On the average order of the lattice rest of a convex body

by

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1. Introduction. Let $B_s$ denote a compact convex subset of $\mathbb{R}^s$, $s \geq 2$, which contains the origin as an inner point. Suppose that the boundary $\partial B_s$ of $B_s$ is an $(s - 1)$-dimensional surface of class $C^{6s+13}$ with finite nonzero Gaussian curvature throughout. For $x > 0$ define $A_{B_s}(x)$ as the number of points of the lattice $\mathbb{Z}^s$ in the “blown up” domain $xB_s$, i.e. $A_{B_s}(x) = \# (xB_s \cap \mathbb{Z}^s)$, and $P_{B_s}(x)$ as the “lattice rest” (1)

\[ P_{B_s}(x) = A_{B_s}(x) - \text{vol}(B_s) x^s. \]

For such a general convex body $B_s$ E. Hlawka [3] proved that

\[ P_{B_s}(x) = O(x^{s(s−1)/(s+1)}) \]

and

\[ P_{B_s}(x) = \Omega(x^{(s−1)/2}). \]

In the last years both estimates have been improved (see [11, 12, 16–18] and for planar domains [5, 15]).

In the present article we study the mean square of $P_{B_s}(x)$. For planar domains this problem has been attacked by W. G. Nowak [14] who proved

\[ \int_0^X |P_{B_s}(x)|^2 \, dx = O(X^2). \]

This bound is best possible since in the special case of a circle $S_2$, say, it

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(1) Note that many authors use $t = x^2$ as parameter. This means that the domain $B_s$ is “blown up” by the factor $x = \sqrt{t}$. 
can be replaced by the asymptotic formula (see I. Kátai [10] (2))

\[ \int_0^X |P_{S_2}(x)|^2 \, dx = c_2 X^2 + O(X(\log X)^2) \quad (c_2 > 0). \]

For \( s \geq 3 \) Nowak’s approach does not give satisfactory bounds. To see what one should expect in higher dimensions let us consider the case of the \( s \)-dimensional unit ball \( S_s \). This case (together with the case of rational ellipsoids) is particularly well understood. For \( s \geq 4 \), A. Walfisz [20] proved

\[ \int_0^X |P_{S_s}(x)|^2 \, dx = c_s X^{2s-3} + O(R_s) \quad (c_s > 0), \]

where \( R_s = X^{2s-5} \) if \( s > 5 \), \( R_5 = X^5(\log X)^2 \) and \( R_4 = X^4 \log X \). In the case \( s = 3 \), V. Jarník [9] obtained the asymptotic formula

\[ \int_0^X |P_{S_3}(x)|^2 \, dx = c_3 X^3 \log X + O(X^3(\log X)^{1/2}) \quad (c_3 > 0). \]

The proof of both results uses the fact that the generating function of the number of lattice points on the sphere is a theta function. It therefore cannot be carried over to our general situation. Using a different method we prove the following higher dimensional analogue to (2).

**Theorem.** Let \( B_s \) denote a compact convex subset of \( \mathbb{R}^s \), \( s \geq 4 \), which contains the origin as an inner point. Suppose that the boundary \( \partial B_s \) of \( B_s \) is an \((s-1)\)-dimensional surface of class \( C^{6s+13} \) with finite nonzero curvature throughout. Then for every fixed \( \varepsilon > 0 \),

\[ \int_0^X |P_{B_s}(x)|^2 \, dx = O(X^{2s-3+\varepsilon}). \]

The theorem shows that spherical balls belong to those smooth convex bodies with nonzero curvature which have (on average) large lattice rest. In general the lattice rest can be much smaller. For instance, V. Jarník [8] proved that for \( s \geq 4 \) almost all ellipsoids

\[ \mathcal{E}_s(a) = \left\{ (x_1, \ldots, x_s) \in \mathbb{R}^s : \sum_{i=1}^s a_i x_i^2 \leq 1 \right\} \quad (a = (a_1, \ldots, a_s) \in \mathbb{R}_+^s) \]

(in the sense of Lebesgue) satisfy \( P_{\mathcal{E}_s(a)}(x) = O(x^{s/2+\varepsilon}) \) for every \( \varepsilon > 0 \).

