow decay of $F(\sigma + it)$ as $t \rightarrow \infty$, especially when σ is close to $\rho_+ = 1$ or larger, see Theorem 5.1. On the other side, note that as $x \searrow 0$, Theorem 6.8 correctly yields $f_{|X|}(x) \sim \frac{1}{\pi}$. In fact, the proof of Theorem 6.4 and Theorem 6.8 works for $x \rightarrow 0$ but not for $x \rightarrow \infty$ due to the fact that Theorem 5.1 yields better estimates for $\sigma < 0$ than for $\sigma > 0$ because $\gamma' > 0$. It would be interesting to find a Mellin transform argument yielding the correct oscillatory asymptotics as $x \rightarrow \infty$.

We give some examples of applying the theorems above to the distributions in Section 3. This is mainly as an illustration of the theorems; we cannot expect to obtain any new results for these classical distributions. Other applications of the theorems are given in Theorems 1.8, 1.9, 1.13, 1.14, 9.1, 9.3, 9.6, 9.7.

Example 6.14. For the exponential distribution in Example 3.2, Theorem 6.1 yields $f(x) \sim e^{-x}$ as $x \rightarrow \infty$ (this is actually an identity for all $x > 0$) and Theorem 6.9 yields, since the poles are at $-n - 1$, with $n = 0, 1, \dots$, $f(x) = \sum_{n=0}^{\infty} (-1)^n x^n / n!$, $x > 0$, again a trival result.

Example 6.15. Similarly, for the Gumbel distribution in Example 3.19, $f(y) = e^{-y-e^{-y}}$ and the asymptotic formula in Theorem 6.7 is actually an equality for all real y .

Example 6.16. Consider the stable distribution in Example 3.10 with $0 < \alpha < 1$. Since $\gamma > 0 > \gamma'$, we can apply Theorem 6.5(i). By (3.16), $F(s) := \mathbb{E} S_\alpha^s$ has simple poles at $s = n\alpha$, $n = 1, 2, \dots$, and, using (A.6),

$$\begin{aligned} \operatorname{Res}_{n\alpha}(F) &= \frac{-\alpha \operatorname{Res}_{1-n}(\Gamma)}{\Gamma(1-n\alpha)} = \frac{(-1)^n \alpha}{(n-1)! \Gamma(1-n\alpha)} \\ &= (-1)^n \frac{\alpha \Gamma(n\alpha) \sin(\pi n\alpha)}{(n-1)! \pi} = (-1)^n \frac{\Gamma(n\alpha+1) \sin(\pi n\alpha)}{\pi n!}. \end{aligned} \quad (6.13)$$

(This includes the case when $n\alpha$ is an integer, in which case $n\alpha$ is not a pole because of cancellation; (6.13) then correctly yields $\operatorname{Res}_{n\alpha}(F) = 0$.) We thus obtain by (6.10)

$$f_{S_\alpha}(x) = \sum_{n=1}^{\infty} (-1)^{n+1} \frac{\Gamma(n\alpha+1) \sin(\pi n\alpha)}{\pi n!} x^{-n\alpha-1}. \quad (6.14)$$

This is the well-known formula for the stable density, see Feller [9, XVII.(6.8)] (with $\gamma = -\alpha$ for the positive case studied here).

In particular, as $x \rightarrow \infty$, (6.14) or Theorem 6.4 (with $\rho_+ = \alpha$) yields

$$f_{S_\alpha}(x) \sim -\operatorname{Res}_\alpha(F)x^{-\alpha-1} = \frac{\alpha}{\Gamma(1-\alpha)}x^{-\alpha-1}, \quad x \rightarrow \infty.$$

As $x \rightarrow 0$, Theorem 6.7 yields rapid convergence to 0:

$$f_{S_\alpha}(x) \sim C_3 x^{-(2-\alpha)/(2-2\alpha)} e^{-c_3 x^{-\alpha/(1-\alpha)}}, \quad x \searrow 0, \quad (6.15)$$

with $c_3 = (1-\alpha)\alpha^{\alpha/(1-\alpha)}$ and $C_3 = (2\pi(1-\alpha))^{-1/2}\alpha^{1/(2-2\alpha)}$.

Example 6.17. Consider the Mittag-Leffler distribution in Example 3.11 with $0 < \alpha < 1$. Since $\gamma, \gamma' > 0$, we can apply Theorem 6.9(i). By (3.17), $F(s) := \mathbb{E} M_\alpha^s$ has simple poles at $s = -n$, $n = 1, 2, \dots$, and, using (A.6), cf. (6.13),

$$\operatorname{Res}_{-n}(F) = \frac{\operatorname{Res}_{1-n}(\Gamma)}{\Gamma(1-n\alpha)} = \frac{(-1)^{n-1}}{(n-1)!\Gamma(1-n\alpha)} = (-1)^{n-1} \frac{\Gamma(n\alpha) \sin(\pi n\alpha)}{(n-1)! \pi}.$$

(Again, this includes the case when $n\alpha$ is an integer, in which case $-n$ is not a pole but the formula correctly yields 0.) We thus obtain by (6.12)

$$\begin{aligned} f_{M_\alpha}(x) &= \sum_{n=1}^{\infty} (-1)^{n-1} \frac{\Gamma(n\alpha) \sin(\pi n\alpha)}{(n-1)! \pi} x^{n-1} \\ &= \sum_{m=0}^{\infty} (-1)^m \frac{\Gamma(m\alpha + \alpha) \sin(\pi(m+1)\alpha)}{m! \pi} x^m. \end{aligned} \quad (6.16)$$

(This is also easily obtained from the stable density (6.14) since $M_\alpha = S_\alpha^{-\alpha}$ and thus $f_{M_\alpha}(x) = \alpha^{-1} x^{-1/\alpha-1} f_{S_\alpha}(x^{-1/\alpha})$.) In particular, in accordance with Theorem 6.11,

$$f_{M_\alpha}(0+) = \frac{\Gamma(\alpha) \sin(\pi\alpha)}{\pi} = \frac{1}{\Gamma(1-\alpha)}.$$

As $x \rightarrow \infty$, Theorem 6.1 yields

$$f_{M_\alpha}(x) \sim C_2 x^{(2\alpha-1)/(2-2\alpha)} e^{-c_2 x^{1/(1-\alpha)}}, \quad x \rightarrow \infty,$$

with $c_2 = (1-\alpha)\alpha^{\alpha/(1-\alpha)}$ and $C_2 = (2\pi(1-\alpha))^{-1/2}\alpha^{(2\alpha-1)/(2-2\alpha)}$. (This also follows from (6.15).)

