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Metrika

International Journal for Theoretical and Applied Statistics

ISSN 0026-1335 Volume 80 Number 3

Metrika (2017) 80:363-375 DOI 10.1007/s00184-017-0609-0





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The variance of the discrepancy distribution of rounding procedures, and sums of uniform random variables

Lothar Heinrich¹ \cdot Friedrich Pukelsheim¹ \cdot Vitali Wachtel¹

Received: 19 May 2016 / Published online: 19 January 2017 © Springer-Verlag Berlin Heidelberg 2017

Abstract When ℓ probabilities are rounded to integer multiples of a given accuracy *n*, the sum of the numerators may deviate from *n* by a nonzero discrepancy. It is proved that, for large accuracies $n \to \infty$, the limiting discrepancy distribution has variance $\ell/12$. The relation to the uniform distribution over the interval [-1/2, 1/2], whose variance is 1/12, is explored in detail.

Keywords Rounding residual · Euler–Maclaurin formula · Invariance principle for rounding residuals · Euler–Frobenius polynomial · Fourier-analytic approach

Mathematics Subject Classification Primary: 62E15 · 62P25; Secondary: 65B15 · 60E10

1 Introduction

Suppose we are given ℓ probabilities, p_1, \ldots, p_ℓ . In printed publications the probabilities are rounded usually to percentages, or to multiples of tenths of a percent. That is, they are converted into integer multiples of n = 100, or of n = 1000. More generally each probability p_j is rounded into a fraction n_j/n , with some integer numerator n_j relative to a given "accuracy" n. The fractions n_j/n provide a valid distribution only

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if the sum of the numerators is equal to the denominator, $n_1 + \cdots + n_\ell = n$. It is well-known that individual rounding of the probabilities p_j may fail to satisfy this equation. Rather, a discrepancy $Z = (n_1 + \cdots + n_\ell) - n$ is observed which may be nonzero. Happacher (2001) calculates the distribution of Z for finite accuracy n when the probability vector $\mathbf{p}^{(\ell)} = (p_1, \ldots, p_\ell)$ is uniformly distributed over the probability simplex. He also shows that these unwieldy distributions converge for large accuracies to the elegant distribution in display (2.1) below, the "discrepancy distribution". Because of symmetry its expectation is zero. Its variance is zero when $\ell = 1$ or $\ell = 2$. Using Maple, Mathematica or similar software (2.1) allows an easy computational verification that the variance is equal to $\ell/12$, for all $\ell \ge 3$ that computers can handle.

In Sect. 2 the variance formula is proved in a rigorous way. Our proof is based on the Euler–Maclaurin formula and complements an alternative approach using characteristic functions (short: CF) that is due to Gawronski and Neuschel (2013) and Janson (2013). Section 3 discusses an invariance principle according to which the discrepancy distribution is the limiting distribution whenever the distribution of the probability vector $\mathbf{p}^{(\ell)} = (p_1, \ldots, p_{\ell})$ is absolutely continuous. The invariance principles explains the universal applicability of the discrepancy distribution and provides a second proof of the variance formula. Section 4 reviews briefly the approach of Gawronski and Neuschel (2013) and Janson (2013). An alternative Fourier-analytic derivation of the CF function of the discrepancy distribution is given in Sect. 5. These results lead to a third proof of the variance formula.

2 Discrepancy variance

When rounding two or more proportions p_1, \ldots, p_ℓ to integer percentages n_1, \ldots, n_ℓ , the resulting percentages do not necessarily sum exactly to 100% points, but possibly leave a positive or negative discrepancy $z = (n_1 + \cdots + n_\ell) - 100$. Considering many and varied sets of $\ell \ge 2$ proportions, we may view the discrepancy to be a random variable *Z* taking its values in \mathbb{Z} , the set of all integers. Happacher (2001) shows that, for all practical purposes, the distribution of the discrepancy *Z* is modelled well by $\mathbf{P}(Z = z) = g_\ell(z)$, where the probabilities are given by

$$g_{\ell}(z) = \sum_{k=0}^{\ell} \frac{(-1)^k}{(\ell-1)!} \binom{\ell}{k} \left(\frac{\ell}{2} + z - k\right)_+^{\ell-1}, \quad z \in \mathbb{Z}.$$
 (2.1)

The notation y_{+}^{n} is short for $(y_{+})^{n}$, where $y_{+} = y$ in case y > 0 and $y_{+} = 0$ otherwise.

It is not hard to see (as reviewed below) that the probabilities are symmetric around zero, $g_{\ell}(z) = g_{\ell}(-z)$. Hence all odd moments of Z vanish. In particular the expectation of the discrepancy is zero, that is, the instances when the sum of the rounded percentages is larger than 100 outweigh the instances when the sum is smaller.

What about the discrepancy variance? It is not hard to see (and reviewed below) that the probability $g_{\ell}(z)$ is positive only for $|z| \leq L$, where here and throughout we put

$$L := \left\lfloor \frac{\ell - 1}{2} \right\rfloor.$$

With $\ell = 2$ the discrepancy attains the value z = 0 with probability one, whence the variance is zero. For three or more proportions this section establishes the following.

Theorem 1 For $\ell \geq 3$ the discrepancy variance is $\ell/12$.

