Khinchin-type inequalities via Hadamard's factorisation

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Abstract

We prove Khinchin-type inequalities with sharp constants for type L random variables and all even moments. Our main tool is Hadamard's factorisation theorem from complex analysis, combined with Newton's inequalities for elementary symmetric functions. Besides the case of independent summands, we also treat ferromagnetic dependencies in a nonnegative external magnetic field (thanks to Newman's generalisation of the Lee-Yang theorem). Lastly, we compare the notions of type L, ultra sub-Gaussianity (introduced by Nayar and Oleszkiewicz) and strong logconcavity (introduced by Gurvits), with the latter two being equivalent.

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1 Introduction and results

Motivated by Ising models of Lee-Yang type, Newman in [13] made the following definition: a (real-valued) random variable X is of **type** \mathscr{L} , if (i) for some positive constants C and C', $|\mathbb{E} \exp(zX)| \leq C \exp(C'|z|^2)$ for all $z \in \mathbb{C}$ and (ii) $\mathbb{E} \exp(zX)$ is even with only pure imaginary zeros. Note that (ii) implies that X is symmetric, that is -X has the same distribution as X. Perhaps the simplest

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nontrivial example is a Rademacher random variable (a symmetric random sign) taking the values ± 1 with probability $\frac{1}{2}$. Newman proved that if X is of type \mathscr{L} , then for an even integer $p \geq 2$,

$$||X||_p \le ||G||_p ||X||_2,$$

where G is a standard Gaussian random variable (see [14] or Theorem 5 in [13]). As usual, $||Y||_p = (\mathbb{E}|Y|^p)^{1/p}$ denotes the *p*-norm of a random variable Y.

The goal of this short note is to extend this result to a comparison inequality for moments of arbitrary even orders. As explained later, the assumption of evenness in (ii) may be easily dropped (it is just a matter of cosmetic shifting). Thus we shall say that a random variable X is of **type** \mathscr{L}' if it satisfies (i) and (ii') $\mathbb{E} \exp(zX)$ has only pure imaginary zeros. Our main result is the following theorem.

Theorem 1. Let X be a random variable of type \mathscr{L}' . Then for every even integers $2 \leq p \leq q$, we have

$$\|X\|_{q} \le \frac{\|G\|_{q}}{\|G\|_{p}} \|X\|_{p},\tag{1}$$

where G is a standard Gaussian random variable.

The usefulness of this result in Khinchin inequalities for sums of independent random variables comes from the following evident fact.

Remark 2. Finite sums of independent type \mathscr{L} (resp. \mathscr{L}') random variables are of type \mathscr{L} (resp. \mathscr{L}').

Thus the above theorem immediately gives classical Khinchin inequalities with sharp constants for even moments, the result due to Nayar and Oleszkiewicz from [10]. Furthermore, it automatically extends to Hilbert space coefficients by isometrical embeddings into L_p spaces based on standard Gaussians (Orlicz-Szarek's argument, see Remark 3 in [21]).

Corollary 3. Let $(H, \|\cdot\|)$ be a separable (real or complex) Hilbert space. If X_1, \ldots, X_n are independent type \mathscr{L}' random variables, then for every vectors v_1, \ldots, v_n in H, the sum $X = \sum_{j=1}^n X_j v_j$ satisfies (1) for all positive even integers $p \leq q$, where we denote $\|X\|_p = (\mathbb{E}\|X\|^p)^{1/p}$.

Thanks to Newman's generalisation of the Lee-Yang theorem and Griffiths' method of a "ghost" spin (see [3, 12, 15]), we moreover obtain (1) when X is a positive linear combination of random variables with ferromagnetic dependencies $J = (J_{jk})$ in a nonnegative external magnetic field $h = (h_j)$.

Corollary 4. Let μ_1, \ldots, μ_n be Borel probability measures on \mathbb{R} , each one of type \mathscr{L} . Suppose that (X_1, \ldots, X_n) is a random vector in \mathbb{R}^n whose law ρ on \mathbb{R}^n is of the form

$$d\rho(x_1, \dots, x_n) = Z^{-1} \exp\left(\sum_{j=1}^n h_j x_j + \sum_{j,k=1}^n J_{jk} x_j x_k\right) d\mu_1(x_1) \dots d\mu_n(x_n)$$
(2)

with $h_j \ge 0$, $J_{jk} \ge 0$ for all $j, k \le n$, where Z is the normalising constant. Then for every nonnegative a_1, \ldots, a_n , the sum $X = \sum_{j=1}^n a_j X_j$ satisfies (1) for all positive even integers $p \le q$.

For a certain subclass of the type \mathscr{L} random variables (cf. the section below devoted to examples), we are also able to derive sharp moment comparison between the second and *p*-th moment, $p \geq 3$.