Example 6.18. The Lévy area in Example 3.20 has the moment generating function (3.37) with simple poles at $(n + \frac{1}{2})\pi$, $n \in \mathbb{Z}$. Theorem 6.5 yields, for $y > \varkappa = 0$,

$$f(y) = \sum_{n=0}^{\infty} (-1)^n e^{-(n+\frac{1}{2})\pi y} = \frac{e^{-\pi y/2}}{1 + e^{-\pi y}} = \frac{1}{2 \cosh(\pi y/2)}$$

while Theorem 6.9 yields, for $y < 0$,

$$f(y) = \sum_{n=1}^{\infty} (-1)^{n-1} e^{(n-\frac{1}{2})\pi y} = \frac{e^{\pi y/2}}{1 + e^{\pi y}} = \frac{1}{2 \cosh(\pi y/2)}.$$

The two sums thus sum to the same analytic expression; hence A has the density $1/2 \cosh(\pi y/2)$ for $-\infty < y < \infty$. (For a more elegant proof of this, see e.g. Protter [28, p. 91].)

7. Brownian supremum process area

We consider the integral $\mathcal{A} = \mathcal{A}(1)$ of the Brownian supremum process defined in (1.5).

Remark 7.1. Let $L(t)$ denote the local time of $B(t)$ at 0. It is well-known that the processes $S(t)$ and $L(t)$, have the same distribution [30, Chapter VI.2]:

$$\{S(t)\}_{t \geq 0} \stackrel{d}{=} \{L(t)\}_{t \geq 0}.$$

Consequently, $\mathcal{A}(T) \stackrel{d}{=} \int_0^T L(t) dt$, so we obtain the same results for this integral.

Let ψ denote the Laplace transform of \mathcal{A} :

$$\psi(s) := \mathbb{E} e^{-s\mathcal{A}}. \quad (7.1)$$

Janson and Petersson [19] proved the following formula for the Laplace transform of a variation of ψ , or in other words, a *double* Laplace transform of \mathcal{A} : For all $\alpha, \lambda > 0$,

$$\int_0^\infty \psi(\alpha t^{3/2}) e^{-\lambda t} dt = \int_0^\infty \left(1 + \frac{3\alpha t}{\sqrt{8\lambda}}\right)^{-2/3} e^{-\lambda t} dt. \quad (7.2)$$

Janson and Petersson [19] used (7.2) to compute the integer moments $\mathbb{E} \mathcal{A}^n$, $n \in \mathbb{N}$; Theorem 1.6 extends their formula to all real and complex moments.

Proof of Theorem 1.6. Consider for convenience $X := \frac{\sqrt{8}}{3}\mathcal{A}$. Taking $\alpha = \sqrt{8}/3$ in (7.2), we find, for $\lambda > 0$,

$$\int_0^\infty \mathbb{E} e^{-t^{3/2}X} e^{-\lambda t} dt = \int_0^\infty \left(1 + \frac{t}{\sqrt{\lambda}}\right)^{-2/3} e^{-\lambda t} dt. \quad (7.3)$$

Denote the common value of the integrals in (7.3) by $G(\lambda)$.

Let $-1 < s < 0$, and integrate $\lambda^s G(\lambda)$. From the left hand side in (7.3) we obtain, using Fubini's theorem a couple of times, the standard Gamma integral (A.7), and the change of variables $t^{3/2} = u$,

$$\begin{aligned} \int_0^\infty G(\lambda) \lambda^s d\lambda &= \int_0^\infty \int_0^\infty \mathbb{E} e^{-t^{3/2}X} e^{-\lambda t} \lambda^s dt d\lambda \\ &= \Gamma(s+1) \mathbb{E} \int_0^\infty e^{-t^{3/2}X} t^{-s-1} dt \\ &= \Gamma(s+1) \mathbb{E} \frac{2}{3} \int_0^\infty e^{-uX} u^{-2s/3-1} du \\ &= \frac{2}{3} \Gamma(s+1) \Gamma(-2s/3) \mathbb{E} X^{2s/3}. \end{aligned} \quad (7.4)$$

Similarly, from the right hand side of (7.3), using the changes of variables $t = \lambda^{1/2}x$ and $\lambda = u^{2/3}$, and the standard Gamma and Beta integrals (A.7) and

(A.9), still assuming $-1 < s < 0$,

$$\begin{aligned}
\int_0^\infty G(\lambda)\lambda^s d\lambda &= \int_0^\infty \int_0^\infty \left(1 + \frac{t}{\sqrt{\lambda}}\right)^{-2/3} e^{-\lambda t} \lambda^s dt d\lambda \\
&= \int_0^\infty \int_0^\infty (1+x)^{-2/3} e^{-\lambda^{3/2}x} \lambda^{s+1/2} dx d\lambda \\
&= \frac{2}{3} \int_0^\infty \int_0^\infty (1+x)^{-2/3} e^{-ux} u^{2s/3} dx du \quad (7.5) \\
&= \frac{2}{3} \Gamma(2s/3 + 1) \int_0^\infty (1+x)^{-2/3} x^{-2s/3-1} dx \\
&= \frac{2}{3} \Gamma(2s/3 + 1) \frac{\Gamma(-2s/3)\Gamma(2s/3 + 2/3)}{\Gamma(2/3)}.
\end{aligned}$$

Setting the right hand sides of (7.4) and (7.5) equal, we find after some cancellations,

$$\mathbb{E} X^{2s/3} = \frac{\Gamma(2s/3 + 1)\Gamma(2s/3 + 2/3)}{\Gamma(2/3)\Gamma(s + 1)},$$

for $-1 < s < 0$, and thus, replacing s by $3s/2$,

$$\mathbb{E} X^s = \frac{\Gamma(s + 1)\Gamma(s + 2/3)}{\Gamma(2/3)\Gamma(3s/2 + 1)}, \quad (7.6)$$

for $-2/3 < s < 0$.

