

Benford's Law, Values of L -Functions and the $3x + 1$ Problem, or: Why the IRS should care about Number Theory

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Five Colleges Number Theory Seminar
January 31, 2012

Introduction

Interesting Question

For a nice data set, such as the Fibonacci numbers, stock prices, street addresses of Five Colleges Number Theory Seminar participants, ..., what percent of the leading digits are 1?

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Plausible answers: 10%, 11%, about 30%.

Summary

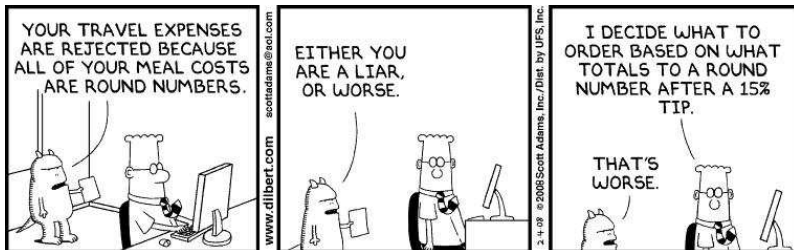
- State Benford's Law.
- Discuss examples and applications.
- Sketch proofs.
- Describe open problems.

Caveats!

- A math test indicating fraud is *not* proof of fraud: unlikely events, alternate reasons.

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 - ◇ **Many streets of different sizes: close to Benford.**

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
- many hierarchical Bayesian models

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Applications

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

General Theory

Mantissas (or Significands)

Mantissa: $x = M_{10}(x) \cdot 10^k$, k integer.

$M_{10}(x) = M_{10}(\tilde{x})$ if and only if x and \tilde{x} have the same leading digits.

Key observation: $\log_{10}(x) = \log_{10}(\tilde{x}) \bmod 1$ if and only if x and \tilde{x} have the same leading digits. Thus often study $y = \log_{10} x$.

Equidistribution and Benford's Law

Equidistribution

$\{y_n\}_{n=1}^{\infty}$ is equidistributed modulo 1 if probability $y_n \bmod 1 \in [a, b]$ tends to $b - a$:

$$\frac{\#\{n \leq N : y_n \bmod 1 \in [a, b]\}}{N} \rightarrow b - a.$$

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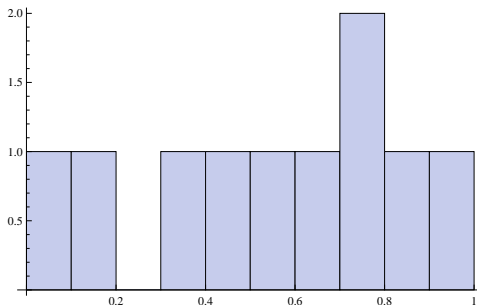
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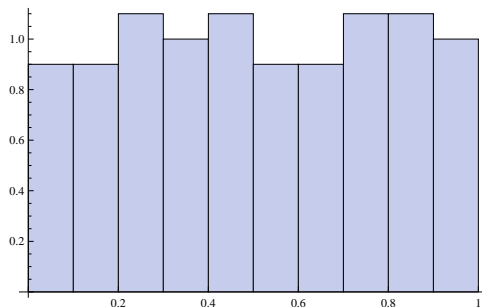
- Thm: $\beta \notin \mathbb{Q}$, $n\beta$ is equidistributed mod 1.
- Examples: $\log_{10} 2, \log_{10} \left(\frac{1+\sqrt{5}}{2}\right) \notin \mathbb{Q}$.

Example of Equidistribution: $n\sqrt{\pi} \bmod 1$



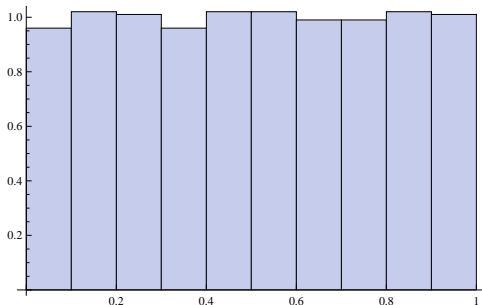
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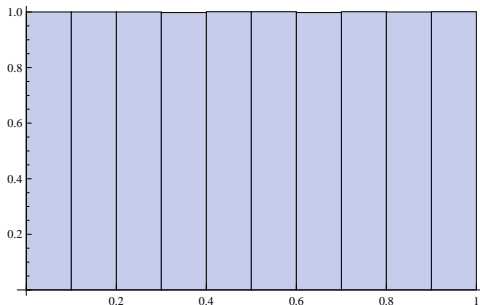
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$n\sqrt{\pi} \bmod 1$ for $n \leq 1000$