Theorem 5. Let X be type \mathscr{L} random variable with characteristic function of the form $\phi_X(t) = e^{-at^2/2} \prod_{j=1}^{\infty} (1-b_jt^2)$ with a > 0, $b_j \ge 0$, $\sum b_j \le a$. Let $\sigma = \sqrt{\operatorname{Var}(X)}$. Then for every $p \ge 3$,

$$\mathbb{E}|\sigma Z_1|^p \le \mathbb{E}|X|^p \le \mathbb{E}|\sigma Z_0|^p,$$

where Z_0 is a standard Gaussian random variable and Z_1 is a random variable with density $(2\pi)^{-1/2}x^2e^{-x^2/2}$.

First we prove Theorem 1 (relying on the product form of the moment generating function $\mathbb{E} \exp(zX)$ and exploiting Newton's inequalities for the elementary symmetric functions) and show how to obtain Corollaries 3 and 4. Then we provide several examples of type \mathscr{L} random variables and prove Theorem 5 (using techniques from [2, 11], relying on convexity properties of certain functions related to $|\cdot|^p$, hence the restriction $p \geq 3$). We conclude with a discussion comparing type \mathscr{L} , ultra sub-Gaussianity (introduced in [10]) and strong logconcavity (introduced in [4]).

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2 Even moments

We recall that a nonnegative sequence $(a_n)_{n\geq 0}$ is called log-concave if it is supported on a contiguous interval (which may be infinite) and satisfies $a_n^2 \geq$ $a_{n-1}a_{n+1}, n \ge 1$. We remark that log-concavity implies the inequalities $a_k a_l \ge a_{k+j}a_{l-j}$ for all $k \ge l, 0 \le j \le l$. We refer for instance to the classical survey [20] by Stanley.

2.1 Proof of Theorem 1 for Rademacher sums

It is instructive to first consider the case when X from Theorem 1 is of the form $\sum_{j=1}^{n} a_j \varepsilon_j$, where a_1, \ldots, a_n are real coefficients and $\varepsilon_1, \varepsilon_2, \ldots$ are i.i.d. Rademacher random variables (symmetric random signs, that is for each j, $\mathbb{P}(\varepsilon_j = -1) = \mathbb{P}(\varepsilon_j = 1) = \frac{1}{2}$). The proof in this case already encapsulates almost all the key ideas needed for the general case and at the same time provides a new, short, elementary and self-contained proof of the main result of [10], the aforementioned classical Khinchin inequalities with sharp constants for even moments.

Let $X = \sum_{j=1}^{m} a_j \varepsilon_j$ and for $k \ge 0$, define

$$r_k = \frac{\mathbb{E}X^{2k}}{\mathbb{E}G^{2k}} \tag{3}$$

(so $r_0 = 1$ and $r_k = \frac{\mathbb{E}X^{2k}}{(2k)!} 2^k k!$). Note that it suffices to show that the sequence $(r_k)_{k=0}^{\infty}$ is log-concave. This is because then in particular, by monotonicity of slopes, the sequence $\frac{\log r_k - \log r_0}{k}$ is nonincreasing and in view of $r_0 = 1$, this gives that the sequence $(r_k^{1/k})_{k\geq 0}$ is nonincreasing, which is exactly (1).

Plainly, since X is symmetric and thanks to independence, for $t \ge 0$,

$$\sum_{k=0}^{\infty} \frac{\mathbb{E}X^{2k}}{(2k)!} 2^k t^k = \mathbb{E}\exp(\sqrt{2t}X) = \prod_{j=1}^m \mathbb{E}\exp(\sqrt{2t}a_j\varepsilon_j) = \prod_{j=1}^m \cosh(\sqrt{2t}a_j).$$

The main point is that since $\cosh z = \prod_{n=1}^{\infty} \left(1 + \frac{z^2}{(n-\frac{1}{2})^2 \pi^2}\right), z \in \mathbb{C}$, the product on the right hand side takes the form

$$\prod_{l=1}^{m} \prod_{n=1}^{\infty} \left(1 + \frac{2a_l^2}{(n-\frac{1}{2})^2 \pi^2} t \right) = \prod_{j=1}^{\infty} (1+\alpha_j t) = \sum_{k=0}^{\infty} \sigma_k t^k,$$

with nonnegative α_j , where

$$\sigma_k = \sigma_k(\{\alpha_j\}) = \sum_{\substack{A \subset \{\alpha_j\} \\ |A| = k}} \prod_{a \in A} a$$

is the kth elementary symmetric function of the (multi)set $\{\alpha_j\}$. Comparing the coefficients thus gives

$$r_k = \frac{\mathbb{E}X^{2k}}{(2k)!} 2^k k! = \sigma_k k!.$$

Thanks to this expression for r_k , Newton's inequalities allow to finish the argument.