We use Theorem 2.1 to extend the domain of validity of (7.6). Denote, as usual, the right hand side of (7.6) by $F(s)$, and note that $s = -2/3$ is *not* a pole of $F(s)$; it is a removable singularity since the poles in the numerator and denominator at $-2/3$ cancel. The first pole of $G(s)$ on the negative real axis is $s = -1$. This can also be seen by the functional equation $\Gamma(z + 1) = z\Gamma(z)$, which enables us to rewrite (7.6) as

$$\left(\frac{\sqrt{8}}{3}\right)^s \mathbb{E} \mathcal{A}^s = \mathbb{E} X^s = \frac{\Gamma(s + 1)\Gamma(s + 5/3)}{\Gamma(5/3)\Gamma(3s/2 + 2)}, \quad (7.7)$$

where the right hand side clearly has a pole at -1 but not in $(-1, \infty)$. Hence $\rho_- = -1$ and $\rho_+ = \infty$, so (7.6) and (7.7) hold for $\operatorname{Re} s > -1$ by Theorem 2.1, while $\mathbb{E} \mathcal{A}^s = \infty$ for $s \leq -1$.

Next, the triplication and duplication formulas (A.4) and (A.3) yield

$$\begin{aligned}
\Gamma(s + 1)\Gamma(s + 4/3)\Gamma(s + 5/3) &= 2\pi 3^{-3s-3+1/2} \Gamma(3s + 3) \\
&= \pi^{1/2} 2^{3s+3} 3^{-3s-5/2} \Gamma(3s/2 + 3/2)\Gamma(3s/2 + 2),
\end{aligned}$$

and thus

$$\frac{\Gamma(s + 1)\Gamma(s + 5/3)}{\Gamma(3s/2 + 2)} = \pi^{1/2} 2^{3s} 3^{-5/2} \frac{\Gamma(3s/2 + 3/2)}{\Gamma(s + 4/3)} \left(\frac{2^3}{3^3}\right)^s.$$

The third formula for $\mathbb{E}\mathcal{A}^s$ follows by substituting this into (7.7), and using $\Gamma(1/3)\Gamma(2/3) = \pi/\sin(\pi/3) = 2\pi/\sqrt{3}$ from (A.6). (The constant factor can always be found by setting $s = 0$, see Remark 1.2.)

Similarly, the final formula follows by applying the duplication formula (A.3) to $\Gamma(s+4/3)$ and the triplication formula (A.4) to $\Gamma(3s/2+3/2)$. (Alternatively, we may use (7.6), applying the duplication formula to $\Gamma(s+1)$ and $\Gamma(s+2/3)$ and the triplication formula to $\Gamma(3s/2+1)$.) \square

We have $\rho_+ = \infty$ and $\rho_- = -1$ for \mathcal{A} . Further, the parameters in (2.4)–(2.8) are, from any of the expressions in Theorem 1.6: $\gamma = \gamma' = 1/2$, $\delta = 1/6$, $\varkappa = -\frac{1}{2}\log 3$, $C_1 = \pi^{-1/2}\Gamma(1/3) = 2\sqrt{\pi/3}/\Gamma(2/3)$. Theorem 5.7 thus yields

$$\mathbb{E}\mathcal{A}^s \sim \frac{\Gamma(1/3)}{\pi^{1/2}} s^{1/6} e^{\frac{1}{2}s \log s - (\frac{1}{2}\log 3 + \frac{1}{2})s} = \frac{\Gamma(1/3)}{\pi^{1/2}} s^{1/6} \left(\frac{s}{3e}\right)^{s/2} \quad (7.8)$$

as $s \rightarrow \infty$, found for integer s in [19].

Proof of Theorem 1.8. The existence of the density function $f_{\mathcal{A}}(x)$ follows from Theorem 5.4. The explicit formulas are obtained from Theorem 6.9 as follows. We use the last expression in Theorem 1.6 for $F(s) := \mathbb{E}\mathcal{A}^s$, where there is no cancellation of poles. The poles are thus given by, for $n \in \mathbb{Z}_{\geq 0}$, $s/2 + 1/2 = -n$ and $s/2 + 5/6 = -n$, i.e. $s = -2n - 1$ and $s = -2n - 5/3$; all poles are simple. The same formula yields the residues, using (A.11) and (A.6),

$$\begin{aligned} \text{Res}_{-2n-1}(F) &= \frac{\Gamma(1/3)}{2^{1/3}\pi} \cdot 2 \frac{(-1)^n}{n!} \frac{\Gamma(1/3-n)}{\Gamma(1/6-n)} \cdot \left(\frac{2}{3}\right)^{-n-1/2} \\ &= (-1)^n \frac{2^{1/6} 3^{1/2} \Gamma(1/3)}{\pi} \cdot \frac{\Gamma(n+5/6)}{n! \Gamma(n+2/3)} \cdot \frac{\sin(\pi/6-n\pi)}{\sin(\pi/3-n\pi)} \cdot \left(\frac{3}{2}\right)^n \\ &= (-1)^n \frac{2^{1/6} \Gamma(1/3)}{\pi} \cdot \frac{\Gamma(n+5/6)}{n! \Gamma(n+2/3)} \cdot \left(\frac{3}{2}\right)^n, \\ \text{Res}_{-2n-5/3}(F) &= \frac{\Gamma(1/3)}{2^{1/3}\pi} \cdot 2 \frac{(-1)^n}{n!} \frac{\Gamma(-1/3-n)}{\Gamma(-1/6-n)} \cdot \left(\frac{2}{3}\right)^{-n-5/6} \\ &= (-1)^n \frac{3^{1/3} \Gamma(1/3)}{2^{1/6}\pi} \cdot \frac{\Gamma(n+7/6)}{n! \Gamma(n+4/3)} \cdot \left(\frac{3}{2}\right)^n. \end{aligned}$$

Consequently, by Theorem 6.9, in particular (6.12),

$$\begin{aligned} f_{\mathcal{A}}(x) &= \frac{2^{1/6} \Gamma(1/3)}{\pi} \sum_{n=0}^{\infty} (-1)^n \frac{\Gamma(n+5/6)}{n! \Gamma(n+2/3)} \left(\frac{3}{2}\right)^n x^{2n} \\ &\quad + \frac{3^{1/3} \Gamma(1/3)}{2^{1/6}\pi} \sum_{n=0}^{\infty} (-1)^n \frac{\Gamma(n+7/6)}{n! \Gamma(n+4/3)} \left(\frac{3}{2}\right)^n x^{2n+2/3}. \end{aligned}$$

By the definition of ${}_1F_1$, and simplifying the constants using (A.3) and (A.6), this can be written as

$$f_{\mathcal{A}}(x) = \frac{2^{1/2}}{\pi^{1/2}} {}_1F_1\left(\frac{5}{6}; \frac{2}{3}; -\frac{3}{2}x^2\right) + \frac{2^{-1/6} 3^{1/3}}{\Gamma(5/6)} x^{2/3} {}_1F_1\left(\frac{7}{6}; \frac{4}{3}; -\frac{3}{2}x^2\right).$$

