Benford’s law and Theil transform of financial data

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Abstract

Among econophysics investigations, studies of religious groups have been of interest. On one hand, the present paper concerns the Antoinist community financial reports, - a community which appeared at the end of the 19-th century in Belgium. Several growth-decay regimes have been previously found over different time spans. However, there is common suspicion about sect finances. In that spirit, the Antoinist community yearly financial reports, income and expenses, are hereby examined along the so-called Benford’s law. The latter is often used as a test about possible accounting wrongdoings. On the other hand, Benford’s law is known to be invariant under scale and base transformation. Therefore, as a further test, of both such data and Benford’s law use, the yearly financial reports are nonlinearly remapped through a sort of Theil transformation, i.e. based on a log-transformation. The resulting data is again analyzed along the Benford’s law scheme. Bizarre, puzzling, features are seen. However, it is emphasized that such a non-linear transformation can shift the argument toward a more objective conclusion. In an Appendix, some brief discussion is made on why the original Theil mapping should not be used. In a second Appendix, an imperfect Benford’s law-like form, - better suited for anomalous distributions, is presented.

Keywords: income; expenses; religious community; Benford’s laws; Theil map; time series.

1 Introduction

The econophysics of religious movements and sects has already attracted researchers [1][2], but it could receive more attention. The finances, and more

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generally the economics of religious movements and sects are often questioned and the source of major political concerns, not discounting the ethics of their portfolios and related matters [3]. Yet, Iannaccone [4] has argued that many standard features of religious institutions exist to reduce (or at least appear to reduce) the risk of fraud and misinformation. In pioneering work [5], Iannaccone has indicated the three main lines of research on such topics, and has emphasized that some religious behavior can be interpreted from an economic perspective, applying microeconomic theory and techniques to explain sect patterns among individuals, groups, and cultures. This suggests an econophysics approach complementary to sociological ones.

The Antoinist community in Belgium and France had much quick growth, in the number of adepts, but not so much anymore [6, 7]. It can be argued that money was not the cause of the growth [6, 7], though the group got a legal Etablissement d’utilité publique (Organization of Public Utility) tax free status. In compensation, it had to publish financial reports in the official Moniteur Belge journal. Such a data for income and expenses is studied here below. The data acquisition is explained in Sect. 3.

One technique to investigate the correct report of financial data is the use of Benford’s law, outlined in Sect. 2. Newcomb and later Benford [8, 9] observed that the occurrence of significant digits in many data sets is not uniform but tends to follow a logarithmic distribution such that the smaller digits appear as first significant digits more frequently than the larger ones, i.e.,

\[ N_d = N \log_{10}(1 + \frac{1}{d}), \quad d = 1, 2, 3, \ldots, 9 \]  

where \( N \) is the total number of considered 1-st digits for checking the law, in short, the number of data points, and \( N_d \) is the number of the so observed integer \( d \) ( = 1, 2, 3, ..., 9) being the starting one (1-st) in the data set list.

Whence, the nowadays called Benford’s law could be used to identify falsely created data, e.g. in corporations financial statements [10], or to verify the (non)reliability of macroeconomic data of countries, - e.g., of too late interest in the recent case of Greece or Belgium [11].

In social sciences, Benford’s law has been used to detect election frauds and anomalies, e.g. in USA or Iran [12, 13]. Closely related to our subject, Mir investigated whether regularities or anomalies exist in numerical data on the country-wise adherent distribution of seven major world religions along Benford’s law [14].

