

Data Forensics via:

Digital Development Pattern

Benford's Law is a statement about the digital configuration of the **entire data set**.

Could we then naively conclude that any **sub-set of the data** on any segment of the x-axis is also Benford?

Does the ***whole*** endow its Benford property to its ***parts***?

NO!

Local digit configuration changes and develops as we move from the left to the right on the x-axis!

By some amazing coincidence...

Digital
Behavior = Political
Narratives

LEFT

Material equality



CENTER

Some inequality



FAR RIGHT

Extreme inequality

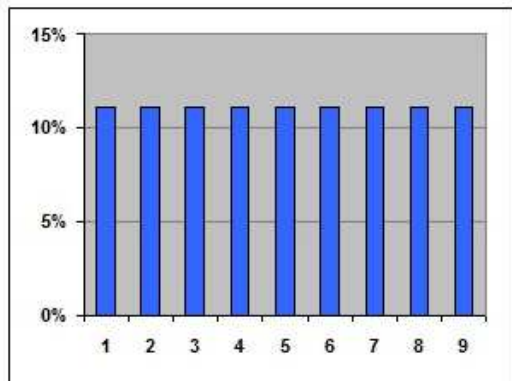


LEFT

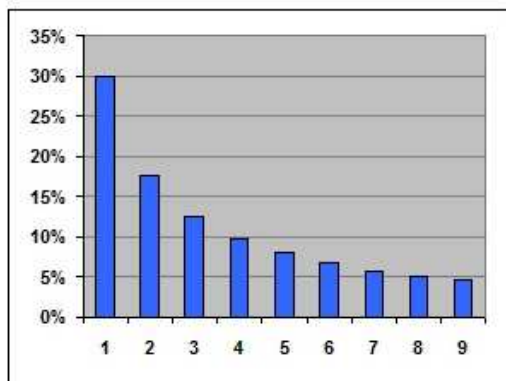
CENTER

FAR RIGHT

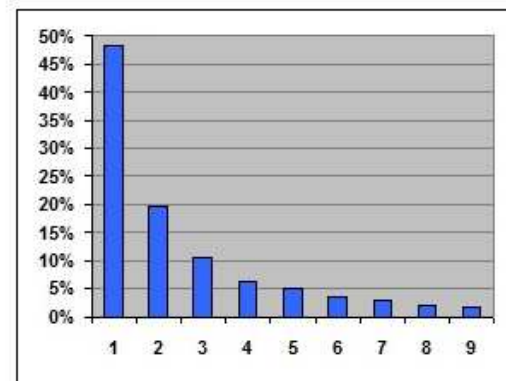
Digital Equality



Benford



Extreme digital inequality

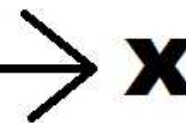


0

small values

middle values

big values



This can be seen on sub-intervals between

**... 0.001, 0.01, 0.1, 1, 10, 100,
1000, 10000, 100000, ...**

All **honest data should come with
Digital Development Pattern!**



Let us check this **fact of life** empirically:



Canford Audio PLC

(Honest Data)

The entire catalog of **Canford Audio PLC** in the **U.K.** which manufactures and retails **15,194** electronic items.

<http://www.canford.co.uk/>

Canford Audio PLC - UK



Catalog



Price: \$20.79



Price: \$15.64



Price: \$31.99

Price: \$6.38

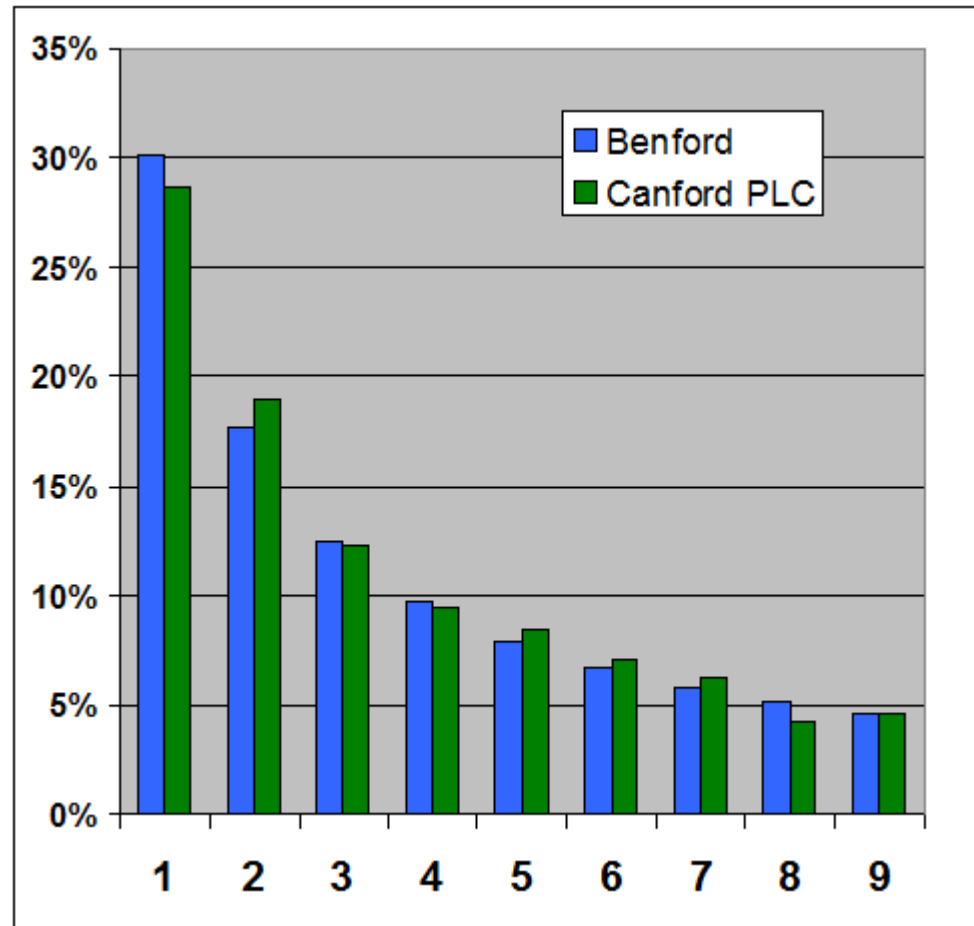


Price: \$25.50

Canford Audio PLC

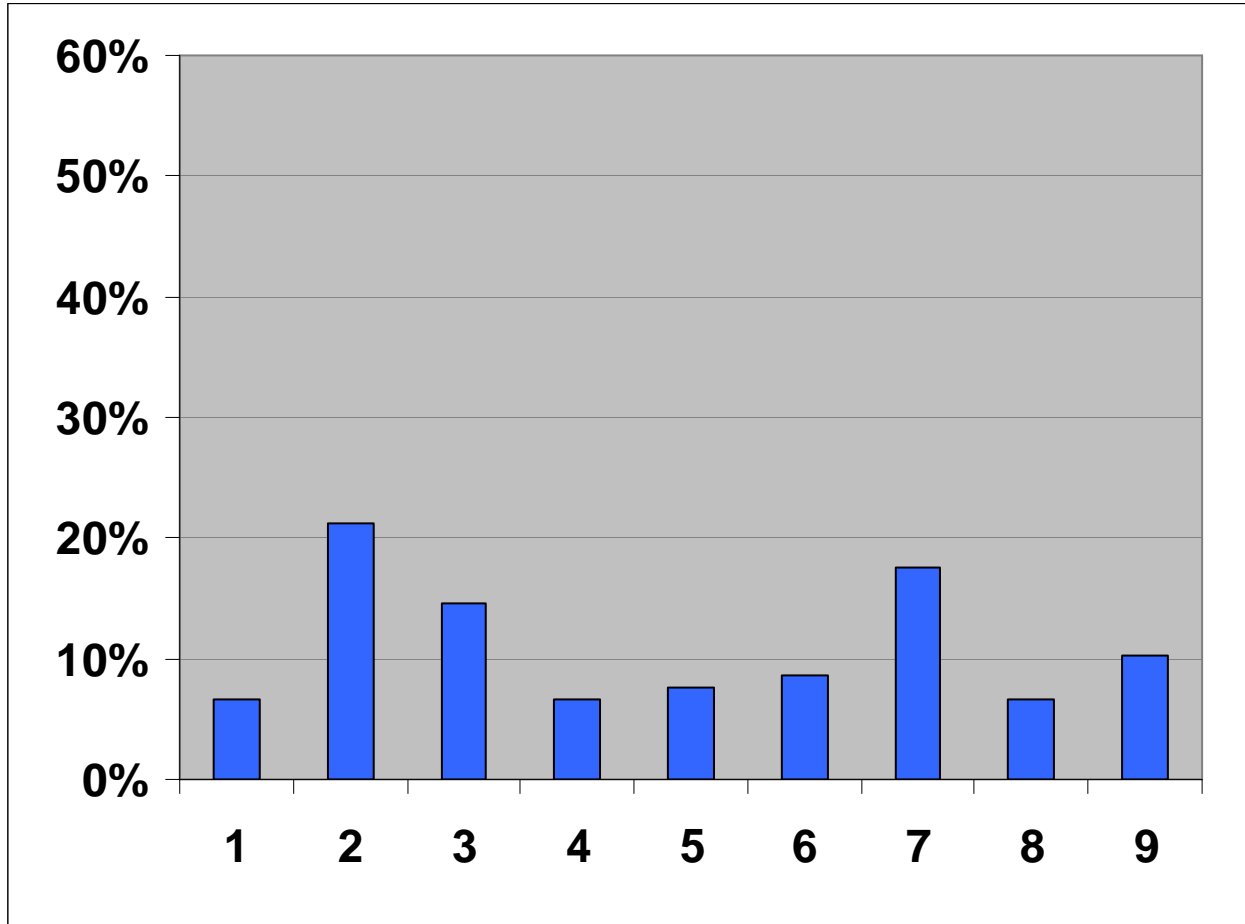
<i>Benford</i>	<i>Canford</i>
30.1%	28.6%
17.6%	19.0%
12.5%	12.3%
9.7%	9.4%
7.9%	8.5%
6.7%	7.1%
5.8%	6.3%
5.1%	4.2%
4.6%	4.6%

Canford Audio PLC

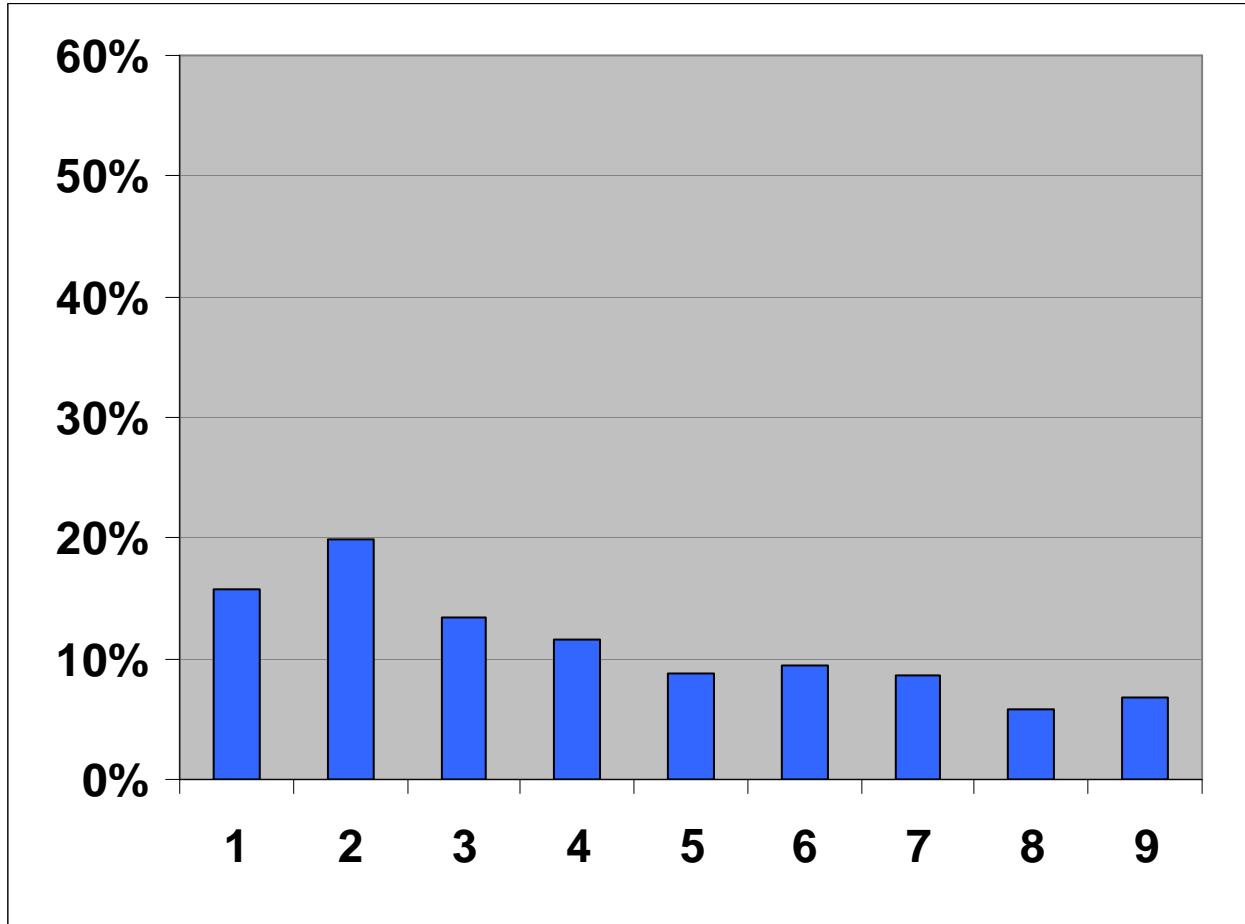


Yet... on local sub-intervals
it develops from left to right:

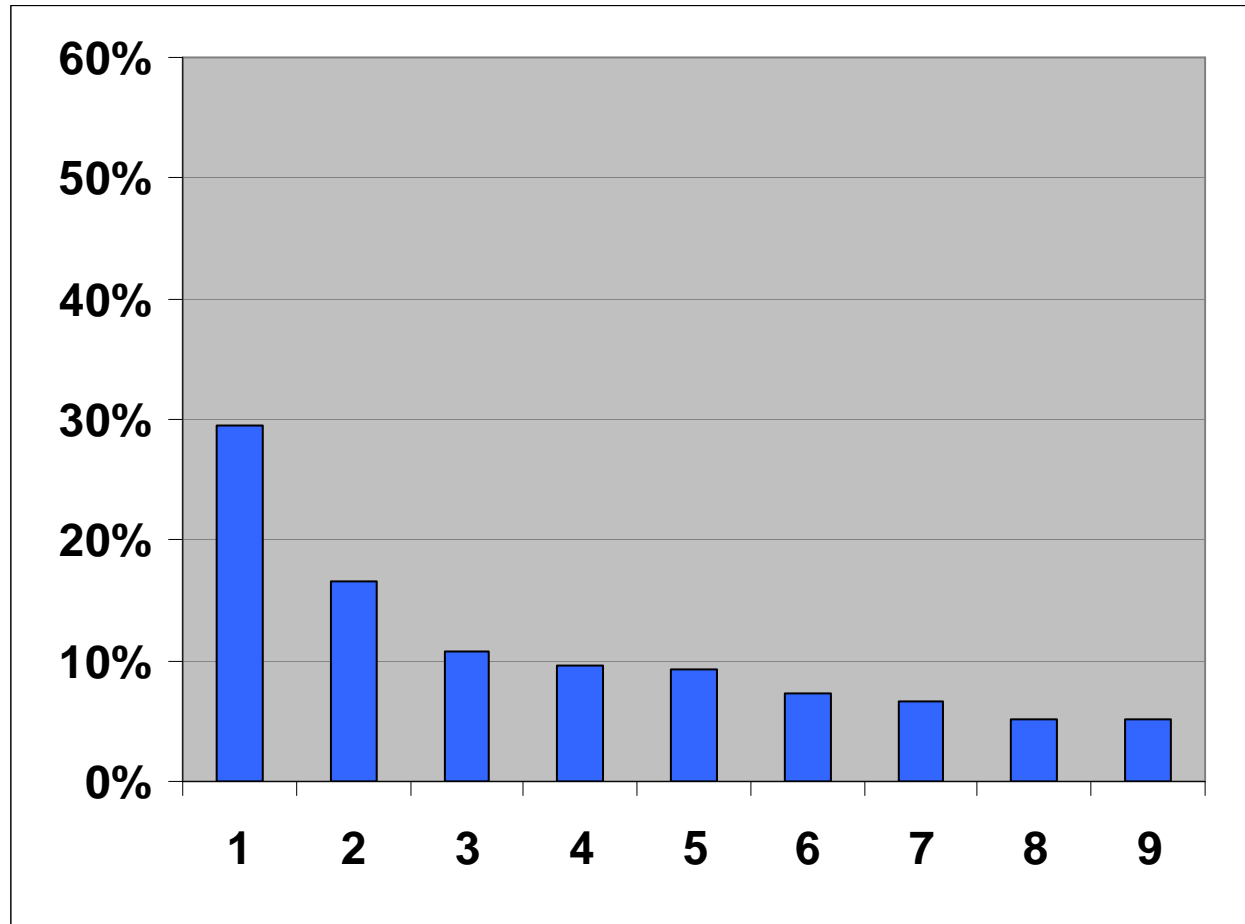
0.1 – 1



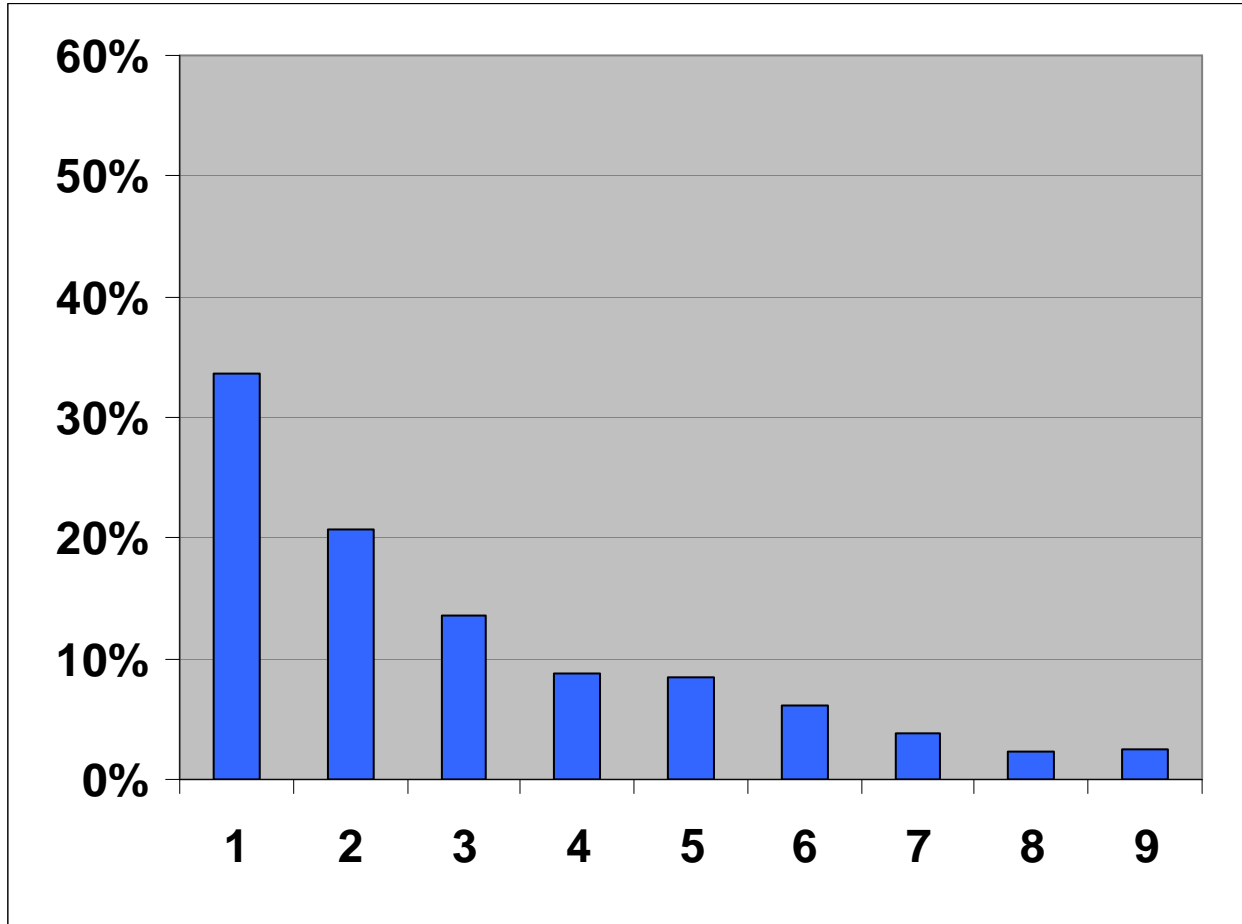
1 – 10



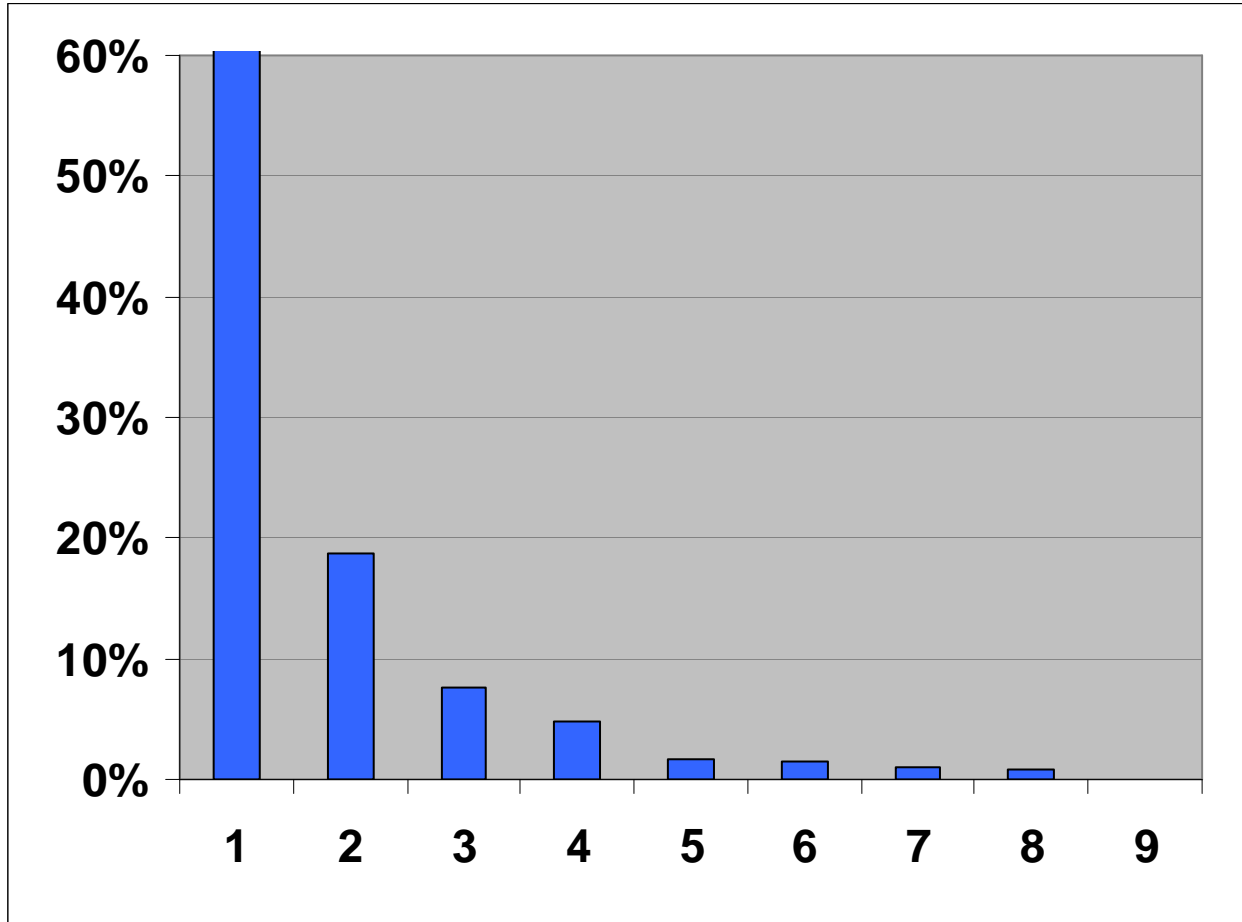
10 – 100



100 – 1000



1000 – 10000





NASDAQ USA Market Capitalization

(Honest Data)

Oct 9, 2016

2,889 companies

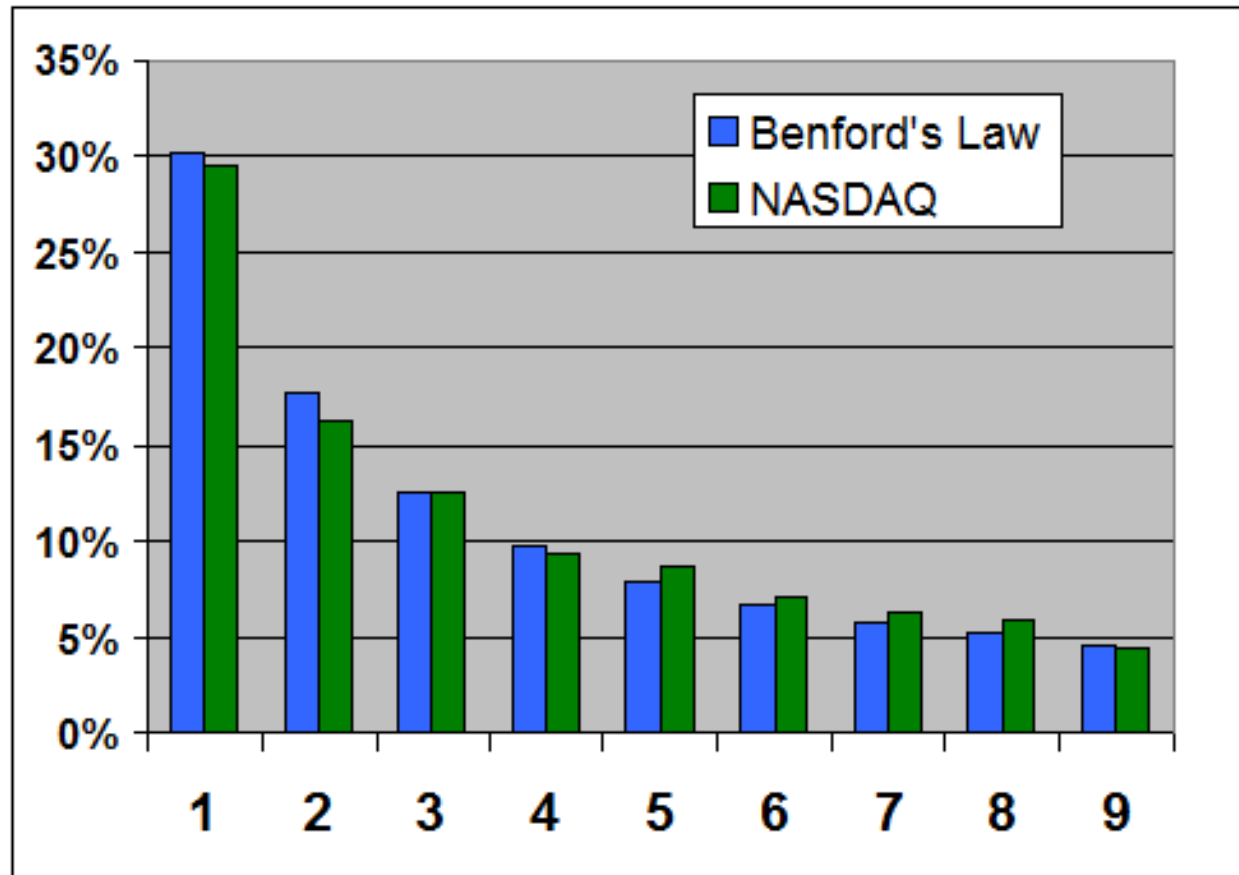
<http://www.nasdaq.com/screening/companies-by-industry.aspx?exchange=NASDAQ>



NASDAQ Market Capitalization

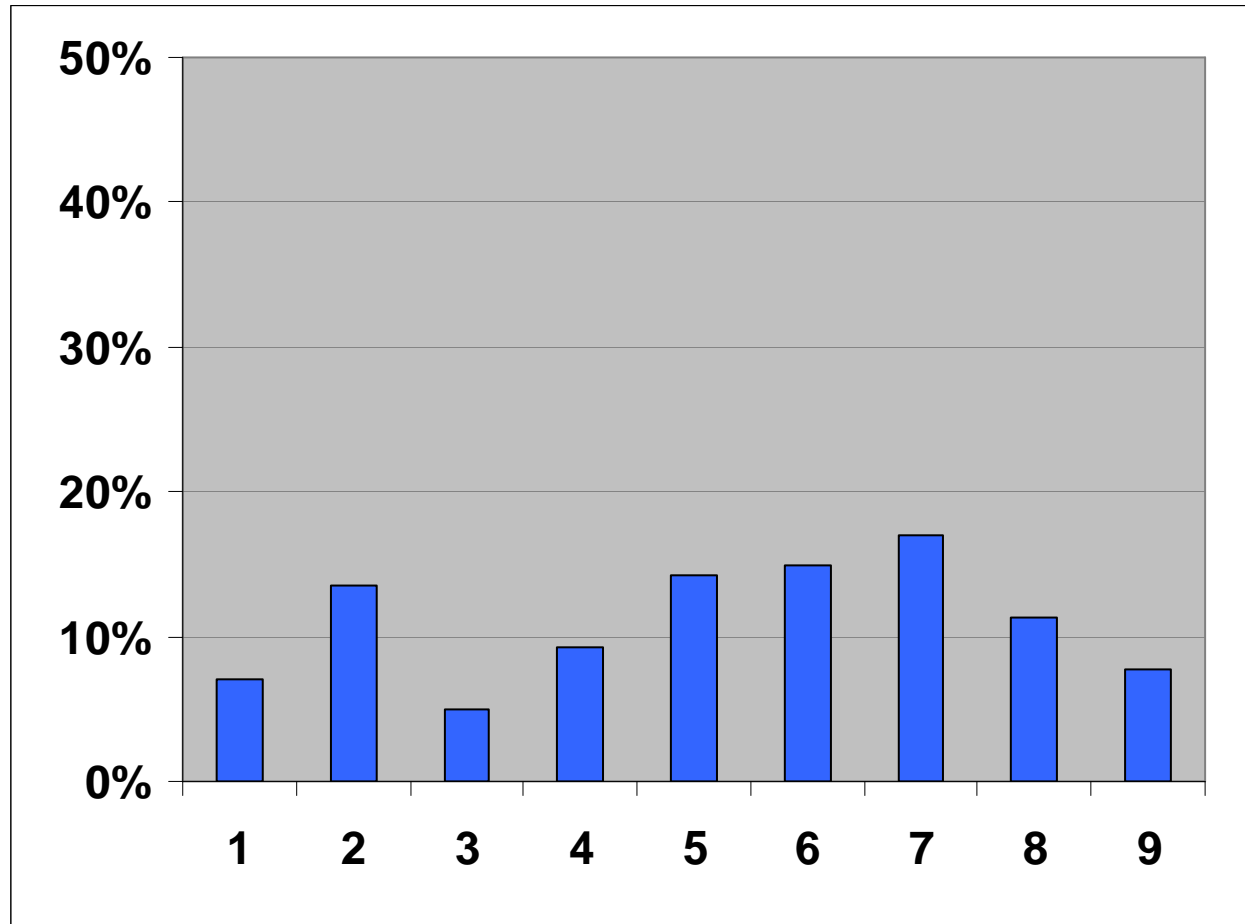
<i>Benford</i>	<i>NASDAQ</i>
30.1%	29.6%
17.6%	16.3%
12.5%	12.5%
9.7%	9.3%
7.9%	8.6%
6.7%	7.2%
5.8%	6.3%
5.1%	5.9%
4.6%	4.4%

NASDAQ Market Capitalization

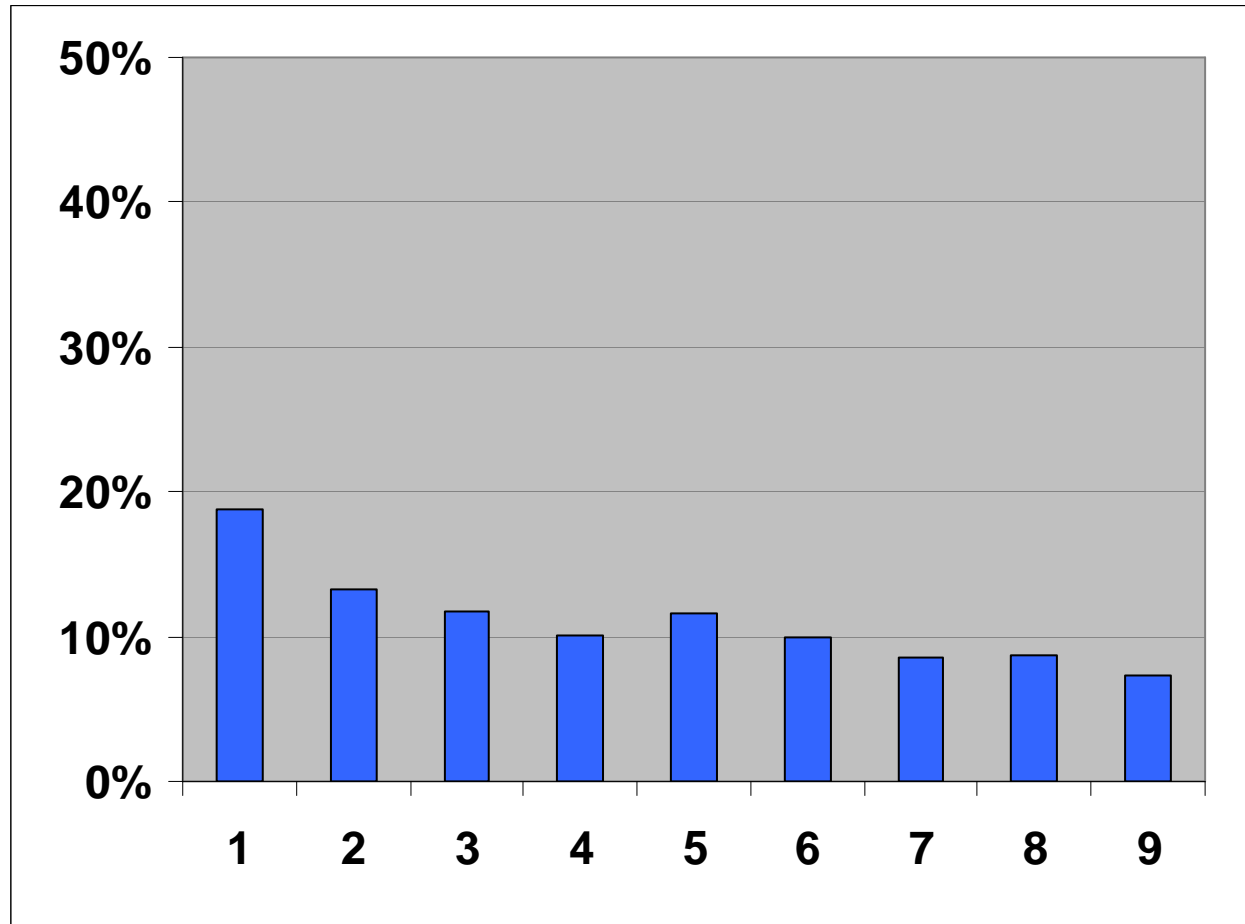


Yet... on local sub-intervals
it develops from left to right:

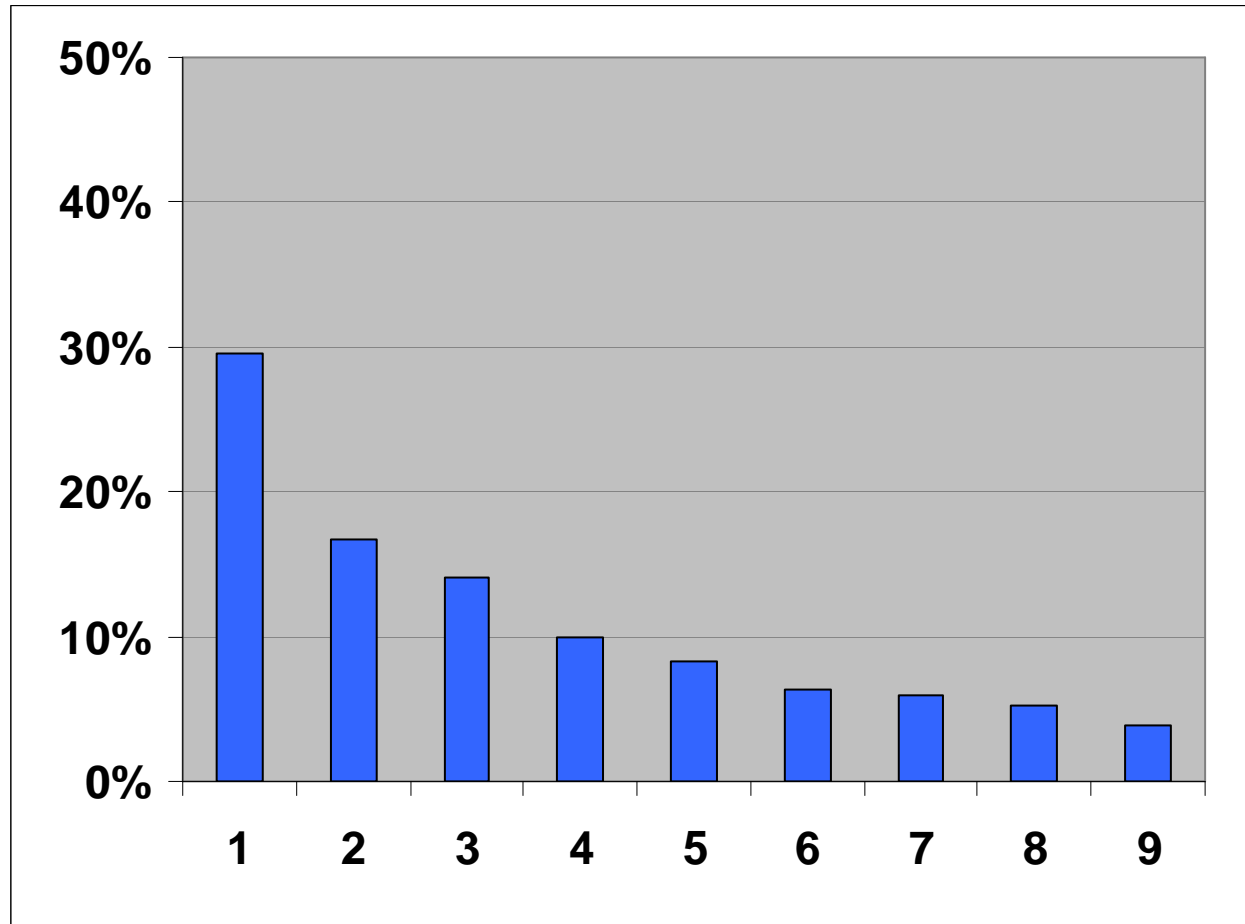
\$1 Million – \$10 Million



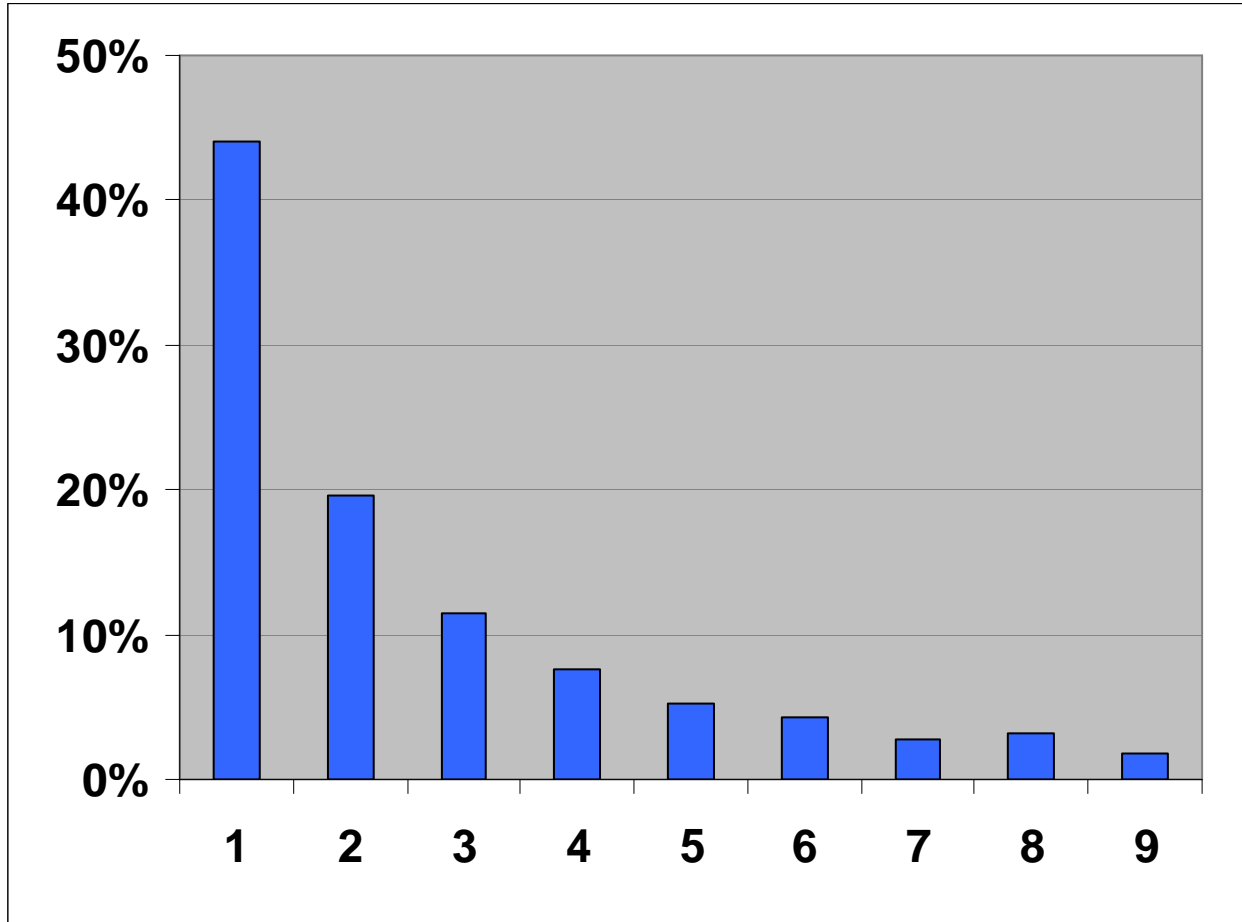
\$10 Million – \$100 Million



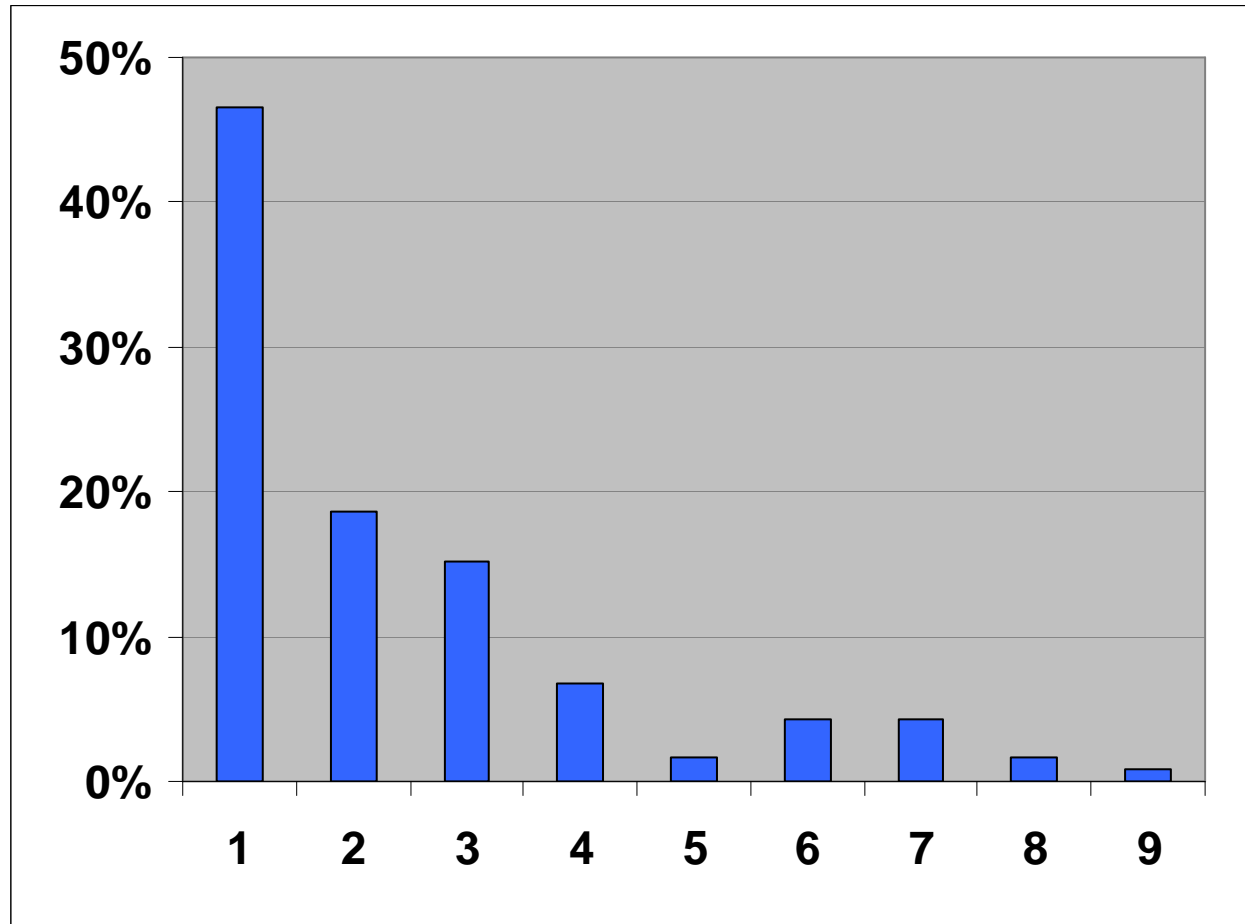
\$100 Million – \$1 Billion



\$1 Billion– \$10 Billion



\$10 Billion – \$100 Billion





Oklahoma State Expenses

(Honest Data)

The State Of Oklahoma in the USA
966,990 payments in 2011.

<https://data.ok.gov/dataset/state-oklahoma-vendor-payments-fiscal-year-2011>

Oklahoma

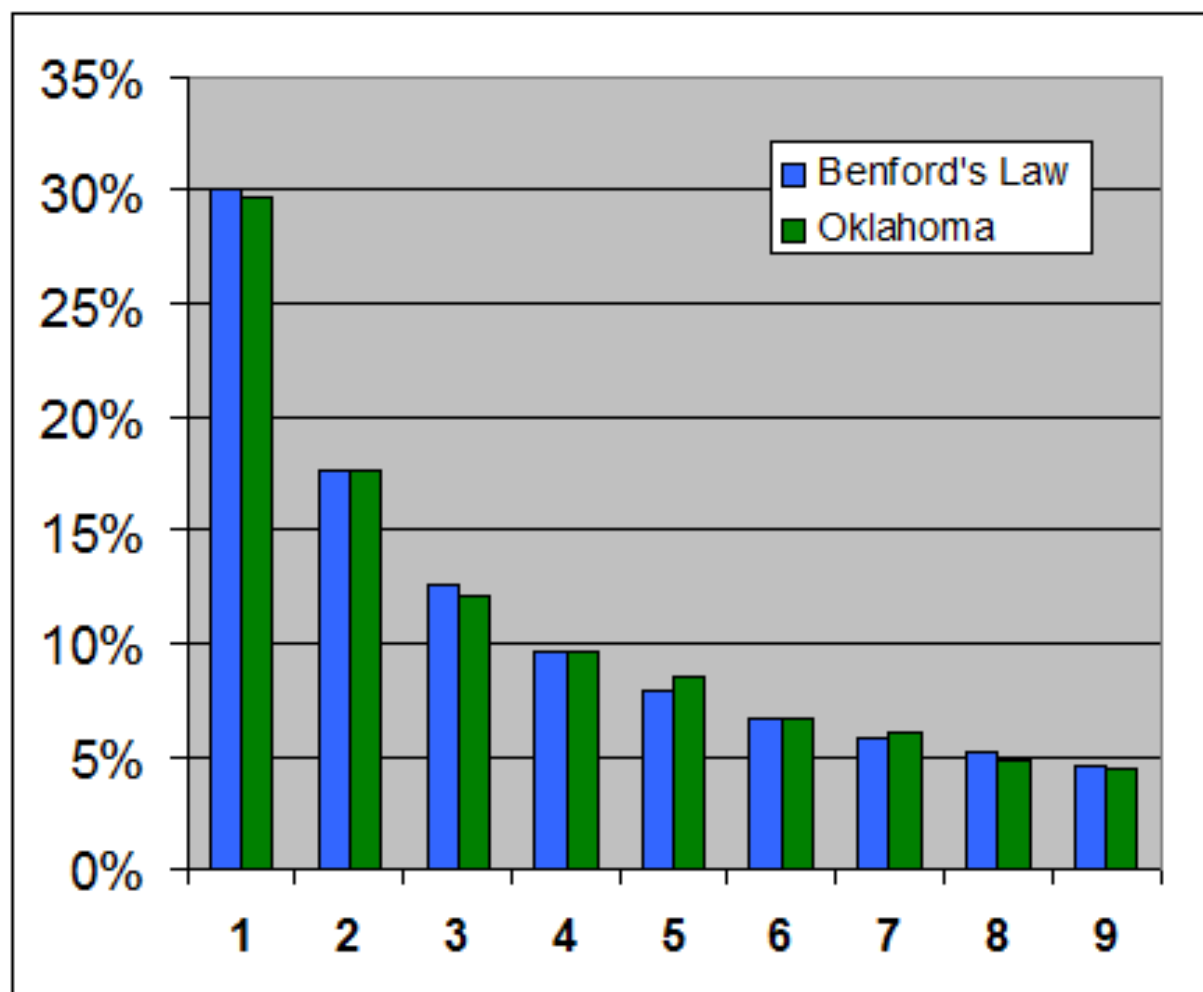


Oklahoma overall 1st digit configuration:

Superb compliance with Benford!