By Kummer's transformation [1, (13.1.27)], this equals

$$e^{-\frac{3}{2}x^2} \left(\frac{2^{1/2}}{\pi^{1/2}} {}_1F_1\left(-\frac{1}{6}; \frac{2}{3}; \frac{3}{2}x^2\right) + \frac{2^{-1/6}3^{1/3}}{\Gamma(5/6)} x^{2/3} {}_1F_1\left(\frac{1}{6}; \frac{4}{3}; \frac{3}{2}x^2\right) \right),$$

which can be rewritten as the two last formulas in the theorem by the definition of U [1, (13.1.3)], see also [1, (13.1.29)], again using (A.3) and (A.6) to simplify constants. \square

Proof of Theorem 1.9. Immediate by Theorem 6.1. \square

Remark 7.2. The hypergeometric function ${}_1F_1(a; b; x)$ and $U(a; b; x)$ satisfies Kummer's equation $xF'' + (b-x)F' - aF = 0$ [1, (13.1.1)], and it follows easily that $f_{\mathcal{A}}$ satisfies the differential equation

$$xf_{\mathcal{A}}''(x) + (3x^2 + \frac{1}{3})f_{\mathcal{A}}'(x) + 5xf_{\mathcal{A}}(x) = 0, \quad x > 0. \quad (7.9)$$

We guess that it also is possible to derive this equation directly from (7.2) by manipulations of Laplace transforms, but we have not pursued this.

8. A hashing variable

As said in Section 1, when studying the maximum displacement in hashing with linear probing, Petersson [24, Theorem 5.1] found as a limit a random variable \mathcal{M} with the distribution

$$\mathbb{P}(\mathcal{M} > x) = \psi(x^{3/2}) = \mathbb{E} e^{-x^{3/2}\mathcal{A}}, \quad x > 0, \quad (8.1)$$

where \mathcal{A} is the Brownian supremum area studied in Section 7. Lemma 1.11 shows that this type of relation preserves moments of Gamma type; hence \mathcal{M} has moments of Gamma type, but we have postponed the proof until now.

Proof of Lemma 1.11. For $x \geq 0$,

$$\mathbb{P}(T^{1/\alpha}/Z^{1/\alpha} > x) = \mathbb{P}(T > x^\alpha Z) = \mathbb{E}(\mathbb{P}(T > x^\alpha Z \mid Z)) = \mathbb{E} e^{-x^\alpha Z}$$

which shows that (1.10) and (1.11) are equivalent.

If (1.10) or (1.11) holds, and thus both hold, then, for $s > -\alpha$, using (3.2),

$$\mathbb{E} V^s = \mathbb{E} T^{s/\alpha} \mathbb{E} Z^{-s/\alpha} = \Gamma(s/\alpha + 1) \mathbb{E} Z^{-s/\alpha}. \quad \square$$

Proof of Theorem 1.12. By (8.1) and Lemma 1.11, with $\alpha = 3/2$,

$$\mathbb{E} \mathcal{M}^s = \Gamma(2s/3 + 1) \mathbb{E} \mathcal{A}^{-2s/3},$$

and the result follows from Theorem 1.6; for the last formula we also use (A.3). \square

Note also that Lemma 1.11 yields the representation

$$\mathcal{M} \stackrel{d}{=} T^{2/3} \mathcal{A}^{-2/3}, \quad (8.2)$$

where $T^{2/3}$ has a Weibull distribution with parameter $3/2$, cf. Example 3.9, and is independent of \mathcal{A} .

For \mathcal{M} , the parameters in (2.4)–(2.8) are, from any of the expressions in Theorem 1.12: $\gamma = 1$, $\gamma' = 1/3$, $\delta = 2/3$, $\varkappa = \frac{1}{3} \log 2$, $C_1 = 2^{7/6} 3^{-2/3} \Gamma(1/3) = 2^{13/6} 3^{-7/6} \pi / \Gamma(2/3)$. Furthermore, the function $F(s) := \mathbb{E} \mathcal{M}^s$ (extended to all of the complex plane) has residue $-3/\sqrt{2\pi}$ at $\rho_+ = 3/2$, and $\sqrt{2/\pi}$ at $\rho_- = -3/2$. (See the proofs below for the other residues.)

Proof of Theorem 1.13. As in the proof of Theorem 1.8, the existence of the density function $f_{\mathcal{M}}(x)$ follows from Theorem 5.4 and the explicit formulas are obtained from Theorem 6.9. We use the third expression in Theorem 1.6 for $F(s) := \mathbb{E} \mathcal{M}^s$. The poles $\rho < 0$ all come from the factor $\Gamma(1 + 2s/3)$ and are thus given by, for $n \in \mathbb{Z}_{\geq 0}$, $1 + 2\rho/3 = -n$, i.e. $\rho = -3n/2 - 3/2$. All poles are simple and we find using (A.11) the residues

$$\operatorname{Res}_{-\frac{3}{2}n - \frac{3}{2}}(F) = \frac{\Gamma(1/3)}{2^{1/3}\pi} \cdot \frac{\Gamma(1 + n/2) \Gamma(4/3 + n/2)}{\Gamma(7/6 + n/2)} \cdot \left(\frac{3}{2}\right)^{-1/2 - n/2} \cdot \frac{3}{2} \frac{(-1)^n}{n!}.$$

Consequently, by Theorem 6.9, in particular (6.12),

$$f_{\mathcal{M}}(x) = \frac{3^{1/2}\Gamma(1/3)}{2^{5/6}\pi} \sum_{n=0}^{\infty} (-1)^n \frac{\Gamma(1 + n/2) \Gamma(4/3 + n/2)}{\Gamma(7/6 + n/2) n!} \left(\frac{2}{3}\right)^{n/2} x^{3n/2 + 1/2}.$$