Theorem 1 focuses on variances because of its use in Theorem 2. However, the proof of Theorem 1 shows that the statement extends to all even moments of order less than ℓ , as pointed out in the Remark at the end of this section and stated more explicitly by (5.4) at the end of Sect. 5.

Before turning to the proof of Theorem 1 we take the time for some preliminary remarks. The theorem states that a sum which appears to be rather elementary admits a simple evaluation,

$$\sum_{z \in \mathbb{Z}} z^2 g_{\ell}(z) = \sum_{z = -L}^{L} z^2 g_{\ell}(z) = \frac{\ell}{12}.$$
(2.2)

Identity (2.2) looks like an innocuous exercise for an introductory probability course. The appearance is deceiving, we do not know of an easy proof of (2.2). Our first proof builds on the Euler–Maclaurin formula, see Abramowitz and Stegun (1970) (p. 806, formula 23.1.30).

The Euler–Maclaurin formula establishes a relation between a sum—the left hand side of (2.2)—and an integral. The right hand side of (2.2) becomes an integral by noting that it is the variance of $V_1 + \cdots + V_\ell$, where V_1, \ldots, V_ℓ are independent and identically distributed random variables whose common distribution is uniform over the interval [-1/2, 1/2].

The (well-known) Lebesgue density of $V_1 + \cdots + V_\ell$ is obtained by way of the convolution lemma,

$$g_{\ell}(x) = \int g_{\ell-1}(y)g_1(x-y)\,dy = \int_{x-1/2}^{x+1/2} g_{\ell-1}(y)\,dy. \tag{2.3}$$

Starting from the indicator function $g_1(x) = \mathbb{1}_{[-1/2, 1/2]}(x)$ for $\ell = 1$, the density for $\ell \ge 2$ is found to be

$$g_{\ell}(x) = \sum_{k=0}^{\ell} \frac{(-1)^{k}}{(\ell-1)!} {\ell \choose k} \left(\frac{\ell}{2} + x - k\right)_{+}^{\ell-1}, \quad x \in \mathbb{R}.$$
 (2.4)

This is the same function as in (2.1), except that the domain of definition is extended from \mathbb{Z} in (2.1), to \mathbb{R} in (2.4).

The interrelation between (2.1) and (2.4) entails three useful implications. Firstly, the probabilities in (2.1) add to unity as they should,

$$\sum_{z \in \mathbb{Z}} g_{\ell}(z) = \sum_{z \in \mathbb{Z}} \int_{z-1/2}^{z+1/2} g_{\ell-1}(y) \, dy = \int_{\mathbb{R}} g_{\ell-1}(y) \, dy = 1.$$

Secondly, the function g_{ℓ} is symmetric. This is obvious for g_1 . Assuming $g_{\ell-1}$ is symmetric, $g_{\ell-1}(y) = g_{\ell-1}(-y)$, so is g_{ℓ} , as seen by

$$g_{\ell}(x) = \int_{x-1/2}^{x+1/2} g_{\ell-1}(y) \, dy = \int_{x-1/2}^{x+1/2} g_{\ell-1}(-y) \, dy$$
$$= \int_{-x-1/2}^{-x+1/2} g_{\ell-1}(y) \, dy = g_{\ell}(-x).$$

Thirdly, the Lebesgue density in (2.4) takes positive values on the open interval $(-\ell/2, \ell/2)$ for $\ell \ge 2$. Indeed, for $x \le -\ell/2$ we have $\ell/2 + x - k \le 0$. With all positive parts in (2.4) vanishing we get $g_{\ell}(x) = 0$. For $x \ge \ell/2$ symmetry entails $g_{\ell}(x) = g_{\ell}(-x) = 0$. Using the recursive definition (2.3) it is easily seen by induction that $g_{\ell}(x) > 0$ for $x \in (-\ell/2, \ell/2)$ and $\ell \ge 2$. From $g_{\ell}(z) = \int_{z-1/2}^{z+1/2} g_{\ell-1}(y) dy$ it follows that the probabilities in (2.1) are positive only for the integers $z \in \{0, \pm 1, \ldots, \pm L\}$. Upon introducing $f_{\ell}(x) = x^2 g_{\ell}(x)$ we may write identity (2.2) as

$$\sum_{z=-L}^{L} f_{\ell}(z) = \int_{-\ell/2}^{\ell/2} f_{\ell}(x) \, dx.$$
(2.5)

The Euler–Maclaurin formula relates sum and integral whenever f_{ℓ} is a smooth function. However, $g_{\ell}(x)$ in (2.4) is smooth only piecewise, on the integer intervals [z, z + 1] when ℓ is even, and on the shifted intervals [z - 1/2, z + 1/2] when ℓ is odd. Therefore the proof of Theorem 1 treats even ℓ and odd ℓ separately.

Proof of Theorem 1 Parts I and II deal with ℓ even. Part I restricts attention to integer intervals [z, z+1] to apply the Euler–Maclaurin formula. Part II aggregates the interval results in order to establish the desired identity (2.5). Part III handles odd ℓ .