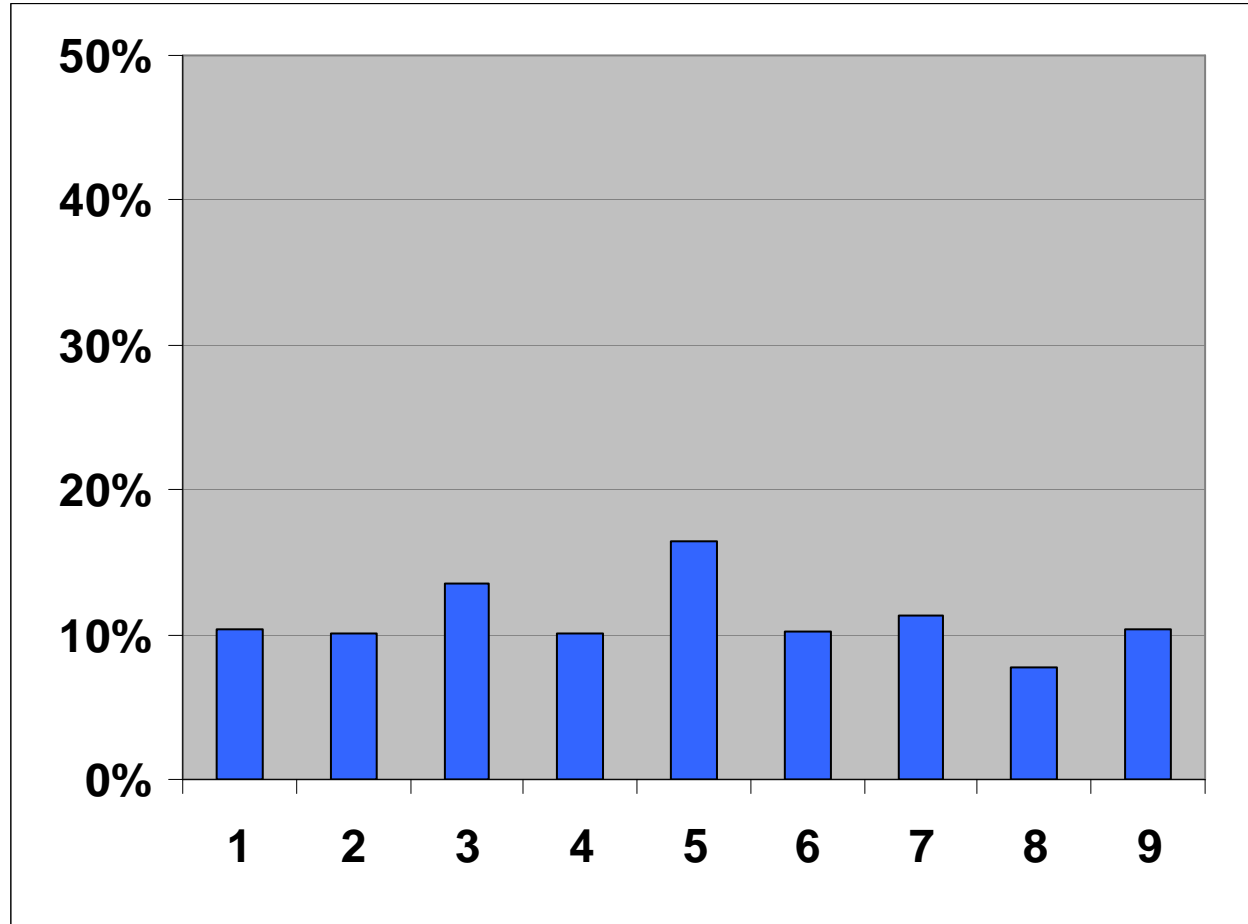
Benford	Oklahoma
30.1%	29.7%
17.6%	17.7%
12.5%	12.1%
9.7%	9.7%
7.9%	8.5%
6.7%	6.6%
5.8%	6.1%
5.1%	4.9%
4.6%	4.5%

Oklahoma overall 1st digit configuration:

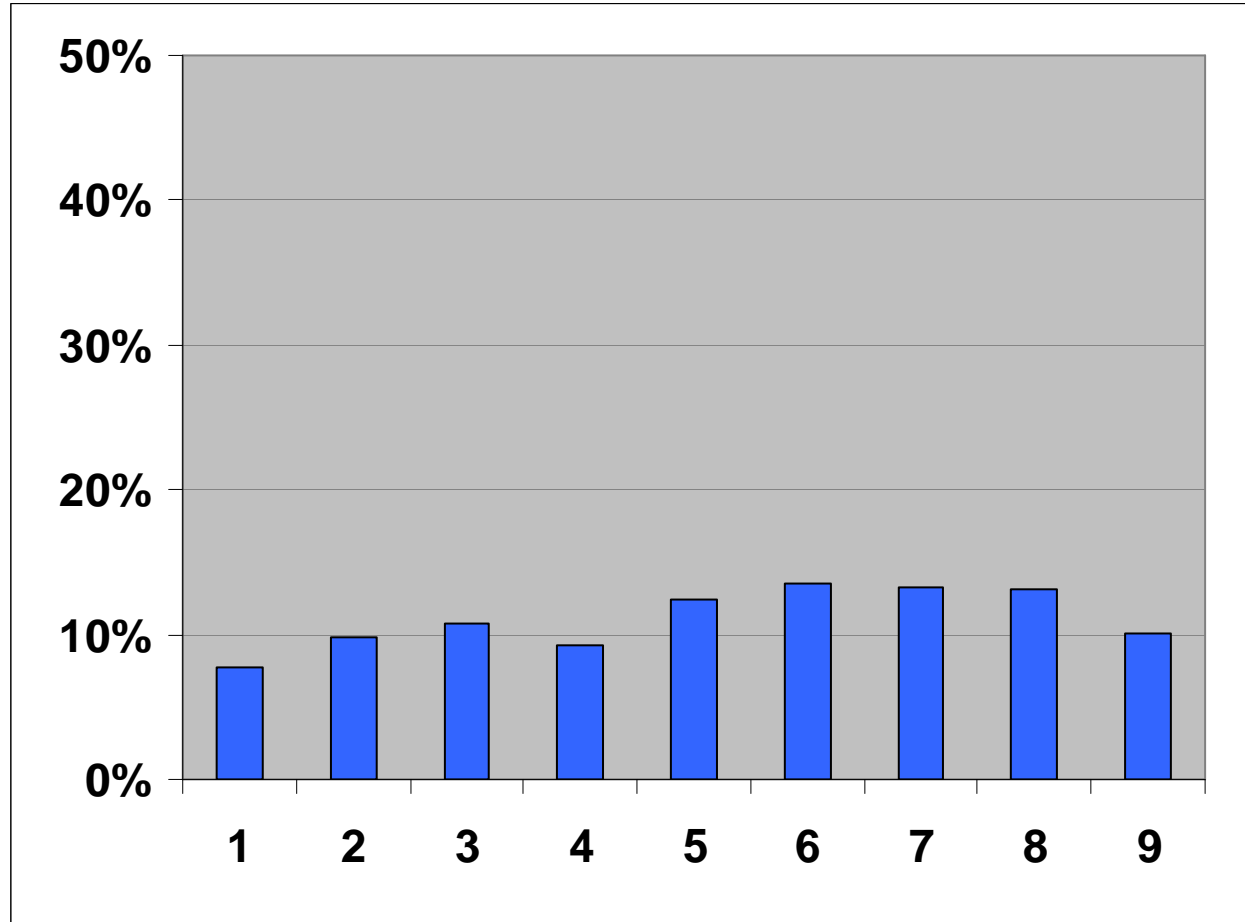


Yet... on local sub-intervals
it develops from left to right:

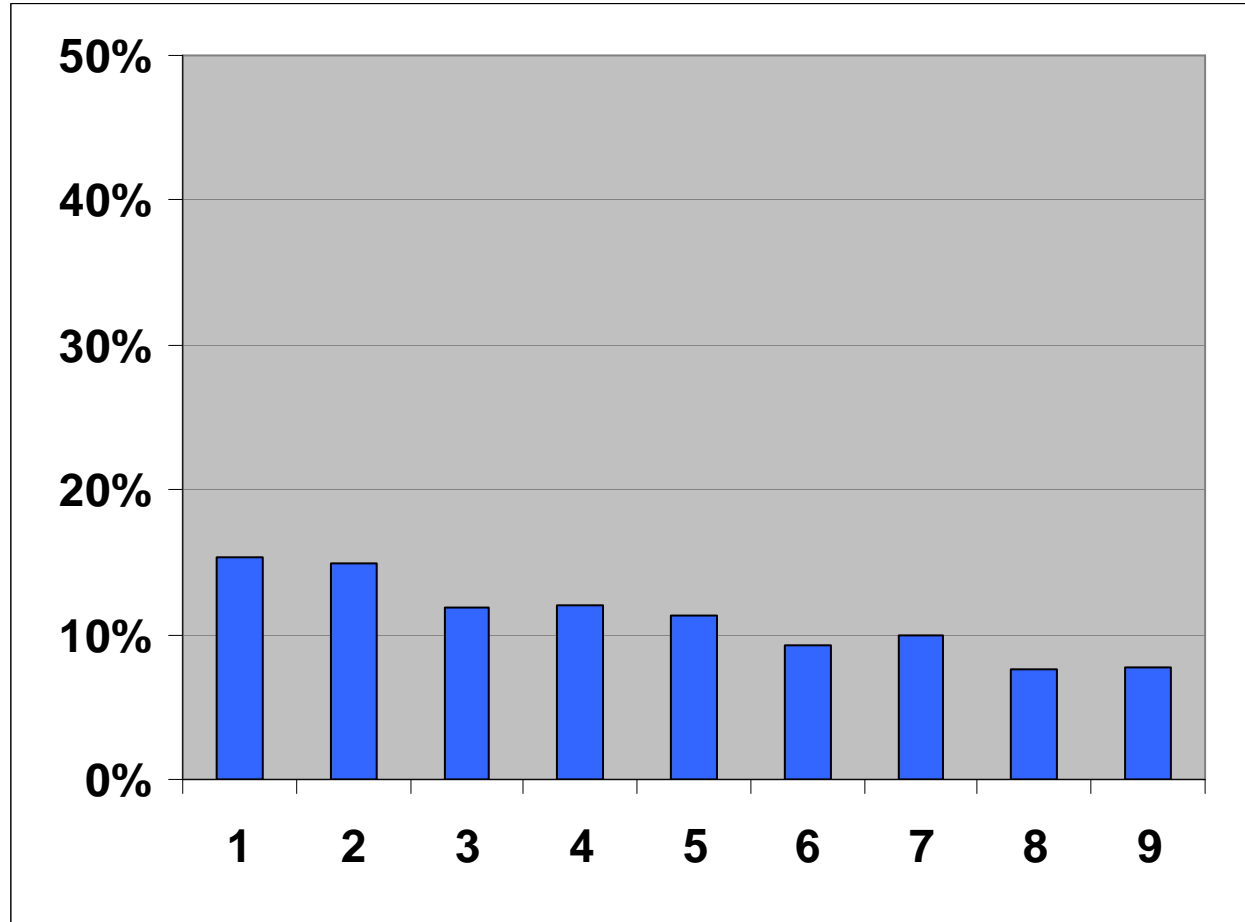
\$0.1 – \$1



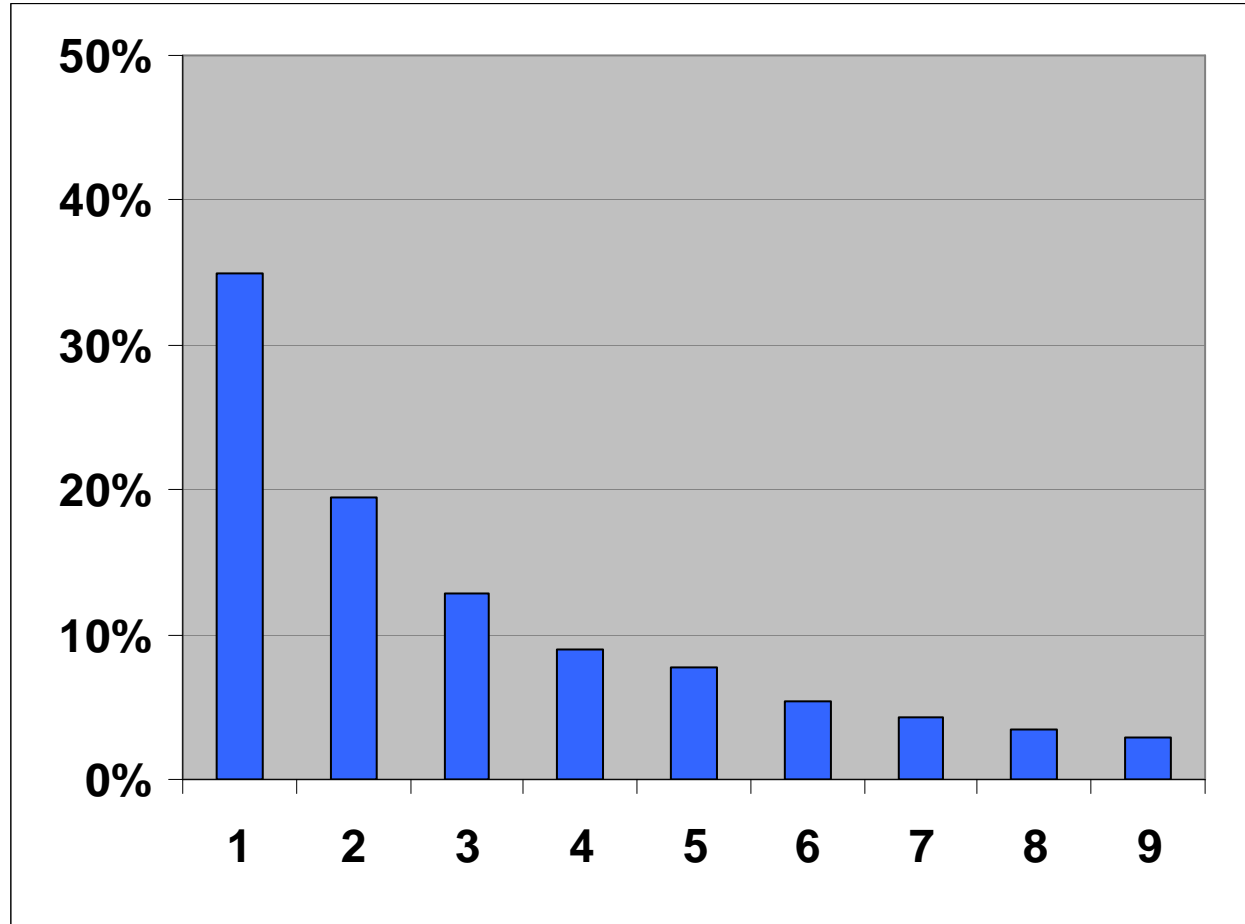
\$1 - \$10



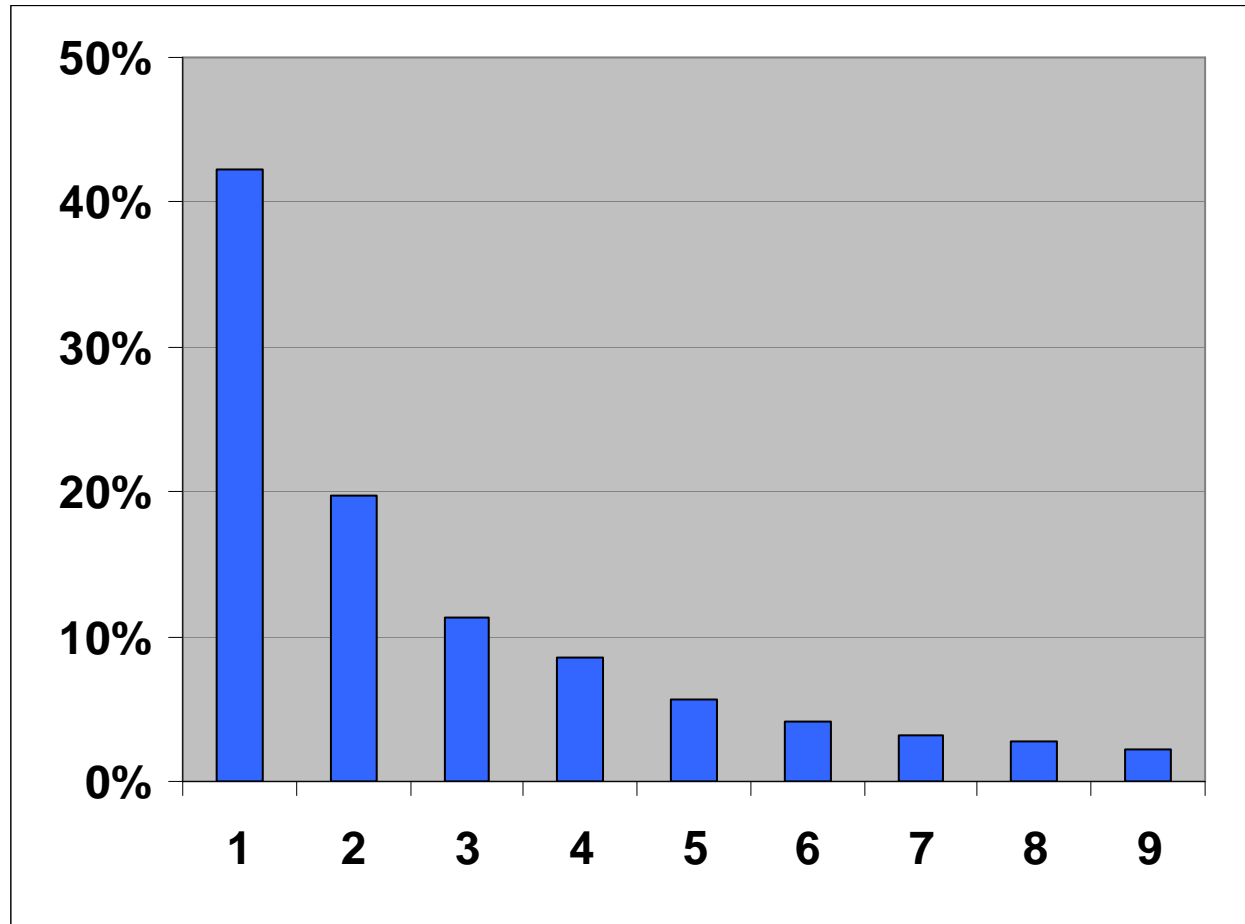
\$10 – \$100



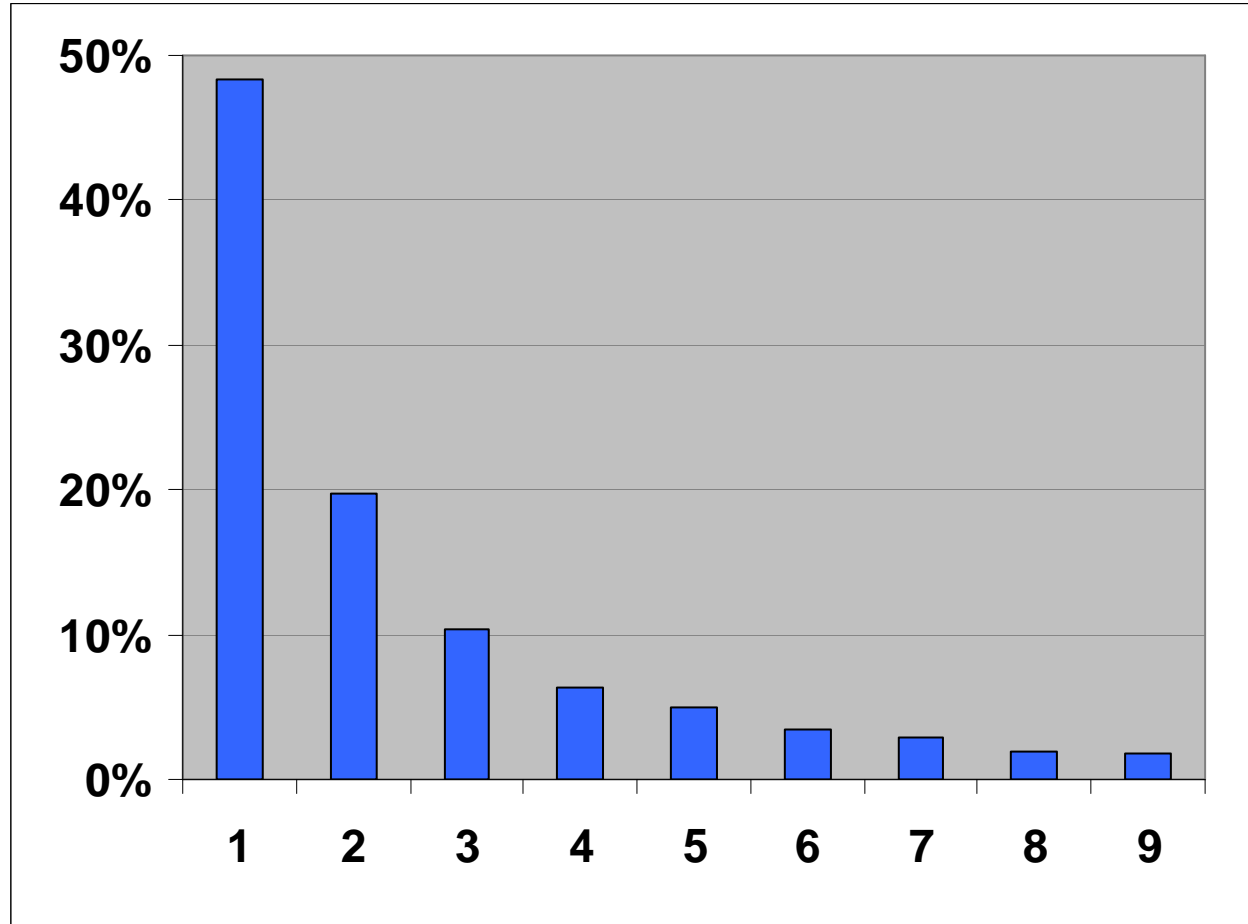
\$100 – \$1000



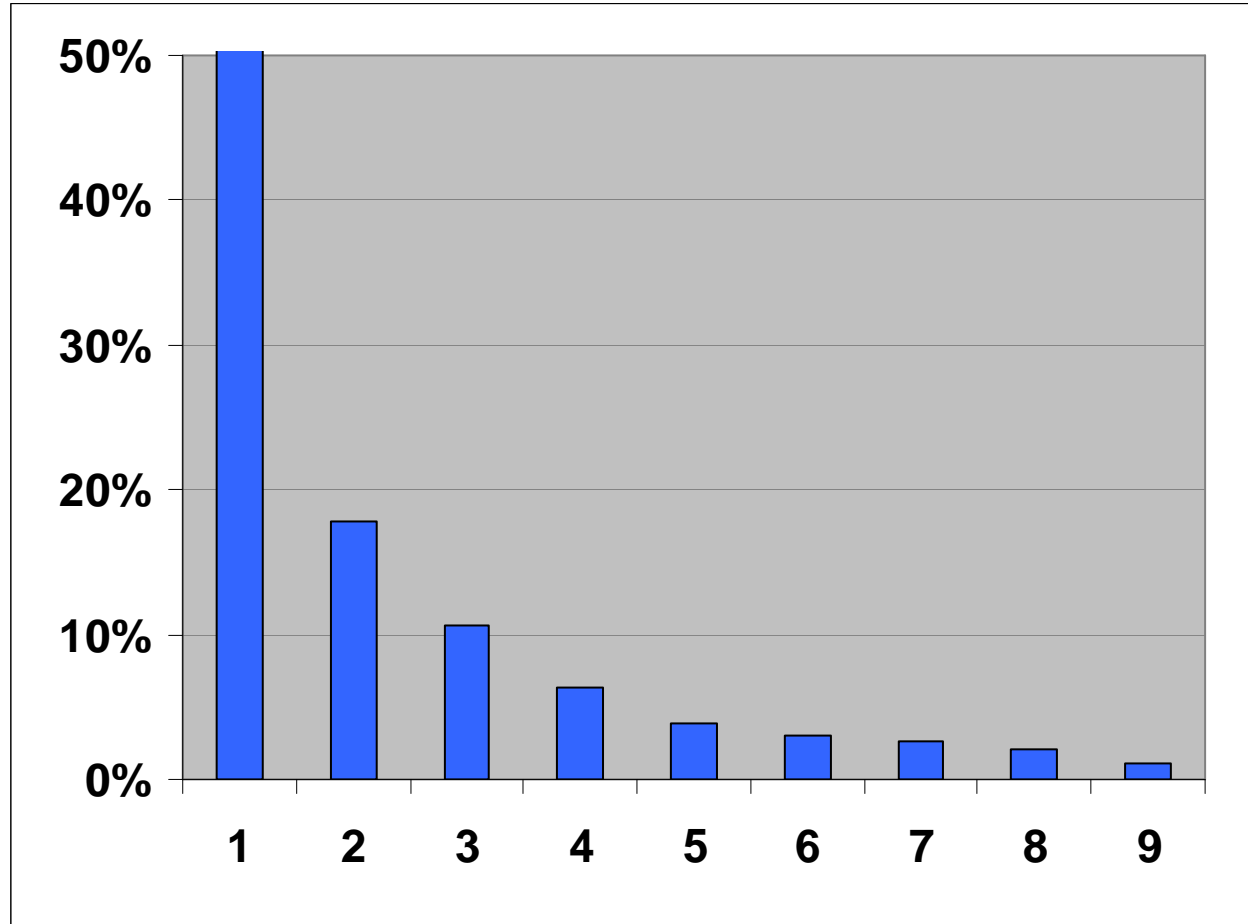
\$1000 – \$10,000



\$10,000 – \$100,000



\$100,000 – \$1,000,000

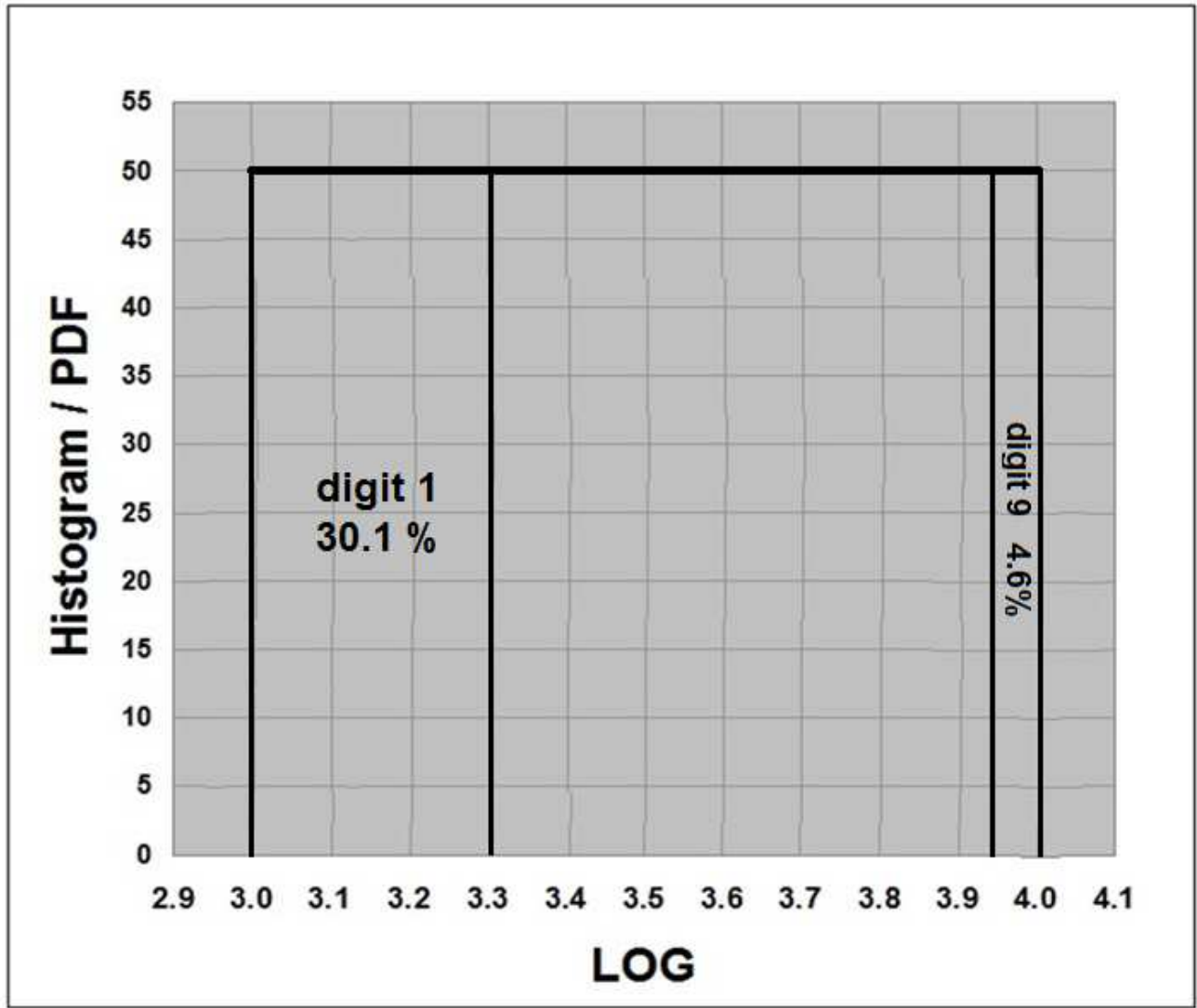




The **mathematical** Basis of
Digital Development Pattern
using **mantissa** consideration:

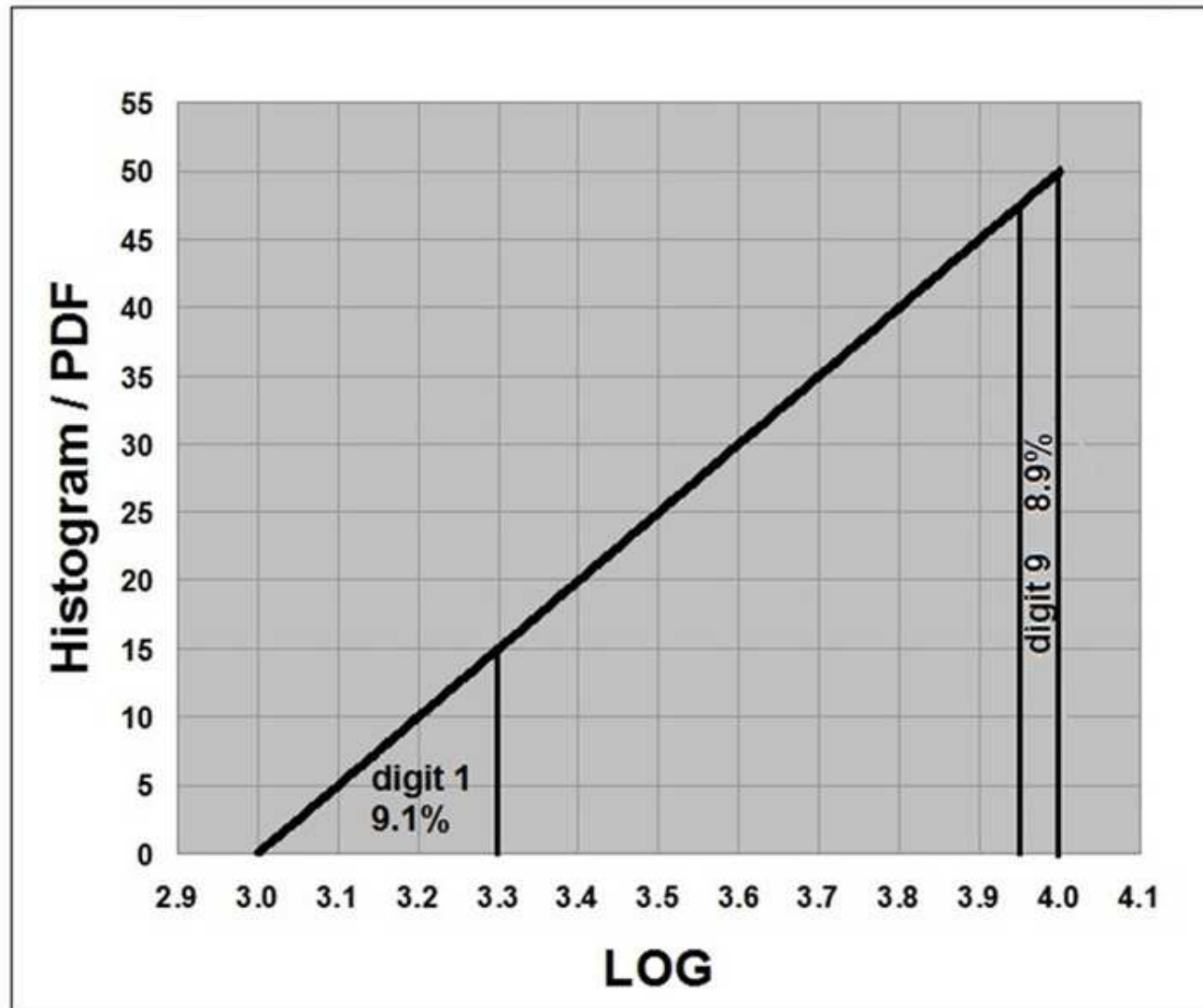
Uniform LOG Yields **Benford**

30.1	17.6	12.5	9.7	7.9	6.7	5.8	5.1	4.6
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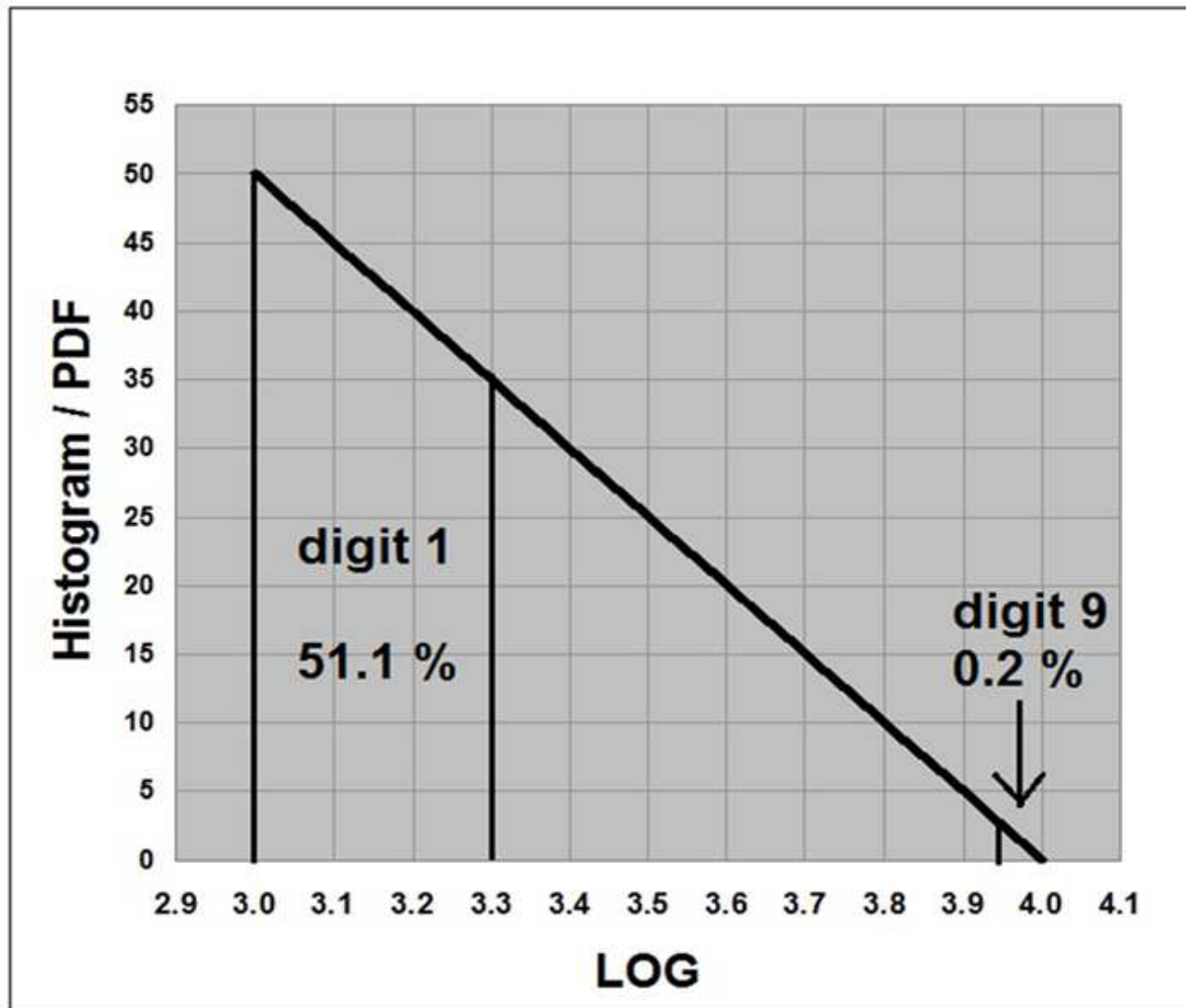
Rising LOG Yields **Near Digital Equality**