The method used to prove (5) seems to be new. Starting from a trivial bound for the mean square of \( P_{B_s^\ast}(x) \), where \( B_s^\ast \) denotes the polar body of

\[ \text{(2)} \text{ Actually Kátai considers the mean square of } P_{S_2}(\sqrt{t}). \text{ Using integration by parts it is easy to see that his formulation is equivalent to ours. The same remark applies to (3) and (4).} \]
$B_s$, it produces a nontrivial bound for the mean square of $P_{B_s}(x)$. Since $B_s^{**} = B_s$, the same method applied to $B_s^*$ gives a better bound for the mean square of $P_{B_s^*}(x)$. Iterating this process yields (5). Unfortunately, the iteration process produces better bounds only if $s \geq 4$. The interesting case $s = 3$ remains open.

2. Geometrical preliminaries. Let $B_s$ denote a compact convex subset of $\mathbb{R}^s$, $s \geq 2$, which contains the origin as an inner point. The polar body $B_s^*$ of $B_s$ is defined by

$$B_s^* = \{ x \in \mathbb{R}^s : \sup_{y \in B_s} \langle x, y \rangle \leq 1 \},$$

where $\langle \cdot, \cdot \rangle$ denotes the standard scalar product in $\mathbb{R}^s$. The polar body of a compact convex body which contains the origin as an inner point is again a compact convex body which contains the origin as an inner point. Furthermore, $B_s^{**} = B_s$ (cf. [13], p. 65 ff). In order to apply the iterative process simultaneously to $B_s$ and $B_s^*$ we have to check that, if $\partial B_s$ is smooth with nonzero curvature throughout, then the same is true for $\partial B_s^*$. The following lemma is a multi-dimensional analogue of Huxley [6], Lemma 4, where only planar domains are considered.

**Lemma 1.** Let $x = x(t), t = (t_1, \ldots, t_{s-1})$, be a regular (local) parametrization of $\partial B_s$ of class $C^k$, $k \geq 2$, and $u = u(t)$ the unit outward normal vector of the tangent hyperplane at $x(t)$. Suppose that the Gaussian curvature $\kappa(x)$ of $\partial B_s$ at $x$ is nonzero. Then

$$x^*(t) = \langle x(t), u(t) \rangle^{-1} u(t)$$

is a regular (local) parametrization of $\partial B_s^*$ of class $C^{k-1}$ and

$$(6) \quad |\kappa^*(x^*)\kappa(x)| = (\|x\| \cdot \|x^*\|)^{-s-1},$$

where $\kappa^*(x^*)$ denotes the Gaussian curvature of $\partial B_s^*$ at $x^*$. Furthermore, the spherical map of $\partial B_s$ (e.g. the map which sends a point $x$ of $\partial B_s$ to the endpoint of the outward normal vector $u$ on the unit sphere $\partial S_s$) and the spherical map of $\partial B_s^*$ are one-to-one and of class $C^{k-1}$.

**Proof.** Here regular means that the vectors $v_i = \partial x/\partial t_i, 1 \leq i \leq s-1$, form a basis of the $(s-1)$-dimensional tangent space $T_x$ at $x$. Then $u$ is (as a function of $v_i, 1 \leq i \leq s-1$) of class $C^{k-1}$. From the fact that the pole of every tangent hyperplane (with respect to the unit sphere) is a point on $\partial B_s^*$ (cf. [13], p. 67) we obtain

$$x^* = x^*(t) = \langle x, u \rangle^{-1} u \in \partial B_s^*.$$ 

This shows that $x^*$ is of class $C^{k-1}$. Note that $\langle x, u \rangle > 0$ since the origin is an inner point of $B_s$. 

The Gaussian curvature \( \kappa(x) \) of \( \partial B_s \) at \( x \) is defined as the determinant of the linear map \( L_x : T_x \rightarrow T_x \) which maps \( v \in T_x \) to \( L_x(v) = \nabla_v u \), where \( \nabla_v u \) denotes the derivative of the normal vector \( u \) with respect to the direction \( v \). Its sign depends on the orientation of the surface \( \partial B_s \). The linear map is determined by \( L_x(v_i) = \partial u/\partial t_i \in T_x \). It can be extended linearly from \( T_x \) to the entire \( s \)-dimensional space by setting \( L_x(u) = u \). This does not change the absolute value of the determinant. Hence, if the matrix \( B \in \mathbb{R}^{s \times s} \) is defined by