When analyzing financial data, a transformation of the raw data can be made through what we call a Theil map or Theil transformation [15, 16, 17, 18], in Sect. 3.3. In order to compare the income \( x_i \) of \( M \) individuals in a country over a time interval and thereby to improve the resolution for changes in high incomes, Theil used the index [19]

\[ Th = \frac{1}{M} \sum_{i} \frac{x_i}{\langle x_i \rangle} \ln \frac{x_i}{\langle x_i \rangle}, \]  

(2)
summing $\frac{x_i}{\langle x_i \rangle}$ over the different times $i$ in the time interval and where $\langle x_i \rangle$ is the series mean value over the appropriate time interval. However, this induces negative and positive values of the (log-transformed) data, depending on the ratio $x_i/\langle x_i \rangle$, thus impairing a test of the validity of Benford’s law. Therefore, in Sect. 3.3 we will simply transform the relevant data through the map

$$y_i \equiv x_i \ln(x_i).$$

(3)

Understand two arguments for doing so. On one hand, questions may be raised whether Benford’s law applies after a log-transform of the raw data [21, 22, 23, 24], while, on the other hand, one might also see some bizarre data reports through such a nonlinear transformation.

Therefore, the paper goes as follows: as mentioned above, Sect. 2 briefly introduces Benford’s law. In Sect. 3 the raw data acquisition is recalled, i.e. yearly expenses and income over about 80 years, Sect. 3.1. The data of interest is displayed (i) through histograms for the first four digits, in Sect. 3.2 and (ii) similarly after a Theil transformation, in Sect. 3.3. A discussion of such histograms is found in Sect. 3.4.

Sect. 5 serves for a conclusion emphasizing (i) the interest of a Theil transformation to study data along Benford’s law concepts, and (ii) the complexity of studying a community through its financial history.

A brief discussion of log-transformed data is given in Appendix A. Moreover, since there is some apparent small deviation in the analyzed data, from the 1-st digit Benford law, an attempt to a better fit of the raw and Theil mapped data through a so called imperfect 1-st- digit Benford law is given in Appendix B.

2 Benford’s law

Benford’s law [8, 9, Eq.(1)], is known as the first digit law or the law of the leading digits. According to Eq.(1), in a given data set the probability of occurrence of a certain digit as the first (1-st) significant digit decreases logarithmically as the value of the digit increases from 1 to 9. Thus, digit 1 should appear as the first significant digit about 30.103% times, and similarly 9 should appear about 4.576% times.

Benford’s law, Eq.(1), holds for data sets in which the occurrence of numbers is free from any restriction: it does not apply to a list of telephone numbers, zip codes, ID-card numbers, car license plate numbers, .... It has been found that tampered, unrelated or fabricated numbers usually do not follow Benford’s law [25]. Thus, significant deviations from the Benford distribution may indicate fraudulent or corrupted data [26].

On the other hand, much theoretical work has been done on the matter. It has been discussed by many, in particular in [27, 28] that base-invariance implies Benford’s law. Whence, the law can be statistically derived along rigorous lines.

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1 The Theil index can be also transformed into a Theil entropy [20] but this is not used here either because it also induces negative and positive values of the (log-transformed) data, thus impairing a test of the validity of Benford’s law.
It is also possible to extend the law to digits beyond the first \[29\]. In particular, the probability of encountering a number starting with a string of digits \( n \) is given by

\[
\log_{10} (n+1) - \log_{10} (n) = \log_{10} \left(1 + \frac{1}{n}\right) = \log_{10} \left(\frac{n+1}{n}\right),
\]

as its results from mere conditional probability algebra.\[2\] This result can be used to find the probability that a particular digit occurs at a given position within a number. E.g., the probability that \( d \) (\( d = 0, 1, \ldots, 9 \)) is encountered as the \( n \)-th (\( n > 1 \)) digit is

\[
\sum_{k=10^{n-2}}^{10^{n-1}-1} \log_{10} \left(1 + \frac{1}{10k+d}\right) = \log_{10} \left[\prod_{k=10^{n-2}}^{10^{n-1}-1} \left(\frac{10k+d}{10k+d+1}\right)\right].
\]

For instance, the probability that a ”2” is encountered as the second digit is \[29\]

\[
\log_{10} \left(1 + \frac{1}{12}\right) + \log_{10} \left(1 + \frac{1}{22}\right) + \cdots + \log_{10} \left(1 + \frac{1}{92}\right) \approx 0.1088.
\]

The distribution of the \( n \)-th digit, as \( n \) increases, exponentially approaches a uniform distribution (UD) \[29\]. In practice, applications of Benford’s law for fraud detection routinely use more than the first digit \[29\]. Indeed, the above can be generalized to forecast how many times any digit, or any combination, should be found at some rank in the number. The \textit{ad hoc} table for the probable position of the digit \( d \) up to rank 4 is given by Nigrini, \[29\], and requoted in \[30\].

