

# Large spread does not imply Benford's Law

A. Berger

Mathematical and Statistical Sciences  
University of Alberta, Edmonton, CANADA

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## Abstract

Sharp universal bounds are given for the distance between normalised Lebesgue measure on  $\mathbb{R}/\mathbb{Z}$  and the distribution of  $\log X \bmod 1$ , where  $X$  is uniform. The results dispel the popular belief that a random variable obeys Benford's Law (at least approximately) whenever its spread is large.

For every real number  $x$ , the largest integer not larger than  $x$  will be denoted by  $\lfloor x \rfloor$ , and  $\llbracket x \rrbracket := x - \lfloor x \rfloor$  is the fractional (or non-integer) part of  $x$ . The base  $\gamma$  logarithm ( $\gamma > 1$ ) of  $x > 0$  is  $\log_\gamma x$ ; if used without a subscript,  $\log$  symbolises the natural logarithm. The sets of natural, non-negative integer, integer, positive real and real numbers are  $\mathbb{N}$ ,  $\mathbb{N}_0$ ,  $\mathbb{Z}$ ,  $\mathbb{R}^+$  and  $\mathbb{R}$ , respectively.

Given any probability measure  $\mu$  on  $\mathbb{R}$ , denote by  $F_\mu$  its distribution function, that is,  $F_\mu(x) = \mu(\cdot - \infty, x]$ . For every measurable map  $T : \mathbb{R} \rightarrow \mathbb{R}$  the probability measure  $T\mu$  is defined as  $T\mu(B) := \mu(T^{-1}(B))$  for all Borel sets  $B$ . Specifically,  $\llbracket \mu \rrbracket$  is, for every  $\mu$ , concentrated on  $[0, 1]$ . The uniform distribution on  $[a, b]$  with  $a < b$  is denoted by  $U_{a,b}$ . Thus

$$F_{U_{a,b}}(x) = \begin{cases} 0 & \text{if } x < a, \\ \frac{x-a}{b-a} & \text{if } a \leq x < b, \\ 1 & \text{if } x \geq b. \end{cases}$$

Given any two probability measures  $\mu, \nu$  on  $\mathbb{R}$ , their Kolmogorov-Smirnov distance  $d_\infty(\mu, \nu)$  is

$$d_\infty(\mu, \nu) = \sup_{x \in \mathbb{R}} |F_\mu(x) - F_\nu(x)|;$$

see e.g. [5] for some details on this metric. Recall that  $\mu$  with  $\mu(\mathbb{R}^+) = 1$ , or a real random variable with distribution  $\mu$ , satisfies *Benford's Law* base  $\gamma$  if and only if  $\log_\gamma \mu$  is uniform modulo one [2], i.e., if  $d_\infty(\llbracket \log_\gamma \mu \rrbracket, U_{0,1})$  equals zero. Contrary to what [4, p.63] may suggest, the uniform distribution  $U_{a,b}$  with  $a \geq 0$  does not even approximately satisfy Benford's Law for any base  $\gamma$ , no matter how large  $b - a$  is.

**Theorem 1.** *For all  $\gamma > 1$  and  $0 \leq a < b$ ,*

$$d_\infty(\llbracket \log_\gamma U_{a,b} \rrbracket, U_{0,1}) \geq C_\gamma > 0, \tag{1}$$

where

$$C_\gamma = \frac{1 - \gamma + \log \gamma + (\gamma - 1) \log(\gamma - 1) - (\gamma - 1) \log \log \gamma}{2(\gamma - 1) \log \gamma}.$$

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*E-mail address:* aberger@math.ualberta.ca

*Proof.* To verify (1), assume first that  $a = 0$ , and let  $l := \lfloor \log_\gamma b \rfloor$  and  $\delta := \lceil \log_\gamma b \rceil$ . Thus  $b = \gamma^{l+\delta}$ , and

$$F_{\log_\gamma U_{0,b}}(x) = \begin{cases} \frac{\gamma^x}{b} & \text{if } x < l + \delta, \\ 1 & \text{if } x \geq l + \delta, \end{cases}$$

from which it follows that, for all  $0 \leq x \leq 1$ ,

$$F_{\lceil \log_\gamma U_{0,b} \rceil}(x) = \sum_{k \in \mathbb{Z}} (F_{\log_\gamma U_{0,b}}(x+k) - F_{\log_\gamma U_{0,b}}(k)) = \begin{cases} \frac{\gamma^{1+x-\delta} - \gamma^{1-\delta}}{\gamma - 1} & \text{if } 0 \leq x < \delta, \\ 1 - \frac{\gamma^{1-\delta} - \gamma^{x-\delta}}{\gamma - 1} & \text{if } \delta \leq x \leq 1. \end{cases}$$

