

# Poisson Summation and Benford's Law: From values of $L$ -functions to the $3x + 1$ problem to products of random variables

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- Applications of Poisson Summation:
  - ◊ L-functions, RMT,  $3x + 1$ .
  - ◊ Products Random Variables
    - (i) Fejér series
    - (ii) Mellin transforms
  - ◊ Sharp errors:  $X_i \sim \text{Exp}(1)$ ,  $\Xi_n = X_1 \cdots X_n$ , error  $\leq .057^n \log_{10} s$ .

## Benford's Law: Newcomb (1881), Benford (1938)

### Statement

For many data sets, probability of observing a first digit of  $d$  base  $B$  is  $\log_B \left( \frac{d+1}{d} \right)$ .

First 60 values of  $2^n$  (only displaying 30)

|     |        |           | digit | #  | Obs Prob | Benf Prob |
|-----|--------|-----------|-------|----|----------|-----------|
| 1   | 1024   | 1048576   |       |    |          |           |
| 2   | 2048   | 2097152   | 1     | 18 | .300     | .301      |
| 4   | 4096   | 4194304   | 2     | 12 | .200     | .176      |
| 8   | 8192   | 8388608   | 3     | 6  | .100     | .125      |
| 16  | 16384  | 16777216  | 4     | 6  | .100     | .097      |
| 32  | 32768  | 33554432  | 5     | 6  | .100     | .079      |
| 64  | 65536  | 67108864  | 6     | 4  | .067     | .067      |
| 128 | 131072 | 134217728 | 7     | 2  | .033     | .058      |
| 256 | 262144 | 268435456 | 8     | 5  | .083     | .051      |
| 512 | 524288 | 536870912 | 9     | 1  | .017     | .046      |

## Examples

- recurrence relations
- special functions (such as  $n!$ )
- iterates of power, exponential, rational maps
- products of random variables
- L-functions, characteristic polynomials
- iterates of the  $3x + 1$  map
- differences of order statistics
- hydrology and financial data

## Applications

- analyzing round-off errors
- determining the optimal way to store numbers
- detecting tax fraud and data integrity

## Logarithms and Benford's Law

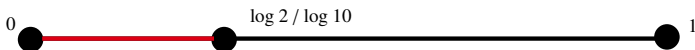
### Fundamental Equivalence

Data set  $\{x_i\}$  is Benford base  $B$  if  $\{y_i\}$  is equidistributed mod 1, where  $y_i = \log_B x_i$ .

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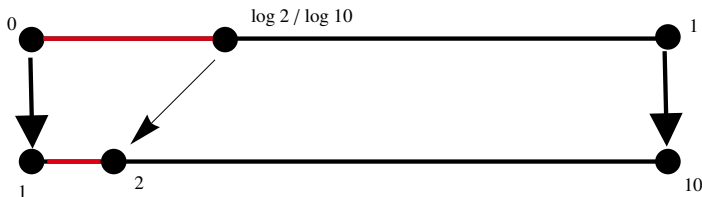
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### Proof:

- $x = M_B(x) \cdot B^k$  for some  $k \in \mathbb{Z}$ .
- $\text{FD}_B(x) = d$  iff  $d \leq M_B(x) < d + 1$ .
- $\log_B d \leq y < \log_B(d + 1)$ ,  $y = \log_B x \bmod 1$ .
- If  $Y \sim \text{Unif}(0, 1)$  then above probability is  $\log_B \left( \frac{d+1}{d} \right)$ .

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### Kronecker-Weyl Theorem

If  $\beta \notin \mathbb{Q}$  then  $n\beta \bmod 1$  is equidistributed.  
(Thus if  $\log_B \alpha \notin \mathbb{Q}$ , then  $\alpha^n$  is Benford.)

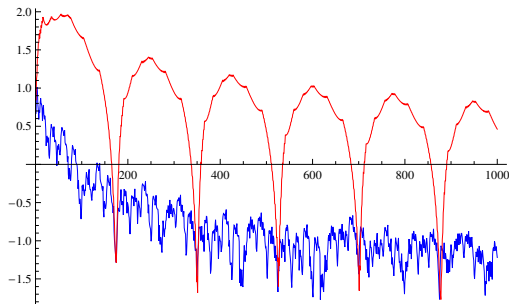
# Logarithms and Benford's Law

$\chi^2$  values for  $\alpha^n$ ,  $1 \leq n \leq N$  (5% 15.5).