An extensive bibliography, from 1881 up to 2006, on Benford’s law papers including theories, applications, generalizations and warnings can be found in a H"urlimann’s unpublished work \[31\].

\section{Antoinism community yearly budget data}

\subsection{Financial data}

A community like the Antoinists exist for about more than a century \[6, 7\]. The time range of the financial data set below examined goes from 1922 till 2002, i.e. when it was mandatory to report the community finances. The data has been extracted from the Belgian yearly official journal, the \textit{Moniteur Belge}, when it was available in the archives of the Antoinist Cult Library in Jemeppe-sur-Meuse, Belgium. A few of these journals are missing, on various years but ca. \[1960-65\] mainly, without any known reason. However, this gap in data points does not appear \textit{a posteriori}, i.e. after the subsequent data analysis, to

\[\text{For example, the probability that a number starts with the digits 123 is } \log_{10}(1+1/123) = \log_{10}(124/123) \approx 0.0035.\]
Table 1: Three growth phase regimes found from best fits in [1] on reported yearly expenses and income data of the Belgium Antoinist community. The time intervals of the three regimes found in [1] have been slightly adapted here, as explained in the main text, for presently checking the Benford’s law validity; $N_I$ is the corresponding number of data points in the regime $I$.
such column stacking displays. The histograms for the 2-nd, 3-rd and 4-th digit are shown in Figs. 1(b)-(d) and Figs. 2(b)-(d). The "theoretical", expected, Benford’s law for the first two digits, Eq.(1) and Eq.(5) respectively, are shown by darkened triangles.

The corresponding cases for the data resulting from the normalization, $x_i/ \langle x_i \rangle$, taking into account the average $\langle x_i \rangle$, either over the full time interval or the respective $\langle x_i \rangle$ in each regime (or in each time interval), are not shown nor considered. Indeed, it is easily proved $[21, 22]$ that Benford’s law is insensitive to such a change of scale, i.e., multiplying (or dividing) the data by any positive scalar, - here some average of the raw data over a time interval. Such an operation leads to identical probability distribution functions of digits $[21, 22]$.

3.3 Benford histogram of Theil transformed data

As discussed in the introduction, there are two reasons to investigate some transformed data along Benford’s law concepts, i.e. (i) to predict theoretical differences, when a non-linear transformation is used, and (ii) to observe whether one can deduce anomalies which would not be seen, if only some raw data is analyzed. Since it is known $[21, 22]$ that a mere base or scale transformation is leaving Benford’s law invariant, it is of interest to examine more elaborate transformations.

Recently, theoretical investigations on non-linear data transformation, like $\log[\log(x)]$, $\sqrt{x}$, $x^2$, have been reported $[23, 24]$, though in another practical context. Our study along a Theil map, Eq. (3), as done here, will thus serve as a complementary information to such publications.

The distributions of 1-st to 4-th digit of the transformed, expenses and income, data according to $x \log_{10} x$ are given in the four subfigures of Figs. 3-4. A comment on the case $\log|x|/ < x >$ is found in Appendix A.

3.4 Benford histograms analysis

First, it is examined whether the distributions of the first digits match the distribution specified by Benford’s Law (BL), Eq.(1). Second, it is examined whether the first digits occur equally often at the second rank.