Splitting the sum into two parts, for $n = 2k$ and $n = 2k + 1$, and using $(2k)! = \pi^{-1/2} 2^{2k} \Gamma(k + 1/2) k!$ and $(2k + 1)! = \pi^{-1/2} 2^{2k+1} k! \Gamma(k + 3/2)$, both instances of (A.3), we obtain, as usual using (A.3) and (A.6),

$$\begin{aligned} f_{\mathcal{M}}(x) &= \frac{3^{1/2}\Gamma(1/3)}{2^{5/6}\pi^{1/2}} \sum_{k=0}^{\infty} \frac{\Gamma(4/3 + k)}{\Gamma(7/6 + k) \Gamma(1/2 + k)} \left(\frac{1}{6}\right)^k x^{3k + 1/2} \\ &\quad - \frac{\Gamma(1/3)}{2^{4/3}\pi^{1/2}} \sum_{k=0}^{\infty} \frac{\Gamma(11/6 + k)}{\Gamma(5/3 + k) k!} \left(\frac{1}{6}\right)^k x^{3k + 2} \\ &= \frac{3^{1/2}\Gamma(1/3)}{2^{5/6}\pi^{1/2}} \frac{\Gamma(4/3)}{\Gamma(7/6) \Gamma(1/2)} x^{1/2} {}_2F_2\left(\frac{4}{3}, 1; \frac{7}{6}, \frac{1}{2}; \frac{x^3}{6}\right) \\ &\quad - \frac{\Gamma(1/3)}{2^{4/3}\pi^{1/2}} \frac{\Gamma(11/6)}{\Gamma(5/3)} x^2 {}_1F_1\left(\frac{11}{6}; \frac{5}{3}; \frac{x^3}{6}\right) \\ &= \frac{2^{1/2}}{\pi^{1/2}} x^{1/2} {}_2F_2\left(\frac{4}{3}, 1; \frac{7}{6}, \frac{1}{2}; \frac{x^3}{6}\right) - \frac{5}{8} x^2 {}_1F_1\left(\frac{11}{6}; \frac{5}{3}; \frac{x^3}{6}\right). \quad \square \end{aligned}$$

Proof of Theorem 1.14. As remarked above, the residue at $\rho_+ = 3/2$ is $-3/\sqrt{2\pi}$, and the next pole is at $5/2$, which yields (1.14) by Theorem 6.4.

More precisely, by the last expression in Theorem 1.12, there are, on the positive real axis, poles when $1/2 - s/3 = -n$ or $5/6 - s/3 = -n$ for integer $n \geq 0$, i.e., $s = 3n + 3/2$ and $s = 3n + 5/2$. The residues are, using (A.11) and (A.6),

$$\begin{aligned} \operatorname{Res}_{3n+3/2}(F) &= -3 \frac{(-1)^n}{n!} \frac{\Gamma(1/3)}{2^{1/3}\pi^{3/2}} \cdot \frac{\Gamma(1+n)\Gamma(3/2+n)\Gamma(1/3-n)}{\Gamma(1/6-n)} \cdot 6^{n+1/2} \\ &= (-1)^{n+1} \frac{2^{1/6}3\Gamma(1/3)}{\pi^{3/2}} \cdot \frac{\Gamma(3/2+n)\Gamma(5/6+n)}{\Gamma(2/3+n)} \cdot 6^n, \\ \operatorname{Res}_{3n+5/2}(F) &= -3 \frac{(-1)^n}{n!} \frac{\Gamma(1/3)}{2^{1/3}\pi^{3/2}} \cdot \frac{\Gamma(4/3+n)\Gamma(11/6+n)\Gamma(-1/3-n)}{\Gamma(-1/6-n)} \cdot 6^{n+5/6} \\ &= (-1)^{n+1} \frac{2^{1/2}3^{4/3}\Gamma(1/3)}{\pi^{3/2}} \cdot \frac{\Gamma(11/6+n)\Gamma(7/6+n)}{n!} \cdot 6^n. \end{aligned}$$

By Theorem 6.4, there is an asymptotic expansion $-\sum_{\rho>0} \operatorname{Res}_{\rho}(F)x^{-\rho-1}$, which by the definition of the (generalized) hypergeometric series can be written as in (1.15), yet again using (A.11) and (A.6). \square

9. Triangular and diagonal Pólya urns

A generalized Pólya urn contains balls of several different colours. At each time $n \geq 1$, one of the balls is drawn at random, and a set of new balls, depending on the colour of the drawn ball, is added to the urn. We consider for simplicity only the case of two colours, say black and white; the replacement rule may then be described by a matrix $\begin{pmatrix} a & b \\ c & d \end{pmatrix}$, meaning that if the drawn ball is black [white], it is replaced together with a black and b white balls [c black and d white balls]. It is here natural to let a, b, c, d be non-negative integers, but in fact, the model can be defined (and the results below hold) for arbitrary real $a, b, c, d \geq 0$, see [15; 16]. (Further, under certain conditions some of the entries can be negative too, but that case is not interesting here.) Different values of the parameters yield a variety of different limit laws for the numbers B_n and W_n of black and white balls in the urn after n steps, see e.g. [10; 15; 16] and the references given there. We are here interested in the special case of a *triangular urn*, meaning that the replacement matrix is triangular, say $b = 0$. We start with $B_0 = b_0 \geq 0$ black and $W_0 = w_0 \geq 0$ white balls, and assume $w_0 > 0$ (otherwise, there will never be any white balls).

9.1. Balanced triangular urns

Assume that the urn is triangular and *balanced*, meaning that the total number of added balls does not depend on the drawn ball, i.e., $a = c + d$; we further assume that $a, c, d > 0$; thus $a > d > 0$ and $c = a - d$. In this case, it is shown by Puyhaubert [29], Flajolet, Dumas and Puyhaubert [10, Section 7] and (with a different proof) Janson [16, Theorems 1.3(v) and 1.7] that $W_n/n^{d/a} \xrightarrow{d} W$

for a random variable W with moments of Gamma type given by

$$\mathbb{E} W^s = d^s \frac{\Gamma((b_0 + w_0)/a)}{\Gamma(w_0/d)} \cdot \frac{\Gamma(s + w_0/d)}{\Gamma(ds/a + (b_0 + w_0)/a)}, \quad \operatorname{Re} s > -\frac{w_0}{d}. \quad (9.1)$$

In the special case $(b_0, w_0) = (c, d)$, and thus $b_0 + w_0 = a$, this simplifies to $d^s \Gamma(s+1)/\Gamma(ds/a+1)$, so W/d has a Mittag-Leffler distribution with parameter $d/a \in (0, 1)$, see (3.17).