Example of Equidistribution: $n\sqrt{\pi} \bmod 1$



$n\sqrt{\pi} \bmod 1$ for $n \leq 10,000$

Logarithms and Benford's Law

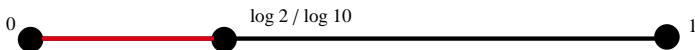
Fundamental Equivalence

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

Logarithms and Benford's Law

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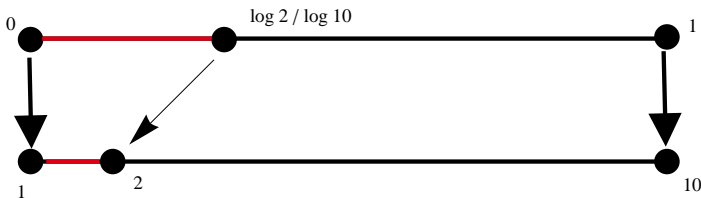
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$$a_{n+1} = a_n + a_{n-1}.$$

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- Most linear recurrence relations Benford.

$$a_{n+1} = 2a_n$$

$$a_{n+1} = 2a_n - a_{n-1}$$

Take $a_0 = a_1 = 1$ or $a_0 = 0, a_1 = 1$.

Digits of 2^n

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
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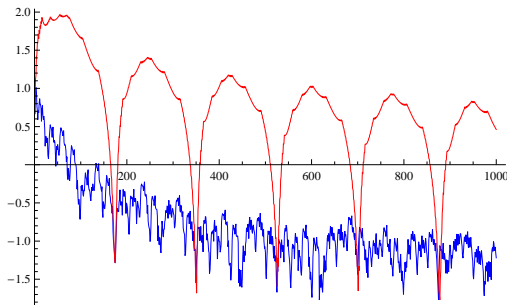
Logarithms and Benford's Law

χ^2 values for α^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

$\log_{10}(\chi^2)$ vs N for π^n (red) and e^n (blue),
 $n \in \{1, \dots, N\}$. Note $\pi^{175} \approx 1.0028 \cdot 10^{87}$, (5%
 and 8 d.f., $\log_{10}(\chi^2) \approx .44$).



Applications

Applications for the IRS: Detecting Fraud

93-4670

1040 Department of the Treasury - Internal Revenue Service
U.S. Individual Income Tax Return 1992

For the year 1992, 1-1-92 to 12-31-92, or other tax year beginning 1992, ending

OMB Use only—Do not write or stamp in this space
OMB No. 1545-0047

Label
Your social security number
SEARCHED SERIALIZED INDEXED FILED
FAR Privacy Act and Paperwork Reduction Act Notice, see page 4.

Use the IRS label.
Otherwise, please print in type

WILLIAM J. CLINTON
HILLARY RODHAM CLINTON
THE WHITE HOUSE
1600 PENNSYLVANIA AVENUE N.W.
WASHINGTON, DC 20500

Do you want \$1 to go to the fund?
If joint return, does your spouse want \$1 to go to the fund?

1 Single
2 Married filing joint return (even if only one had income)
3 Married filing separate return. Enter spouse's SSN above and full name below.
4 Head of household.
5 Qualifying widow(er) with dependent child.
6 Dependent.
7 Beneficiary of an estate.
8 Trust beneficiary.

Exemptions

Dependent's name (Last, first, and last name)	SSN	31 Page 104 (a) dependent's name (last, first, and last name)	Relationship to you	Qualifies as an exempt person (1992)	No. of tax credits for this person
CHYLSA	081	DAUSHEWER	DAUGHTER	1	1

Income

7	Wages, salaries, tips, etc. (Attach Form W-2)	7	237,659
8	Taxable interest income. Attach Schedule B if over \$400	8	7,269
9	Dividend or capital gain distributions. Attach Schedule D if over \$400	9	743
10	Taxable refunds, credits, or offsets of state and local income taxes	10	1,404
11	Alimony received	11	
12	Business income or loss. Attach Schedule C or C-EZ	12	16,336
13	Capital gain or loss. Attach Schedule D	13	
14	Capital gain distributions not reported on line 13	14	
15	Other gains or losses. Attach Form 4797	15	
16	Total IRA distributions	16	
17	Total pensions and annuities	17	
18	Rents, royalties, partnerships, estates, trusts, etc. Attach Schedule E	18	1,328
19	Farmland income or loss. Attach Schedule F	19	
20	Unemployment compensation	20	
21	Social Security benefits	21	
22	Other income (Losses from other sources are shown on other lines)	22	22,400
23	Add the amounts in the far right column for lines 7 through 22. This is your total income	23	32,400
24	Subtract the amount in the far right column for line 23 from line 23. This is your total income	24	297,177

Adjustments to income

25	Spouse's IRA deduction	25	250
26	Overseas self-employment tax	26	28
27	Self-employed health insurance deduction	27	28
28	Keogh retirement plan and self-employed SEP deduction	28	6,480
29	Penalty on early withdrawal of savings	29	
30	Alimony paid. Attach Form 1041	30	
31	Add lines 25 through 30. These are your total adjustments	31	6,480
32	Subtract line 31 from line 24. This is your adjusted gross income.	32	290,697

AGI 290,697 From Table (100)

1079

Detecting Fraud

Bank Fraud

- Audit of a bank revealed huge spike of numbers

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- Write-off limit of \$5,000. Officer had friends applying for credit cards, ran up balances just under \$5,000 then he would write the debts off.