| $N$  | $\chi^2(\gamma)$ | $\chi^2(e)$ | $\chi^2(\pi)$ |
|------|------------------|-------------|---------------|
| 100  | 0.72             | 0.30        | 46.65         |
| 200  | 0.24             | 0.30        | 8.58          |
| 400  | 0.14             | 0.10        | 10.55         |
| 500  | 0.08             | 0.07        | 2.69          |
| 700  | 0.19             | 0.04        | 0.05          |
| 800  | 0.04             | 0.03        | 6.19          |
| 900  | 0.09             | 0.09        | 1.71          |
| 1000 | 0.02             | 0.06        | 2.90          |

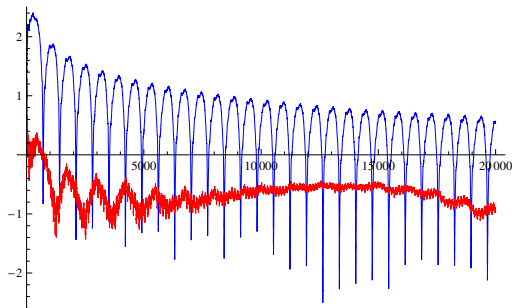
## Logarithms and Benford's Law: Base 10

$\log(\chi^2)$  vs  $N$  for  $\pi^n$  (red) and  $e^n$  (blue),  
 $n \in \{1, \dots, N\}$ . Note  $\pi^{175} \approx 1.0028 \cdot 10^{87}$ , (5%,  
 $\log(\chi^2) \approx 2.74$ ).



## Logarithms and Benford's Law: Base 20

$\log(\chi^2)$  vs  $N$  for  $\pi^n$  (red) and  $e^n$  (blue),  
 $n \in \{1, \dots, N\}$ . Note  $e^3 \approx 20.0855$ , (5%,  
 $\log(\chi^2) \approx 2.74$ ).



## Benford Good Processes

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- Poisson Summation Formula:  $f$  nice:

$$\sum_{\ell=-\infty}^{\infty} f(\ell) = \sum_{\ell=-\infty}^{\infty} \widehat{f}(\ell),$$

$$\text{Fourier transform } \widehat{f}(\xi) = \int_{-\infty}^{\infty} f(x) e^{-2\pi i x \xi} dx.$$

## Benford Good Process

$X_T$  is **Benford Good** if there is a nice  $f$  st

$$\text{CDF}_{\vec{Y}_{T,B}}(y) = \int_{-\infty}^y \frac{1}{T} f\left(\frac{t}{T}\right) dt + E_T(y) := G_T(y)$$

and monotonically increasing  $h$  ( $h(|T|) \rightarrow \infty$ ):

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- **Small translated error:**  $\mathcal{E}(a, b, T) =$   
 $\sum_{|\ell| \leq Th(T)} [E_T(b + \ell) - E_T(a + \ell)] = o(1).$

## Main Theorem

### Theorem (Kontorovich and M–, 2005)

*$X_T$  converging to  $X$  as  $T \rightarrow \infty$  (think spreading Gaussian). If  $X_T$  is Benford good, then  $X$  is Benford.*

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- **Examples**

- ◇  $L$ -functions
- ◇ characteristic polynomials (RMT)
- ◇  $3x + 1$  problem
- ◇ geometric Brownian motion.

## Sketch of the proof

- **Structure Theorem:**
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  - ◇ main term is something nice spreading out
  - ◇ apply Poisson summation
- **Control translated errors:**
  - ◇ hardest step
  - ◇ techniques problem specific

## Sketch of the proof (continued)

$$\sum_{\ell=-\infty}^{\infty} \mathbb{P} \left( \boldsymbol{a} + \ell \leq \overrightarrow{\boldsymbol{Y}}_{T,B} \leq \boldsymbol{b} + \ell \right)$$

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$$\begin{aligned}
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 &= \int_a^b \sum_{|\ell| \leq Th(T)} \frac{1}{T} f \left( \frac{t}{T} \right) dt + \mathcal{E}(a, b, T) + o(1)
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 &= \int_a^b \sum_{|\ell| \leq Th(T)} \frac{1}{T} f \left( \frac{t}{T} \right) dt + \mathcal{E}(a, b, T) + o(1) \\
 &= \widehat{f}(0) \cdot (b - a) + \sum_{\ell \neq 0} \widehat{f}(T\ell) \frac{e^{2\pi i b \ell} - e^{2\pi i a \ell}}{2\pi i \ell} + o(1).
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# *L*-functions and Random Matrix Theory

## Good *L*-functions ( $\zeta(s)$ , full level cusp form)

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$L$ -function is **good** if:

- Euler product:

$$L(s, f) = \sum_{n=1}^{\infty} \frac{a_f(n)}{n^s} = \prod_{p \text{ prime}} \prod_{j=1}^d (1 - \alpha_{f,j}(p)p^{-s})^{-1}.$$