Claim. The sequence $(\sigma_k k!)_{k\geq 0}$ is log-concave.

Proof. Fix $k \ge 1$. By Newton's inequalities,

$$\frac{\sigma_k(\{\alpha_j\}_{j=1}^n)^2}{\binom{n}{k}^2} \ge \frac{\sigma_{k-1}(\{\alpha_j\}_{j=1}^n)}{\binom{n}{k-1}} \frac{\sigma_{k+1}(\{\alpha_j\}_{j=1}^n)}{\binom{n}{k+1}}.$$

Letting $n \to \infty$ gives the claim.

2.2 Proof of Theorem 1

We proceed with the proof of Theorem 1 in the general case when X is type \mathscr{L}' . First we argue that it suffices to assume that X is of type \mathscr{L} and then give the proof in this case.

Type \mathscr{L} vs. \mathscr{L}' . We begin with a straightforward observation.

Lemma 6. If X is a random variable of type \mathscr{L}' , then there is $c \in \mathbb{R}$ such that X - c is of type \mathscr{L} .

Proof. As in Proposition 2 in [13], by the Hadamard factorization theorem,

$$\mathbb{E}\exp(zX) = \exp(bz^2/2 + cz) \prod_j (1 + \alpha_j z^2/2)$$
(4)

for some $c \in \mathbb{R}$, $b \ge 0$ and $0 < \alpha_1 \le \alpha_2 \le \ldots$ with $\sum \alpha_j < \infty$ (the set $\{\alpha_j\}$ may be empty, finite or infinite countable). Thus Y = X - c is of type \mathscr{L} .

Reduction to type \mathscr{L} . Let X be a random variable of type \mathscr{L}' . Thus Y = X - c is of type \mathscr{L} . Since Y is a symmetric random variable,

$$\mathbb{E}|X|^p = \mathbb{E}|Y+c|^p = \mathbb{E}|Y+c\varepsilon|^p \tag{5}$$

where ε is a Rademacher random variable independent of Y. Note that $Y + c\varepsilon$ is of type \mathscr{L} (as a sum of two independent random variables of type \mathscr{L}). Assuming the theorem holds for all random variables of type \mathscr{L} , we thus conclude it for X because X has the same moments as $Y + c\varepsilon$.

Proof for type \mathscr{L} . Let X be a random variable of type \mathscr{L} . Let r_k be given by (3) and as in the Rademacher case, it suffices to show that the sequence $(r_k)_{k=0}^{\infty}$

is log-concave. Note that (4) holds with c = 0, so applying it to $z = \sqrt{2t}$ with $t \ge 0$, we obtain

$$\sum_{k=0}^{\infty} \frac{r_k}{k!} t^k = \sum_{k=0}^{\infty} \frac{\mathbb{E}X^{2k}}{(2k)!} 2^k t^k = \mathbb{E} \exp(\sqrt{2t}X) = \exp(bt) \prod_j (1+\alpha_j t).$$

Moreover, as in the Rademacher case, $\prod_j (1 + \alpha_j t) = \sum_{k=0}^{\infty} \sigma_k t^k$, where $\sigma_k = \sigma_k(\{\alpha_j\})$ is the *k*th elementary symmetric function. Expanding $\exp(bt)$, the right hand side then becomes

$$\left(\sum_{k=0}^{\infty} \frac{b^k}{k!} t^k\right) \left(\sum_{k=0}^{\infty} \sigma_k t^k\right) = \sum_{k=0}^{\infty} \left(\sum_{j=0}^{k} \frac{b^{k-j}}{(k-j)!} \sigma_j\right) t^k.$$

By comparing the coefficients,

$$r_k = k! \sum_{j=0}^k \frac{b^{k-j}}{(k-j)!} \sigma_j = \sum_{j=0}^k \binom{k}{j} b^{k-j} \sigma_j j!.$$

The sequence $(b^k)_{k=0}^{\infty}$ is trivially log-concave and, as noted in the Rademacher case, the sequence $(\sigma_k k!)_{k=0}^{\infty}$ is also log-concave. Walkup's theorem (see, e.g. [5, 9, 10, 22]) which says that the binomial convolution preserves log-concavity yields that the sequence $(r_k)_{k\geq 0}$ is also log-concave and the proof of Theorem 1 is complete.

Incidentally, since $(r_n)_{n\geq 0}$ being log-concave means that X is ultra sub-Gaussian (see Section 5 and [10]), we have obtained the following result.

Theorem 7. If a random variable X is of type \mathscr{L}' , then the sequence $(r_n)_{n=0}^{\infty}$ defined in (3) is log-concave. In particular, type \mathscr{L} random variables are ultra sub-Gaussian.