All poles of $F(s) := \mathbb{E} W^s$ are on the negative real axis, so $\rho_+ = \infty$. In general, (9.1) shows that there is a pole at $-w_0/d$, but if $b_0 = 0$, then this singularity is removable and the first pole on the negative real axis is $-w_0/d - 1$. We thus have $\rho_- = -w_0/d$ when $b_0 > 0$, but $\rho_- = -w_0/d - 1$ when $b_0 = 0$. In fact, if $b_0 = 0$, so we start with only w_0 white balls, the first drawn ball is necessarily white, and thus urn after the first draw contains c black and $w_0 + d$ white balls. Thus the limit random variable W is the same for the initial conditions $(0, w_0)$ and $(c, w_0 + d)$, and we may without loss of generality assume that $b_0 > 0$.

By (9.1), we have $\gamma = \gamma' = 1 - d/a = c/a$, $\delta = w_0/d - (b_0 + w_0)/a$, $\varkappa = -\frac{d}{a} \log \frac{d}{a} + \log d = (c \log d + d \log a)/a$, and $C_1 = (a/d)^{(b_0 + w_0)/a - 1/2} \Gamma((b_0 + w_0)/a)/\Gamma(w_0/d)$.

The function $F(s)$ in (9.1) has simple poles at $s = -w_0/d - n$, $n = 0, 1, \dots$, (except that some of these may in fact be removable singularities) and Theorems 5.4 and 6.9 yield by a straightforward calculation of the residues, using (A.11) and (A.6), the following:

Theorem 9.1. *The limit variable W for a balanced triangular urn $\binom{a}{c \ d}$ with $a = c + d$ and $a, c, d, w_0 > 0$ has a density function f_W on $(0, \infty)$ given by, for $x > 0$,*

$$\begin{aligned} f_W(x) &= \frac{\Gamma((b_0 + w_0)/a)}{\Gamma(w_0/d)} \sum_{n=0}^{\infty} \frac{(-1)^n}{n!} \cdot \frac{d^{-n-w_0/d}}{\Gamma(-dn/a + b_0/a)} x^{n+w_0/d-1} \\ &= \frac{\Gamma((b_0 + w_0)/a)}{\pi d \Gamma(w_0/d)} \sum_{n=0}^{\infty} \frac{(-1)^n}{n!} \Gamma\left(\frac{dn + a - b_0}{a}\right) \sin \frac{\pi(b_0 - dn)}{a} \left(\frac{x}{d}\right)^{n+w_0/d-1}. \end{aligned}$$

In fact, [10] even gives a local limit theorem to this density function.

Remark 9.2. It follows from Theorem 9.1 by comparison with Example 6.17, or more simply directly from (9.1) and (3.17), that in the special case $b_0 = 0$, W/d has a Mittag-Leffler(d/a) distribution conjugated with $x^{w_0/d}$, see Remark 2.11; similarly, in the special case $b_0 = c = a - d$, W/d has a Mittag-Leffler(d/a) distribution conjugated with $x^{(w_0-d)/d}$.

Theorem 9.1 shows immediately that as $x \searrow 0$, the density $f_W(x)$ satisfies $f_W(x) \sim C' x^{w_0/d-1}$ where $C' = \Gamma((b_0 + w_0)/a)(\Gamma(w_0/d)\Gamma(b_0/a))^{-1} d^{-w_0/d} > 0$, provided $b_0 > 0$. For large x , Theorem 6.1 yields:

Theorem 9.3. *As $x \rightarrow \infty$,*

$$f_W(x) \sim C_2 x^{c_1-1} e^{-c_2 x^{a/c}},$$

with $c_1 = (\delta + 1/2)a/c$, $c_2 = ca^{-a/c}d^{-1}$, $C_2 = C_1(2\pi c/a)^{-1/2}(da^{d/c})^{-(\delta+1/2)}$, where δ and C_1 are given above.

Remark 9.4. For non-balanced triangular urns ($a \neq c + d$), limit results are given in [16], but the results are more complicated and we do not believe that the limits have moments of Gamma type. (See for example [16, Theorem 1.6], which gives a complicated integral formula for the moments in the case $a > d > 0$, $c > 0$. In the balanced case, it simplifies to (9.1), but as far as we know, there is no similar simplification in general.

Remark 9.5. The case of triangular urns with three or more colours is not yet fully explored. Limit laws with moments of Gamma type occur in some cases, but presumably not in all. Some such results are given by Puyhaubert [29], see also Flajolet, Dumas and Puyhaubert [10].

9.2. Diagonal urns

In the diagonal case $b = c = 0$ (with $a, d, b_0, x_0 > 0$ to avoid trivialities), there are simple limit results, see [16, Theorem 1.4]. We distinguish between three cases.

(i) If $a = d$, the classical Pólya urn [7; 27], $W_n/n \rightarrow W$ where $W/d \sim B(w_0/a, b_0/a)$. Hence, by Example 3.4, W has moments of Gamma type

$$\mathbb{E} W^s = a^s \frac{\Gamma((b_0 + w_0)/a)}{\Gamma(w_0/a)} \cdot \frac{\Gamma(s + w_0/a)}{\Gamma(s + (b_0 + w_0)/a)}, \quad s > -\frac{w_0}{a}. \quad (9.2)$$

Hence, recalling that $a = d$, (9.1) holds in this case too. We have $\rho_+ = \infty$, $\rho_- = -w_0/a$, $\gamma = \gamma' = 0$, $\delta = -b_0/a$, $\varkappa = \log a$, $C_1 = \Gamma((b_0 + w_0)/a)/\Gamma(w_0/a)$.

(ii) If $a > d$, $W_n/n^{d/a} \xrightarrow{d} W := dU^{-d/a}V$ where $U \sim \Gamma(b_0/a)$ and $V \sim \Gamma(w_0/d)$ are independent. Thus, by (3.1),

$$\mathbb{E} W^s = d^s \frac{\Gamma(b_0/a - ds/a) \Gamma(w_0/d + s)}{\Gamma(b_0/a) \Gamma(w_0/d)}, \quad -\frac{w_0}{d} < \operatorname{Re} s < \frac{b_0}{d}. \quad (9.3)$$

We have $\rho_+ = b_0/d$, $\rho_- = -w_0/d$, $\gamma = 1 + d/a$, $\gamma' = 1 - d/a$, $\delta = b_0/a + w_0/d - 1$, $\varkappa = -\frac{d}{a} \log \frac{d}{a} + \log d$, $C_1 = 2\pi(\Gamma(b_0/a)\Gamma(w_0/d))^{-1}(d/a)^{b_0/a-1/2}$. Theorems 5.4 and 6.9 apply again and yield the following:

Theorem 9.6. *The limit variable W for a diagonal urn $\binom{a \ 0}{0 \ d}$ with $a > d > 0$ and $b_0, w_0 > 0$ has a density function f_W on $(0, \infty)$ given by, for $x > 0$,*

$$f_W(x) = \frac{1}{d\Gamma(b_0/a)\Gamma(w_0/d)} \sum_{n=0}^{\infty} \frac{(-1)^n}{n!} \Gamma\left(\frac{dn + b_0 + w_0}{a}\right) \left(\frac{x}{d}\right)^{n+w_0/d-1}.$$

Again, the asymptotic $f_W(x) \sim C'x^{w_0/d-1}$ as $x \searrow 0$, for some $C' > 0$, is immediate. For large x , we this time use Theorem 6.4, since $\rho_+ < \infty$. The poles of (9.3) on the positive real axis are $(an + b_0)/d$, $n = 0, 1, \dots$, and the residues are easily calculated. This yields a divergent asymptotic expansion, interpreted as in Remark 1.10.