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- meromorphic continuation to  $\mathbb{C}$ , of finite order, at most finitely many poles (all on the line  $\operatorname{Re}(s) = 1$ ).
- Functional equation:  $\omega \in \mathbb{R}$ ,  $G(s)$  prod  $\Gamma$ -fns:

$$e^{i\omega} G(s) L(s, f) = e^{-i\omega} \overline{G(1 - \bar{s}) L(1 - \bar{s})}.$$

## Good *L*-functions

- For some  $\aleph > 0$ ,  $c \in \mathbb{C}$ ,  $x \geq 2$ :

$$\sum_{p \leq x} \frac{|a_f(p)|^2}{p} = \aleph \log \log x + c + O\left(\frac{1}{\log x}\right).$$

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- The  $\alpha_{f,j}(p)$  are (Ramanujan-Petersson) tempered:  $|\alpha_{f,j}(p)| \leq 1$ .
- $N(\sigma, T) = \#\{\rho : L(\rho, f) = 0, \operatorname{Re}(\rho) \geq \sigma, \operatorname{Im}(\rho) \in [0, T]\}$ .  $\exists \beta > 0$

$$N(\sigma, T) = O\left(T^{1-\beta(\sigma-\frac{1}{2})} \log T\right).$$

# Log-Normal Law (Hejhal, Laurinćikas, Selberg)

## Log-Normal Law

$$\frac{\mu(\{t \in [T, 2T] : \log |L(\sigma + it, f)| \in [a, b]\})}{T} =$$

$$\frac{1}{\sqrt{\psi(\sigma, T)}} \int_a^b e^{-\pi u^2 / \psi(\sigma, T)} du + \text{Error}$$

$$\psi(\sigma, T) = \aleph \log \left[ \min \left( \log T, \frac{1}{\sigma - \frac{1}{2}} \right) \right] + O(1)$$

$$\frac{1}{2} \leq \sigma \leq \frac{1}{2} + \frac{1}{\log^\delta T}, \quad \delta \in (0, 1).$$

## Values of *L*-functions and Benford's Law

### Theorem (Kontorovich and M–, 2005)

*$L(s, f)$  a good *L*-function, as  $T \rightarrow \infty$ ,  
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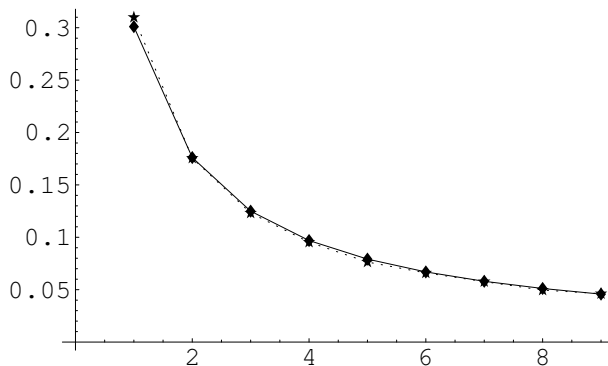
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- Montgomery-Vaughan:

$$\int_T^{2T} \sum a_n n^{-it} \overline{\sum b_m m^{-it}} dt = H \sum a_n \bar{b}_n + O(1) \sqrt{\sum n |a_n|^2 \sum n |b_n|^2}.$$

# Riemann Zeta Function

$$\left| \zeta \left( \frac{1}{2} + i \frac{k}{4} \right) \right|, k \in \{0, 1, \dots, 65535\}.$$



## Random Matrices: Preliminaries

- $N \times N$  unitary matrices  $U$  (Haar measure):

$$p_N(U) = \frac{1}{(2\pi)^N N!} \prod_{1 \leq j < m \leq N} |e^{i\theta_j} - e^{i\theta_m}|.$$

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- $\rho_N(x)$  the probability density for  $\log |Z(U, \theta)|$ :

$$\tilde{\rho}_N(x) = \sqrt{Q_2(N)} \rho_N(\sqrt{Q_2(N)} x),$$

variance  $Q_2(N) \sim (\log N)/2$ .

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### Theorem (Kontorovich and M–, 2005)

*As  $N \rightarrow \infty$ , the distribution of digits of the absolute values of the characteristic polynomials of  $N \times N$  unitary matrices (with respect to Haar measure) converges to the Benford probabilities.*

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- Key Ingredient: Keating-Snaith:

$$\tilde{\rho}_N(x)dx = \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx + O\left((\log N)^{-3/2} dx\right).$$

# The $3x + 1$ Problem and Benford's Law

## $3x + 1$ Problem

- Kakutani (conspiracy), Erdős (not ready).
- $x$  odd,  $T(x) = \frac{3x+1}{2^k}$ ,  $2^k || 3x + 1$ .