I. We fix an integer $z \in \mathbb{Z}$. For $x \le z+1$ the term $(\ell/2 + x - k)_+^{\ell-1}$ in (2.4) is positive only for $0 < \ell/2 + x - k \le \ell/2 + z + 1 - k$, that is, $k < \ell/2 + z + 1$ and $k \le \ell/2 + z$. If additionally $x \ge z$ then $\ell/2 + x - k \ge \ell/2 + z - k \ge 0$. The passage to positive parts becomes superfluous, $(\ell/2 + x - k)_+^{\ell-1} = (\ell/2 + x - k)^{\ell-1}$. This shows that on the interval [z, z + 1] the function $g_\ell(x)$ is a polynomial of degree $\ell - 1$, namely,

$$g_{\ell}(x) = \sum_{k=0}^{\ell/2+z} \frac{(-1)^k}{(\ell-1)!} {\ell \choose k} \left(\frac{\ell}{2} + x - k\right)^{\ell-1} \quad \text{for} \quad z \le x \le z+1 \,.$$
(2.6)

Hence on [z, z + 1] the function $f_{\ell}(x) = x^2 g_{\ell}(x)$ is a polynomial of degree $\ell + 1$.

Let $g_{\ell}^{(q)}(z+)$ and $f^{(q)}(z+)$ denote the derivatives of order q at the left endpoint of [z, z+1], and $g_{\ell}^{(q)}(z+1-)$ and $f^{(q)}(z+1-)$ those at the right endpoint of [z, z+1]. The Euler–Maclaurin formula invokes the Bernoulli numbers B_{2k} and the odd derivatives $f^{(2k-1)}$ which we include up to order $\ell - 1$. The formula finishes with a balancing term depending on the $(\ell + 2)$ nd derivative of f_{ℓ} which is zero. Therefore the balancing term disappears, in our application. Thus the Euler–Maclaurin formula yields an identity, Variance of the discrepancy distribution of rounding...

$$\frac{1}{2} f_{\ell}(z) + \frac{1}{2} f_{\ell}(z+1) = \int_{z}^{z+1} f_{\ell}(x) dx + \sum_{k=1}^{\ell/2} \frac{B_{2k}}{(2k)!} \left(f^{(2k-1)}(z+1-) - f^{(2k-1)}(z+) \right). \quad (2.7)$$

The derivatives of f_{ℓ} are obtained from the Leibniz rule,

$$f^{(q)}(x) = \sum_{j=0}^{q} {\binom{q}{j}} \frac{d^{j}}{dx^{j}} x^{2} \frac{d^{q-j}}{dx^{q-j}} g_{\ell}(x)$$

= $x^{2} g_{\ell}^{(q)}(x) + 2q x g_{\ell}^{(q-1)}(x) + q (q-1) g_{\ell}^{(q-2)}(x).$ (2.8)

The derivatives of g_{ℓ} in (2.6) are

$$g_{\ell}^{(q)}(x) = \begin{cases} \sum_{k=0}^{\ell/2+z} \frac{(-1)^{k}}{(\ell-1-q)!} \binom{\ell}{k} \binom{\ell}{2} + x - k \end{pmatrix}^{\ell-1-q} & \text{if } q < \ell-1, \\ \sum_{k=0}^{\ell/2+z} (-1)^{k} \binom{\ell}{k} & \text{if } q = \ell-1, \\ 0 & \text{if } q > \ell-1. \end{cases}$$
(2.9)

For orders $q < \ell - 1$ the *q*th derivative $g_{\ell}^{(q)}(z+)$ at the left endpoint of [z, z+1] coincides with the *q*th derivative $g_{\ell}^{(q)}(z-)$ at the right endpoint of the preceding interval [z-1, z],

$$g_{\ell}^{(q)}(z+) = \sum_{k=0}^{\ell/2+z} \frac{(-1)^k}{(\ell-1-q)!} {\ell \choose k} \left(\frac{\ell}{2} + z - k\right)^{\ell-1-q} \\ = \sum_{k=0}^{\ell/2+z-1} \frac{(-1)^k}{(\ell-1-q)!} {\ell \choose k} \left(\frac{\ell}{2} + z - k\right)^{\ell-1-q} = g_{\ell}^{(q)}(z-).$$

Therefore g_{ℓ} and f_{ℓ} are q times differentiable also at the knots $z \in \mathbb{Z}$, with derivatives $g_{\ell}^{(q)}(z) = g_{\ell}^{(q)}(z+) = g_{\ell}^{(q)}(z-)$ and $f^{(q)}(z) = f^{(q)}(z+) = f^{(q)}(z-)$, as long as $q < \ell - 1$.

II. Now we sum (2.7) over $z \in \mathbb{Z}$. Since (2.1) has support points $-L, \ldots, L$ and (2.4) has support $(-\ell/2, \ell/2)$ aggregation of the Euler–Maclaurin formulas yields

$$\sum_{z=-L}^{L} f_{\ell}(z) = \int_{-\ell/2}^{\ell/2} f_{\ell}(x) dx + \sum_{k=1}^{\ell/2} \frac{B_{2k}}{(2k)!} \sum_{z=-L}^{L} \left(f^{(2k-1)}(z+1-) - f^{(2k-1)}(z+) \right). \quad (2.10)$$

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Thus it suffices to show that

$$\sum_{z=-L}^{L} \left(f^{(2k-1)}(z+1-) - f^{(2k-1)}(z+) \right) = 0, \quad k = 1, \dots, \ell/2.$$
 (2.11)