Proof. It has been shown above that if X is of type \mathscr{L} , then X is ultra sub-Gaussian. Consequently, by Lemma 6 and (5), the same holds if X is of type \mathscr{L}' .

2.3 Proof of Corollary 3

First suppose that H is real Hilbert space. Fix an orthonormal basis e_1, e_2, \ldots in H. Let g_1, g_2, \ldots be i.i.d. standard Gaussian random variables (independent of the X_i). Consider the map

$$H \ni x \mapsto c_q \sum_k \langle x, e_k \rangle g_k \in L_q(\Omega, \mathbb{P}).$$

Since sums of independent Gaussians are Gaussian, this map with $c_q = ||G||_q^{-1}$ is an isometrical embedding. Given vectors v_1, \ldots, v_n in H, for appropriate scalars a_j , we thus have

$$\left\|\sum_{j} X_{j} v_{j}\right\|^{q} = \mathbb{E}_{g} \left|\sum_{j} a_{j} X_{j}\right|^{q}, \qquad a_{j} = c_{q} \left\langle v_{j}, \sum_{k} g_{k} e_{k} \right\rangle.$$
(6)

Using this twice, Theorem 1 applied to $\sum_j a_j X_j$ and Minkowski's inequality, we arrive at

$$\mathbb{E}_{X} \left\| \sum_{j} X_{j} v_{j} \right\|^{q} = \mathbb{E}_{g} \mathbb{E}_{X} \left| \sum_{j} a_{j} X_{j} \right|^{q}$$

$$\leq \left(\frac{\|G\|_{q}}{\|G\|_{p}} \right)^{q} \mathbb{E}_{g} \left(\mathbb{E}_{X} \left| \sum_{j} a_{j} X_{j} \right|^{p} \right)^{q/p}$$

$$\leq \left(\frac{\|G\|_{q}}{\|G\|_{p}} \right)^{q} \left(\mathbb{E}_{X} \left(\mathbb{E}_{g} \left| \sum_{j} a_{j} X_{j} \right|^{q} \right)^{p/q} \right)^{q/p}$$

$$= \left(\frac{\|G\|_{q}}{\|G\|_{p}} \right)^{q} \left(\mathbb{E}_{X} \left\| \sum_{j} X_{j} v_{j} \right\|^{p} \right)^{q/p}.$$

This finishes the proof when H is real. For the complex case, the only required modification of the above argument is in the definition of the isometrical embedding which now reads

$$H \ni x \mapsto c_q \sum_k \left(\operatorname{Re}\langle x, e_k \rangle g_k + \operatorname{Im}\langle x, e_k \rangle g'_k \right) \in L_q(\Omega, \mathbb{P}),$$

where $g_1, g'_1, g_2, g'_2, \ldots$ are i.i.d. standard Gaussian random variables. Identity (6) holds with $a_j = c_q \left(\operatorname{Re}\langle v_j, \sum_k g_k e_k \rangle + \operatorname{Im}\langle v_j, \sum_k g'_k e_k \rangle \right)$ and the rest of the argument is unchanged.

2.4 Proof of Corollary 4

Let (X_1, \ldots, X_n) be a random vector with distribution given by (2) and let ε be an independent Rademacher random variable. By Proposition 6 from [15], the vector $(Y_0, Y_1, \ldots, Y_n) = (\varepsilon, \varepsilon X_1, \ldots, \varepsilon X_n)$ has distribution ρ' of the form

$$d\rho'(x_0, x_1, \dots, x_n) = Z'^{-1} \exp\left(\sum_{j,k=0}^n J'_{jk} x_j x_k\right) d\mu_0(x_0) d\mu_1(x_1) \dots d\mu_n(x_n),$$

where μ_0 is the distribution of ε , $J'_{0,0} = 0$, $J'_{0,k} = J'_{k,0} = h_k/2$, $J'_{jk} = J_{jk}$, $j,k \ge 1$, so of the form (2) with $h \equiv 0$. Therefore, by Theorem 2 from [14], for every $a_0, a_1, \ldots, a_n \ge 0$, the sum $S = \sum_{j=0}^n a_j Y_j = a_0 \varepsilon + \sum_{j=1}^n a_j \varepsilon X_j$ is of type \mathscr{L} and in particular, S satisfies (1). Hence, taking $a_0 = 0$ yields that $\sum_{j=1}^n a_j X_j$ also satisfies (1).

3 Examples of type \mathscr{L} random variables

We list some examples of probability distributions of type \mathscr{L} . This is based on several well-known sufficient conditions for Fourier transforms of positive functions to have only real zeros. We mention merely in passing that the development of such conditions, related to the study of the Laguerre-Pólya class, was initiated mainly with connections to the Riemann hypothesis and refer to the comprehensive survey [1]. In what follows, X is a symmetric random variable.