In order to do so, $\chi^2$ tests have been used respectively on the 1-st and 2-nd digit through:

$$\chi^2_{1BL} = \sum_{i=1}^{9} \frac{(d_{i1} - d_{i1B})^2}{d_{i1B}},$$

$$\chi^2_{2BL} = \sum_{i=0}^{9} \frac{(d_{i2} - d_{i2B})^2}{d_{i2B}},$$

where $d_{i1B}$ and $d_{i2B}$ are the theoretically expected values according to BL. The results are given in Table 2.
For estimating the result interest, i.e. "feeling the numbers", one can make a comparable test of the data with respect to a simple uniform distribution (UD), i.e.

\[ \chi^2_{1UD} = \sum_{i=1}^{9} \frac{(d_{i1} - d_1)^2}{d_1} \]  

\[ \chi^2_{2UD} = \sum_{i=0}^{9} \frac{(d_{i2} - d_2)^2}{d_2} \]  

where \( d_1 \) and \( d_2 \) is the number expected from a uniform distribution at the 1-st and 2-nd rank respectively; here obviously, \( d_1 = \frac{64}{9} \) while \( d_2 = \frac{64}{10} \). The results are given in Table 2.

First, the fits to the Benford’s law for the raw data and after the Theil mapping can be contrasted, for both cases, expenses or income. It appears that the BL \( \chi^2 \) value for the expenses test is smaller in 7 cases out of 8; the "not like others" case is the 1BL Theil map. This is exactly the opposite for the UD \( \chi^2 \) test.

In the cases of 1BL and 1UD, these statistics may be compared to the \( \chi^2 \)-distribution (\( \chi^2_8 \)) with 8 degrees of freedom. That distribution has a critical value of 15.5 for a 0.05-level test \[32\]. For 2BL and 2UD, these statistics may be compared to the \( \chi^2 \)-distribution with 9 degrees of freedom (\( \chi^2_9 \)), which has a critical value of 16.9 for a 0.05-level test.

Because all of the statistics reported in Table 2 for UD greatly exceed those values, the hypothesis that the UD is obeyed can be at once disregarded \[24\]. The situation of BL is more surprising. Indeed, according to this goodness fit test, the 1BL does not seem to be obeyed for the raw data, neither for expenses nor income. This is somewhat surprising. But one can observe that the 1-st digit is much more often 1 than "should be expected". In fact, the curvature of the envelope of the histogram even changes sign and curls up for the last digits, markedly smoothly in the case of the expenses, but in a less smooth way in the case of the income data. In both cases, the upturn occurs around digit 5. Due to this observation, an attempt to generalize Benford’s law taking into account such curling effect is given in Appendix B.

However the BL holds for the 2-nd digit and for the Theil mapping, according to this goodness fit test. It is remarkable that already the distribution of the second digit tends toward a flat distribution: the "almost equally occurring digits" are 0,1,2,3.

We have not studied the distributions of the 3-rd and 4-th digit neither in the raw data nor in the Theil mapped data. Indeed, to find a quasi uniform distribution for such cases might not be expected due to the small number of data points \[33\]. Fig. 3(d) is nevertheless remarkable, i.e. the 4-th digit distribution for the expenses raw data, indicates a sort of uniform background on which regular peaks are superposed. This bizarre behavior (of the 4-th digit!) might cast some doubt about the exactness of the content of the reported data \[25\].
Table 2: Values of the $\chi^2$ for a 1-st digit Benfords Law (1BL) test and 2nd-digit Benford’s Law (2BL) test of the raw and the Theil mapped data of reported yearly income and expenses of the Belgium Antoinist community during the 20-th century. The corresponding $\chi^2$ assuming a uniform distribution (UD) is given for comparison. Recall the $\chi^2$-distribution critical value for a 0.05-level test $[32]$: $\chi^2_8 = 15.5$ and $\chi^2_9 = 16.9$ with 8 and 9 degrees of freedom respectively.

Finally, note that the Benford test after the Theil mapping of the raw data gives $\chi^2$ values twice lower than for the raw data. In some sense one could have guessed such a feature for the 1BL. Indeed a mere logarithmic transformation would induce an accumulation of first digits representing the power of 10. If the change in budget is mild over several years, a peak would be found at some digit. However note that the transformation is $x \log_{10}(x)$, thereby mixing digits. The $\chi^2$ test indicates some interesting though unknown up to now feature of the Theil mapping for testing, e.g., faked data.