\[
(7) \quad \left( \frac{\partial u}{\partial t_1}, \ldots, \frac{\partial u}{\partial t_{s-1}}, u \right) = B(v_1, \ldots, v_{s-1}, u),
\]

then \( |\det B| = |\kappa(x)| > 0 \). It follows that \( \partial u/\partial t_1, \ldots, \partial u/\partial t_{s-1}, u \) are linearly independent. This implies that the tangent vectors

\[
(8) \quad v_i^* = \frac{\partial x^*}{\partial t_i} = (x, u)^{-1} \partial u/\partial t_i - (x, u)^{-2}(x, \partial u/\partial t_i) \quad (1 \leq i \leq s-1)
\]

are linearly independent, proving that the parametrization \( x^*(t) \) of \( \partial B^*_s \) is regular. Since \( \langle v_i^*, x \rangle = 0 \) for \( 1 \leq i \leq s-1 \), we find that \( u^* = \|x\|^{-1}x \) is normal to the tangent space \( T^*_x \) of \( \partial B^*_s \) at \( x^* \). Representing \( u^* \) as

\[
u^* = \|x\|^{-1}x = \sum_{i=1}^{s-1} \alpha_i \frac{\partial u}{\partial t_i} + \langle x, u \rangle \frac{\|x\|}{\langle x, u \rangle} \quad u \]

with some \( \alpha_i \in \mathbb{R} \) we conclude that

\[
u^* = \sum_{i=1}^{s-1} \alpha_i v_i^* + \langle x, u \rangle^{-1} \left( \sum_{i=1}^{s-1} \alpha_i \frac{\langle x, \partial u/\partial t_i \rangle}{\langle x, u \rangle} + \langle x, u \rangle^2 \right) \quad u \]

\[
= \sum_{i=1}^{s-1} \alpha_i v_i^* + \frac{\|x\|}{\langle x, u \rangle} \quad u.
\]

Together with (7) and (8) this yields

\[
\det(v_1^*, \ldots, v_{s-1}^*, u^*) = \det \left( v_1^*, \ldots, v_{s-1}^*, \|x\| u \right) = \frac{\|x\|}{\langle x, u \rangle} \det \left( \langle x, u \rangle^{-1} \frac{\partial u}{\partial t_1}, \ldots, \langle x, u \rangle^{-1} \frac{\partial u}{\partial t_{s-1}}, u \right) = \|x\| \det(B) \det(v_1, \ldots, v_{s-1}, u).
\]

If \( B^* \in \mathbb{R}^{s \times s} \) denotes the matrix defined by

\[
\left( \frac{\partial u^*}{\partial t_1}, \ldots, \frac{\partial u^*}{\partial t_{s-1}}, u^* \right) = B^*(v_1^*, \ldots, v_{s-1}^*, u^*),
\]
then $|\det B^*| = |\kappa^*(x^*)|$. Moreover, since
$$
\frac{\partial u^*_i}{\partial t_i} = \frac{\partial}{\partial t_i} (||x||^{-1} x) = ||x||^{-1} v_i - ||x||^{-2} (x, v_i) u^*
$$
we obtain
$$
\det \left( \frac{\partial u^*_i}{\partial t_1}, \ldots, \frac{\partial u^*_i}{\partial t_{s-1}}, u^* \right) = ||x||^{-s} \det (v_1, \ldots, v_{s-1}, x) = ||x||^{-s} \det (v_1, \ldots, v_{s-1}, (x, u) u)
$$
and
$$
|\kappa^*(x^*)| = \left| \det \left( \frac{\partial u^*_i}{\partial t_1}, \ldots, \frac{\partial u^*_i}{\partial t_{s-1}}, u^* \right) \right| \cdot |\det (v_1^*, \ldots, v_{s-1}^*, u^*)|^{-1}
$$
$$
= \left( \frac{(x, u)}{||x||} \right)^{s+1} |\kappa(x)|^{-1}.
$$
This proves (6). Finally, we note that the spherical map of the boundary of a compact convex domain with nonvanishing Gaussian curvature is always one-to-one (cf. [19], p. 105). It follows from the above discussion that the maps $x \mapsto u$ and $x^* \mapsto u^*$ are of class $C^{k-1}$.