For $k < \ell/2$ the order is $q = 2k - 1 < \ell - 1$ for which f_{ℓ} is q times differentiable. Thus (2.11) is a telescope sum and simplifies to a plain difference,

$$\sum_{z=-\ell}^{\ell} \left(f^{(q)}(z+1) - f^{(q)}(z) \right) = f^{(q)}(\ell+1) - f^{(q)}(-\ell) = 0.$$

For $k = \ell/2$ with ensuing order $q = \ell - 1$ we must evaluate the sum

$$S = \sum_{z=-L}^{L} \left(f^{(\ell-1)}(z+1-) - f^{(\ell-1)}(z+) \right).$$

The Leibniz rule (2.8) includes lower order derivatives $g_{\ell}^{(\ell-2)}$ and $g_{\ell}^{(\ell-3)}$. They, too, lead to telescope sums that vanish and hence contribute nothing to *S*. As for the $(\ell-1)$ st derivative, (2.9) gives $g_{\ell}^{(\ell-1)}(z+) = g_{\ell}^{(\ell-1)}(z+1-) = \sum_{k=0}^{\ell/2+z} (-1)^k {\ell \choose k}$. We obtain

$$S = \sum_{z=-L}^{L} \left((z+1)^2 - z^2 \right) \sum_{k=0}^{\ell/2+z} (-1)^k \binom{\ell}{k} = \sum_{k=0}^{\ell} (-1)^k \binom{\ell}{k} \sum_{z=k-\ell/2}^{\ell/2} \left((z+1)^2 - z^2 \right)$$

with

$$\sum_{z=k-\ell/2}^{\ell/2} \left((z+1)^2 - z^2 \right) = \left(\frac{\ell}{2} + 1\right)^2 - \left(k - \frac{\ell}{2}\right)^2$$
$$= (\ell+1) + (\ell-1)k - k(k-1)$$

The binomial theorem and multiple uses of the identity $\binom{\ell}{k}k = \ell\binom{\ell-1}{k-1}$ result in

$$S = (\ell+1)(1-1)^{\ell} - \ell(\ell-1)(1-1)^{\ell-1} - \ell(\ell-1)(1-1)^{\ell-2}.$$
 (2.12)

This gives S = 0 whenever $\ell \ge 3$, and establishes (2.11). Now (2.10) reduces to (2.5). The proof of Theorem 1 for even ℓ is complete.

III. In case ℓ is odd we again start out with a fixed integer $z \in \mathbb{Z}$. Since ℓ is odd the polynomial structure of the function g_{ℓ} holds true on the shifted interval,

$$g_{\ell}(x) = \sum_{k=0}^{(\ell-1)/2+z} \frac{(-1)^k}{(\ell-1)!} {\ell \choose k} \left(\frac{\ell}{2} + x - k\right)^{\ell-1} \quad \text{for} \quad z - \frac{1}{2} \le x \le z + \frac{1}{2}.$$

The derivatives of g_{ℓ} are the same as in (2.9), except that the upper summation limit now reads $(\ell - 1)/2$. The function $f_{\ell}(x) = x^2 g_{\ell}(x)$ continues to be a polynomial of degree $\ell + 1$.

As in (2.7) the Euler–Maclaurin formula yields the identity

$$\frac{1}{2}f\left(z-\frac{1}{2}\right)+\frac{1}{2}f\left(z+\frac{1}{2}\right)=\int_{z-1/2}^{z+1/2}f_{\ell}(x)\,dx+\sum_{k=1}^{(\ell+1)/2}\frac{B_{2k}}{(2k)!}\Delta(k),$$

where we have set $\Delta(k) := f^{(2k-1)}(z + 1/2 -) - f^{(2k-1)}(z - 1/2 +)$ for short. The formula permits a refinement by dividing the underlying interval into two equal parts in order to pick up the value of f_{ℓ} at the interval midpoint z,

$$\frac{1}{2}f\left(z-\frac{1}{2}\right) + f_{\ell}(z) + \frac{1}{2}f\left(z+\frac{1}{2}\right) = 2\int_{z-1/2}^{z+1/2} f_{\ell}(x) \, dx + \sum_{k=1}^{(\ell+1)/2} \frac{1}{2^{2k-1}} \frac{B_{2k}}{(2k)!} \Delta(k).$$

Subtraction of the first equation from the second yields the version to be pursued further,

$$f_{\ell}(z) = \int_{z-1/2}^{z+1/2} f_{\ell}(x) \, dx - \sum_{k=1}^{(\ell+1)/2} \left(1 - \frac{1}{2^{2k-1}}\right) \frac{B_{2k}}{(2k)!} \Delta(k)$$

Summation over $z \in \mathbb{Z}$ leads to

$$\sum_{z=-L}^{L} f_{\ell}(z) = \int_{-\ell}^{\ell} f_{\ell}(x) \, dx - \sum_{k=1}^{(\ell+1)/2} \left(1 - \frac{1}{2^{2k-1}}\right) \frac{B_{2k}}{(2k)!} \sum_{z=-L}^{L} \Delta(k). \quad (2.13)$$

We aim to verify that the last sum is zero, that is,

$$\sum_{z=-L}^{L} \left(f^{(2k-1)} \left(z + \frac{1}{2} - \right) - f^{(2k-1)} \left(z - \frac{1}{2} + \right) \right) = 0, \quad k = 1, \dots, \frac{\ell+1}{2}.$$
(2.14)

For $k < (\ell + 1)/2$ the order of the derivative is $q = 2k - 1 < \ell$. Since q and ℓ are odd this forces $q < \ell - 1$, whence the sum in (2.14) is a telescope sum that vanishes.

