

# Central Limit Theorem

Independence & Benford's Law

# CLT: Important Requirements

CLT requires the sampled variables to be independent and identically distributed.

Independence:

Two events are independent if the  
**occurrence of one event** does not affect the  
**chances of the occurrence of the other event**

Is there independence in real life?

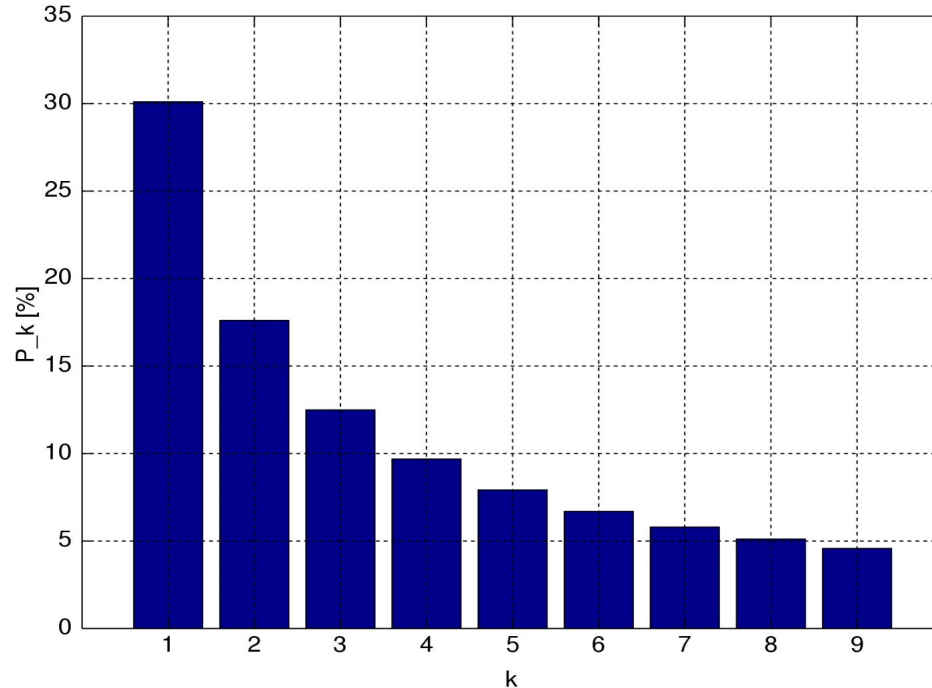
X1 (A, B, C, XI, 陆, 🦂, 4, ♁, weather, ...)

X2 (D, E, F, V, 拾, 🚢, 🏰, ♎, weather, ...)

# Benford's Law

Benford's law is an observation about the leading digits of the numbers found in real-world data sets. It states that the leading digit being 1 is more frequent than 2 and 2 more than 3 and so on.

**Distribution of first digits according to Benford's Law**



$$\Pr(d) = \log(1 + 10/d)$$

# Leading digit of heights of the 58 tallest structures in the world

Leading digit	m		ft		Per Benford's law
	Count	%	Count	%	
1	24	41.4%	16	27.6%	30.1%
2	9	15.5%	8	13.8%	17.6%
3	7	12.1%	5	8.6%	12.5%
4	6	10.3%	7	12.1%	9.7%
5	1	1.7%	10	17.2%	7.9%
6	5	8.6%	4	6.9%	6.7%
7	1	1.7%	2	3.4%	5.8%
8	4	6.9%	5	8.6%	5.1%
9	1	1.7%	1	1.7%	4.6%

## Leading digit of 1st 100 elements of Fibonacci Series

Leading Digit	Frequency
1	0.3
2	0.18
3	0.13
4	0.09
5	0.08
6	0.06
7	0.05
8	0.07
9	0.04

**Benford's Law compliance:** Benford's law is prominent when the data has several orders of magnitude, like exponential distribution. It needs the distribution to be widely spread.

### **Examples that follow Benford's Law:**

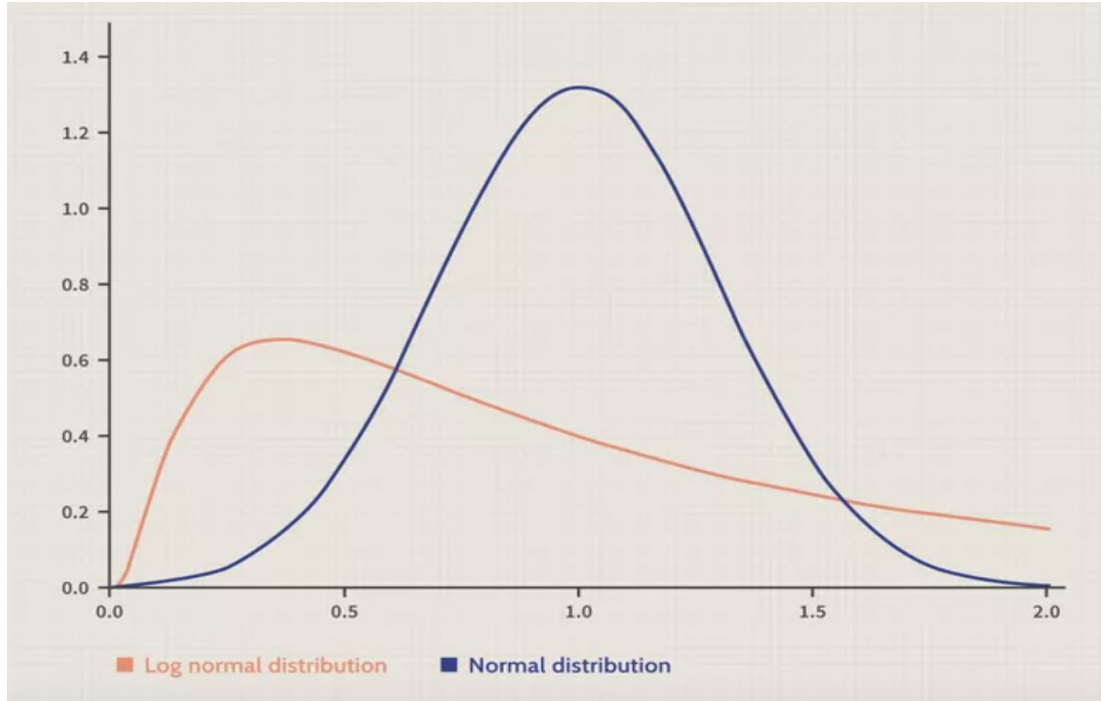
- 1) Fibonacci series
- 2) Factorials
- 3) Powers of two

### **Examples that don't follow Benford's Law:**

- 1) Lists of local telephone numbers
- 2) Distributions that do not span several orders of magnitude will not follow Benford's law  
Ex: Height

## How is CLT related to Benford's Law?

**Central limit theorem** says that multiplying more and more random variables will create a log-normal distribution with larger and larger variance, so eventually it covers many orders of magnitude.



Log-normal distributions are positively skewed with long right tails due to low mean values and high variances in the random variables.



# Applications of Benford's Law

**1) Accounting fraud detection:** Based on a plausible assumption that people who fabricate figures tend to distribute their digits fairly uniformly, a simple comparison of first-digit frequency distribution from the data with the expected distribution according to Benford's law ought to show up any anomalous results.

**2) Criminal trials**

**3) Election data**