**Lemma 2.** Let $B_s$ be a compact convex subset of $\mathbb{R}^s$, $s \geq 2$, which contains the origin as an inner point. Assume that the spherical map of $\partial B_s$ is one-to-one and of class $C^{2s+12}$, and that the Gaussian curvature of $\partial B_s$ is nonzero throughout. Then the Fourier transform of the indicator function $I_{x^*B_s}$,
$$
\hat{I}_{x^*B_s}(k) = \int_{x^*B_s} e^{2\pi i (k, u)} \, du \quad (k \in \mathbb{R}^s),
$$
satisfies, for $k \neq 0$,
$$
\hat{I}_{x^*B_s}(k) = (2\pi)^{-1} x^{(s-1)/2} ||k||^{-s/2} (\alpha_k e^{2\pi i x H(k) - \gamma i} + \alpha_{-k} e^{-2\pi i x H(-k) + \gamma i})
$$
$$
+ O(x^{(s-3)/2} ||k||^{-(s+3)/2}),
$$
where $\gamma = (s+1)\pi/4$, $\alpha_k = \sqrt{\kappa_k}$ ($\kappa_k$ denotes the Gaussian curvature of $\partial B_s$ at the point where the outward normal has direction $k$) and $H$ denotes the distance function of the polar body $B_s^*$.

**Proof.** This is a special case of Hlawka’s Satz 2 in [4] (see also Satz 5 in [3]). Note that the distance function of $B_s^*$ agrees with the “tag”-function of $B_s$ (cf. [13], p. 127) and that $\alpha_k \asymp 1$.

**3. The basic estimate of $P_{B_s}(x)$.** As usual it is easier to estimate a smoothed version of $P_{B_s}$ rather than $P_{B_s}$ itself. To this purpose we introduce a weight function $\delta : \mathbb{R}^s \rightarrow [0, \infty)$ which has the following properties (see [1], p. 88, or [7] for the existence of such a function):
(i) the support of $\delta_1$ lies inside the unit ball $S_s$.

(ii) $\int_{\mathbb{R}^s} \delta_1(y) \, dy = 1$.

(iii) the Fourier transform $\hat{\delta}_1(k) = \int_{\mathbb{R}^s} \delta_1(y)e^{2\pi i (k,y)} \, dy$, $k \in \mathbb{R}^s$, satisfies $\hat{\delta}_1(k) \ll \exp(-\|k\|^{1/2})$.

For $0 < \varepsilon < 1$ set $\delta_\varepsilon(y) = \varepsilon^{-s}\delta_1(\varepsilon^{-1}y)$, $y \in \mathbb{R}^s$, then

$$\text{supp}(\delta_\varepsilon) \subseteq \varepsilon S_s, \quad \hat{\delta}_\varepsilon(0) = 1, \quad \hat{\delta}_\varepsilon(k) = \hat{\delta}_1(\varepsilon k) \ll \exp(-\|\varepsilon k\|^{1/2})\text{.}$$

We use convolution with $\delta_\varepsilon$ to smooth the indicator function $I_{xB_s}$:

$$I_{xB_s} \ast \delta_\varepsilon(z) = \int_{\mathbb{R}^s} I_{xB_s}(y)\delta_\varepsilon(z-y) \, dy \quad (z \in \mathbb{R}^s).$$

Lemma 3. Let $B_s$ be a compact convex subset of $\mathbb{R}^s$ which contains the origin as an inner point. Denote by $\rho > 0$ the radius of any inscribed ball $\rho S_s \subseteq B_s$ and set $x_\pm = x \pm \varepsilon/\rho > 0$. Then for all $z \in \mathbb{R}^s$,

$$I_{x-B_s} \ast \delta_\varepsilon(z) \leq I_{xB_s}(z) \leq I_{x+B_s} \ast \delta_\varepsilon(z).$$

Proof. We first claim that for $d \geq 0$,

$$y \in B_s, \quad z \notin (1+d)B_s \Rightarrow \|z-y\| > d\rho.$$

To prove this (geometrically evident) fact consider the distance function $F(u) = \inf\{\lambda : u \in \lambda B_s\}$ of $B_s$. Then $y \in B_s$ if and only if $F(y) \leq 1$, and $z \notin (1+d)B_s$ if and only if $F(z) > 1+d$. Since the distance function of a convex body is convex we conclude $F(z) = F(z-y+y) \leq F(z-y) + F(y)$. This implies $F(z-y) > d$. Hence $z-y \notin dB_s \supseteq d\rho S_s$, and this proves (10). Since $0 \leq I_{x-B_s} \ast \delta_\varepsilon(x) \leq 1$ the lemma follows if we can show that

(i) $z \notin xB_s \Rightarrow I_{x-B_s} \ast \delta_\varepsilon(z) = 0$,

(ii) $z \in xB_s \Rightarrow I_{x+B_s} \ast \delta_\varepsilon(z) = 1$.