For $k = (\ell + 1)/2$ the order is $q = 2k - 1 = \ell$. The sum in (2.14) becomes

$$S = \sum_{z=-L}^{L} \left(f^{(\ell)} \left(z + \frac{1}{2} - \right) - f^{(\ell)} \left(z - \frac{1}{2} + \right) \right).$$

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Applying the Leibniz rule (2.8) to $f^{(\ell)}$ we find that the first term depends on $g_{\ell}^{(\ell)}$ which is zero throughout. The third term, involving $g_{\ell}^{(\ell-2)}$, leads to another telescope sum that vanishes. Thus *S* is determined by the second term,

$$S = 2\ell \sum_{z=-L}^{L} \sum_{k=0}^{(\ell-1)/2+z} (-1)^k \binom{\ell}{k} = 2\ell \sum_{k=0}^{\ell} (-1)^k \binom{\ell}{k} \sum_{z=k-(\ell-1)/2}^{(\ell+1)/2} 1$$

Since the last sum counts $(\ell + 1) - k$ ones, the result is

$$S = 2(\ell+1)\ell(1-1)^{\ell} + 2\ell^2(1-1)^{\ell-1}.$$

Now S = 0 verifies (2.14), whence (2.13) turns into (2.5). Thus, Theorem 1 is completely proved.

Remark The proof generalizes to functions of the type $f_{\ell}(x) = x^p g_{\ell}(x)$, with $p < \ell$ even. Hence the discrepancy Z and the sum $V_1 + \cdots + V_{\ell}$ share not only the same variance $\ell/12$, but also all even moments up to order $p < \ell$.

The aggregated Euler–Maclaurin identity (2.10) remains valid also when $\ell = 2$. The discrepancy variance on the left is 0, as mentioned in the paragraph prior to Theorem 1. On the other hand (2.12) results in S = -2. Inserting $B_2 = 1/6$ turns the right hand side of (2.10) into 2/12 + (1/6)(-2)/2 = 0, too. In the same way identity (2.13) maintains its validity when $\ell = 1$.

3 Discrepancy representation as a sum of uniform random variables

Happacher (2001) derives the discrepancy distribution (2.1) under the assumption that the vector of proportions $\mathbf{p}^{(\ell)} = (p_1, \ldots, p_\ell)$ follows a uniform distribution on the probability simplex $\Omega_{\ell} = \{(p_1, \ldots, p_\ell) \in (0, 1)^{\ell} \mid p_1 + \cdots + p_{\ell} = 1\}$. This assumption is too specific to justify the universal applicability of the discrepancy distribution (2.1). Rather, the justification originates from an invariance principle that allows to replace the uniform distribution on Ω_{ℓ} by an arbitrary absolutely continuous distribution on Ω_{ℓ} .

The general task is to round proportions p_j to integer multiples n_j of a preordained accuracy n. The accuracy is n = 100 for percentages, n = 1000 for tenths of a percent, etc. An obvious approach is to multiply a proportion by n, and to round the scaled quantity np_j to an integer n_j . We designate the standard rounding function by angle brackets $\langle x \rangle$, as do Abramowitz and Stegun (1970, p. 223). That is, if the fractional part of x > 0 is smaller than one half then x is rounded downwards, $\langle x \rangle = \lfloor x \rfloor$. If the fractional part is larger than or equal to one half then x is rounded upwards, $\langle x \rangle = \lceil x \rceil$.

The rounding procedure gives rise to the rounding residuals $U_j(n) = \langle np_j \rangle - np_j$. As $\langle np_j \rangle = n_j$ the discrepancy Z coincides with the sum of the rounding residuals,

$$Z = (n_1 + \dots + n_{\ell}) - n = U_1(n) + \dots + U_{\ell}(n).$$

The representation has dramatic consequences. The distribution of Z depends an the distributional assumption for the proportions (p_1, \ldots, p_ℓ) on Ω_ℓ only through the induced distribution of the rounding residuals $U_1(n), \ldots, U_\ell(n)$. While every specific distributional assumption for the proportions (p_1, \ldots, p_ℓ) sparks objections as to its universal validity, this is not so for rounding residuals.

Rounding residuals $U_j(n)$ are "known" to be uniformly distributed over the interval [-1/2, 1/2]. This knowledge is scientific commonplace to an extent that every reference which makes use of it can be surpassed by a prior reference that has done so earlier. Examples abound, as Seal (1950) demonstrates in his witty note "Spot the prior reference"; see Happacher (2001) for a reprint of the note. While there exist numerous publications modeling a rounding residual by the uniform distribution, we know of just a handful of sources proposing a rigorous argument how the uniform distribution comes into being.

The work of Happacher (2001) implies that for finite accuracy *n* the discrepancy fails the distribution (2.1), and that the rounding residuals cannot have a uniform distribution. Even when assuming that each rounding residual is uniformly distributed over the interval [-1/2, 1/2] they cannot be jointly independent, because their sum has the discrete distribution (2.1) and not the convolution distribution (2.4).

