

# Benford's Law: From Logarithms to Dynamical Systems

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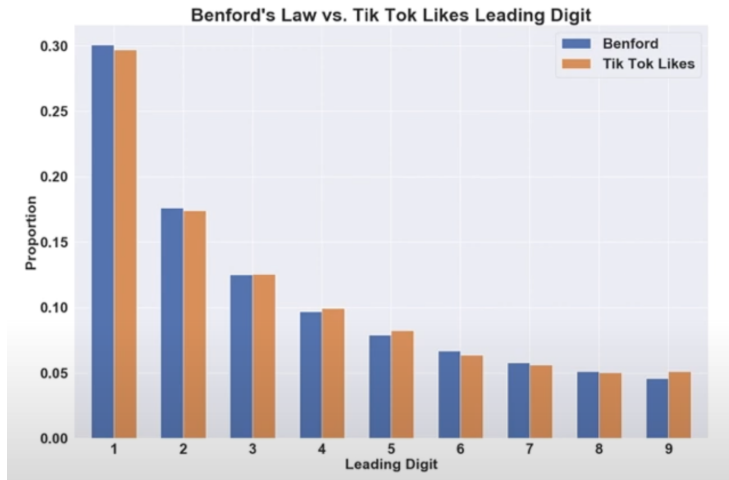
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- Given a collection of positive numbers from a set of random data
- We'd expect that the **leading digits** are distributed evenly...
- **Benford's Law:**
  - Leading digit 1 ~30% of the time
  - Leading digit 9 only ~4.6% of the time
- This talk will connect Benford's Law to logarithms, randomness, and dynamical systems

# Motivation



Data from: <https://www.youtube.com/watch?v=42fGFDNs-0A>

# Historical Origins: Early Observations

- **First Observed by Simon Newcomb (1881):**

- Earlier pages of log tables were significantly more worn out than later pages.
- He made a rule similar to Benford's Law

- **Rediscovered by Frank Benford (1938):**

- Observed the same phenomenon
- **Tested the law** extensively. Tests included data like:
  - Surface areas of 335 rivers
  - Sizes of 3,259 U.S. populations
  - 104 physical constants
  - 1,800 molecular weights
  - Street addresses from *American Men of Science*
- He popularized the law.

# What is Benford's Law?

- **Benford's Law:** In naturally occurring, or extensive numerical datasets, the probability  $P(d)$  that the **leading digit** is  $d$  ( $d \in \{1, 2, \dots, 9\}$ ) is:

$$P(d) = \log_{10} \left( 1 + \frac{1}{d} \right)$$

- **Generalization to any base  $b$ :** The probability that the first digit is  $d \in \{1, 2, \dots, b - 1\}$  is:

$$P_b(d) = \log_b \left( 1 + \frac{1}{d} \right)$$

- **Base 10 Examples:**

- $P(1) = \log_{10}(2) \approx 0.301$  (30.1%)
- $P(2) = \log_{10}(1.5) \approx 0.176$  (17.6%)
- ...
- $P(9) = \log_{10}(10/9) \approx 0.046$  (4.6%)

# What digits appear in $2^n$ ? (Israel Gelfand's Question)

- Seemingly unrelated application:
- Consider the sequence  $2^n$ : 2, 4, 8, 16, 32, 64, 128, 256, 512, 1024, ...
- What are the leading digits of each term?
- **Claim:** The distribution of these leading digits follows Benford's Law.
- To prove, we need a way to analyze the distribution of first digits

# Proof Idea 1: Logarithms Determine First Digits

## Relating First Digit to Logarithms

- Any power  $2^n = d \times 10^k$ ;  $1 \leq d < 10$ ,  $k = \lfloor \log_{10}(2^n) \rfloor$ .
  - Ex.  $2^{14} = 16384 = 1.6384 \times 10^4$
- First digit,  $D$ , is always equal to  $\lfloor d \rfloor$  (floor function)
- Take  $\log_{10}$  :

$$\log_{10}(2^n) = \log_{10}(d \times 10^k)$$

$$n \log_{10}(2) = \log_{10}(d) + k$$

- Let  $X = n \log_{10}(2)$ . Note that  $k = \lfloor X \rfloor$

$$X - k = \log_{10}(d)$$

$$X - \lfloor X \rfloor = \log_{10}(d)$$

$$\boxed{\text{frac}\{X\} = \text{frac}\{n \log_{10}(2)\} = \log_{10}(d)}$$

- Mathematical inequality for floor function:

$$\log_{10}(D) \leq \log_{10}(d) < \log_{10}(D + 1)$$

$$\log_{10}(D) \leq \text{frac}\{n \log_{10}(2)\} < \log_{10}(D + 1)$$

- **Key Insight:** Distribution of the first digit depends **entirely** on how  $\text{frac}\{n \log_{10}(2)\}$  is distributed within  $[0, 1)$