9.1 13.7 13.5 12.6 11.7 10.9 10.1 9.5 8.9

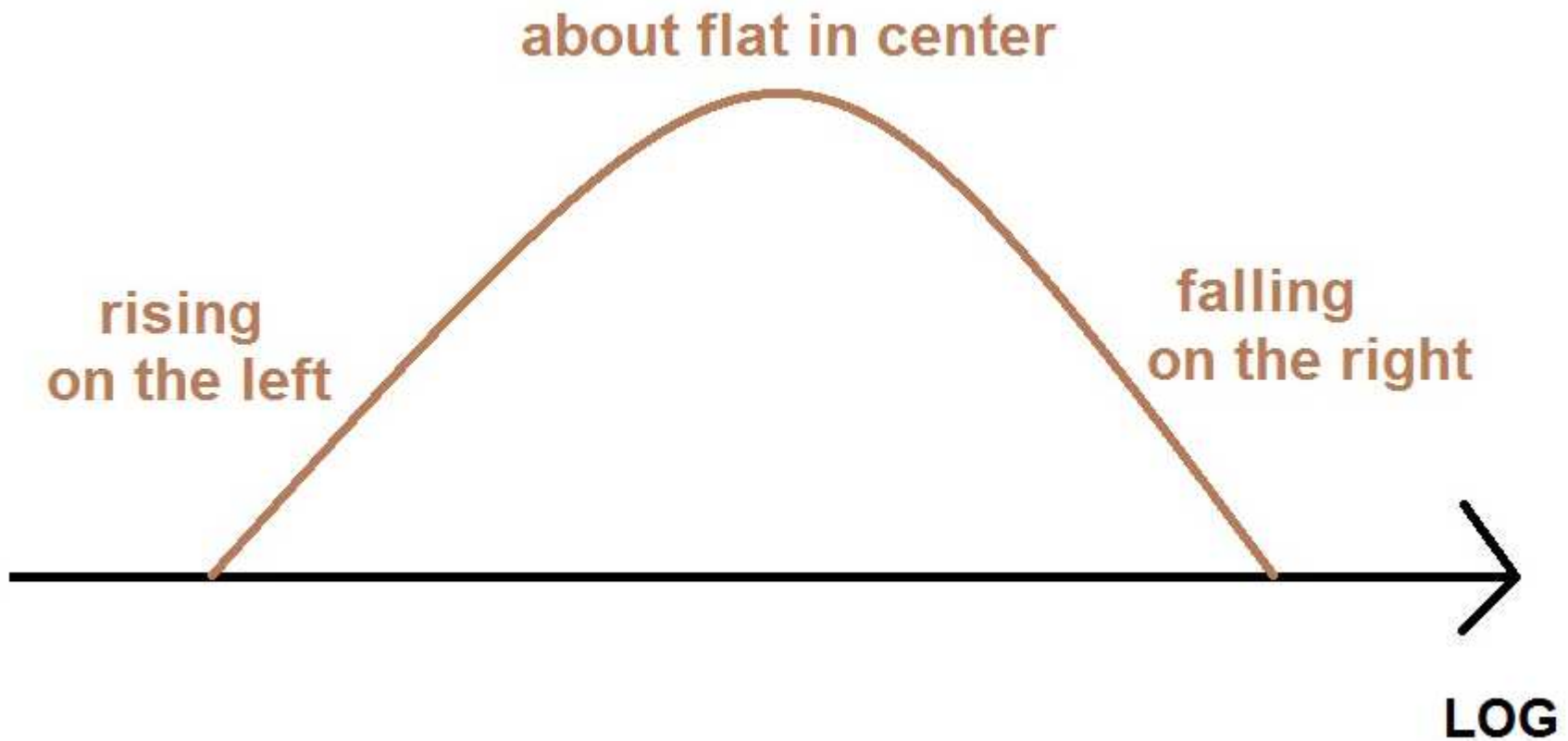


Falling Log Yields **Extreme Digital Skewness**

51.1 21.5 11.5 6.8 4.1 2.5 1.5 0.7 0.2

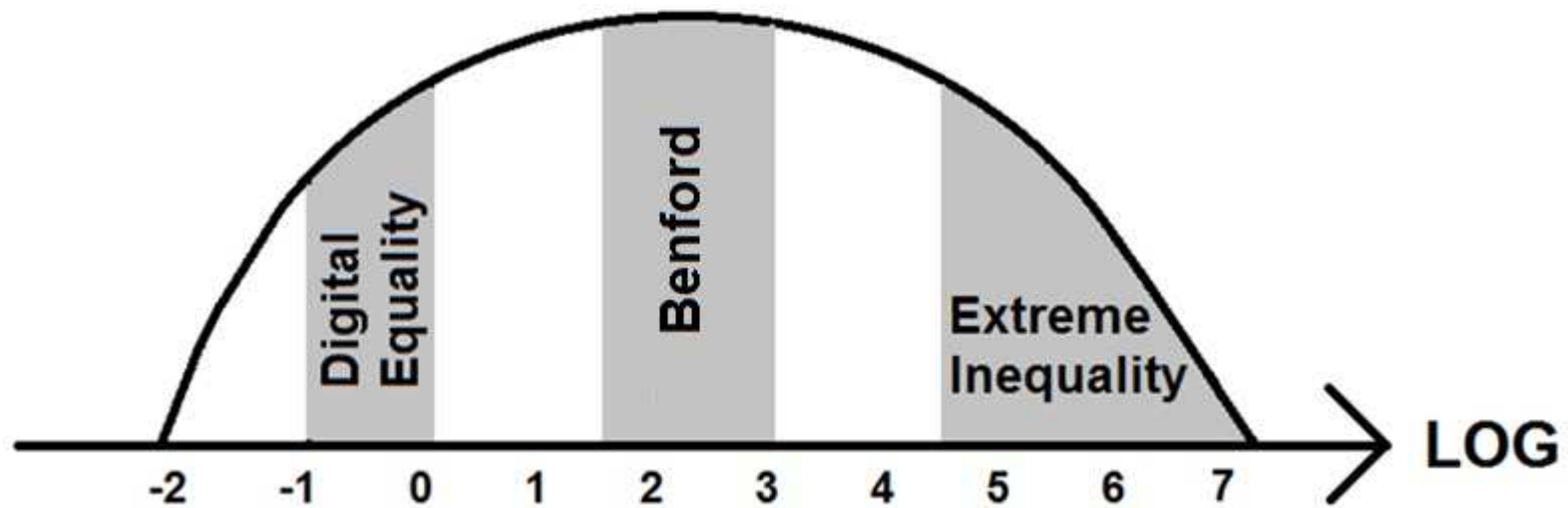


Typical LOG Histograms



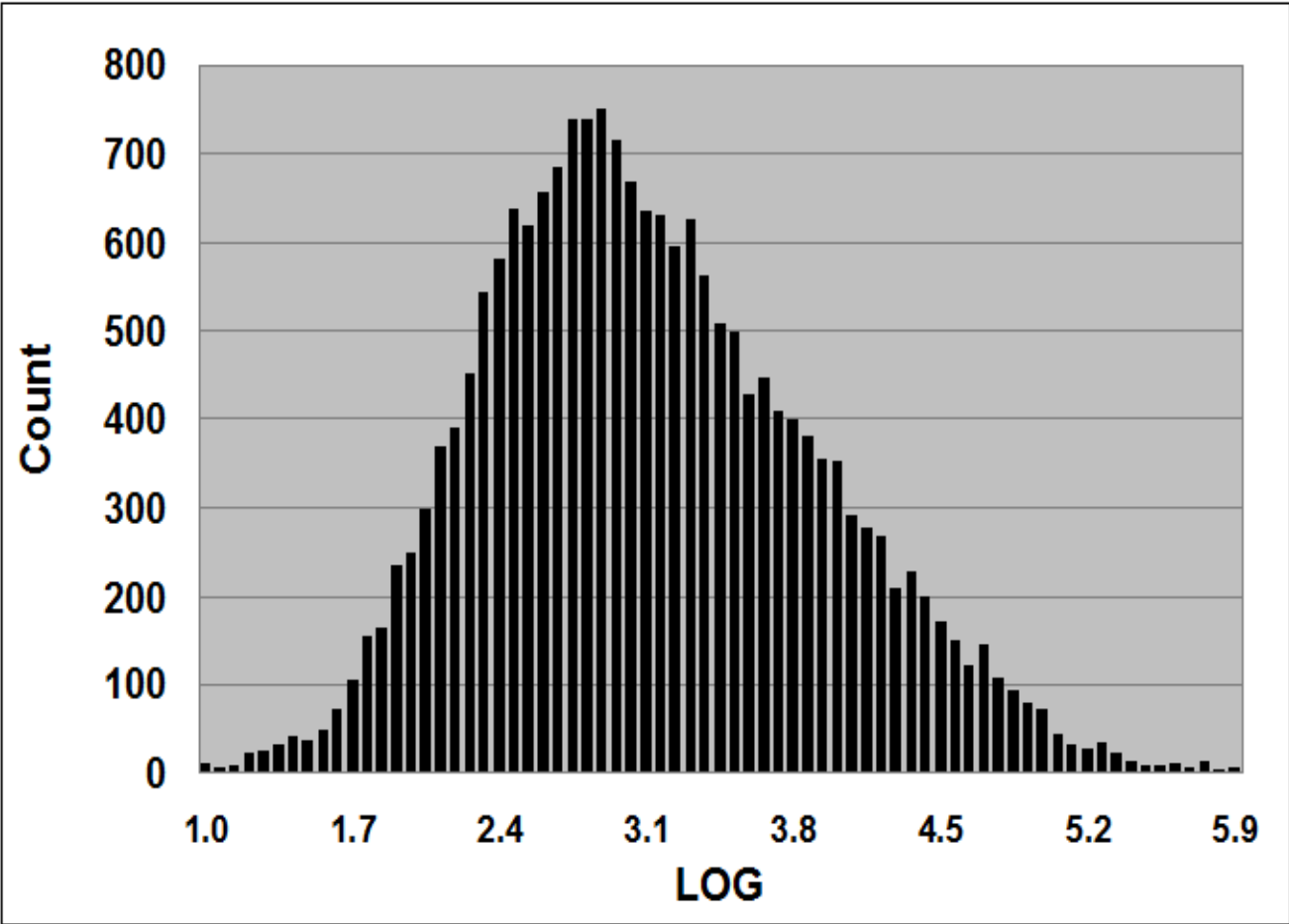
Such differentiated LOG histogram
leads to differentiated digital structure.

Log of data and its digital status throughout:

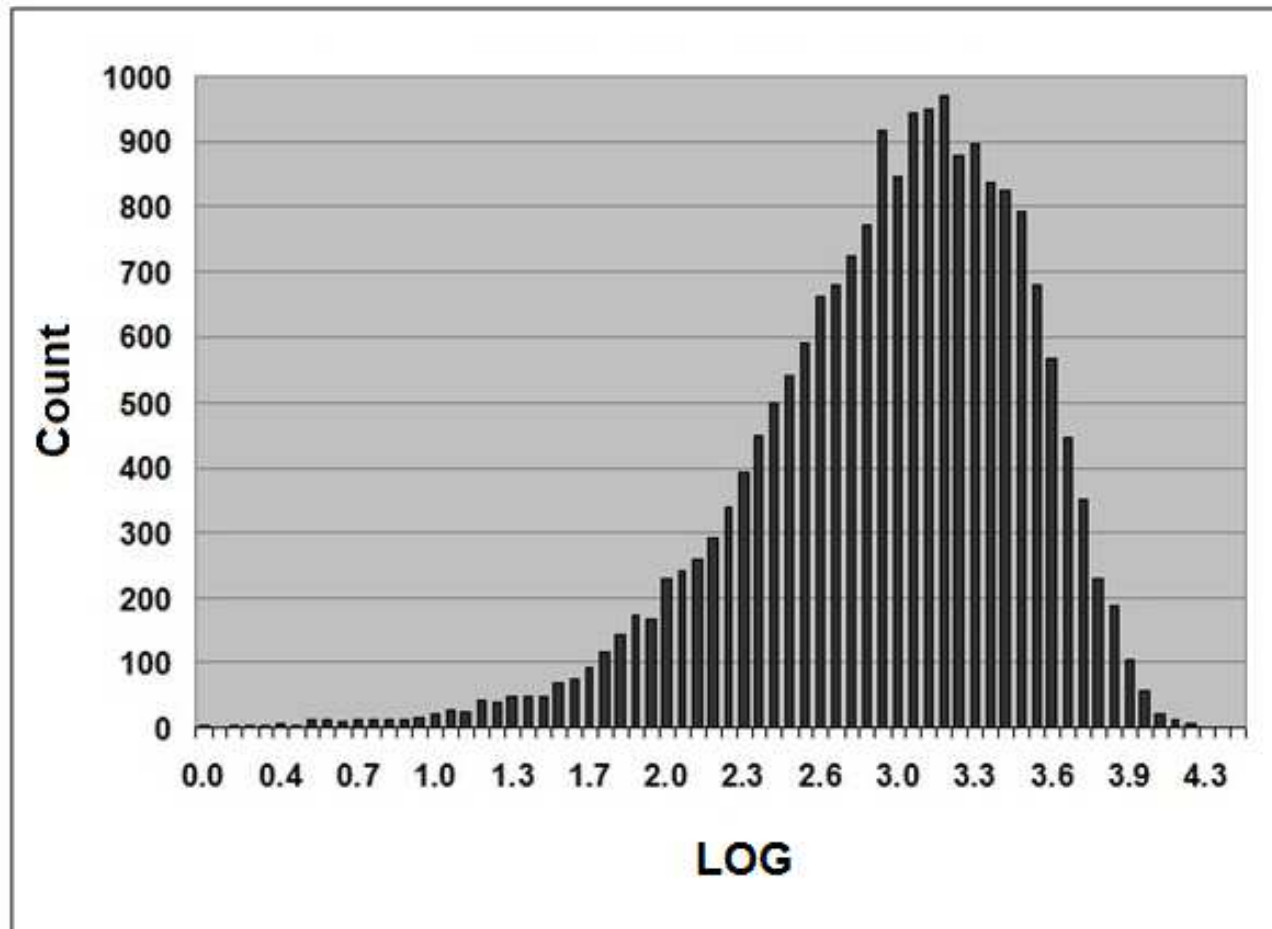


Real-Life examples
of LOG histograms:

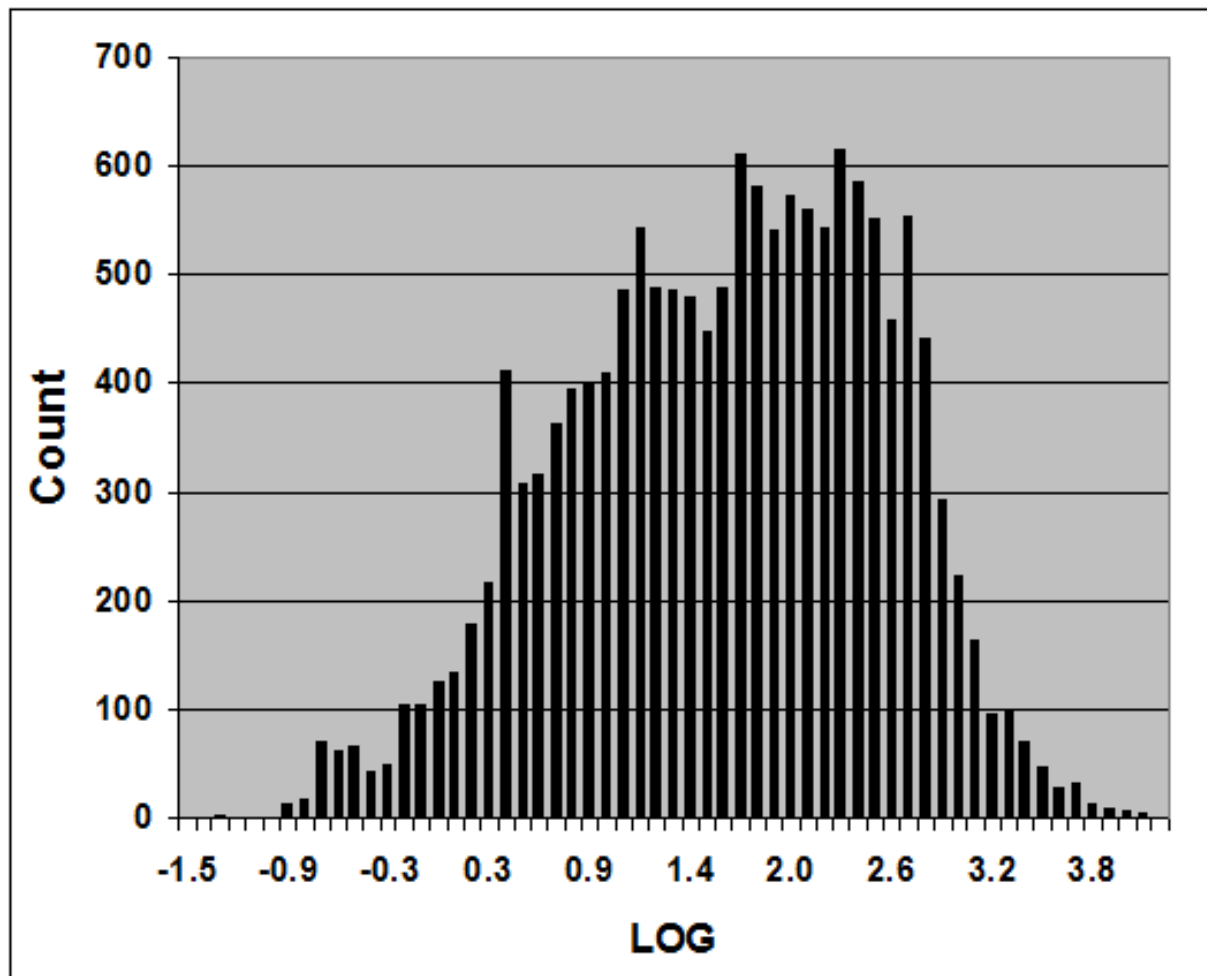
Population of USA Cities and Towns – 2009 Census