Theorem 9.7. *As $x \rightarrow \infty$, the density $f_W(x)$ has an asymptotic expansion*

$$f_W(x) \sim \frac{a}{d^2 \Gamma(b_0/a) \Gamma(w_0/d)} \sum_{n=0}^{\infty} \frac{(-1)^n}{n!} \Gamma\left(\frac{an + b_0 + w_0}{d}\right) \left(\frac{x}{d}\right)^{-an/d - b_0/d - 1}.$$

(iii) If $a < d$, we may interchange the two colours and obtain $n^{-a/d}(nd - W_n) \xrightarrow{d} W := dUV^{-a/d}$, with U and V as above, and aW/d has the distribution in (ii) with the exchanges $a \leftrightarrow d$ and $b_0 \leftrightarrow w_0$.

10. Further examples

We give a couple of further examples, or rather counter examples.

Example 10.1. Let X have a distribution that is a mixture of a point mass at 1 and a uniform distribution on $[0, 1]$, with equal weights; thus $X = 1 - V + VU$ where $V \sim \text{Be}(1/2)$ and $U \sim \text{U}(0, 1)$ are independent. Then, for $\text{Re } s > -1$,

$$\mathbb{E} X^s = \frac{1}{2} \cdot 1^s + \frac{1}{2} \cdot \mathbb{E} U^s = \frac{1}{2} + \frac{1}{2} \frac{1}{s+1} = \frac{s+2}{2(s+1)} = \frac{\Gamma(s+3)\Gamma(s+1)}{2\Gamma(s+2)^2}. \quad (10.1)$$

Equivalently,

$$\mathbb{E} X^s = \frac{s/2+1}{s+1} = \frac{\Gamma(s/2+2)\Gamma(s+1)}{\Gamma(s/2+1)\Gamma(s+2)}. \quad (10.2)$$

Hence X has moments of Gamma type. We have $\rho_+ = \infty$, $\rho_- = -1$, $\gamma = \gamma' = \delta = \varkappa = 0$ and $C_1 = 1/2$. Note that $\mathbb{E} X^{it} \rightarrow 1/2 \neq 0$ as $t \rightarrow \pm\infty$; cf. Remark 5.6.

Example 10.2. Consider $X := T/U$, where $T \sim \text{Exp}(1)$ and $U \sim \text{U}(0, 1)$ are independent. Then, see Remark 1.5 and Examples 3.2 and 3.3, X has moments of Gamma type

$$\mathbb{E} X^s = \frac{\Gamma(s+1)}{1-s} = \frac{\Gamma(s+1)\Gamma(1-s)}{\Gamma(2-s)}, \quad -1 < \text{Re } s < 1. \quad (10.3)$$

Consequently, $\rho_+ = 1$ and $\rho_- = -1$. There is an infinite number of poles on the negative real axis, viz. $-1, -2, \dots$, but the only pole on the positive real axis is 1. We have $\gamma = \gamma' = 1$, $\delta = -1/2$, $\varkappa = 0$, $C_1 = \sqrt{2\pi}$.

It is easy to find the density of $X = T/U$: for $x > 0$,

$$\mathbb{P}(T/U > x) = \mathbb{P}(T > Ux) = \int_0^1 \mathbb{P}(T > ux) \, du = \int_0^1 e^{-ux} \, du = \frac{1 - e^{-x}}{x}$$

and thus X has the density function

$$f(x) = -\frac{d}{dx} \frac{1 - e^{-x}}{x} = \frac{1 - (1+x)e^{-x}}{x^2}, \quad x > 0. \quad (10.4)$$

Since $\gamma, \gamma' > 0$, Theorem 6.9 applies. The residue at $-n-1$ is $(-1)^n/(n!(n+2))$ and (6.12) yields

$$f(x) = \sum_{n=0}^{\infty} \frac{(-1)^n}{(n+2)n!} x^n$$

which, of course, also follows directly from (10.4).

However, in Theorem 6.5, although the sum in (6.10) consists of a single term x^{-2} and thus converges, the sum $x^{-2} \neq f(x)$ for all $x > 0$, as asserted in Remark 6.6. (But the error is exponentially small, and the estimates in Theorem 6.4 apply.)

11. Further remarks

Remark 11.1. Suppose that X is a positive random variable with finite moments (of all positive orders): $\mathbb{E} X^n < \infty$ for $n \geq 0$. If X has moments of Gamma type, then $\rho_+ = \infty$ and (1.1) gives, in particular, a formula for all integer moments $\mathbb{E} X^n$ in terms of Gamma functions. However, the converse does not hold; even if (1.1) holds for every integer $s \geq 0$, it does not necessarily hold for other s . An example is provided by Stieltjes' original example of indeterminacy in the moment problem [32, §55]: Let, for $\lambda \in [-1, 1]$, X_λ have the density function $a(1 + \lambda \sin(x^{1/4})) \exp(-x^{1/4})$ with the normalizing constant $a = 1/24$. Then, using $\sin(y) = (e^{iy} - e^{-iy})/2i$ and (A.7),

$$\begin{aligned} \mathbb{E} X_\lambda^n &= a \int_0^\infty x^n (1 + \lambda \sin(x^{1/4})) e^{-x^{1/4}} dx \\ &= 4a \int_0^\infty y^{4n+3} (1 + \lambda \sin(y)) e^{-y} dy = \frac{1}{6} \Gamma(4n+4), \end{aligned}$$

for any integer $n \geq 0$ and any $\lambda \in [-1, 1]$; thus the variables X_λ have the same integer moments. For $\lambda = 0$, the same calculation applies to non-integer n as well, and shows that $\mathbb{E} X_0^s = \frac{1}{6} \Gamma(4s+4)$, $-1 < s < \infty$, so X_0 has moments of Gamma type. However, this formula cannot hold for any other λ (and s in an interval), by the uniqueness Corollary 2.3.