To prove (i) assume that $z \notin xB_s$ and $y \in x-B_s$; then $z = xz_0$ with $z_0 \notin B_s$, $y = x-y_0$ with $y_0 \in B_s$ and by (10)

$$\|z-y\| = x\left\|\left(1 + \frac{\varepsilon}{\rho x_-}\right)z_0 - y_0\right\| > \varepsilon.$$ 

Hence, for every $y \in \mathbb{R}^s$, $I_{x-B_s}(y)\delta_\varepsilon(z-y) = 0$. This implies $I_{x-B_s} \ast \delta_\varepsilon(z) = 0$. The proof of (ii) is similar. Note that the right hand side in (ii) is equivalent to $I_{(x,B_s) \ast \delta_\varepsilon}(z) = 0$. 

We are now in a position to derive the basic estimate of $P_{B_s}(x)$. By Lemma 3,

$$\sum_{k \in \mathbb{Z}^s} I_{x-B_s} \ast \delta_\varepsilon(k) \leq A_{B_s}(x) \leq \sum_{k \in \mathbb{Z}^s} I_{x+B_s} \ast \delta_\varepsilon(k).$$
The multi-dimensional Poisson summation formula (see [2]) yields
\[ \hat{\delta}_\varepsilon(k) = \hat{\sum}_{k \in \mathbb{Z}^s} \hat{f}(k) \leq A_{S_\varepsilon}(x) \leq \hat{\sum}_{k \in \mathbb{Z}^s} \hat{f}(k). \]

Since \( \hat{\delta}_\varepsilon(0) = 1 \) and
\[ \hat{\sum}_{k \in \mathbb{Z}^s} \hat{f}(k) = \text{vol}(B_s)x_s^s = \text{vol}(B_s)x_s^s + O(x^{s-1} \varepsilon) \]
we obtain
\[ P_{S_\varepsilon}(x) \ll \left| \sum_{k \in \mathbb{Z}^s} \hat{\sum}_{k \in \mathbb{Z}^s} \hat{f}(k) \right| + \left| \sum_{k \in \mathbb{Z}^s} \hat{\sum}_{k \in \mathbb{Z}^s} \hat{f}(k) \right| + x^{s-1} \varepsilon, \]
where \( Z^*_s = \mathbb{Z}^s \setminus \{0\} \) for short. In the right hand side we insert the asymptotic expansion of Lemma 2. For \( s \geq 4, X \leq x \leq 2X \) and \( \varepsilon \geq X^{-1} \) the contribution of the error term is at most
\[ X^{(s-3)/2} \sum_{k \in \mathbb{Z}^s} \|k\|^{-(s+3)/2} \hat{\delta}_\varepsilon(k) \ll (X \varepsilon^{-1})^{(s-3)/2} \ll X^{s-1} \varepsilon. \]
Here we used the estimate
\[ (11) \sum_{k \in \mathbb{Z}^s} \|k\|^{-\alpha} \exp(-\varepsilon\|k\|^{1/2}) = \int_{1/2}^{\infty} u^{-\alpha} \exp(-\varepsilon u^{1/2}) \, dA_{S_\varepsilon}(u) \ll \varepsilon^{\alpha-s}, \]
which is valid for \( \alpha < s \) and \( 0 < \varepsilon < 1 \). It can be proved by using integration by parts and \( A_{S_\varepsilon}(u) \ll u^s \). Thus, for \( s \geq 4, X \leq x \leq 2X \) and \( X^{-1} \leq \varepsilon < 1 \),
\[ (12) P_{S_\varepsilon}(x) \ll X^{(s-1)/2}(|S_+(x)| + |S_-(x)|) + X^{s-1} \varepsilon, \]
where
\[ S_\pm(x) = \sum_{k \in \mathbb{Z}^s} \alpha_{\pm k} \|k\|^{-(s+1)/2} \hat{\delta}_\varepsilon(k)e^{2\pi i x \pm H(k)}. \]
This is the basic estimate of \( P_{S_\varepsilon}(x) \). We remark that (12) remains true if \( s = 3 \) (in this case (11) is used with \( \alpha = s \); thus the bound \( \varepsilon^{\alpha-s} \) in (11) has to be replaced by \( |\log \varepsilon| \)). Moreover, the trivial estimate of \( S_\pm(x) \) together with (11) and the choice \( \varepsilon = X^{-(s-1)/(s+1)} \) yields (1).