4 Discussion

As the starting point of a discussion, recall that the present work has two aims. One is a test of Benford’s law on specific data: the question being whether the data is reliable, or concurrently whether Benford’s law is valid for such a case. The second aim is to touch upon the question whether Benford’s law applies after a log-transform of the raw data, - thus a non-linear transformation, a question raised e.g., in [21, 22, 23, 24]. This is in line with recent considerations in statistical physics. Indeed, some time ago, physicists have been attracted to study financial time series and to provide or present simple model equations or algorithms for the data evolution; more exactly making hypotheses on the microscopic and dynamical causes of the measured statistical characteristics; e.g. see [34, 35, 36].

The finances of religious communities and sects in particular have been the object of rumors, scandals, criticisms, etc, for centuries and in various media or groups. Some global data is sometimes freely available [37, 38], - however without any (statistically based) reliable control. Whence econo-physics tests and considerations are of interest. Pay attention that finances of sects have been already discussed along statistical mechanics concepts [1, 2]. Ausloos looked
at Antoinist cult community[1] because the data is legally available. In [2], partial distribution functions of adherents were found to follow laws as in economic activities and scientific research, suggesting that religious activities are governed by universal growth mechanisms. However, one might wonder whether the financial reports are faked.

Thus, it has been argued, as here above, that it is of interest to analyze data along Benford’s law lines. It is much agreed that Benford’s law is definitively not exact[39]. Thus, some deviation from the Benford distribution would not provide a conclusive proof of manipulation, just as conformity does not prove cleanliness of the data. Nevertheless, Benford’s law(s) may be considered useful as an aid in analytical procedures of testing the exactness of financial reports, like those of such religious movements[26].

Note that this empirical law has also been used in other cases in order to find out whether the data is reliable, as briefly mentioned by Mir[14] and by Pietronero[40]. In physics and applied mathematics, the law has been applied to numerical data on physical constants[41], atomic spectra[42], values of radioactive decay half lives[43], decay width of hadrons[44], magnitude and depth of earthquakes[45], mantissa distribution of pulsars[46], solutions or nonlinear differential equation systems[47], or appearance of numbers on the internet[48]. The law can be used in optimizing the size of computer files[49] or enhancing the computing speed[50].

Thus, if the test is conclusive, analyzers are happy, but if not, this induces more questions and reflexions. The apparent lack of agreement with Benford’s law for the 1-st digit of the raw data only, as found in the previous sections is somewhat frightening. One could stop at this level, sending the Antoinists treasurer and hierarchy to court for falsification. However an acceptable goodness fit test for the other cases, - disregarding the uniform distribution, of course, turns the case otherwise. This leads to basically two sets of questions, economic and financial ones, about the specificity of the data, resulting from an accumulation of items:

1. Most likely, in order to resolve such a puzzle, one should demand more information on the items leading to the final sum of income and expenses. What was really accounted for? Although the reported income and expenses pertained to a concluded year, it might occur that some rounding factor accumulated so much in a few cases as modifying the first digits of items and finally the global report. It does not mean that the Antoinist community has been cheating when reporting, - why should it?

2. The anomalies might be only the result of sloppy, deliberate manipulation or unintentional but lazy accounting, quite in contrast from data manipulation by Governments[11].

3. A third hypothesis might have a more fundamental aspect: indeed, non-conformity with Benford’s law should not be qualified as a reliable sign of poor quality of macroeconomic data, but could rather be based on marked structural shifts in the data set[51]. However, this is only
throwing the stone toward closer inspection, much outside an econophysics investigation.

In fact, a subsequent question rests upon the measure of the tail of the distribution. It is known that it is not anomalous that the tail of a distribution function often appears well measured on log-log plots.

At this stage of the conclusion, it should be emphasized that the Benford’s law test of the resulting data after a Theil mapping of the raw data gives $\chi^2$ values in favor of the Benford’s law validity for the Antoinists financial data reports. This $\chi^2$ test result suggests to consider more often the Theil mapping for testing whether there is some faked data, within a Benford’s law framework. This non-linear transformation should be further examined by mathematicians (and physicists) involved with research on the whereabouts of Benford’s law.