Note that $X_0 \stackrel{d}{=} Z^4$, where Z has the Gamma distribution $\Gamma(4)$, cf. Example 3.1. A similar example is provided by N^6 (or $|N|^\alpha$ for any real number $\alpha > 4$) with $N \sim N(0, 1)$, see [2]; indeed $|N|^\alpha \stackrel{d}{=} 2^{\alpha/2} Z^{\alpha/2}$ with $Z \sim \Gamma(1/2)$, see Examples 3.5 and 3.6, and cZ^β with $c > 0$ and $Z \sim \Gamma(\gamma)$ is not determined by its (integer) moments for any $\gamma > 0$ and $\beta > 2$, see e.g. [14, Section 4.10], so it too provides a counter example. See also [31].

Remark 11.2. Many of the examples in Section 3 are infinitely divisible, for example the Gamma distribution $\Gamma(\alpha)$, W_α [5, p. 26], \tilde{P}_α and thus P_α [5, p. 26], L_α [25]. We do not know whether there are any interesting connections between moments of Gamma type and infinite divisibility.

Remark 11.3. It is possible to consider, more generally, moments of the form (1.1) where a_j, b_j, a'_k, b'_k may be complex (and appearing in conjugate pairs to make the function real for real s). We have not pursued this extension and do not know whether there are any interesting results or examples for this class. A trivial example is the following.

Let X have a two-point distribution with $\mathbb{P}(X = x_1) = \mathbb{P}(X = x_2) = 1/2$, where $0 < x_1 < x_2 < \infty$. Then $Y := \log X$ too has a two-point distribution with $\mathbb{P}(Y = y_1) = \mathbb{P}(Y = y_2) = 1/2$ where $y_j = \log x_j$, $j = 1, 2$. Let $d := \mathbb{E} Y = (y_1 + y_2)/2$ and $\beta := (y_2 - y_1)/2\pi$; thus $y_1, y_2 = d \pm \pi\beta$. Then, using (A.6),

$$\mathbb{E} X^s = \mathbb{E} e^{sY} = e^{sd} \cosh(\pi\beta s) = e^{sd} \sin\left(\frac{\pi}{2} + \pi\beta si\right) = \frac{\pi e^{sd}}{\Gamma(\frac{1}{2} + i\beta s)\Gamma(\frac{1}{2} - i\beta s)}.$$

Appendix A: Some standard formulas

For the readers' (and our own) convenience we here collect some well-known formulas for the Gamma function, see e.g. [1, Chapter 6]. Recall that Γ is a meromorphic function in the complex plane, with simple poles at the non-negative integers $0, -1, -2, \dots$ and no zeros, so $1/\Gamma$ is an entire function.

$$\Gamma(s) = \int_0^\infty t^{s-1} e^{-t} dt, \quad \operatorname{Re} s > 0; \quad (\text{A.1})$$

$$\Gamma(z+1) = z\Gamma(z); \quad (\text{A.2})$$

$$\Gamma(2z) = \pi^{-1/2} 2^{2z-1} \Gamma(z)\Gamma(z + \frac{1}{2}); \quad (\text{A.3})$$

$$\Gamma(3z) = (2\pi)^{-1} 3^{3z-1/2} \Gamma(z)\Gamma(z + \frac{1}{3})\Gamma(z + \frac{2}{3}); \quad (\text{A.4})$$

$$\Gamma(mz) = (2\pi)^{-(m-1)/2} m^{mz-1/2} \prod_{j=0}^{m-1} \Gamma(z + \frac{j}{m}) \quad (\text{A.5})$$

$$\Gamma(z)\Gamma(1-z) = \frac{\pi}{\sin(\pi z)}; \quad (\text{A.6})$$

$$\int_0^\infty t^{s-1} e^{-at} dt = a^{-s} \Gamma(s), \quad \operatorname{Re} s > 0, \operatorname{Re} a > 0; \quad (\text{A.7})$$

$$\int_0^1 t^{s-1} (1-t)^{u-1} dt = \frac{\Gamma(s)\Gamma(u)}{\Gamma(s+u)}, \quad \operatorname{Re} s, \operatorname{Re} u > 0; \quad (\text{A.8})$$

$$\int_0^\infty t^{s-1} (1+t)^{-v} dt = \frac{\Gamma(s)\Gamma(v-s)}{\Gamma(v)}, \quad \operatorname{Re} v > \operatorname{Re} s > 0; \quad (\text{A.9})$$

Equation (A.7) yields by Fubini–Tonelli a relation between the Laplace transform and negative moments for any positive random variable X :

$$\int_0^\infty t^{s-1} \mathbb{E} e^{-tX} dt = \mathbb{E} \int_0^\infty t^{s-1} e^{-tX} dt = \Gamma(s) \mathbb{E} X^{-s}, \quad s > 0. \quad (\text{A.10})$$

The residue $\operatorname{Res}_{-n}(\Gamma) = (-1)^n/n!$ (an easy consequence of (A.2)). Thus,

more generally, for any complex $a \neq 0$ and b ,

$$\operatorname{Res}_{z=-(n+b)/a}(\Gamma(az+b)) = a^{-1} \frac{(-1)^n}{n!}. \quad (\text{A.11})$$

Stirling's formula says that for all complex z in a sector $|\arg z| < \pi - \varepsilon$ avoiding the negative real axis

$$\log \Gamma(z) = (z - \frac{1}{2}) \log z - z + \log \sqrt{2\pi} + O(|z|^{-1}), \quad (\text{A.12})$$

where the logarithm $\log z$ is the principal value with imaginary part in $(-\pi, \pi)$. (Here, $\varepsilon > 0$ is arbitrary, but the implicit constant in the O term depends on ε .) By differentiating (A.12) twice we find, for $|\arg z| < \pi - \varepsilon$ (e.g., for $\operatorname{Re} z > 0$),

$$\frac{d}{dz}(\log \Gamma(z)) = \log z + O(|z|^{-1}), \quad (\text{A.13})$$

$$\frac{d^2}{dz^2}(\log \Gamma(z)) = \frac{1}{z} + O(|z|^{-2}). \quad (\text{A.14})$$

(Note that also the error term may be differentiated since the functions are analytic in a larger sector and we may use Cauchy's estimate for the derivative.)

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