It seems natural to resort to an asymptotic approach. Janson (2013) assumes that the accuracies *n* are uniformly distributed over a finite range $\{1, ..., N\}$. Then he lets the range tend to infinity, $N \rightarrow \infty$. In the present paper we assume an absolutely continuous distribution for the vector of proportions, as do Diaconis and Freedman (1979) and Janson (2013). Then we let the accuracy tend to infinity, $n \rightarrow \infty$. In this setting Heinrich et al. (2004) prove that the limiting distribution of the rounding residual becomes uniform if the underlying distribution admits a Riemann integrable density. Janson (2013) shows that the conclusion remains valid if the underlying distribution is absolutely continuous. Only recently did we spot the prior reference Tuckey (1938) who establishes the same result. The state of the art is summarized by the invariance principle for rounding residuals as given in the following Theorem 2. A precursor to the invariance principle is Theorem 3 in Diaconis and Freedman (1979).

Theorem 2 Assume that, for $\ell \geq 2$, the proportions $\mathbf{p}^{(\ell)} = (p_1, \ldots, p_\ell)$ follow an absolutely continuous distribution on the probability simplex Ω_ℓ . Then the vector of rounding residuals $\mathbf{U}^{(\ell)}(n) = (U_1(n), \ldots, U_\ell(n))$ for accuracy n and the vector of proportions $\mathbf{p}^{(\ell)}$ jointly converge in distribution,

$$\left(\mathbf{U}^{(\ell)}(n), \mathbf{p}^{(\ell)}\right) \xrightarrow{\text{in distribution}} (\mathbf{U}^{(\ell)}, \mathbf{p}^{(\ell)}),$$

where the components of the limit random vector $\mathbf{U}^{(\ell)} = (U_1, \ldots, U_\ell)$ are uniformly distributed over the interval [-1/2, 1/2], exchangeable, and independent of $\mathbf{p}^{(\ell)}$. Omitting an arbitrary component U_j , the remaining $\ell - 1$ variables U_k , $k \neq j$, are independent. If $\ell \geq 3$ then the variables U_1, \ldots, U_ℓ are uncorrelated.

Proof Theorem 2 coincides with Theorem 6.10 in Pukelsheim (2014) where the assertions are proved, except for uncorrelatedness. But uncorrelatedness is an immediate

consequence of the preceding statement. If $\ell \geq 3$ then any two variables are independent and hence uncorrelated.

Theorem 2 provides a second, one-line proof for the variance formula in Theorem 1. Assuming $\ell \ge 3$ uncorrelatedness yields $\operatorname{Var}(Z) = \operatorname{Var}(U_1 + \cdots + U_\ell) = \ell/12 + \ell(\ell-1) \operatorname{Cov}(U_1, U_2) = \ell/12.$

Theorem 2 provides a solid justification for the commonplace assumption that rounding residuals follow a uniform distribution. Theorem 2 also justifies the universal applicability of the discrepancy distribution (2.1). Indeed, from $Z = U_1 + \cdots + U_\ell$ we see that the discrepancy Z attains a value $z \in \mathbb{Z}$ if and only if $U_1 + \cdots + U_{\ell-1} = z - U_\ell$. Evidently we have $z - U_\ell \in [z - 1/2, z + 1/2]$. This yields

$$\left\{Z = z\right\} = \left\{U_1 + \dots + U_{\ell-1} \in \left[z - \frac{1}{2}, z + \frac{1}{2}\right]\right\}.$$
(3.1)

When for $n \to \infty$ the limit distributions take over, the probability of the event on the right hand side in (3.1) becomes $\int_{z-1/2}^{z+1/2} g_{\ell-1}(y) dy = g_{\ell}(z)$ as stipulated by (2.1). This shows that for large accuracies the distribution of the discrepancy Z is given by (2.1).

Standard rounding permits yet another representation. Since the right hand side in (3.1) may be expressed as $\{\langle U_1 + \cdots + U_{\ell-1} \rangle = z\}$, the discrepancy satisfies

$$Z = \langle U_1 + \cdots + U_{\ell-1} \rangle \, .$$

That is, the discrepancy Z behaves as if standard rounding is applied to the sum of $\ell - 1$ copies of uniform random variables.

4 Euler–Frobenius distributions

The shifted discrepancy Z + L is a discrete random variable that is nonnegative. Hence its probability generating function is a polynomial. Gawronski and Neuschel (2013) identify this polynomial to be an Euler–Frobenius polynomial and study the induced distributions. Janson (2013) calls them Euler–Frobenius distributions, and provides many additional results. The distribution of the discrepancy Z is the Euler–Frobenius distribution $\mathscr{E}_{\ell-1,\ell/2}$.

Gawronski and Neuschel (2013) (p. 7) and Janson (2013) (p. 10) also calculate the CF of Euler–Frobenius distributions. The discrepancy distribution turns out to have CF

$$\varphi_{\ell}(s) = i^{-\ell} e^{-i\ell s/2} \left(e^{is} - 1 \right)^{\ell} \sum_{k=-\infty}^{\infty} \frac{e^{-\pi ik\ell}}{(s+2\pi k)^{\ell}}.$$
(4.1)

Thus, by calculating the negative second derivative of the CF (4.1) at s = 0, the variance is found to be $\ell/12$. This approach provides another proof of Theorem 1. Section 5 concludes with an alternative derivation of the CF (4.1) in addition to those in Gawronski and Neuschel (2013) and Janson (2013).