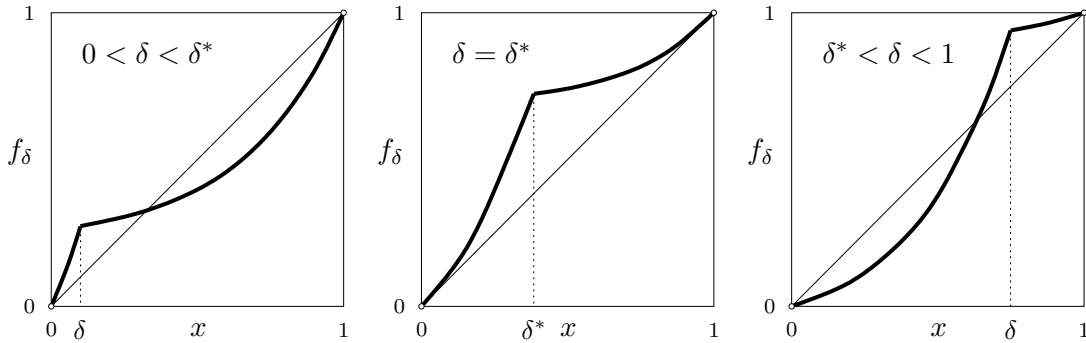
Notice that  $F_{\lceil \log_\gamma U_{0,b} \rceil}$  does not depend on  $l$ . For the sake of brevity, let  $f_\delta(x) := F_{\lceil \log_\gamma U_{0,b} \rceil}(x)$  and  $g_\delta(x) := f_\delta(x) - x$  for all  $x$  and  $\delta$ . Note that  $f_\delta$  is convex on  $[0, \delta]$  and on  $[\delta, 1]$ , and  $f_0(x) = \frac{\gamma^x - 1}{\gamma - 1}$ , whereas for  $0 < \delta < 1$

$$f'_\delta(0+) = f'_\delta(1-) = \frac{\gamma^{1-\delta} \log \gamma}{\gamma - 1}.$$

In the latter case, there exists a unique  $0 < \delta^* < 1$  such that  $f'_{\delta^*}(0+) = f'_{\delta^*}(1-) = 1$ ; explicitly  $\gamma^{1-\delta^*} = (\gamma - 1)/\log \gamma$ , and thus

$$\delta^* = \frac{\log \gamma - \log(\gamma - 1) + \log \log \gamma}{\log \gamma}.$$

Consequently, for  $0 < \delta < 1$  the graph of  $f_\delta$  can have three qualitatively different forms.



Note that  $g_\delta$  always attains its maximum at  $x = \delta$ . This suggests introducing the auxiliary function  $\psi$  according to

$$\psi(\delta) := g_\delta(\delta) = f_\delta(\delta) - \delta = \frac{\gamma - \gamma^{1-\delta}}{\gamma - 1} - \delta.$$

It follows from

$$\psi'(\delta) = \frac{\gamma^{1-\delta} \log \gamma}{\gamma - 1} - 1 = \gamma^{\delta^* - \delta} - 1,$$

that  $\psi$  is concave, with  $\psi(0) = \psi(1) = 0$  and  $\psi'(\delta^*) = 0$ . Hence  $\psi(\delta) > 0$  for all  $0 < \delta < 1$ , and

$$\max_{0 \leq \delta \leq 1} \psi(\delta) = \psi(\delta^*) = \frac{\gamma - \gamma^{1-\delta^*}}{\gamma - 1} - \delta^* = \frac{\gamma \log \gamma - \gamma + 1}{(\gamma - 1) \log \gamma} - \delta^* = 2C_\gamma.$$