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- $7 \rightarrow_1 11 \rightarrow_1 17 \rightarrow_2 13 \rightarrow_3 5 \rightarrow_4 1 \rightarrow_2 1$ ,  
 2-path (1, 1), 5-path (1, 1, 2, 3, 4).  
*m*-path:  $(k_1, \dots, k_m)$ .

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 &= \log a_n + \log \left( \frac{3}{4} \right).
 \end{aligned}$$

Geometric Brownian Motion, drift  $\log(3/4) < 1$ .

## Structure Theorem: Sinai, Kontorovich-Sinai

$$\mathbb{P}(A) = \lim_{N \rightarrow \infty} \frac{\#\{n \leq N: n \equiv 1, 5 \pmod{6}, n \in A\}}{\#\{n \leq N: n \equiv 1, 5 \pmod{6}\}}.$$

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## 3x + 1 and Benford

### Theorem (Kontorovich and M–, 2005)

*As  $m \rightarrow \infty$ ,  $x_m / (3/4)^m x_0$  is Benford.*

### Theorem (Lagarias-Soundararajan 2006)

*$X \geq 2^N$ , for all but at most  $c(B)N^{-1/36}X$  initial seeds the distribution of the first  $N$  iterates of the  $3x + 1$  map are within  $2N^{-1/36}$  of the Benford probabilities.*

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$$C = \log_B 2 \text{ of irrationality type } \kappa < \infty:$$

$$\#\{k \in I_\ell : \overline{kC} \in [a, b]\} = M(b-a) + O(M^{1+\epsilon-1/\kappa}).$$

## Irrationality Type

### Irrationality type

$\alpha$  has irrationality type  $\kappa$  if  $\kappa$  is the supremum of all  $\gamma$  with

$$\lim_{q \rightarrow \infty} q^{\gamma+1} \min_p \left| \alpha - \frac{p}{q} \right| = 0.$$

- Algebraic irrationals: type 1 (Roth's Thm).
- Theory of Linear Forms:  $\log_B 2$  of finite type.

## Linear Forms

### Theorem (Baker)

$\alpha_1, \dots, \alpha_n$  algebraic numbers height  $A_j \geq 4$ ,  
 $\beta_1, \dots, \beta_n \in \mathbb{Q}$  with height at most  $B \geq 4$ ,

$$\Lambda = \beta_1 \log \alpha_1 + \dots + \beta_n \log \alpha_n.$$

If  $\Lambda \neq 0$  then  $|\Lambda| > B^{-C\Omega \log \Omega'}$ , with  
 $d = [\mathbb{Q}(\alpha_i, \beta_j) : \mathbb{Q}]$ ,  $C = (16nd)^{200n}$ ,  
 $\Omega = \prod_j \log A_j$ ,  $\Omega' = \Omega / \log A_n$ .

Gives  $\log_{10} 2$  of finite type, with  $\kappa < 1.2 \cdot 10^{602}$ :

$$|\log_{10} 2 - p/q| = |q \log 2 - p \log 10| / q \log 10.$$

## Quantified Equidistribution

### Theorem (Erdős-Turan)

$$D_N = \frac{\sup_{[a,b]} |N(b-a) - \#\{n \leq N : x_n \in [a,b]\}|}{N}$$

*There is a  $C$  such that for all  $m$ :*

$$D_N \leq C \cdot \left( \frac{1}{m} + \sum_{h=1}^m \frac{1}{h} \left| \frac{1}{N} \sum_{n=1}^N e^{2\pi i h x_n} \right| \right)$$

## Proof of Erdős-Turan

Consider special case  $x_n = n\alpha$ ,  $\alpha \notin \mathbb{Q}$ .

- Exponential sum  $\leq \frac{1}{|\sin(\pi h\alpha)|} \leq \frac{1}{2||h\alpha||}$ .
- Must control  $\sum_{h=1}^m \frac{1}{h||h\alpha||}$ , see irrationality type enter.
- type  $\kappa$ ,  $\sum_{h=1}^m \frac{1}{h||h\alpha||} = O(m^{\kappa-1+\epsilon})$ , take  $m = \lfloor N^{1/\kappa} \rfloor$ .