(a) Let X be integer-valued with $\mathbb{P}(X=0) = p_0$ and $\mathbb{P}(X=-k) = \mathbb{P}(X=k) = p_k, k = 1, ..., n$ for nonnegative $p_0, ..., p_n$ with $p_0 + 2\sum_{k=1}^n p_k = 1$.

If $\frac{1}{2}p_0 \leq p_1 \leq \cdots \leq p_n$, then $\mathbb{E}\cos(zX) = p_0 + \sum_{k=1}^n (2p_k)\cos(kz)$ has only real zeros, as it follows from the Eneström-Kakeya theorem (see, e.g. Problem III.204 in [16]). As a result, X is of type \mathscr{L} . In particular, if X is uniform on $\{-n, \ldots, 1, 1, \ldots, n\}$ with a possible atom at 0 satisfying $\mathbb{P}(X=0) \leq \frac{1}{n+1}$, then X is of type \mathscr{L} (the last class was also recently studied in [6] in the context of Khinchin inequalities).

By the symmetry of X, the polynomial $Q(w) = \mathbb{E}w^{X+n}$ is self-inversive (the sequence of its coefficients is a palindrome, in other words, $w^{2n}Q(1/w) = Q(w)$). In particular, all its roots are symmetric with respect to the unit circle, that is if w_0 is a root of Q, then so is $1/w_0$. For instance, if for some $\alpha \geq 1$,

$$\frac{1}{2}p_0^{\alpha} + \sum_{k=1}^{n-1} p_k^{\alpha} \le \left(\frac{2}{n-2}\right)^{\alpha-1} p_n^{\alpha},$$

where n is the number of nonzero coefficients of Q, then Q has zeros only on the unit circle, so X is of type \mathscr{L} , which follows from the main result of [18]. In general, the Jury test provides an efficient algorithm to determine whether Q has zeros only on the unit circle, given its coefficients, see [8].

- (b) Let X take values in [-1,1] and have a density f (which is even). Each of the following conditions implies that X is of type \mathscr{L} .
 - (i) f is nondecreasing on (0, 1).
 - (ii) f is C^2 with f' < 0 and f'' < 0 on (0, 1).

(iii) $f(t) = h(t)^{\alpha}$, where $\alpha > -1$ and h is an entire even function which is real-valued on the real line, h(1) = 0, h(0) > 0 and h'(iz) is in the Laguerre-Pólya class. In particular, $f(t) = \text{const} \cdot (1 - t^{2m})^{\alpha}$ for a nonnegative integer m.

Moreover, if X has a density on \mathbb{R} of the following form, then it is of type \mathscr{L} .

(iv) $f(t) = \text{const} \cdot e^{-t^{2m}}$ for a nonnegative integer m.

(v)
$$f(t) = (2\pi)^{-1/2} e^{-t^2/2} (1-b+bt^2), \ 0 \le b \le 1.$$

Condition (i) is justified again by the Eneström-Kakeya theorem combined with a limit argument (see, e.g. Problem III.205 in [16]), (ii) is due to Pólya (see, e.g. Problem V.173 in [17]), (iii) is due to Ilieff (see [7]), which gives (iv) by a limit argument (see also Problem V.170 and 171 in [17]), (v) is justified by a direct computation of the moment generating function which is $(1+bz^2)e^{z^2/2}$. Moreover, if the density of X is of the form $f(t) = \text{const} \cdot e^{-|t|^{\alpha}}$ with $\alpha \geq 2$, $\alpha \notin \{2, 4, ...\}$, then its characteristic function has infinitely many non-real zeros, in particular X is not of type \mathscr{L} (see the solution of Problem V.171 in [17]).

We refer to [13] for additional examples important in statistical mechanics.

4 Type \mathscr{L} random variables with "enough Gaussianity"

The goal in this section is to prove Theorem 5. We need a little bit of preparation. For a real parameter b, consider the following function

$$\phi_b(t) = e^{-t^2/2}(1 - bt^2), \qquad t \in \mathbb{R},$$

which is in L_1 . By the Fourier inversion formula, we have,

$$f_b(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \phi_b(t) e^{-itx} dt = (1 - b + bx^2) \frac{e^{-x^2/2}}{\sqrt{2\pi}}, \qquad x \in \mathbb{R}$$

Clearly, f_b is nonnegative on \mathbb{R} if and only if $b \in [0, 1]$. Thus ϕ_b is the characteristic function of a random variable precisely for every $b \in [0, 1]$ and then its density is f_b and variance is the coefficient $2\phi_b(it)[t^2] = 1 + 2b$. Throughout, Z_b is a random variable with characteristic function ϕ_b . In particular, Z_0 is standard Gaussian and Z_1 has the bimodal density $(2\pi)^{-1/2}x^2e^{-x^2/2}$.