Assume first that  $0 \leq \delta \leq \delta^*$ . In this case, the function  $g_\delta$  has a non-positive minimum at  $x = 1 + \delta - \delta^* > \delta$ , with

$$g_\delta(1 + \delta - \delta^*) = 1 - \frac{\gamma^{1-\delta} - \gamma^{1-\delta^*}}{\gamma - 1} - 1 - \delta + \delta^* = \psi(\delta) - \psi(\delta^*),$$

showing that  $\max_{0 \leq x \leq 1} g_\delta(x) = \psi(\delta)$  as well as  $-\min_{0 \leq x \leq 1} g_\delta(x) = \psi(\delta^*) - \psi(\delta)$ . Similarly, if  $\delta^* < \delta < 1$  then  $g_\delta$  has a negative minimum at  $x = \delta - \delta^* < \delta$ , with

$$g_\delta(\delta - \delta^*) = \frac{\gamma^{1-\delta^*} - \gamma^{1-\delta}}{\gamma - 1} - \delta + \delta^* = \psi(\delta) - \psi(\delta^*).$$

For all  $0 \leq \delta < 1$ , therefore,

$$\max_{0 \leq x \leq 1} g_\delta(x) = \psi(\delta), \quad -\min_{0 \leq x \leq 1} g_\delta(x) = \psi(\delta^*) - \psi(\delta),$$

and consequently

$$\max_{0 \leq x \leq 1} |g_\delta(x)| = \max\{\psi(\delta), \psi(\delta^*) - \psi(\delta)\} \geq \frac{1}{2}\psi(\delta^*) = C_\gamma,$$

which establishes (1) for the case  $a = 0$ .

To verify (1) for  $a > 0$  assume for the time being that  $\log_\gamma a = k \in \mathbb{Z}$ , and let  $l := \lfloor \log_\gamma b \rfloor$  and  $\delta := \lfloor \log_\gamma b \rfloor$  as before; for convenience set  $m := l - k \in \mathbb{N}_0$ . A short computation confirms that

$$f_{m,\delta}(x) := F_{\lfloor \log_\gamma U_{a,b} \rfloor}(x) = \begin{cases} \frac{\gamma^x - 1}{\gamma - 1} \cdot \frac{\gamma^{m+1} - 1}{\gamma^{m+\delta} - 1} & \text{if } 0 \leq x < \delta, \\ 1 - \frac{\gamma - \gamma^x}{\gamma - 1} \cdot \frac{\gamma^m - 1}{\gamma^{m+\delta} - 1} & \text{if } \delta \leq x \leq 1. \end{cases}$$

Notice that  $f_{m,\delta} \rightarrow f_\delta$  uniformly on  $[0, 1]$  as  $m \rightarrow \infty$ . Let again  $g_{m,\delta}(x) := f_{m,\delta}(x) - x$  and observe that

$$g_{m,\delta}(x) - g_\delta(x) = f_{m,\delta}(x) - f_\delta(x) = \Delta_{m,\delta}(x),$$

where  $\Delta_{m,\delta}$  is given by

$$\Delta_{m,\delta}(x) = \begin{cases} \frac{\gamma^x - 1}{\gamma - 1} \cdot \frac{\gamma^{1-\delta} - 1}{\gamma^{m+\delta} - 1} & \text{if } 0 \leq x < \delta, \\ \frac{\gamma^{1-\delta} - \gamma^{x-\delta}}{\gamma - 1} \cdot \frac{\gamma^\delta - 1}{\gamma^{m+\delta} - 1} & \text{if } \delta \leq x \leq 1. \end{cases}$$

Obviously,  $\Delta_{m,\delta} \geq 0$  with  $\Delta_{m,\delta}(0) = \Delta_{m,\delta}(1) = 0$ , and  $\Delta_{m,0} = 0$  for all  $m \geq 1$ . Furthermore, for  $0 < \delta < 1$  the function  $\Delta_{m,\delta}$  is convex and increasing on  $[0, \delta]$ , and concave and decreasing on  $[\delta, 1]$ . Since both  $g_\delta$  and  $\Delta_{m,\delta}$  attain their respective maximal value at  $x = \delta$ ,

$$\max_{0 \leq x \leq 1} g_{m,\delta}(x) = g_{m,\delta}(\delta) = g_\delta(\delta) + \Delta_{m,\delta}(\delta) = \psi(\delta) + \Delta_{m,\delta}(\delta).$$

If  $0 \leq \delta \leq \delta^*$  then, with the appropriate  $0 \leq \xi \leq 1$ ,

$$\begin{aligned} \max_{0 \leq x \leq 1} g_{m,\delta}(x) - \min_{0 \leq x \leq 1} g_{m,\delta}(x) &= g_{m,\delta}(\delta) - g_{m,\delta}(\xi) \\ &\geq g_{m,\delta}(\delta) - g_{m,\delta}(1 + \delta - \delta^*) \\ &= \psi(\delta) + \Delta_{m,\delta}(\delta) - g_\delta(1 + \delta - \delta^*) - \Delta_{m,\delta}(1 + \delta - \delta^*) \\ &= \psi(\delta^*) + \Delta_{m,\delta}(\delta) - \Delta_{m,\delta}(1 + \delta - \delta^*) \\ &\geq \psi(\delta^*) \\ &= 2C_\gamma. \end{aligned}$$