## $3x + 1$ Data: random 10,000 digit number, $2^k \parallel 3x + 1$

80,514 iterations ( $(4/3)^n = a_0$  predicts 80,319);  
 $\chi^2 = 13.5$  (5% 15.5).

| Digit | Number | Observed | Benford |
|-------|--------|----------|---------|
| 1     | 24251  | 0.301    | 0.301   |
| 2     | 14156  | 0.176    | 0.176   |
| 3     | 10227  | 0.127    | 0.125   |
| 4     | 7931   | 0.099    | 0.097   |
| 5     | 6359   | 0.079    | 0.079   |
| 6     | 5372   | 0.067    | 0.067   |
| 7     | 4476   | 0.056    | 0.058   |
| 8     | 4092   | 0.051    | 0.051   |
| 9     | 3650   | 0.045    | 0.046   |

## 3x + 1 Data: random 10,000 digit number, 2|3x + 1

241,344 iterations,  $\chi^2 = 11.4$  (5% 15.5).

| Digit | Number | Observed | Benford |
|-------|--------|----------|---------|
| 1     | 72924  | 0.302    | 0.301   |
| 2     | 42357  | 0.176    | 0.176   |
| 3     | 30201  | 0.125    | 0.125   |
| 4     | 23507  | 0.097    | 0.097   |
| 5     | 18928  | 0.078    | 0.079   |
| 6     | 16296  | 0.068    | 0.067   |
| 7     | 13702  | 0.057    | 0.058   |
| 8     | 12356  | 0.051    | 0.051   |
| 9     | 11073  | 0.046    | 0.046   |

**$5x + 1$  Data: random 10,000 digit number,  $2^k \parallel 5x + 1$**

27,004 iterations,  $\chi^2 = 1.8$  (5% 15.5).

| Digit | Number | Observed | Benford |
|-------|--------|----------|---------|
| 1     | 8154   | 0.302    | 0.301   |
| 2     | 4770   | 0.177    | 0.176   |
| 3     | 3405   | 0.126    | 0.125   |
| 4     | 2634   | 0.098    | 0.097   |
| 5     | 2105   | 0.078    | 0.079   |
| 6     | 1787   | 0.066    | 0.067   |
| 7     | 1568   | 0.058    | 0.058   |
| 8     | 1357   | 0.050    | 0.051   |
| 9     | 1224   | 0.045    | 0.046   |

## 5x + 1 Data: random 10,000 digit number, 2|5x + 1

241,344 iterations,  $\chi^2 = 3 \cdot 10^{-4}$  (5% 15.5).

| Digit | Number | Observed | Benford |
|-------|--------|----------|---------|
| 1     | 72652  | 0.301    | 0.301   |
| 2     | 42499  | 0.176    | 0.176   |
| 3     | 30153  | 0.125    | 0.125   |
| 4     | 23388  | 0.097    | 0.097   |
| 5     | 19110  | 0.079    | 0.079   |
| 6     | 16159  | 0.067    | 0.067   |
| 7     | 13995  | 0.058    | 0.058   |
| 8     | 12345  | 0.051    | 0.051   |
| 9     | 11043  | 0.046    | 0.046   |

# Products of Random Variables

## Preliminaries

- $X_1 \cdots X_n \Leftrightarrow Y_1 + \cdots + Y_n \bmod 1$ ,  $Y_i = \log_B X_i$
- Density  $Y_i$  is  $g_i$ , density  $Y_i + Y_j$  is

$$(g_i * g_j)(y) = \int_0^1 g_i(t)g_j(y - t)dt.$$

- $h_n = g_1 * \cdots * g_n$ ,  $\widehat{g}(\xi) = \widehat{g}_1(\xi) \cdots \widehat{g}_n(\xi)$ .
- Dirac delta functional:  $\int \delta_\alpha(y)g(y)dy = g(\alpha)$ .

## Fourier input

- Fejér kernel:

$$F_N(x) = \sum_{n=-N}^N \left(1 - \frac{|n|}{N}\right) e^{2\pi i n x}.$$

- Fejér series:

$$T_N f(x) = (f * F_N)(x) = \sum_{n=-N}^N \left(1 - \frac{|n|}{N}\right) \widehat{f}(n) e^{2\pi i n x}.$$

- Lebesgue's Theorem:  $f \in L^1([0, 1])$ . As  $N \rightarrow \infty$ ,  $T_N f$  converges to  $f$  in  $L^1([0, 1])$ .
- $T_N(f * g) = (T_N f) * g$ : convolution assoc.

## Modulo 1 Central Limit Theorem

### Theorem (M– and Nigrini 2007)

$\{Y_m\}$  independent continuous random variables on  $[0, 1)$  (not necc. i.i.d.), densities  $\{g_m\}$ .

$Y_1 + \cdots + Y_M \bmod 1$  converges to the uniform distribution as  $M \rightarrow \infty$  in  $L^1([0, 1])$  iff  $\forall n \neq 0$ ,  
 $\lim_{M \rightarrow \infty} \widehat{g}_1(n) \cdots \widehat{g}_M(n) = 0$ .

## Generalizations

- Levy proved for i.i.d.r.v. just one year after Benford's paper.
- Generalized to other compact groups, with estimates on the rate of convergence.
  - ◊ Stromberg:  $n$ -fold convolution of a regular probability measure on a compact Hausdorff group  $G$  converges to normalized Haar measure in weak-star topology iff support of the distribution not contained in a coset of a proper normal closed subgroup of  $G$ .