In fine, in light of theoretical and practical considerations along Theil’s ideas, about income, and recent work in macro- and micro-econophysics on the matter, it seems relevant to emphasize how useful it can be to observe data along various "lenses".

5 Conclusions

A definite conclusion in scientific work is not always possible. We do stress that it would be regrettable that all conclusions of scientific papers be "definite conclusions". Yet some practical information should be necessarily outlined after some scientific analysis. The present approach has to be taken as leading to a warning on results, in some economic sense. The present financial case presents such an ambiguity, indeed, - at least at first sight. However, it seems that one method, as used here, i.e. a non-linear transformation of the data, can lead to more (or maybe less, in other cases) confidence on the data reliability.

In brief, two ingredients, yearly income and expenses, of a religious movement financial reports have been here investigated along Benford’s law concept. Moreover, we have also transformed the reported data through a non-linear mapping before again testing Benford’s law. Note also that we have been testing another distribution, the uniform distribution, beside the Benford one. This allows us to have some convincing argument about the validity of the main approach. The analysis of the non-linearly transformed data strengthens the feeling toward an objective conclusion.

One practical conclusion, however, is that complementary accounting techniques to Benford’s law should be considered before deciding whether some data is faked. Our attempt to use a non-linear transformation, through a sort of Theil map, suggests to consider a physicist technique. We doubt that it will be the case in accounting and political economy. But it might be used in scientific circles, before being considered adequate in other fields.

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Appendix A: Log transform and Benford law

One can directly take the natural log of the yearly income and expenses of the Belgium Antoinist community from the yearly budgets reported in the Moniteur Belge, i.e., studying $\log(x_i)$ digit distributions.

One can also "normalize" the raw data, in order to define yearly relative income and expenses, $x_i/\langle x_i \rangle$, taking into account the average expenses and income, e.g., over the whole time range or within the three time regimes, as done in Fig. [3]. The overall relative income and expenses data are also displayed. It is readily seen, as could be expected, that the corresponding normalization only results in a reduction of the older/low values of course.

"Finally", one could be studying $\log([x_i/\langle x_i \rangle])$. The transformed data is shown in Fig. [6].

Obviously, to divide by the average income or the average expenses merely result in a change of scale for the data. However, such a change induces through the log transformation a set of values greater or smaller than zero (here, ca. 1975) which is incompatible with the validity presumption of Benford’s law. Nevertheless, note the accumulation of data points near integers, in particular +2, +1, and also -1.

For the same reasons, the "official Theil mapping", Eq. (2), leading also to a set of negative terms, was not studied within this Benford’s law framework.

Therefore, we have preferred studying the data resulting from the non linear transformation $x_i \log(x_i)$, a sort of Theil mapping, Eq. (3).

Appendix B: Imperfect Benford’s law

It seems that raw data does not always well agreed with Benford’s law at "large" 1st-digit ($d \geq 5$). Sometimes a curl up feature can be seen, as in Fig. 1(a) and in Fig. 2(a). The exact reason is unknown. It might be due to psychological effects in business, i.e., it is better to pretend to have goods less expensive that they could be, thus reducing the sale price toward a lower decade. This increases the number of 9’s in contrast to the number of 1’s as a first digit. The same can be thought about tax evasion. Therefore, it is of interest to slightly modify the 1st-digit Benford’s law to take such an effect into account.

There are several ways of course to generalize Benford’s law, keeping the writing as simple as possible, and with the same sort of analytical form. For completeness, let us point out that Benford’s law was already "generalized" in a specific and ad hoc way [52] as

$$p_{r,q}(x) = N_{r,q} \log_{10}(1 + \frac{1}{r + x^q}), \quad x = 1, 2, ..., 9$$