Lemma 8. For every a > 0 and $b_1, b_2, \ldots \ge 0$ with $\sum b_j \le a$, the function

$$e^{-at^2/2} \prod_j (1-b_j t^2)$$

is the characteristic function of a type $\mathscr L$ random variable.

Proof. Let X_1, X_2, \ldots be i.i.d. copies of Z_1 and let Z_0 be an independent standard Gaussian. Then $\sum \sqrt{b_j}X_j + \sqrt{a - \sum b_j}Z_0$ (which converges in L_2 if infinitely many b_j are positive) is the desired random variable.

Lemma 9. For $\lambda \in (0,1)$, let g_{λ} be the density of $\sqrt{\lambda}X_1 + \sqrt{1-\lambda}X_2$, where X_1, X_2 are independent copies of Z_1 . Then for every $0 < \lambda_1 < \lambda_2 < \frac{1}{2}$, the function $g_{\lambda_2} - g_{\lambda_1}$ on $(0, +\infty)$ has exactly two zeros and the sign pattern +-+.

Proof. By a direct computation,

$$g_{\lambda}(x) = \left(x^2 + \lambda(1-\lambda)(3-6x^2+x^4)\right) \frac{e^{-x^2/2}}{\sqrt{2\pi}}$$

so $g_{\lambda_2} - g_{\lambda_1}$ has the same sign as $(\lambda_2 - \lambda_1)(1 - \lambda_1 - \lambda_2)(3 - 6x^2 + x^4)$.

Lemma 10. Let X_1, X_2, \ldots be i.i.d. copies of Z_1 and let Y be a symmetric random variable independent of the X_j . Then the function

$$\Psi(b_1,\ldots,b_n) = \mathbb{E}|\sqrt{b_1}X_1 + \cdots + \sqrt{b_n}X_n + Y|^p$$

is Schur-concave on $[0+\infty)^n$.

Proof. We use the technique of interlacing densities (see, e.g. [2] or [11]). Let $h(x) = |x+1|^p + |x-1|^p$. It suffices to show that for every $0 < \lambda_1 < \lambda_2 < \frac{1}{2}$, we have

$$\int_0^\infty h(x)(g_{\lambda_2(x)} - g_{\lambda_1}(x)) \mathrm{d}x \ge 0,$$

where g_{λ} is as in Lemma 9. For arbitrary α, β ,

$$\int (\alpha x^2 + \beta)(g_{\lambda_2}(x) - g_{\lambda_1}(x)) \mathrm{d}x = 0.$$

This is because g_{λ} is a probability density, so $\int g_{\lambda}(x) dx = 1$ and by its definition, regardless of the value of λ , it has the same variance as Z_1 , so $\int x^2 g_{\lambda}(x) dx =$ Var (Z_1) . Thus the desired inequality is equivalent to

$$\int_0^\infty \tilde{h}(|x|)(g_{\lambda_2(x)} - g_{\lambda_1}(x)) \mathrm{d}x \ge 0,$$

with $\tilde{h}(x) = h(x) + \alpha x^2 + \beta$. Let x_1, x_2 be the zeros of $g_{\lambda_2(x)} - g_{\lambda_1}(x)$. Choose α and β such that \tilde{h} has zeros at x_1 and x_2 . Since for $p \geq 3$, $\tilde{h}(\sqrt{x})$ is convex on $(0, +\infty)$, \tilde{h} on $(0, +\infty)$ has no other zeros and the sign pattern + - +. Thus the integrand is pointwise nonnegative, hence the result.

Theorem 5 is a rather straighforward consequence of the above lemma.

Proof of Theorem 5. By a standard approximation argument of truncating the infinite product if necessary, we can assume that only finitely many b_j are nonzero, say $b_1, \ldots, b_n > 0$. By homogeneity, we can also assume that $\sum b_j = 1$. Set a = 1+v with $v \ge 0$. Then X has the same distribution as $\sum \sqrt{b_j}X_j + \sqrt{v}Z_0$, where X_1, X_2, \ldots are i.i.d. copies of Z_1 and Z_0 is an independent standard Gaussian. Note that

$$\sigma^{2} = \operatorname{Var}(X) = \sum b_{j} \operatorname{Var}(X_{j}) + v \operatorname{Var}(Z_{0}) = \operatorname{Var}(X_{1}) + v = 3 + v.$$

By Lemma 10,

$$\mathbb{E}|X|^p \le \mathbb{E}\left|\sum_{j=1}^n \frac{1}{\sqrt{n}} X_j + \sqrt{v} Z_0\right|^p.$$

Moreover, also by Lemma 10, the right hand side is nondecreasing as n increases. Letting $n \to \infty$ and invoking the central limit theorem gives the upper bound.