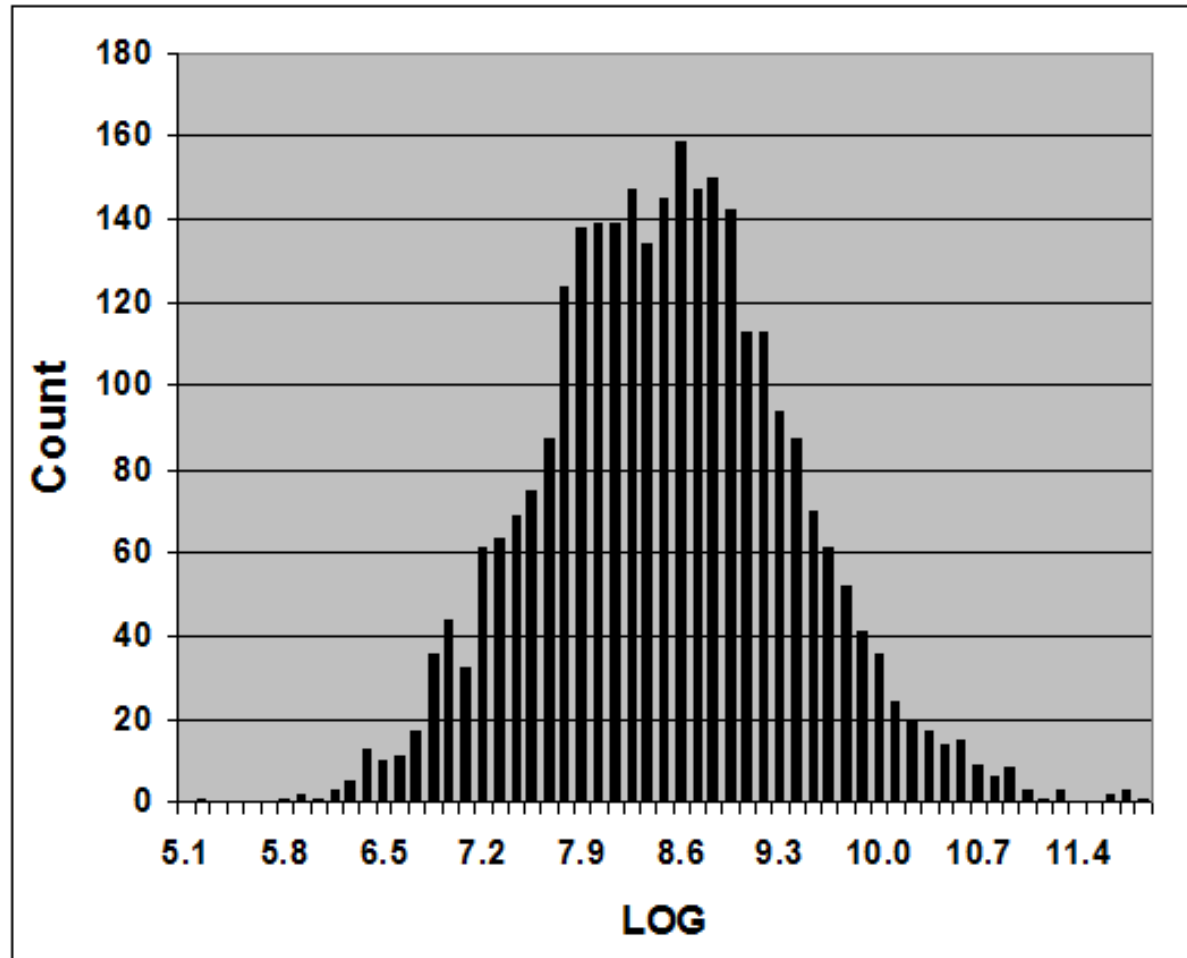
Time between all Earthquake – 2012 Global Data



Canford Audio UK - Catalog



NASDAQ Market Capitalization





Orthodox Data Forensics via Benford's Law

Two Pitfalls



**Data itself
is not Benford**



**Fraudster knows about
Benford's Law and adjusts**

Question: How could we detect fraud in these 2 cases?

Answer: Via Digital Development Pattern!

Fake Data Concocted to Appear as if Benford – Unusually Steady, No Develop

From: To:	0.1 1	1 10	10 100	100 1,000	Entire Range
1	29.2%	31.2%	30.6%	29.7%	30.6%
2	16.8%	16.5%	18.5%	19.4%	17.8%
3	12.9%	13.5%	11.7%	13.3%	12.6%
4	9.1%	9.6%	9.3%	9.2%	9.4%
5	8.3%	7.7%	8.3%	7.4%	8.0%
6	7.3%	6.8%	7.1%	6.2%	6.9%
7	5.8%	5.3%	5.5%	5.5%	5.4%
8	6.4%	4.9%	5.1%	5.2%	5.1%
9	4.2%	4.5%	3.9%	4.1%	4.2%

Fake Data Concocted to Appear as if Benford – Zigzag and Chaotic ‘Develop’

From: To:	10 100	100 1,000	1,000 10,000	10,000 100,000	Entire Range
1	21.3%	44.1%	16.7%	33.6%	30.8%
2	18.4%	20.1%	15.2%	21.8%	18.6%
3	11.1%	11.3%	14.1%	13.7%	12.6%
4	10.7%	6.2%	12.0%	8.8%	9.1%
5	9.6%	4.5%	11.1%	8.3%	7.9%
6	8.6%	4.5%	8.4%	6.1%	6.5%
7	7.9%	3.9%	7.3%	3.1%	5.3%
8	6.3%	2.9%	8.2%	2.5%	5.0%
9	6.1%	2.7%	7.0%	2.1%	4.4%

Fake Data Concocted to Appear as if Benford – Inverse Development

From: To:	1 10	10 100	100 1,000	1,000 10,000	Entire Range
1	38.9%	31.5%	26.2%	12.3%	29.1%
2	22.4%	18.5%	14.7%	11.8%	17.4%
3	12.3%	14.1%	13.5%	11.3%	13.2%
4	9.3%	8.5%	12.1%	10.5%	9.9%
5	6.2%	7.3%	9.1%	11.4%	8.1%
6	4.1%	6.8%	8.2%	11.2%	7.2%
7	3.3%	5.9%	6.0%	11.1%	6.0%
8	2.3%	4.2%	5.2%	10.1%	4.8%
9	1.2%	3.2%	5.0%	10.3%	4.2%



Development Compliance Test

An Exact Measure of Development

Via

A Measure of 'Over-Skewness'

ES12 \equiv A measure of skewness over and above the normal Benford skewed configuration

ES12 \equiv Excess Sum of digits 1 and 2

ES12 \equiv [observed digits 1 & 2] – [Benford allocation for 1 and 2]

ES12 \equiv [observed digits 1 & 2] – [**30.1% + 17.6%**]

ES12 \equiv [observed digits 1 & 2] – [**47.7%**]

For example, if digits within sub-interval are:

{44.1%, 19.6%, 11.5%, 7.6%, 5.3%, 4.3%, 2.7%, 3.2%, 1.8%}

ES12 \equiv [observed digits 1 & 2] – [47.7%]

ES12 = [44.1% + 19.6%] – [47.7%]

ES12 = [63.7%] – [47.7%]

ES12 = +16.0%

- ES12 is negative $\Rightarrow \approx$ digital equality

0 ES12 is \approx zero $\Rightarrow \approx$ Benford

+ ES12 is positive \Rightarrow extreme inequality

Canford Catalog

From	0.1	1	10	100	1,000
To	1	10	100	1,000	10,000
Digit 1 :	6.7%	15.7%	29.5%	33.6%	63.9%
Digit 2 :	21.2%	19.9%	16.6%	20.6%	18.8%
Digit 3 :	14.5%	13.4%	10.7%	13.7%	7.6%
Digit 4 :	6.7%	11.6%	9.6%	8.8%	4.8%
Digit 5 :	7.7%	8.8%	9.3%	8.5%	1.7%
Digit 6 :	8.7%	9.4%	7.3%	6.1%	1.5%
Digit 7 :	17.6%	8.6%	6.7%	3.8%	0.9%
Digit 8 :	6.7%	5.8%	5.2%	2.4%	0.8%
Digit 9 :	10.2%	6.8%	5.2%	2.5%	0.0%
Data Proportion	4.0%	21.0%	35.7%	34.8%	4.4%
ES12:	-19.8%	-12.1%	-1.6%	6.6%	35.0%

Nasdaq Market Capitalization

From	1,000,000	10,000,000	100,000,000	1,000,000,000	10,000,000,000
To	10,000,000	100,000,000	1,000,000,000	10,000,000,000	100,000,000,000
Digit 1 :	7.1%	18.7%	29.6%	44.1%	46.6%
Digit 2 :	13.5%	13.3%	16.7%	19.6%	18.6%
Digit 3 :	5.0%	11.8%	14.1%	11.5%	15.3%
Digit 4 :	9.2%	10.0%	9.9%	7.6%	6.8%
Digit 5 :	14.2%	11.6%	8.3%	5.3%	1.7%
Digit 6 :	14.9%	9.9%	6.3%	4.3%	4.2%
Digit 7 :	17.0%	8.6%	5.9%	2.7%	4.2%
Digit 8 :	11.3%	8.7%	5.3%	3.2%	1.7%
Digit 9 :	7.8%	7.4%	3.8%	1.8%	0.8%
Data Proportion	4.9%	25.9%	42.8%	21.7%	4.1%
ES12:	-27.1%	-15.7%	-1.4%	16.0%	17.5%

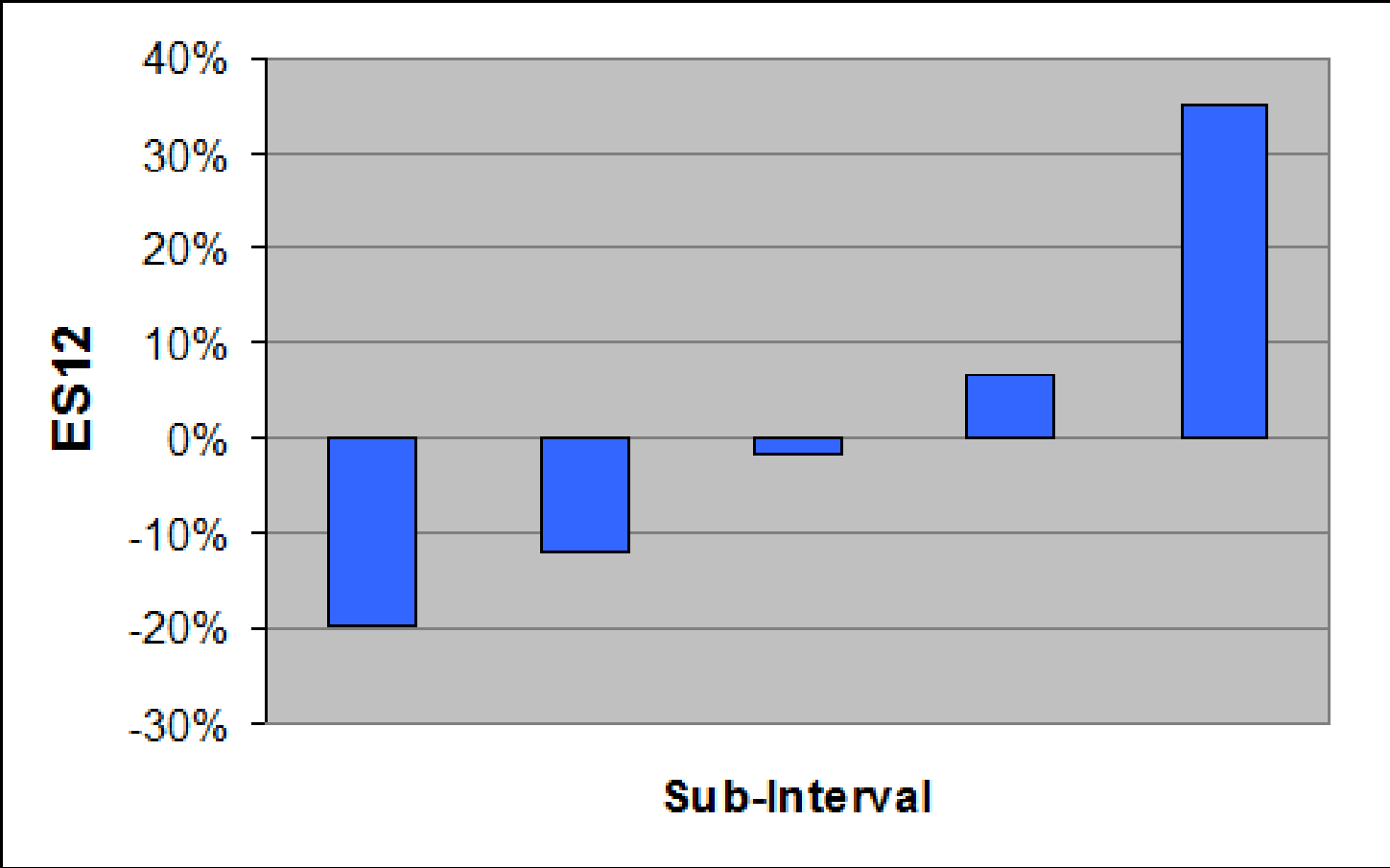
Oklahoma State Expenses

From	0.1	1	10	100	1,000	10,000	100,000
To	1	10	100	1,000	10,000	100,000	1,000,000
Digit 1 :	10.3%	7.7%	15.3%	35.0%	42.3%	48.4%	52.5%
Digit 2 :	10.1%	9.9%	14.9%	19.5%	19.7%	19.8%	17.8%
Digit 3 :	13.5%	10.8%	11.9%	12.8%	11.3%	10.4%	10.6%
Digit 4 :	10.0%	9.3%	12.0%	9.0%	8.6%	6.4%	6.3%
Digit 5 :	16.4%	12.5%	11.3%	7.7%	5.7%	5.0%	3.9%
Digit 6 :	10.2%	13.5%	9.2%	5.4%	4.1%	3.4%	3.1%
Digit 7 :	11.4%	13.3%	9.9%	4.3%	3.2%	2.9%	2.6%
Digit 8 :	7.8%	13.1%	7.7%	3.4%	2.8%	2.0%	2.0%
Digit 9 :	10.3%	10.0%	7.8%	3.0%	2.2%	1.7%	1.1%
Data Proportion	0.2%	4.0%	30.8%	43.5%	16.8%	4.1%	0.6%
ES12:	-27.2%	-30.1%	-17.5%	6.8%	14.3%	20.5%	22.6%

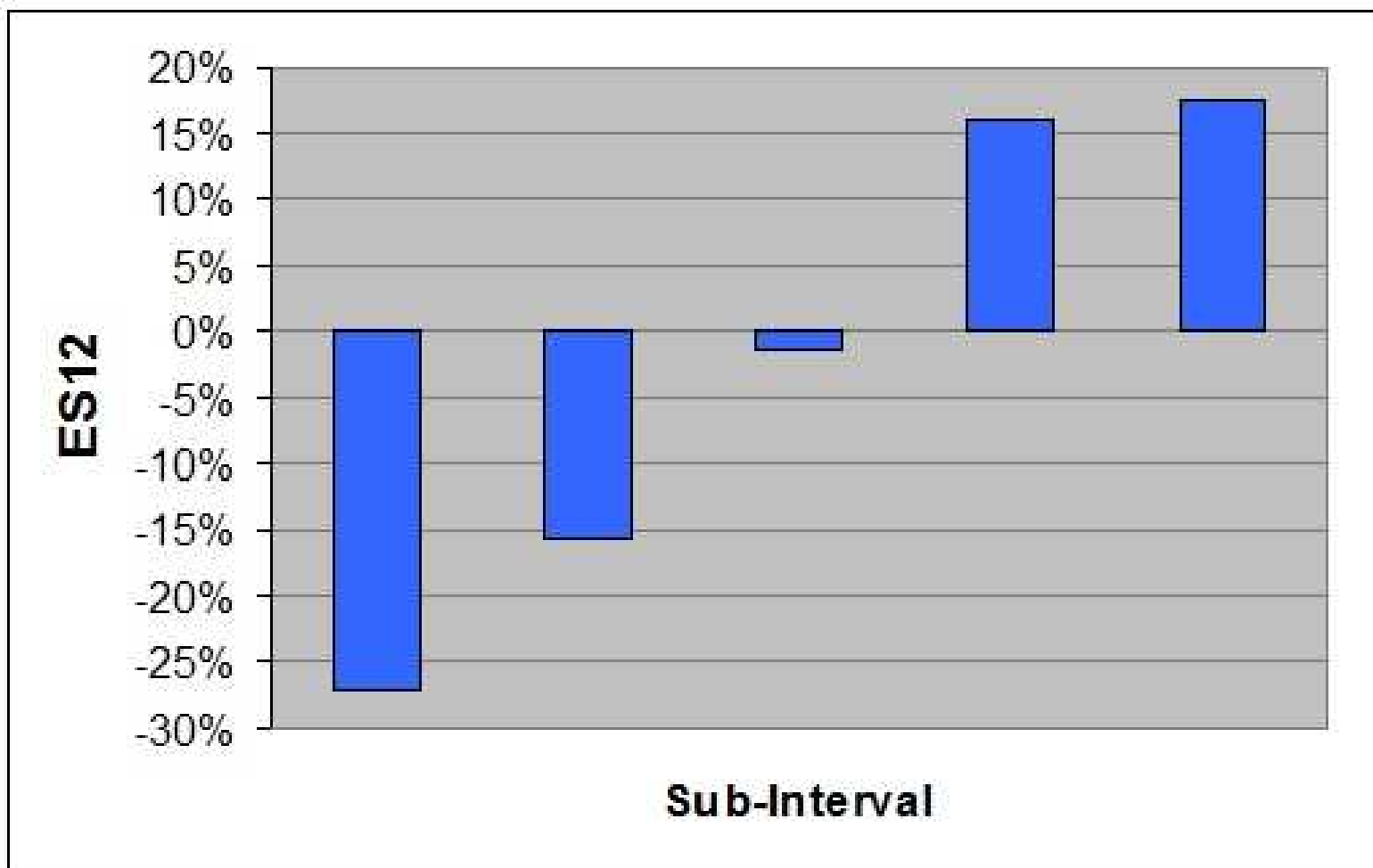
Now we can **visualize**
development clearly!