Benford Good Processes

Poisson Summation and Benford's Law: Definitions

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- Poisson Summation Formula: f nice:

$$\sum_{l=-\infty}^{\infty} f(l) = \sum_{l=-\infty}^{\infty} \hat{f}(l),$$

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

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)

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- **Examples**

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

Sketch of the proof

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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)

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

Sketch of the proof (continued)

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 = & \sum_{|\ell| \leq Th(T)} [\mathbf{G}_T(\mathbf{b} + \ell) - \mathbf{G}_T(\mathbf{a} + \ell)] + o(1)
 \end{aligned}$$

Sketch of the proof (continued)

$$\begin{aligned}
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 = & \int_{\mathbf{a}}^{\mathbf{b}} \sum_{|\ell| \leq Th(T)} \frac{1}{T} f \left(\frac{t}{T} \right) dt + \mathcal{E}(\mathbf{a}, \mathbf{b}, T) + o(1)
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 = & \int_a^b \sum_{|l| \leq Th(T)} \frac{1}{T} f\left(\frac{t}{T}\right) dt + \mathcal{E}(\mathbf{a}, \mathbf{b}, T) + o(1) \\
 = & \widehat{f}(0) \cdot (\mathbf{b} - \mathbf{a}) + \sum_{l \neq 0} \widehat{f}(Tl) \frac{e^{2\pi i b l} - e^{2\pi i a l}}{2\pi i l} + o(1).
 \end{aligned}$$

L-functions and Random Matrix Theory

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

L -function is **good** if:

- Euler product:

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

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

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

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

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

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 $T \rightarrow \infty$,
 $L(\sigma_T + it, f)$ is Benford.

Ingredients

- Approximate $\log L(\sigma_T + it, f)$ with

$$\sum_{n \leq x} \frac{c(n)\Lambda(n)}{\log n} \frac{1}{n^{\sigma_T + it}}.$$

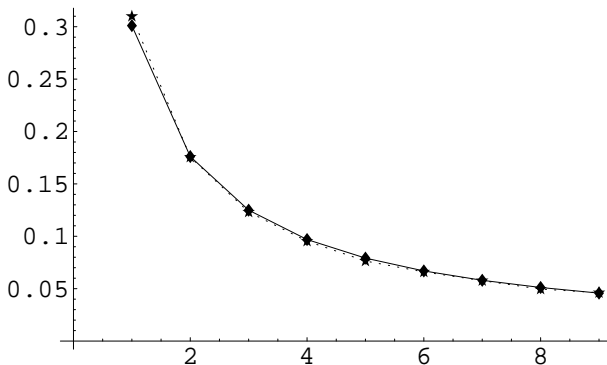
- study moments $\int_T^{2T} |\cdot|^k, k \leq \log^{1-\delta} T.$

- 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):

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

- characteristic polynomial:

$$Z(U, \theta) = \det(I - Ue^{-i\theta}) = \prod (1 - e^{i(\theta_n - \theta)}).$$

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

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

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

Random Matrices and Benford's Law

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.

- 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

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- x odd, $T(x) = \frac{3x+1}{2^k}$, $2^k \parallel 3x + 1$.

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 2-path (1, 1), 5-path (1, 1, 2, 3, 4).
m-path: (k_1, \dots, k_m) .

Heuristic Proof of $3x + 1$ Conjecture

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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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$$6 \cdot 2^{k_1 + \dots + k_m} p + q.$$

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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

α has irrationality type κ if κ is the supremum of all γ with

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

- 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

$$d = [\mathbb{Q}(\alpha_i, \beta_j) : \mathbb{Q}], \quad C = (16nd)^{200n},$$

$$\Omega = \prod_j \log A_j, \quad \Omega' = \Omega / \log A_n.$$

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 κ , $\sum_{h=1}^m \frac{1}{h\|h\alpha\|} = O(m^{\kappa-1+\epsilon})$, 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

Conclusions

Conclusions and Future Investigations

- See many different systems exhibit Benford behavior.
- Ingredients of proofs (logarithms, equidistribution).
- Applications to fraud detection / data integrity.