For the lower bound, again thanks to Lemma 10,

$$\mathbb{E}|X|^p \ge \mathbb{E}|Z_1 + \sqrt{v}Z_0|^p.$$

The random variable $\sigma^{-1/2}(Z_1 + \sqrt{v}Z_0)$ has the same distribution as $(1 + 2b)^{-1/2}Z_b$ with $b = \frac{1}{1+v}$, by comparing the characteristic functions. By a direct computation (using the density of Z_b),

$$\mathbb{E}|(1+2b)^{-1/2}Z_b|^p = \frac{2^{p/2}\Gamma\left(\frac{p+1}{2}\right)}{\sqrt{\pi}}\frac{1+pb}{(1+2b)^{p/2}}$$

and we check that the right hand side as a function of b with $p \ge 3$ fixed is decreasing, which finishes the proof.

5 Ultra sub-Gaussianity and related notions

Here we only consider real-valued random variables. Recall that after [10], a random variable X is called **ultra sub-Gaussian** if it is symmetric, has all

moments finite and the sequence $(a_n)_{n=0}^{\infty}$ defined by $a_0 = 1$, $a_n = \frac{\mathbb{E}X^{2n}}{\mathbb{E}G^{2n}}$, $n \ge 1$, is log-concave. Note that then the function

$$f_X(t) = \mathbb{E}\exp(\sqrt{t}X) = \sum_{n=0}^{\infty} \frac{\sqrt{t^{2n}}\mathbb{E}X^{2n}}{(2n)!} = \sum_{n=0}^{\infty} \frac{\mathbb{E}X^{2n}}{(2n-1)!!} \frac{t^n}{2^n n!} = \sum_{n=0}^{\infty} \frac{a_n}{n!} (t/2)^n$$

is $C^{\infty}(\mathbb{R})$ and the inequalities $a_n^2 \ge a_{n-1}a_{n+1}$, $n \ge 1$, are equivalent to

$$(f_X^{(n)}(0))^2 \ge f_X^{(n-1)}(0)f_X^{(n+1)}(0), \qquad n \ge 1.$$
 (7)

After [4], an entire function $f: \mathbb{C} \to \mathbb{C}$ is called **strongly log-concave** if for every $n \ge 0$, $f^{(n)}$ is either zero or $\log f^{(n)}$ is concave on \mathbb{R}_+ . Thus we shall say that a random variable is **strongly log-concave** if it is symmetric, $z \mapsto$ $\mathbb{E} \exp(zX)$ defines an entire function and $f_X^{(n)}$ is log-concave for every $n \ge 0$, where $f_X(t) = \mathbb{E} \exp(\sqrt{t}X)$. Then, in particular, inequalities (7) hold and thus X is also ultra sub-Gaussian. Remarkably, the converse is also true by the following result of Gurvits (see Proposition 4.2(ii) in [4]).

Theorem 11 (Gurvits, [4]). If a nonnegative sequence $(a_n)_{n=0}^{\infty}$ is log-concave, then the power series $f(t) = \sum_{n=0}^{\infty} \frac{a_n}{n!} t^n$ defines a C^{∞} function which is log-concave on \mathbb{R}_+ .

Gurvits obtains this as a by-product of his characterisation of *shifts* preserving log-concavity. He also indicates a very different, short argument (see [4, 5]): by Shephard's theorem (see [19]), if a finite sequence $(a_n)_{n=0}^{n_0}$ is log-concave, then there are simplices K_1, K_2 in \mathbb{R}^N such that $p_N(t) = \sum_{n=0}^N {N \choose n} a_n t^n$ is of the form $\operatorname{vol}_N(K_1 + tK_2)$. By the Brunn-Minkowski inequality, $p_N(t)^{1/N}$ is concave, which gives $(p_N(t)')^2 \geq \frac{N}{N-1}p_N(t) \cdot p_N(t)''$. Fix s > 0, let t = s/N with $N \to \infty$ to get $(p'(s))^2 \geq p'(s)p''(s)$ with $p(s) = \sum_{n=0}^{n_0} \frac{a_n}{n!} s^n$, allowing to conclude the result. We present a third, rather elementary proof.

Proof of Theorem 11. Thanks to log-concavity, (a_n) is majorised by a geometric sequence, so the radius of convergence of $\sum \frac{a_n}{n!} t^n$ is ∞ , hence f is C^{∞} on \mathbb{R} .