The same argument applies for  $\delta^* < \delta < 1$  with  $1 + \delta - \delta^*$  replaced by  $\delta - \delta^*$ . Thus

$$\max_{0 \leq x \leq 1} g_{m,\delta}(x) - \min_{0 \leq x \leq 1} g_{m,\delta}(x) \geq 2C_\gamma \quad (2)$$

holds for all  $m \in \mathbb{N}_0$  and  $0 \leq \delta < 1$ , and this in turn implies (1) since

$$\max_{0 \leq x \leq 1} |g_{m,\delta}(x)| \geq \frac{1}{2} (\max_{0 \leq x \leq 1} g_{m,\delta}(x) - \min_{0 \leq x \leq 1} g_{m,\delta}(x)) \geq C_\gamma.$$

Overall, therefore, the proof is complete if  $a = 0$  or  $\log_\gamma a \in \mathbb{Z}$ .

Finally, assume that  $a > 0$  does not satisfy  $\log_\gamma a \in \mathbb{Z}$ , that is, the number  $\tau := \llbracket \log_\gamma a \rrbracket$  lies strictly between 0 and 1. Note that

$$\llbracket \log_\gamma U_{a,b} \rrbracket = \llbracket \log_\gamma U_{a\gamma^{-\tau}, b\gamma^{-\tau}} + \tau \rrbracket,$$

and clearly  $\log_\gamma(a\gamma^{-\tau}) \in \mathbb{Z}$ . It is readily verified that, for every non-atomic probability measure  $\mu$  on  $\mathbb{R}$  and every  $t \in \mathbb{R}$ ,

$$F_{\llbracket \mu+t \rrbracket}(x) = \begin{cases} F_{\llbracket \mu \rrbracket}(x+1-\llbracket t \rrbracket) - F_{\llbracket \mu \rrbracket}(1-\llbracket t \rrbracket) & \text{if } 0 \leq x < \llbracket t \rrbracket, \\ F_{\llbracket \mu \rrbracket}(x-\llbracket t \rrbracket) + 1 - F_{\llbracket \mu \rrbracket}(1-\llbracket t \rrbracket) & \text{if } \llbracket t \rrbracket \leq x \leq 1, \end{cases}$$

and therefore also

$$G_{\llbracket \mu+t \rrbracket}(x) = \begin{cases} G_{\llbracket \mu \rrbracket}(x+1-\llbracket t \rrbracket) - G_{\llbracket \mu \rrbracket}(1-\llbracket t \rrbracket) & \text{if } 0 \leq x < \llbracket t \rrbracket, \\ G_{\llbracket \mu \rrbracket}(x-\llbracket t \rrbracket) - G_{\llbracket \mu \rrbracket}(1-\llbracket t \rrbracket) & \text{if } \llbracket t \rrbracket \leq x \leq 1, \end{cases}$$

where generally  $G_\mu(x) := F_\mu(x) - x$ . In particular,

$$\max_{0 \leq x \leq 1} G_{\llbracket \mu+t \rrbracket}(x) - \min_{0 \leq x \leq 1} G_{\llbracket \mu+t \rrbracket}(x) = \max_{0 \leq x \leq 1} G_{\llbracket \mu \rrbracket}(x) - \min_{0 \leq x \leq 1} G_{\llbracket \mu \rrbracket}(x), \quad (3)$$

which merely expresses the intuitively obvious fact that the *span* (i.e. the difference between maximal and minimal value) of  $G_{\llbracket \mu \rrbracket}$  is not affected by the rotation caused by adding (modulo one) any number  $t$ . With the notation introduced earlier,  $G_{\llbracket \log_\gamma U_{a\gamma^{-\tau}, b\gamma^{-\tau}} \rrbracket} = g_{m,\delta}$ , where  $m = \llbracket \log_\gamma b/a \rrbracket$  and  $\delta = \llbracket \log_\gamma b/a \rrbracket$ . Combining (2) and (3) for  $\mu = \log_\gamma U_{a\gamma^{-\tau}, b\gamma^{-\tau}}$  and  $t = \tau$  therefore yields

$$\max_{0 \leq x \leq 1} G_{\llbracket \log_\gamma U_{a,b} \rrbracket}(x) - \min_{0 \leq x \leq 1} G_{\llbracket \log_\gamma U_{a,b} \rrbracket}(x) \geq 2C_\gamma.$$