# Proof Idea 2: The Circle Rotation Analogy

## Connecting to Dynamical Systems: Irrational Circle Rotation

- Consider the sequence of fractional parts:  $x_n = \text{frac}\{na\}$ ;  $a = \log_{10}(2)$ .
- This sequence is generated by the **dynamical system** of rotation on a circle:
  - *Space*: The circle  $\mathbb{T} = \mathbb{R}/\mathbb{Z}$  (or interval  $[0, 1)$ )
  - *Transformation*:  $T(x) = \{x + a\}$  (Rotation by  $a$ )

# Proof Idea 3: Input from Dynamical Systems Theory

## Applying Weyl's Theorem

- **Key Fact:**  $\alpha = \log_{10}(2)$  is *irrational*.
- **Weyl's Equidistribution Theorem:** For irrational  $\alpha$ , the orbit  $\{n\alpha\}$  is **uniformly distributed** in  $[0, 1)$ .
- **Conclusion:** Proportion of  $n$  where  $x_n = \text{frac}\{n\alpha\}$  falls into  $[\log_{10}(D), \log_{10}(D + 1))$  is just length.

$$P(D) = \text{Length} = \log_{10}(D + 1) - \log_{10}(D) = \log_{10}\left(1 + \frac{1}{D}\right)$$

- This is exactly Benford's Law!

- Key property arising from Benford's Law: **scale invariance**.
- If dataset  $\{x_i\}$  follows Benford's law, then scaled dataset  $\{c \cdot x_i\}$  (for  $c > 0$ ) should also follow.
- **Analogy:** Imagine converting units of a data set, like feet to meters, dollars to euros, miles to kilometers. The distribution of first digits shouldn't change. (And it doesn't!)

## Mathematical expression:

- If  $X$  is a r.v., Benford's Law holds if  $\text{frac}\{\log_{10} X\}$  is uniform on  $[0, 1)$
- Scale by  $c$ :  $\log_{10}(cX) = \log_{10} c + \log_{10} X$
- Take fractions:  $\text{frac}\{\log_{10}(cX)\} = \text{frac}\{\log_{10} c\} + \text{frac}\{\log_{10} X\}$ .
- **What this means:** Adding a constant  $\log_{10} c$  and taking the fractional part corresponds to *rotation* on a circle.
- If original distribution  $\text{frac}\{\log_{10} X\}$  was uniform, the rotated distribution remains uniform.

- Benford's Law appears naturally in many systems with growth, multiplicative, or chaotic behavior
  - $3^n$
  - Fibonacci sequence (ratio of terms approaches  $\phi$ )
  - Logistic growth for population modeling
- Textbook: Nillsen discusses equivalence/connection of a *Kronecker system* (irrational rotations on a circle) and a *Benford system* (related to multiplication by 10 modulo 1) (Section 3.18)

# Randomness: Benford vs. Normal Numbers

- **Normal Numbers (Borel):** Numbers that exhibit maximum digit randomness.
  - Normal Number: All digits appear with equal asymptotic frequency ( $1/b$ ) in a *single* number's base- $b$  expansion.
  - *Borel's Theorem (1909)*: Almost all real numbers (Lebesgue measure) are normal to every base  $b \geq 2$
- **Distinction:**
  - Normality  $\rightarrow$  Distribution of **all** digits in a **single** number.
  - Benford  $\rightarrow$  Distribution of the **first** digit across a **set** of numbers.
- **Analogy: “Logarithmic” Normality**
  - Benford's Law: the fractional part of numbers' logarithms are uniformly distributed on  $[0, 1)$
  - *Uniformity of logarithms* parallels the *uniformity of digits* in Borel's NNT

# Applications of Benford's Law

- **Fraud Detection:** One of the major applications
  - Financial statements, tax returns, credit card transactions, election results
  - Human-fabricated data often doesn't follow Benford's Law
- **Validating Scientific Data:** Checking if experimental results or physical constants conform to Benford's Law
- **Natural Phenomena:** Population sizes, river lengths, stock market data, earthquake magnitudes
  - **Important:** Benford's Law applies more strongly to natural data sets when the data **spans several orders of magnitude**

# Takeaways and Conclusion

- Benford's Law is an interesting, counter-intuitive statistical law that seems to pop up all the time. Governs what leading digit looks like in a variety of datasets
- Fundamentally connected to properties of logarithms and scale invariance
- Deeply connected to dynamical systems (Weyl's Theorem - irrational increment "rotations" are uniformly spread on a circle)
- Broad connections to randomness, recurrence, and chaotic behavior in mathematical systems



# References



[Rodney V. Nillsen \(2010\)](#)

Randomness and Recurrence in Dynamical Systems

*Section 3.14-3.19*



[Minding the Data \(2020\)](#)

Benford's Law in Real Life | Finding Real World Examples

[YouTube Video](#)



[Wikipedia](#)

Benford's law

[Wikipedia entry](#)

# Questions?