4. The iterative estimation process

**Lemma 4.** Let \( \mathcal{B} = \mathcal{B}_s \) or \( \mathcal{B} = \mathcal{B}_s^* \). Assume that for some \( \gamma > 2s - 3 \) and \( X \geq 1 \),
\[ (13) \int_0^X |P_B(x)|^2 \, dx \ll X^\gamma. \]
Furthermore, let \( R \geq U > 0 \) and \( V \geq R^{(\gamma-1)/2} \) be real parameters and set
\[ M(U, V) = \# \{ j \in \mathbb{Z} : 0 \leq j \leq R/U, |P_B(R + jU)| \geq V \}. \]
Then
\[ M(U, V) \ll R^7V^{-2}(R^sV^{-1} + U^{-1}). \]
Note that for \( V \leq R^{(7-1)/2} \) the trivial bound \( M(U, V) \ll R/U \) is sharper.

**Proof.** Since \( A_{\mathcal{B}_s}(u) \) is increasing it follows that there is a constant \( c > 0 \) such that for \( R \leq v \leq u \leq 2R \),
\[ P_{\mathcal{B}}(v) - P_{\mathcal{B}}(u) \leq \text{vol}(\mathcal{B})(u^s - v^s) \leq \frac{c}{2} (u - v) R^{s-1}. \]
Hence, if \( t \in [R, 2R] \) is a value with \( |P_{\mathcal{B}}(t)| \geq V \), there exists an interval \( I \) of length \( V(cR_{s-1})^{-1} \), containing \( t \) as an endpoint, such that \( |P_{\mathcal{B}}(u)| \geq V/2 \) for all \( u \in I \). For all \( j \) counted in \( M(U, V) \) we choose (disjoint) intervals \( J_j \) of length \( |J_j| = \min(V(cR_{s-1})^{-1}, U/2) \) with \( |P_{\mathcal{B}}(u)| \geq V/2 \) for all \( u \in J_j \). Then trivially
\[ V^2 M(U, V) \min(V R^{1-s}, U) \ll \int_0^{3R} |P_{\mathcal{B}}(u)|^2 \, du \ll R^\gamma \]
and
\[ M(U, V) \ll R^7V^{-2} \max(V^{-1}R^{s-1}, U^{-1}) \ll R^7V^{-2}(R^{s-1}V^{-1} + U^{-1}). \]

In the following we assume that (13) is true for \( P_{\mathcal{B}^*} \). The basic estimate (12) is used to bound the mean square of \( P_{\mathcal{B}^*} \). Since
\[ \int_X^{2X} e^{2\pi i \beta x} \, dx \ll \min(X, |eta|^{-1}) \quad (\beta \in \mathbb{R}), \]
it follows that
\[ \int_X^{2X} |S_\pm(x)|^2 \, dx \ll \sum_{k, m \in \mathbb{Z}^*} (\|k\| \cdot \|m\|)^{-\frac{s+1}{2}} |\widehat{\delta_\varepsilon}(k)| \cdot |\widehat{\delta_\varepsilon}(m)| \min(X, |H(k) - H(m)|^{-1}). \]
Hence by (12),
\[ \int_X^{2X} |P_{\mathcal{B}^*}(x)|^2 \, dx \ll X^{s-1}(X \Sigma_1 + \Sigma_2) + X^{2s-1} \varepsilon^2, \]
where
\[ \Sigma_1 = \sum_{\substack{k, m \in \mathbb{Z}^* \\mid |H(k) - H(m)| < X^{-1}}} (\|k\| \cdot \|m\|)^{-\frac{s+1}{2}} |\widehat{\delta_\varepsilon}(k)| \cdot |\widehat{\delta_\varepsilon}(m)| \]
and
\[ \Sigma_2 = \sum_{|H(k) - H(m)| \geq X^{-1}} \sum_{k, m \in \mathbb{Z}^*} (\|k\| \cdot \|m\|)^{-(s+1)/2} |\tilde{\delta}_e(k)| \cdot |\tilde{\delta}_e(m)| \cdot |H(k) - H(m)|^{-1}. \]