This completes the proof.  $\square$

**Remark 2.** (i) As the above argument shows, the constant  $C_\gamma$  in (1) is best possible: For every  $C > C_\gamma$  there exist  $a, b$  with  $0 < a < b$  such that  $d_\infty(\llbracket \log_\gamma U_{a,b} \rrbracket, U_{0,1}) < C$ .

(ii) It follows from the first part of the proof of Theorem 1 that, for all  $\gamma > 1$  and  $b > 0$ ,

$$d_\infty(\llbracket \log_\gamma U_{0,b} \rrbracket, U_{0,1}) = \Psi(\log_\gamma b),$$

with the continuous, 1-periodic function  $\Psi : x \mapsto \max\{\psi(\llbracket x \rrbracket), 2C_\gamma - \psi(\llbracket x \rrbracket)\} = C_\gamma + |\psi(\llbracket x \rrbracket) - C_\gamma|$ .

(iii) Note that  $\gamma \mapsto C_\gamma$  is monotonically increasing, with  $\lim_{\gamma \rightarrow 1+} C_\gamma = 0$  and  $\lim_{\gamma \rightarrow \infty} C_\gamma = \frac{1}{2}$ . For  $\gamma = 10$ , the most important special case in view of Benford's Law, one finds  $C_{10} \approx 0.13442$ .

(iv) Satisfactory though it may be, Theorem 1 has a small shortcoming: For every probability measure  $\mu$  on  $\mathbb{R}$ , the measure  $\llbracket \mu \rrbracket$  naturally lives on  $\mathbb{T} = \mathbb{R}/\mathbb{Z}$  rather than on  $[0, 1]$ , but  $d_\infty$  is unsuitable for measures on  $\mathbb{T}$ . Specifically,  $\mathbb{T}$  is a compact metric space when endowed with the metric  $d(x + \mathbb{Z}, y + \mathbb{Z}) := \min_{k \in \mathbb{Z}} |x - y + k|$ , and consequently  $\mathcal{P}(\mathbb{T})$ , the space of all probability measures on  $\mathbb{T}$  with the topology of weak convergence, is compact and metrizable [3]. A natural metric inducing this topology is the Kantorovich–Wasserstein distance  $d_K$  defined as

$$d_K(\mu, \nu) := \sup \left\{ \left| \int_{\mathbb{T}} f \, d\mu - \int_{\mathbb{T}} f \, d\nu \right| : f \in C_{\mathbb{R}}(\mathbb{T}), \text{Lip } f \leq 1 \right\}.$$

Unlike  $d_\infty$ , the metric  $d_K$  on  $\mathcal{P}(\mathbb{T})$  is invariant under isometries of  $\mathbb{T}$ : If  $T : \mathbb{T} \rightarrow \mathbb{T}$  is any isometry, then  $d_K(T\mu, T\nu) = d_K(\mu, \nu)$  for all  $\mu, \nu \in \mathcal{P}(\mathbb{T})$ . Explicit practicable formulae for  $d_K$  have been derived in [1]. A truly satisfactory variant of Theorem 1, therefore, would consider  $[\log_\gamma U_{a,b}]$  an element of  $\mathcal{P}(\mathbb{T})$  and provide a lower bound for its distance from  $\lambda_{\mathbb{T}}$ , the uniform distribution on  $\mathbb{T}$ . Such a result can indeed be achieved using parts of the proof of Theorem 1 even though the necessary calculations are significantly more involved. The final result, however, is even slightly simpler than (1): For all  $\gamma > 1$  and  $0 \leq a < b$ ,

$$d_K([\log_\gamma U_{a,b}], \lambda_{\mathbb{T}}) \geq \log_\gamma \frac{1+\sqrt{\gamma}}{2} - \frac{1}{4} =: \frac{1}{4}\Phi\left(\frac{1}{4}\log \gamma\right) > 0, \quad (4)$$

where  $\Phi$  is the real-analytic odd function  $\Phi(x) = x^{-1} \log \cosh x$ ; as in the case of (1), the inequality (4) is best possible in the sense of (i).

## References

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