5 An alternative Fourier-analytic approach

To complete this paper we present a further way to obtain the CF (4.1) which seems to be more direct and different from the methods used in Gawronski and Neuschel (2013) and Janson (2013). Among others our approach is related with so-called sincintegrals which recently attracted much interest due to their unexpected properties, see e.g. Schmid (2014), Almkvist and Gustavsson (2014). The sinc-function is defined as follows: $\operatorname{sin}(t) := \sin(t)/t$ for $t \in \mathbb{R} \setminus \{0\}$ and $\operatorname{sinc}(0) := 1$. It is easily verified that the CF $u_{\ell}(t) := \mathbf{E} \exp\{it(V_1 + \cdots + V_{\ell})\}$ of independent and uniformly on [-1/2, 1/2] distributed random variables V_1, \ldots, V_{ℓ} , see Sect. 2, can be expressed by $u_{\ell}(t) = (\operatorname{sinc}(t/2))^{\ell}$. Since $u_{\ell}(t)$ is absolutely integrable for $\ell \ge 2$, the Fourier inversion theorem, see e.g. Taylor (1996) (p. 271) combined with $\operatorname{sinc}(-t) = \operatorname{sinc}(t)$, allows to express the symmetric density of $V_1 + \cdots + V_{\ell}$ given in (2.4) as Fourier integral

$$g_{\ell}(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-itx} u_{\ell}(t) dt = \frac{1}{\pi} \int_{-\infty}^{\infty} e^{2itx} \left(\operatorname{sinc}(t)\right)^{\ell} dt$$
$$= \frac{1}{\pi} \int_{-\infty}^{\infty} \cos(2tx) \left(\operatorname{sinc}(t)\right)^{\ell} dt > 0 \quad \text{for} \quad |x| < \frac{\ell}{2}.$$
(5.1)

The right hand integral disappears for $|x| \ge \ell/2$. For $x \in \{0, \pm 1, ..., \pm L\}$ we get the symmetric lattice distribution (2.1) of the discrepancy introduced in Sect. 1. Its CF $\psi_{\ell}(s)$ is defined by

$$\psi_{\ell}(s) := \sum_{z=-L}^{L} e^{isz} g_{\ell}(z) = \frac{1}{\pi} \int_{-\infty}^{\infty} \sum_{z=-L}^{L} e^{i(s+2t)z} \left(\operatorname{sinc}(t)\right)^{\ell} dt \,.$$
(5.2)

We may rewrite this CF as follows:

Theorem 3 For $\ell \geq 2$ and all $s \in \mathbb{R}$ we have the identity

$$\psi_{\ell}(s) = \frac{1}{\pi} \int_{-\infty}^{\infty} \frac{\sin((2L+1)(\frac{s}{2}+t))}{\sin(\frac{s}{2}+t)} \left(\operatorname{sinc}(t)\right)^{\ell} dt = \sum_{k \in \mathbb{Z}} \left(\operatorname{sinc}(\frac{s}{2}+k\pi)\right)^{\ell} = \varphi_{\ell}(s).$$

Proof The first equality of Theorem 3 is immediately seen by inserting the Dirichlet kernel $\sum_{z=-L}^{L} e^{2ixz} = \sin((2L+1)x)/\sin(x)$, see e.g. Taylor (1996) (p. 162), into the right hand integral of (5.2) for x = s/2 + t. The third equality can be checked simply by rewriting all members of the doubly infinite series (4.1) by inserting Eulers formula $e^{ix} = \cos(x) + i \sin(x)$. It remains to verify the middle equality between the improper integral on the left and the double-sided infinite series on the right for any real *s*. It is rapidly seen that both $\psi_{\ell}(s)$ as well as $\varphi(s)$ (as uniformly convergent series for $\ell \ge 2$) are even and 2π -periodic functions that are infinitely often differentiable and take the value 1 at s = 0. Here, $\psi_{\ell}(0) = 1$ is obvious for a CF but it also follows in the special case a = 2L + 1, $k = \ell$ from the "sinc integral"

$$\int_{-\infty}^{\infty} \frac{\sin(at)}{\sin(t)} \left(\operatorname{sinc}(t)\right)^k dt = \pi \quad \text{for any real} \quad a > 0 \quad \text{and} \quad k = 1, \dots, \lfloor a \rfloor + 1,$$

see Schmid (2014) (pp. 15–17). Now, we prove that $\psi_{\ell}(s)$ and $\varphi_{\ell}(s)$ have the same Fourier expansions. For doing this we show that the corresponding coefficients at $\cos(js)$ are identical for all $j \in \mathbb{Z}$. The expansion (5.2) reveals that the *j*th Fourier coefficient of $\psi_{\ell}(s)$ coincides with $g_{\ell}(j)$ as expressed in (5.1) for $j \in \mathbb{Z}$, where $g_{\ell}(j) = 0$ for |j| > L. Assuming an expansion $\varphi_{\ell}(s) = \sum_{j \in \mathbb{Z}} c_{\ell}(j) \cos(js)$ we can calculate the coefficients $c_{\ell}(j)$ as follows:

$$c_{\ell}(j) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \cos(js) \varphi_{\ell}(s) ds = \frac{1}{\pi} \int_{-\pi/2}^{\pi/2} \cos(2js) \sum_{k \in \mathbb{Z}} (\sin(s+k\pi))^{\ell} ds$$
$$= \frac{1}{\pi} \sum_{k \in \mathbb{Z}} \int_{-\pi/2+k\pi}^{\pi/2+k\pi} \cos(2j(s-k\pi)) (\sin(s))^{\ell} ds$$
$$= \frac{1}{\pi} \int_{-\infty}^{\infty} \cos(2js) (\sin(s))^{\ell} ds .$$