## Theorem (M– and Nigrini 2007)

*$\{Y_m\}$  indep. discrete random variables on  $[0, 1)$ , not necc. identically distributed, densities*

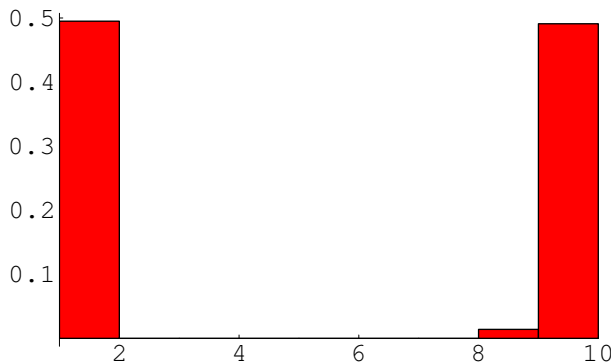
$$g_m(x) = \sum_{k=1}^{r_m} w_{k,m} \delta_{\alpha_{k,m}}(x), w_{k,m} > 0, \sum_{k=1}^{r_m} w_{k,m} = 1.$$

*Assume that there is a finite set  $A \subset [0, 1)$  such that all  $\alpha_{k,m} \in A$ .  $Y_1 + \dots + Y_M \bmod 1$  converges weakly to the uniform distribution as  $M \rightarrow \infty$  iff  $\forall n \neq 0, \lim_{M \rightarrow \infty} \widehat{g}_1(n) \cdots \widehat{g}_M(n) = 0$ .*

# Distribution of digits (base 10) of 1000 products

$X_1 \cdots X_{1000}$ , where  $g_{10,m} = \phi_{11^m}$ .

$\phi_m(x) = m$  if  $|x - 1/8| \leq 1/2m$  (0 otherwise).



## Proof of Modulo 1 CLT

- Density of sum is  $h_\ell = g_1 * \cdots * g_\ell$ .
- Suffices show  $\forall \epsilon$ :  

$$\lim_{M \rightarrow \infty} \int_0^1 |h_M(x) - 1| dx < \epsilon.$$
- Lebesgue's Theorem:  $N$  large,

$$\|h_1 - T_N h_1\|_1 = \int_0^1 |h_1(x) - T_N h_1(x)| dx < \frac{\epsilon}{2}.$$

- Claim: above holds for  $h_M$  for all  $M$ .

## Proof of claim

$$T_N h_{M+1} = T_N(h_M * g_{M+1}) = (T_N h_M) * g_{M+1}$$

$$\begin{aligned}
 \|h_{M+1} - T_N h_{M+1}\|_1 &= \int_0^1 |h_{M+1}(x) - T_N h_{M+1}(x)| dx \\
 &= \int_0^1 |(h_M * g_{M+1})(x) - (T_N h_M) * g_{M+1}(x)| dx \\
 &= \int_0^1 \left| \int_0^1 (h_M(y) - T_N h_M(y)) g_{M+1}(x-y) dy \right| dx \\
 &\leq \int_0^1 \int_0^1 |h_M(y) - T_N h_M(y)| g_{M+1}(x-y) dx dy \\
 &= \int_0^1 |h_M(y) - T_N h_M(y)| dy \cdot 1 < \frac{\epsilon}{2}.
 \end{aligned}$$

## Proof of Modulo 1 CLT (continued)

Show  $\lim_{M \rightarrow \infty} \|h_M - 1\|_1 = 0$ .

Triangle inequality:

$$\|h_M - 1\|_1 \leq \|h_M - T_N h_M\|_1 + \|T_N h_M - 1\|_1.$$

Choices of  $N$  and  $\epsilon$ :

$$\|h_M - T_N h_M\|_1 < \epsilon/2.$$

Show  $\|T_N h_M - 1\|_1 < \epsilon/2$ .

$$\begin{aligned} \|T_N h_M - 1\|_1 &= \int_0^1 \left| \sum_{\substack{n=-N \\ n \neq 0}}^N \left(1 - \frac{|n|}{N}\right) \widehat{h}_M(n) e^{2\pi i n x} \right| dx \\ &\leq \sum_{\substack{n=-N \\ n \neq 0}}^N \left(1 - \frac{|n|}{N}\right) |\widehat{h}_M(n)| \end{aligned}$$

$$\widehat{h}_M(n) = \widehat{g}_1(n) \cdots \widehat{g}_M(n) \xrightarrow{M \rightarrow \infty} 0.$$

For fixed  $N$  and  $\epsilon$ , choose  $M$  large so that

$$|\widehat{h}_M(n)| < \epsilon/4N \text{ whenever } n \neq 0 \text{ and } |n| \leq N.$$

## Where's the Poisson Summation?

- Mellin transform and Fourier transform related by **logarithmic** change of variable.
- Poisson summation from collapsing to modulo 1 random variables.