(11)

where $N_{r,q}$ is a normalization factor which makes $p_{r,q}(x)$ a probability distribution; $r$ and $q$ are model parameters to precisely describe the distribution for different cases. Obviously, when $r = 0$ and $q = 1$, one recovers the Benford’s law distribution. Note that if $r = 1$, one might reanalyze other data sets, like zip
Table 3: Values of the "Imperfect Benford’s Law" parameters, $N_s$, and $s$, obtained for the fits with Eq. (12), to the raw or Theil mapped reported yearly income and expenses of the Belgium Antoinist community during the 20-th century. The $\chi^2$ result is given together with the resulting value of the theoretical histogram surface $S$

|       | expenses | expenses | income | income |
|-------|----------|----------|--------|--------|
| raw   | 61       | 63       | 62     | 63     |
| Theil | 0.0031   | 0.0012   | 0.0021 | 0.0011 |
| $\chi^2$ | 16.11   | 8.58     | 22.73  | 7.08   |
| $S$   | 64.09    | 64.24    | 64.13  | 64.14  |

code lists, telephone number lists, ... for which the first digit can be 0. Thus, in Eq. (11), one would let $x \in [0, 9]$ without any log-divergence problem.

However, the distribution curvature is unchanged when using Eq. (11). In order to have a curl up, i.e. a positive slope at "high" 1-st digit, another approach is mandatory. In order to keep the same analytic log-form, as Eq. (1), we have found out that the most simple form is

$$p_s(x) = N_s \log_{10}(\frac{1}{x} + 1 + s \, x), \; x = 1, 2, ..., 9 \quad (12)$$

where $N_s$ is a normalization factor which makes $p_s(x)$ a probability distribution; $s$ being a model parameter to precisely fit the data. The minimum of the distribution, $\log_{10}(1 + 2\sqrt{s})$, occurs at $1/\sqrt{s}$. Obviously, when $s = 0$, one recovers the Benford’s law distribution. For illustration, see Fig. 7.

As a test, the 1st-digit histogram of the raw and Theil mapped data for both income and expenses of the reported Antoinist community budget are shown in Fig. 8. For ease in the fit, we have let $N_s$ be constrained to be an integer. In so doing the total surface $S$ of the histogram is not exactly equal to the number of data points, i.e. 64. The $s$, $N_s$, $S$, and $\chi^2$ values for the best fits are given in Table 3. Those values of course depend on the accuracy with which one constrains the nonlinear fits. Nevertheless, the fits are well improved in the case of the expenses. Again, the Theil mapping markedly improves the overall fits.
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Figure 1: Benford histogram of the first four digits of reported expenses in the yearly budget of the Belgian Antoinist community: the previously found three growth-decay time interval regimes are distinguished and summed up. The expected Benford’s laws for the first two digits are shown by triangles.
Figure 2: Benford histogram of the first four digits of reported income in the yearly budget of the Belgian Antoinist community: the previously found three growth-decay time interval regimes are distinguished and summed up. The expected Benford’s laws for the first two digits are shown by triangles.
Figure 3: Benford histogram of the first four digits of Theil transformed expenses in the yearly budget of the Belgian Antoinist community: the previously found three growth-decay time interval regimes are distinguished and summed up. The expected Benford’s laws for the first two digits are shown by triangles.
Figure 4: Benford histogram of the first four digits of Theil transformed income in the yearly budget of the Belgian Antoinist community: the previously found three growth-decay time interval regimes are distinguished and summed up. The expected Benford’s laws for the first two digits are shown by triangles.
Figure 5: Yearly relative income and expenses of the Belgium Antoinist community from the yearly budgets reported in the Moniteur Belge; three time regimes can be distinguished; the overall relative data is also shown.
Figure 6: Natural log of the yearly relative income (inc) and expenses (exp) of the Belgium Antoinist community adapted from the yearly budgets reported in the *Moniteur Belge*; although the average expenses ($\langle \text{exp}_i \rangle$) and income ($\langle \text{inc}_i \rangle$) are taken over the whole time range, three time regimes can be distinguished.
Figure 7: Imperfect Benford law, Eq. (12)
Figure 8: Plots of the raw and Theil mapped data histogram of the 1st-digit in the reported expenses and income for the yearly budget of the Belgian Antoinist community comparing the expected and imperfect Benford’s laws; the best fit parameters of the latter are given in Table 3.