To estimate \( \Sigma_1 \) and \( \Sigma_2 \) we build blocks of the form \( R < H(k) \leq 2R, \ R = 2^l, \ l \geq 0 \), and (in the second sum) \( U \leq |H(k) - H(m)| < 2U, \ U = X^{-1}2^n, \ n \geq 0 \). In the first sum the summation condition implies \( H(k) \asymp H(m) \asymp R \).

Since \( H(k) \asymp \|k\| \) it follows that with some constant \( c > 0 \),

\[ \Sigma_1 \ll \sum_{R = 2^l \geq 1} R^{-s-1} \exp(-c\varepsilon R^{1/2}) N(R, X^{-1}) + 1, \]

where
\[ N(R, U) = \#\{(k, m) \in \mathbb{Z}^{s \times s} : R < H(k) \leq 2R, H(k) - U < H(m) \leq H(k) + U\}. \]

In the second sum we find \( H(k) \asymp H(m) \asymp R \) if \( U \leq R/4 \), thus
\[ \Sigma_2 \ll \sum_{R} \sum_{X^{-1} \leq U \leq R/4} R^{-s-1} \exp(-c\varepsilon R^{1/2}) U^{-1} N(R, 2U) \]
\[ + \sum_{R} R^{-(s+3)/2} \exp(-c\varepsilon R^{1/2}) A_{B^*_s}(2R) \sum_{m \in \mathbb{Z}^*} \|m\|^{-(s+1)/2} |\tilde{\delta}_e(m)|. \]

By (11) the innermost sum of the second term is \( O(\varepsilon^{-(s+1)/2}) \). Since \( A_{B^*_s}(2R) \ll R^s \) and

\[ \sum_{l \geq 0, (R = 2^l)} R^\alpha \exp(-c\varepsilon R^{1/2}) \ll \begin{cases} \varepsilon^{-\alpha}, & \alpha > 0, \\ \log \varepsilon, & \alpha = 0, \end{cases} \]

the whole second term is \( O(\varepsilon^{2-s}) \). Hence
\[ \Sigma_2 \ll \sum_{R} \sum_{X^{-1} \leq U \leq R} R^{-s-1} \exp(-c\varepsilon R^{1/2}) U^{-1} N(R, 2U) + \varepsilon^{2-s} \]

and by (14) and (15),

\[ \int_X^{2X} |P_{B^*_s}(x)|^2 \, dx \ll X^{s-1} \sum_{R} \sum_{X^{-1} \leq U \leq R} R^{-s-1} \exp(-c\varepsilon R^{1/2}) U^{-1} N(R, 2U) \]
\[ + X^{s-1} \varepsilon^{2-s} + X^{2s-1} \varepsilon^2. \]

To estimate \( N(R, U) \) we cover the interval \( (R, 2R] \) by intervals of the form
\((R_j, R_{j+1}], R_j = R + jU, 0 \leq j < R/U + 1\), and define \((3)\)

\[
K_j = \#\{k \in \mathbb{Z}^s : R_j < H(k) \leq R_{j+1}\}.
\]

Then by Cauchy’s inequality,

\[
N(R, U) \ll \sum_j \#\{(k, m) \in \mathbb{Z}^{s \times s} : R_j < H(k) \leq R_{j+1}, R_j - 1 < H(m) \leq R_{j+2}\}
\]

\[
\ll \sum_j K_j(K_{j-1} + K_j + K_{j+1}) \ll \sum_{j<R/U+1} K_j^2.
\]

Since

\[
K_j^2 = (A_{B^*_s}(R + (j + 1)U) - A_{B^*_s}(R + jU))^2
\ll (R^{s-1}U + |P_{B^*_s}(R + (j + 1)U)| + |P_{B^*_s}(R + jU)|)^2
\ll R^{2s-2}U^2 + |P_{B^*_s}(R + (j + 1)U)|^2 + |P_{B^*_s}(R + jU)|^2
\]

we obtain

\[
N(R, U) \ll R^{2s-1}U + \sum_{0 \leq j \ll R/U} |P_{B^*_s}(R + jU)|^2.
\]