We have $f^{(k)}(t) = \sum_{n \ge 0} \frac{a_{n+k}}{n!} t^n$. The log concavity of f is equivalent to $(f')^2 \ge f'' f$, that is

$$\left(\sum_{n\geq 0}\frac{a_{n+1}}{n!}t^n\right)^2 - \left(\sum_{n\geq 0}\frac{a_n}{n!}t^n\right)\left(\sum_{n\geq 0}\frac{a_{n+2}}{n!}t^n\right) \geq 0.$$

The left hand side is a power series with coefficient in front of t^n equal to

$$\sum_{k+l=n} \frac{a_{k+1}}{k!} \cdot \frac{a_{l+1}}{l!} - \sum_{k+l=n} \frac{a_k}{k!} \cdot \frac{a_{l+2}}{l!}.$$

It is enough to show that this expression is non-negative. Multiplying both sides by n!, we see that it is enough to prove the inequality

$$\sum \binom{n}{k} a_{k+1} a_{n-k+1} - \sum \binom{n}{k} a_k a_{n-k+2} \ge 0$$

with the usual convention that $\binom{n}{k} = 0$ if k < 0 or k > n. The sums are over \mathbb{Z} . Shifting the summation index in the first sum we get an equivalent form

$$\sum \left(\binom{n}{k-1} - \binom{n}{k} \right) a_k a_{n-k+2} \ge 0.$$

For n = 0 this is just $a_1^2 \ge a_0 a_2$ and for n = 1, it is $2a_1 a_2 \ge a_0 a_3 + a_1 a_2$, which is just $a_1 a_2 \ge a_0 a_3$. By log-concavity, both hold. Thus in what follows we can assume that $n \ge 2$.

Let $b_k = a_k a_{n-k+2}$ (we set $a_k = 0$ for k < 0). Our goal is to prove

$$\sum \left(\binom{n}{k-1} - \binom{n}{k} \right) b_k \ge 0.$$

Note that $b_k = b_{n-k+2}$. From log concavity we know that the sequence b_k is nondecreasing, $b_{k-1} \leq b_k$ for $k \leq \frac{n}{2} + 1$. Since b_k is symmetric about $k = \frac{n}{2} + 1$, we can write the desired inequality in an equivalent form

$$\left(\binom{n}{\frac{n}{2}} - \binom{n}{\frac{n}{2}-1}\right)b_{\frac{n}{2}+1} + \sum_{k<\frac{n}{2}+1}\left(2\binom{n}{k-1} - \binom{n}{k} - \binom{n}{k-2}\right)b_k \ge 0$$

with the first term present only if n is even. To prove this, we set

$$s_k = 2\binom{n}{k-1} - \binom{n}{k} - \binom{n}{k-2}, \qquad k < \frac{n}{2} + 1$$

and additionally when n is even,

$$s_{n/2+1} = \binom{n}{\frac{n}{2}} - \binom{n}{\frac{n}{2} - 1}$$

and use the following elementary lemma (Lemma 6 in [10])

Lemma 12 (Nayar-Oleszkiewicz, [10]). If $b_0 \leq b_1 \leq \ldots \leq b_m$ and real numbers s_0, s_1, \ldots, s_m are such that $\sum_{k=0}^m s_k = 0$ and the sequence $(\operatorname{sgn}(s_k))$ is nondecreasing, then $\sum_{k=0}^m s_k b_k \geq 0$.

The lemma is applied with $m = \lfloor n/2 + 1 \rfloor$ and the sequence (s_k) defined above (which plainly satisfies $\sum_{k=0}^{m} s_k = 0$). It remains to check that $(sgn(s_k))$ is a nondecreasing sequence. For $n \ge 2$ the inequality $s_k \le 0$ is equivalent to

$$-4k^2 + 4kn + 8k - n^2 - 3n - 2 \le 0.$$

The maximum of this parabola is attained for $k = \frac{n}{2} + 1$ and thus for $k \leq \frac{n}{2} + 1$ this quadratic function is increasing. This shows that $(\text{sgn}(s_k))$ is non-decreasing, which finishes the proof.

This gives that the classes of strongly log-concave and ultra sub-Gaussian random variables are the same. As observed earlier in Theorem 7, type \mathscr{L} random variables are ultra sub-Gaussian. The converse is false: it is enough to consider X with 3 atoms: $\mathbb{P}(X=0) = 1 - \beta$, $\mathbb{P}(X=\pm 1) = \beta/2$. Then $\mathbb{E} \exp(zX) = (1-\beta) + \beta \cosh(z)$. The equation $(1-\beta) + \beta \cosh(z) = 0$ has only purely imaginary zeroes only if $\beta > 1/2$. On the other hand, this random variables is ultra sub-Gaussian for any $\beta \in (0, 1)$. Another examples are random variables with densities proportional to $e^{-|x|^{\alpha}}$: they are ultra sub-Gaussian for every $\alpha \ge 2$ (see Corollary 2 in [10]), whereas they are of type \mathscr{L} only for integral α (see Section 3). This discussion is summarised in Figure 1.



Figure 1: Implications between ultra sub-Gaussian (USG), strongly log-concave (SLC) and type \mathscr{L} random variables.

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