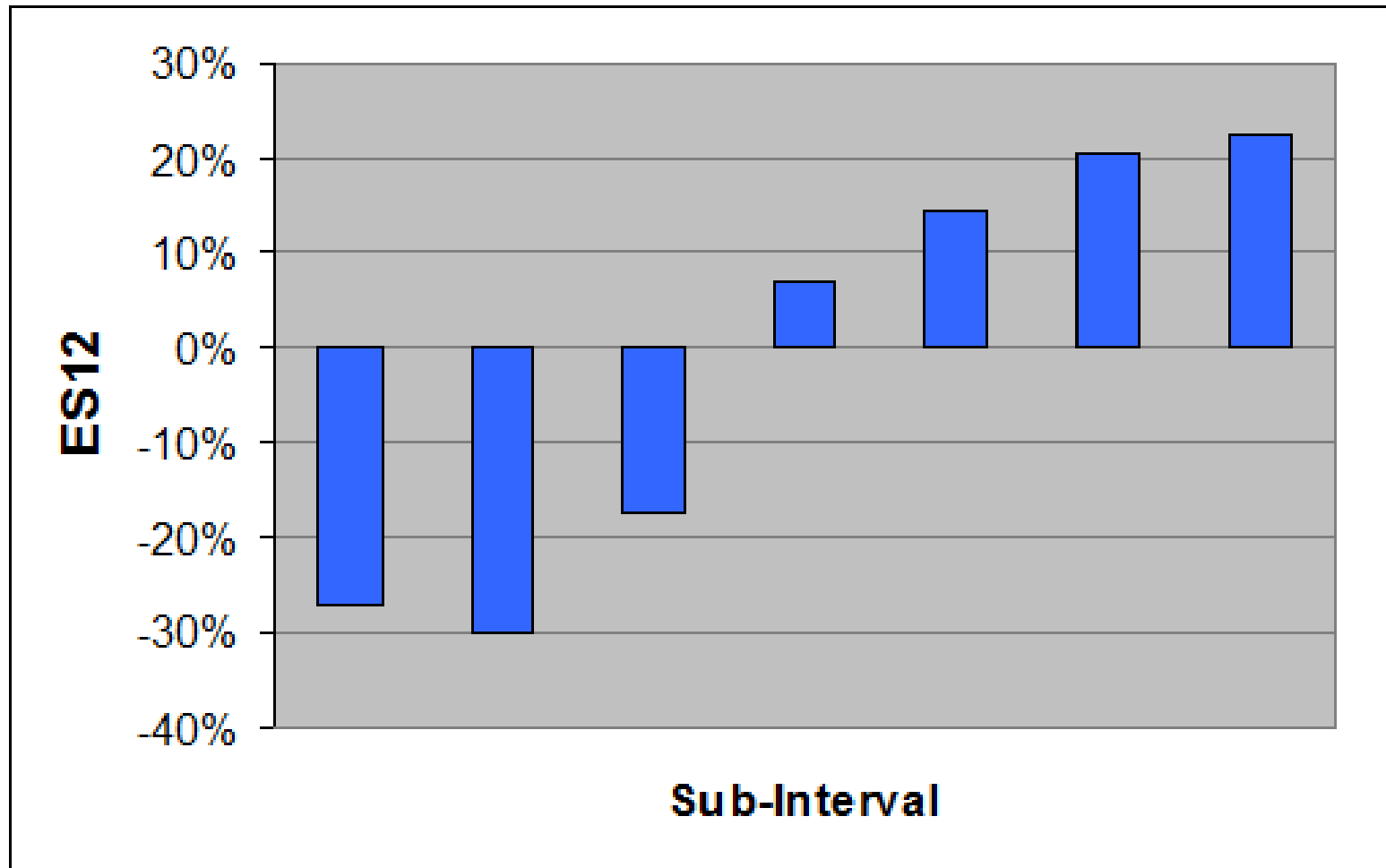
Canford Catalog



Nasdaq Market Capitalization



Oklahoma State Expenses



Question:

What is the minimum level of development so that data does not raise suspicion?

Namely: What is the threshold or cutoff point differentiating *'development-compliance'* from *'development-non-compliance'*?

Answer:

Simple Linear Regression is performed on **ES12**, requiring left-most point to be **below -15%** and right-most point to be **above +15%**, in order to declare data as development-compliance.

Firstly, any sub-interval with less than **0.1%** of overall data is eliminated, together with any sub-interval further out.

$N_j \equiv$ Rank number of sub-interval

$J = 1, 2, 3, \dots, L$

L being the last on the right.

1 being the first on the left.

$ES12_j$ being the dependent variable.

N_j being the independent variable.

Regression is modeled on:

$$ES12_j = mN_j + b$$

where **m** is the slope and **b** is the intercept.

$$ES12_left \equiv m\mathbf{1} + b$$

$$ES12_right \equiv m\mathbf{L} + b$$

Development Test:

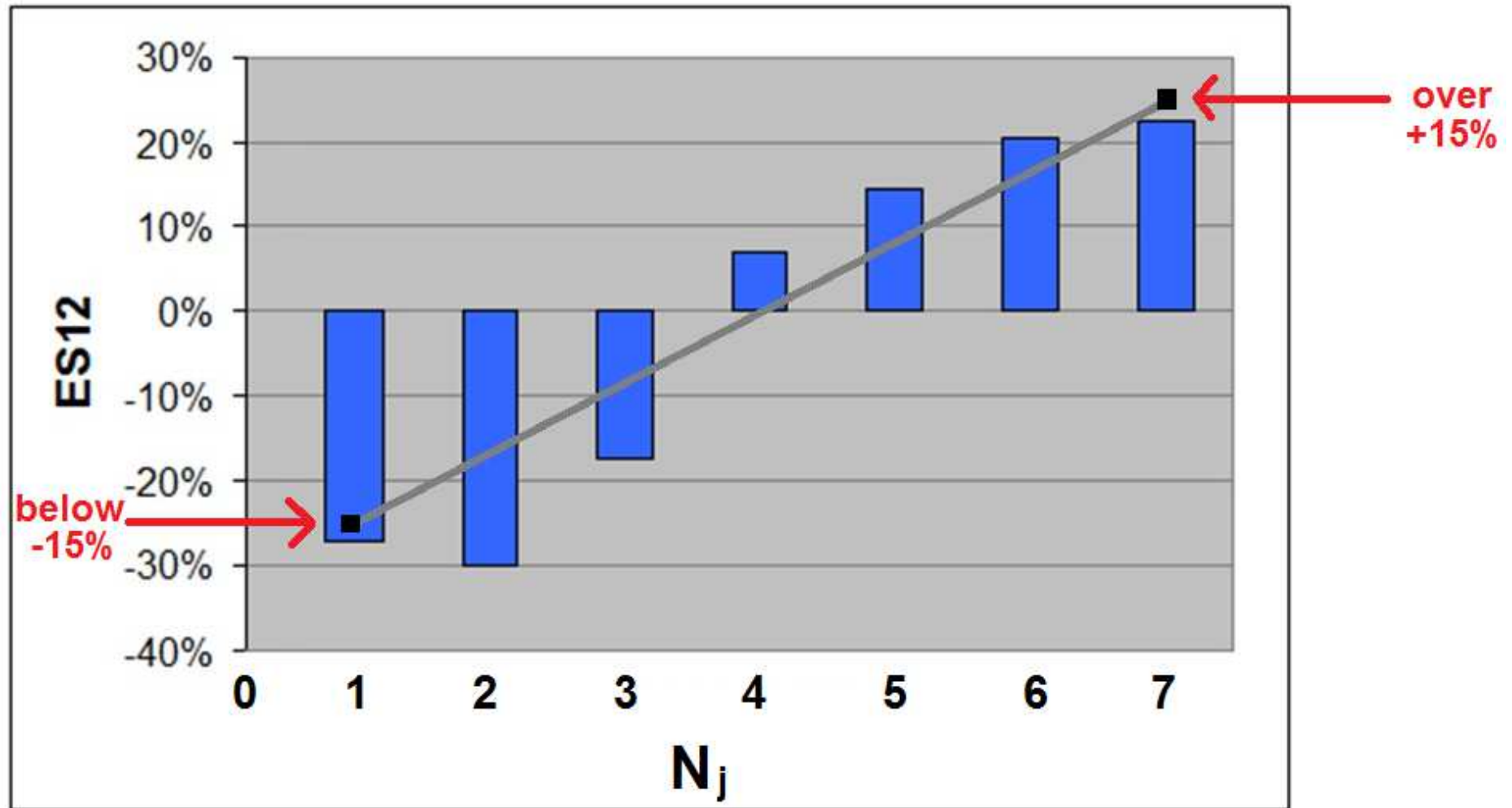
Reported data passes honesty test if the following two conditions are true:

$$ES12_left < -15\%$$

AND

$$ES12_right > +15\%$$

Oklahoma State Expenses easily passes Development Test!



The constraint of this test constitutes the **EMPIRICAL LAW OF DEVELOPMENT**, true across practically all random and statistical data types, Benford as well as non-Benford ones.

Failure to pass the test strongly hints at the possibility of fraudulent activity in data reporting.

Remarkably, **Digital Development Pattern** is found even in non-Benford data!

Remarkably, **Digital Development Pattern** is found even in data with low order of magnitude!

Hence...

**Digital Development Pattern
is even more prevalent
than Benford's Law itself!**

!!!

**Hence digital forensics can still
be perform on data sets which
are not Benford to begin with!**

Just about the **ONLY EXCEPTIONS** are **exponential growth series** and the **k/x distribution** which come with flat and horizontal LOG histogram and steady digital configuration of the Benford type throughout.

**funnily,
strangely,
surprisingly,
shockingly,**



**k/x IS WITHOUT DEVELOPMENT!
It's steady!**

Limited Applicability of k/x Distribution:

In spite of its Benfordness & uniqueness, the k/x distribution is **NOT** the appropriate model for typical everyday data sets!



k/x distribution has very **limited relevance** to real life empirical application of Benford's Law.

Theoretically though **k/x** plays a crucial role.



Patent has been granted for the
above **forensic algorithm** by the
US Patent Office in 2015.

Patent # 9,058,285.

The
United
States
of
America



**The Director of the United States
Patent and Trademark Office**

*Has received an application for a patent for
a new and useful invention. The title and
description of the invention are enclosed.
The requirements of law have been com-
plied with, and it has been determined that
a patent on the invention shall be granted
under the law.*

Therefore, this

United States Patent

*Grants to the person(s) having title to this
patent the right to exclude others from mak-
ing, using, offering for sale, or selling the
invention throughout the United States of
America or importing the invention into the
United States of America, and if the inven-
tion is a process, of the right to exclude oth-
ers from using, offering for sale or selling
throughout the United States of America, or
importing into the United States of
America, products made by that process,
for the term set forth in 35 U.S.C. 154(a)(2)
or (c)(1), subject to the payment of mainte-
nance fees as provided by 35 U.S.C. 41(b).
See the Maintenance Fee Notice on the
inside of the cover.*

Michelle K. Lee

Director of the United States Patent and Trademark Office



US009058285B2

(12) **United States Patent**
Kossovsky

(10) **Patent No.:** US 9,058,285 B2
(45) **Date of Patent:** Jun. 16, 2015

(54) **METHOD AND SYSTEM FOR FORENSIC DATA ANALYSIS IN FRAUD DETECTION EMPLOYING A DIGITAL PATTERN MORE PREVALENT THAN BENFORD'S LAW**

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(*) **Notice:** Subject to any disclaimer, the term of this patent is extended or adjusted under 35 U.S.C. 154(b) by 236 days.

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G06F 17/00 (2006.01)
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G06F 17/10 (2006.01)
G06Q 40/00 (2012.01)
G06F 08/00 (2006.01)

ABSTRACT

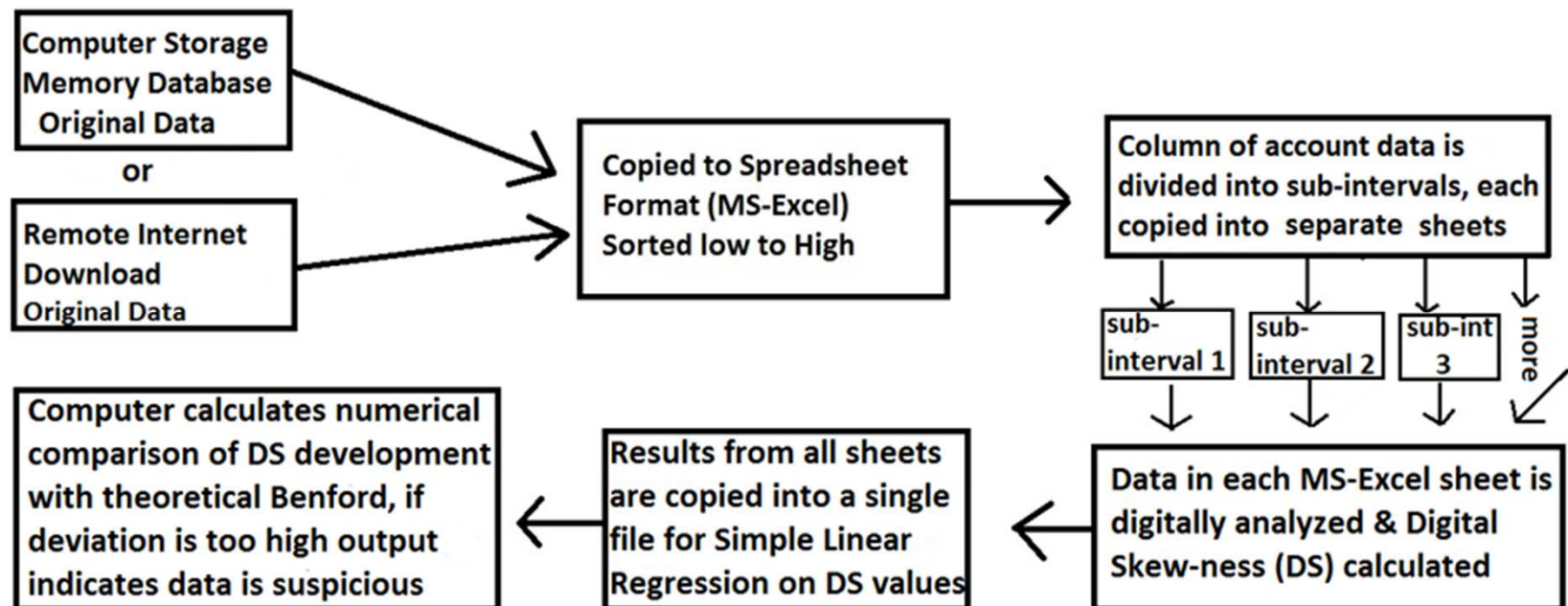
(57) A computerized system for a digital method for the detection of fraud and/or anomalous transactions is disclosed based on a novel statistical interpretation of Benford's Law and a unique set of computer implementations outlining the development of digital distributions from the low-value region on the right side of a data set to the high-value region on the left side of a data set.

Patent's Title:

“Method and system for Forensic Data Analysis in fraud detection employing a digital pattern more prevalent than Benford’s Law”.

Link on US Patent Office website:

<http://patft.uspto.gov/netacgi/nph-Parser?Sect1=PTO2&Sect2=HITOFF&p=1&u=%2Fnetacgi%2FPTO%2Fsearch-bool.html&r=1&f=G&l=50&co1=AND&d=PTXT&s1=9058285&OS=9058285&RS=9058285>



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