## Preliminaries

- $\Xi_1, \dots, \Xi_n$  nice independent r.v.'s on  $[0, \infty)$ .

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$$\int_0^\infty f_2\left(\frac{x}{t}\right) f_1(t) \frac{dt}{t}$$

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◇ Proof:  $\text{Prob}(\Xi_1 \cdot \Xi_2 \in [0, x])$ :

$$\begin{aligned} & \int_{t=0}^\infty \text{Prob}\left(\Xi_2 \in \left[0, \frac{x}{t}\right]\right) f_1(t) dt \\ &= \int_{t=0}^\infty F_2\left(\frac{x}{t}\right) f_1(t) dt, \end{aligned}$$

differentiate.

# Mellin Transform

$$(\mathcal{M}f)(s) = \int_0^{\infty} f(x) x^s \frac{dx}{x}$$

$$(\mathcal{M}^{-1}g)(x) = \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} g(s) x^{-s} ds$$

$$g(s) = (\mathcal{M}f)(s), f(x) = (\mathcal{M}^{-1}g)(x).$$

$$(f_1 \star f_2)(x) = \int_0^{\infty} f_2\left(\frac{x}{t}\right) f_1(t) \frac{dt}{t}$$

$$(\mathcal{M}(f_1 \star f_2))(s) = (\mathcal{M}f_1)(s) \cdot (\mathcal{M}f_2)(s).$$

## Mellin Transform Formulation: Products Random Variables

### Theorem

$X_i$ 's independent, densities  $f_i$ .  $\Xi_n = X_1 \cdots X_n$ ,

$$h_n(x_n) = (f_1 \star \cdots \star f_n)(x_n)$$

$$(\mathcal{M}h_n)(s) = \prod_{m=1}^n (\mathcal{M}f_m)(s).$$

As  $n \rightarrow \infty$ ,  $\Xi_n$  becomes Benford:  $Y_n = \log_B \Xi_n$ ,  
 $|\text{Prob}(Y_n \bmod 1 \in [a, b]) - (b - a)| \leq$

$$(b - a) \cdot \sum_{\ell \neq 0, \ell = -\infty}^{\infty} \prod_{m=1}^n (\mathcal{M}f_i) \left( 1 - \frac{2\pi i \ell}{\log B} \right).$$

## Proof of Mellin formulation of product

$$Y_n = \log_B \Xi_n:$$

$$\text{Prob}(Y_n \leq y) = \text{Prob}(\Xi_n \leq B^y) = H_n(B^y)$$

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Differentiate:  $g_n(y) = h_n(B^y)B^y \log B$ .

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Differentiate:  $g_n(y) = h_n(B^y)B^y \log B$ .

Poisson summation:

$$\sum_{\ell=-\infty}^{\infty} g_n(y + \ell) = \sum_{\ell=-\infty}^{\infty} e^{2\pi i y \ell} \widehat{g}_n(\ell)$$

where  $\widehat{f}$  denotes the Fourier transform of  $f$ :

$$\widehat{f}(\xi) = \int_{-\infty}^{\infty} f(x) e^{-2\pi i x \xi} dx.$$

## Proof of the Mellin formulation for the product (continued)

$\text{Prob}(Y_n \bmod 1 \in [a, b])$  is

$$\begin{aligned} & \sum_{\ell=-\infty}^{\infty} \int_{a+\ell}^{b+\ell} g_n(y) dy \\ &= \int_a^b \sum_{\ell=-\infty}^{\infty} g_n(y + \ell) dy \end{aligned}$$

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 &= \int_a^b \sum_{\ell=-\infty}^{\infty} e^{2\pi i y \ell} \widehat{g}_n(\ell) dy \\
 &= b - a + (b - a) \sum_{\ell \neq 0} |\widehat{g}_n(\ell)|.
 \end{aligned}$$

## Proof of the Mellin formulation for the product (continued)

Must show sum over  $\ell$  tends to zero as  $n \rightarrow \infty$ .

$$\widehat{g}_n(\xi) = \int_{-\infty}^{\infty} g_n(y) e^{-2\pi i y \xi} dy$$

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Must show sum over  $\ell$  tends to zero as  $n \rightarrow \infty$ .

$$\begin{aligned}\widehat{g}_n(\xi) &= \int_{-\infty}^{\infty} g_n(y) e^{-2\pi i y \xi} dy \\ &= \int_{-\infty}^{\infty} h_n(B^y) B^y \log B \cdot e^{-2\pi i y \xi} dy\end{aligned}$$