The remaining sum can be estimated by Lemma 4. Set \(V = R^{(\gamma - 1)/2}2^r \ll R^s, 0 \leq r \ll \log R\). Then

\[
N(R, U) \ll R^{2s-1}U + R^\gamma U^{-1} + \sum_{V \geq R^{(\gamma - 1)/2}} V^2 M(U, V)
\ll R^{2s-1}U + R^\gamma (\log R)U^{-1} + R^s(\gamma - 1)/2.
\]

Together with (16) and (17) this implies

\[
\int_{X}^{2X} |P_{B^*_s}(x)| \, dx \ll X^{s-1} \sum_{R \leq \sigma \leq U} \sum_{X^{-1} \leq U \leq R} \exp(-c(\varepsilon R)^{1/2})
\times (R^{s-2} + R^{\gamma-s-1}(\log R)U^{-2} + R^{(\gamma-3)/2}U^{-1})
\ll X^{s-1} \varepsilon^{2-s} + X^{2s-1} \varepsilon^2.
\]

\((3)\) I learned this trick from M. N. Huxley during the conference on Analytic and Elementary Number Theory in Vienna, July 1996, which has been organized in honour of E. Hlawka’s 80th birthday. Huxley used it to bound the mean square of the Hlawka zeta function of a planar convex body on a vertical line.
\[ \ll X^{s-1} \sum_{R} \exp(-cR^{1/2}) \]
\[ \times \left( R^{s-2} \left( \log X + \log R \right) + R^{\gamma-s-1}(\log R)X^2 + R^{(\gamma-3)/2}X \right) \]
\[ + X^{s-1} \varepsilon^{2-s} + X^{2s-1} \varepsilon^2 \]
\[ \ll X^{s-1}(\log X)\varepsilon^{2-s} + X^{s+1}(\log X)\varepsilon^{s+1-\gamma} + X^s \varepsilon^{(3-\gamma)/2} + X^{2s-1} \varepsilon^2. \]

Note that \( \gamma - s - 1 > 0 \) and \( \gamma - 3 > 0 \) if \( s \geq 4 \). In the case \( s = 3 \) the term \( X^{s+1} \log X \) is too large to give a nontrivial estimate. Balancing the second and the last term yields the optimal choice \( \varepsilon = X^{-(s-2)/(\gamma-s+1)} \geq X^{-1} \). With this choice it is easy to check that the remaining two terms are of smaller order (use \( \gamma > 2s - 3 \)), hence
\[
\int_{X}^{2X} \frac{1}{X} \left| P_{B_s}(x) \right|^2 dx \ll X^{2s-1-2(s-2)/(\gamma-s+1)} \log X.
\]

Summing over \( X2^{-j}, j \geq 1 \), we have proved that for \( s \geq 4 \) the estimate
\[ \int_{0}^{X} \frac{1}{X} \left| P_{B_s}(x) \right|^2 dx \ll X^{\gamma} \]
with \( \gamma > 2s - 3 \) implies the bound
\[ \int_{0}^{X} \frac{1}{X} \left| P_{B_s}(x) \right|^2 dx \ll X^{2s-1-2(s-2)/(\gamma-s+1)} \log X. \]

Under the assumptions of our Theorem, Lemma 1 shows that we can interchange the roles of \( B_s \) and \( B_s^* \). Hence, starting with the trivial bound \( \gamma_0 = 2s + 1 \) for the mean square of \( P_{B_s}(x) \) and \( P_{B_s^*}(x) \), the iteration
\[ \gamma_{n+1} = 2s - 1 - \frac{2(s-2)}{\gamma_n - s + 1} \]
produces better bounds. Since \( \gamma_{n+1} < \gamma_n \) for \( \gamma_n > 2s - 3 \) the iteration converges to its largest fixpoint. The two fixpoints are 2 \( s - 3 \) and \( s + 1 \). Hence, for \( s \geq 4 \), \( \lim_{n \to \infty} \gamma_n = 2s - 3 \). This proves the Theorem.

References


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