Together with (5.1) we have $c_{\ell}(j) = g_{\ell}(j)$ for all $j \in \mathbb{Z}$ which was to be proved. \Box

Theorem 3 and the identity $\operatorname{sinc}(\frac{s}{2} + k\pi) = \operatorname{sinc}(\frac{s}{2})(-1)^k s/(s + 2k\pi)$ for $k \in \mathbb{Z}$ imply that

$$\log \psi_{\ell}(s) = \ell \log \mathbf{E} \exp\{\mathrm{i} s V_1\} + r_{\ell}(s) \quad \text{with} \quad r_{\ell}(s) = \log \left(\sum_{k \in \mathbb{Z}} \left(\frac{(-1)^k s}{s + 2k\pi}\right)^\ell\right).$$
(5.3)

The rules of differentiation yield that the derivatives of $r_{\ell}(s)$ up to order $\ell - 1$ disappear at s = 0. Using this fact the relation (5.3) reveals that the *m*th cumulant of the discrepancy distribution (2.1) is just equal to ℓ times the *m*th cumulant of V_1 (abbreviated by $C_m(V_1)$) for $m = 1, ..., \ell - 1$, see also Theorem 5.3 in Janson (2013).

In the particular case m = 2 this confirms once more the variance formula in Theorem 1. Moreover (5.3) allows to determine the moments $M_m^{(\ell)}$ of the discrepancy distribution (2.1) for even $m \ge 2$ and $\ell \ge m + 1$. In particular, (5.3) yields $M_4^{(\ell)} =$ $\ell C_4(V_1) + 3 \ell C_2(V_1) M_2^{(\ell)}$ when $\ell \ge 5$, whereas for $\ell = 3$ and $\ell = 4$ the fourth moment $M_4^{(\ell)}$ follows directly from (2.1):

$$M_4^{(\ell)} = \sum_{z=-L}^L z^4 g_\ell(z) = \begin{cases} \ell (5\ell-2)/240 & \text{if } \ell \ge 5, \\ 1/3 & \text{if } \ell = 4, \\ 1/4 & \text{if } \ell = 3. \end{cases}$$

A well-known general relationship between cumulants and moments applied to the special case $C_k(V_1) = M_k^{(\ell)} = 0$ for odd $k \ge 1$ and $C_k(V_1) = B_k/k$ for even $k \ge 2$, see e.g. Theorem 5.3 in Janson (2013), leads to the recursion formula

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$$M_m^{(\ell)} = \ell \sum_{k=0}^{\frac{m}{2}-1} \binom{m-1}{2k} C_{m-2k}(V_1) \ M_{2k}^{(\ell)} = \frac{\ell}{m} \sum_{k=0}^{\frac{m}{2}-1} \binom{m}{2k} B_{m-2k} \ M_{2k}^{(\ell)}$$
(5.4)

for all even $m \ge 2$ and $\ell \ge m+1$. Here, B_0, B_1, B_2, \ldots denote the Bernoulli numbers (which already appeared in the Euler–Maclaurin formula in Sect. 2) defined by the generating function $x/(e^x - 1) = \sum_{k=0}^{\infty} B_k x^k/k!$

Acknowledgements The authors would like to thank the anonymous referee for his comments on the original manuscript which led to an improved version of the paper.

References

- Abramowitz M, Stegun IA (1970) Handbook of mathematical functions with formulas. Graphs and mathematical tables. Dover, New York
- Almkvist G, Gustavsson J (2014) More remarkable sinc integrals and sums. arXiv: 1405.1265v2 [math.CA], 1–9
- Diaconis P, Freedman D (1979) On rounding percentages. J Am Stat Assoc 74:359-364
- Gawronski W, Neuschel T (2013) Euler–Frobenius numbers. Integr Transforms Special Funct 24:817–830 Happacher M (2001) The discrepancy distribution of stationary multiplier rules for rounding probabilities. Metrika 53:171–181
- Heinrich L, Pukelsheim F, Schwingenschlögl U (2004) Sainte-Laguë's chi-square divergence for the rounding of probabilities and its convergence to a stable law. Stat Decis 22:43–59

Janson S (2013a) Asymptotic bias of some election methods. Ann Oper Res 215:89-136

- Janson S (2013b) Euler-Frobenius numbers and rounding. Online J Anal Comb 8:1-39
- Pukelsheim F (2014) Proportional representation—apportionment methods and their applications. With a forword by Andrew Duff MEP. Springer, Cham
- Schmid H (2014) Two curious integrals and a graphic proof. Elem Math 69(1):11-17

Seal HL (1950) Spot the prior reference. J Inst Actuar Stud Soc 10:255–258

Taylor CJ (1996) An introduction to measure and probabilty. Springer, New York

Tuckey JW (1938) On the distribution of the fractional part of a statistical variable. Recueil Mathématique Nouvelle Série (Matematicheskii Sbornik Novaya Seriya) 4:461–462