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 &= \int_{-\infty}^{\infty} h_n(B^y) B^y \log B \cdot e^{-2\pi i y \xi} dy \\
 &= \int_0^{\infty} h_n(t) t^{-2\pi i \xi / \log B} dt
 \end{aligned}$$

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 \widehat{g}_n(\xi) &= \int_{-\infty}^{\infty} g_n(y) e^{-2\pi i y \xi} dy \\
 &= \int_{-\infty}^{\infty} h_n(B^y) B^y \log B \cdot e^{-2\pi i y \xi} dy \\
 &= \int_0^{\infty} h_n(t) t^{-2\pi i \xi / \log B} dt \\
 &= (\mathcal{M}h_n) \left( 1 - \frac{2\pi i \xi}{\log B} \right) = \prod_{m=1}^n (\mathcal{M}f_m) \left( 1 - \frac{2\pi i \xi}{\log B} \right).
 \end{aligned}$$

## Proof of Kossovsky's Chain Conjecture for certain densities

### Conditions

- $\{\mathcal{D}_i(\theta)\}_{i \in I}$ : one-parameter distributions, densities  $f_{\mathcal{D}_i(\theta)}$  on  $[0, \infty)$ .

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$$f_m(x_m) = \int_0^\infty f_{\mathcal{D}_{p(m)}(1)}\left(\frac{x_m}{x_{m-1}}\right) f_{m-1}(x_{m-1}) \frac{dx_{m-1}}{x_{m-1}}$$

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$$\lim_{n \rightarrow \infty} \sum_{\substack{\ell = -\infty \\ \ell \neq 0}}^{\infty} \prod_{m=1}^n (\mathcal{M} f_{\mathcal{D}_{p(m)}(1)}) \left(1 - \frac{2\pi i \ell}{\log B}\right) = 0$$

# Proof of Kossovsky's Chain Conjecture for certain densities

## Theorem (JKKKM)

- *If conditions hold, as  $n \rightarrow \infty$  the distribution of leading digits of  $X_n$  tends to Benford's law.*
- *The error is a nice function of the Mellin transforms: if  $Y_n = \log_B X_n$ , then*

$$\begin{aligned}
 & |\text{Prob}(Y_n \bmod 1 \in [a, b]) - (b - a)| \leq \\
 & \left| (b - a) \cdot \sum_{\substack{\ell=-\infty \\ \ell \neq 0}}^{\infty} \prod_{m=1}^n (\mathcal{M} f_{\mathcal{D}_{p(m)}(1)}) \left( 1 - \frac{2\pi i \ell}{\log B} \right) \right|
 \end{aligned}$$

## Example: All $X_i \sim \text{Unif}(0, k)$

- $X_i \sim \text{Unif}(0, k)$ : without loss of generality  $k \in [1, 10)$ .
- $P_n(s) = \text{Prob}(M_{10}(\Xi_n) \leq s)$ .
- $|P_n(s) - \log_{10}(s)| \leq$

$$\frac{k (\log k)^{n-1}}{s \Gamma(n)} + \left( \frac{1}{2.9^n} + \frac{\zeta(n) - 1}{2.7^n} \right) 2 \log_{10} s.$$

## Example: All $X_i \sim \text{Exp}(1)$

- $X_i \sim \text{Exp}(1)$ ,  $Y_n = \log_B \Xi_n$ .
- Needed ingredients:
  - ◇  $\int_0^\infty \exp(-x) x^{s-1} dx = \Gamma(s)$ .
  - ◇  $|\Gamma(1 + ix)| = \sqrt{\pi x / \sinh(\pi x)}$ ,  $x \in \mathbb{R}$ .
- $|P_n(s) - \log_{10}(s)| \leq$

$$\log_B s \sum_{\ell=1}^{\infty} \left( \frac{2\pi^2 \ell / \log B}{\sinh(2\pi^2 \ell / \log B)} \right)^{n/2}.$$

## Example: All $X_i \sim \text{Exp}(1)$ (continued)

### Bounds on the error

- $|P_n(s) - \log_{10} s| \leq$ 
  - ◇  $3.3 \cdot 10^{-3} \log_B s$  if  $n = 2$ ,
  - ◇  $1.9 \cdot 10^{-4} \log_B s$  if  $n = 3$ ,
  - ◇  $1.1 \cdot 10^{-5} \log_B s$  if  $n = 5$ , and
  - ◇  $3.6 \cdot 10^{-13} \log_B s$  if  $n = 10$ .
- Error at most

$$\log_{10} s \sum_{\ell=1}^{\infty} \left( \frac{17.148\ell}{\exp(8.5726\ell)} \right)^{n/2} \leq .057^n \log_{10} s$$

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