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“INDIVIDUAL AND FIELD CITATION DISTRIBUTIONS IN 29 BROAD SCIENTIFIC FIELDS”

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Abstract

Using a large unique dataset consisting of 35.1 million authors and 105.3 million articles published in the period 2000-2016, which are classified into 29 broad scientific fields, we search for regularities at the individual level for very productive authors with citation distributions of a certain size, and for the existence of a macro-micro relationship between the characteristics of a scientific field citation distribution and the characteristics of the individual citation distributions of the authors belonging to the field. Our main results are the following three. Firstly, although the skewness of individual citation distributions varies greatly within each field, their average skewness is of a similar order of magnitude in all fields. Secondly, as in the previous literature, field citation distributions are highly skewed and the degree of skewness is very similar across fields. Thirdly, the skewness of field citation distributions is essentially explained in terms of the average skewness of individual authors, as well as individuals' differences in mean citation rates and the number of publications per author. These results have important conceptual and practical consequences: to understand the skewness of field citation distributions at any aggregate level we must simply explain the skewness of the individual citation distributions of their very productive authors.

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I. INTRODUCTION

At any aggregation level, bibliometric studies using citation counts may reveal statistically significant macro-patterns in the communication process that cannot be seen from the limited perspective of the individual researcher in peer review exercises. In this paper, we search for regularities at the level of individual authors, and the nature of the macro-micro relationship between the field citation distribution and the individual citation distributions of the authors in the field. In the context of science as a system of highly interconnected entities at different levels (individual researchers, research groups, university departments, research institutes, universities), Costas *et al.* (2009) have emphasized the importance in large networked systems of the relations between large-scale attributes and local patterns (i.e. between field and individual citation distributions in our case). More generally, Katz (2016) views the global research system as a complex innovation system exhibiting a variety of scale-invariant properties that are statistically similar at many levels of observation.

Costas *et al.* (2009) study the scaling relationship between the number of citations and the number of scientific publications. Specifically, they investigate whether the scaling behavior identified at the research group level (Van Raan, 2006a, 2006b, 2008) is also observed at the individual level. As for Katz (2016), he studies scale-invariant correlations between the growth of impact and size over time, and between impact and size across fields and sub-fields at a point in time. In this paper, we focus on a key characteristic emphasized since the inception of scientometrics by Price (1965) and Seglen (1992), namely, the skewness of citation distributions according to which a large proportion of articles receives no or few citations while a small percentage of them account for a disproportionate amount of all citations.

At the field level, we should take into account that wide differences in production and citation practices across fields greatly affect the size and the mean of field citation distributions. Similarly, differences in individual productivity and citation impact among authors in a given field give rise to wide differences in the size and mean of individual citation distributions. Therefore, it seems convenient to evaluate the skewness of citation distributions abstracting from size and mean

differences across fields and individual authors. For that purpose, we use the Characteristic Scores and Scales (CSS hereafter) technique for grouping ranked observations into ranked-specific categories (Schubert *et al.*, 1987, Glänzel and Shubert, 1988), which is size- and scale-independent.

We study two topics. Firstly, we focus on individual citation distributions within and between scientific fields. That is, we study how different individual citation distributions are in a given field, and whether –in spite of such differences– their average characteristics are similar across fields. Secondly, we focus on the macro-micro relationship within and between fields. That is, we investigate whether the skewness of the citation distribution of all articles in a given field can be explained in terms of the characteristics of the individual citation distributions of the authors that make up the field in question. Furthermore, we study whether the macro-micro relationship between field and individual citation distributions is similar across fields.

We begin with 15 million distinct articles indexed by Clarivate Analytics, formerly the IP & Science business of Thomson Reuters, and published by 18.5 million distinct authors in the period 2000-2016. Applying a variable citation window from the publication year until 2016, these articles receive 231 million citations. To pursue our study, we must confront the following four methodological problems: (i) the classification of articles into scientific fields; (ii) the identification of the author(s) of each article, (iii) the allocation of authors to fields, and (iv) the attribution of individual responsibility in cases of multiple authorship. We solve these problems as in Ruiz-Castillo & Costas (2014a) –RCC hereafter. (i) We follow a multiplicative strategy to solve the problem of the assignment of a large percentage of articles to several WoS (Web of Science) subject categories. (ii) WoS subject categories are aggregated into 29 broad scientific fields. (iii) A researcher who writes articles in several fields is treated as a set of independent, different authors in the respective fields. (iv) Finally, the problem of multiple authorship is solved in a multiplicative manner. Thus, we end up with a dataset consisting of

35.1 million authors, 105.3 million articles, and 2,102 million citations. In comparison, this dataset is approximately twice as large as the one used in RRC.¹

In RCC, we only studied two characteristics for all authors: their individual productivity, measured by the number of articles *per capita*, and their citation impact, measured by their mean citation rate. It should be noted that, since our aim is the skewness of entire citation distributions at the individual level, in this paper we must ignore authors with few publications. That is, we must restrict our attention to researchers with a citation distribution of a certain size. Specifically, we focus on *very productive* authors with a number of publications above a certain relative benchmark that takes into account that the average number of articles per author varies widely across fields. We also consider *merely productive* authors, defined as those who publish at least five articles during our 16-year period. On average over all fields, these two types of productive authors only represent 5.2% and 9.4% of the population, but are responsible for 38.0% and 47.9% of all publications.

Turning now to field citation distributions, previous research based on large datasets of publications has yielded two important results: independently of the granularity of the classification system used and the length of the citation window, (i) field citation distributions are highly skewed, and (ii) the degree of skewness is very similar across fields (Schubert *et al.*, 1987, Glänzel, 2007, Radicchi *et al.*, 2008, Albarrán and Ruiz-Castillo, 2011, Albarrán *et al.*, 2012, Radicci & Castellano, 2012, Li *et al.*, 2013, and Ruiz-Castillo & Waltman, 2015). However, it should be emphasized that these results refer to a dataset of articles that, by ignoring authors, do not need to contend with the attribution of individual responsibility in cases of co-authorship. Fortunately, we find that the characteristics of field citation distributions before and after addressing the multiple authorship problem are very similar indeed.

The remainder of the paper is organized into five Sections and four appendices. Section II presents the data, the notation, and some descriptive statistics. In order to assess the reliability of our dataset, in Appendix I we compare some of its key characteristics with those of the RCC dataset. Section III contains the within- and between-field results concerning individual citation distributions

¹ Specifically, RCC begin with 7.7 million distinct articles published in the period 2003-2011 by 9.6 million distinct authors, and end up with 17.2 million authors and 48.2 million articles.

among very productive authors. After Appendix II establishes that the characteristics of field citation distributions are independent of the co-authorship problem, Section IV presents the within- and between-field results concerning the macro-micro relationship between field and individual citation distributions with the help of some illustrative examples presented in Appendix III. Appendix IV studies the robustness of our results when we consider merely productive authors. Section V discusses the main findings of the paper, while Section VI offers some concluding comments.

II. DATA, DESCRIPTIVE STATISTICS, AND METHODS

II.1. The construction of the dataset

Since we wish to address a homogeneous population, we only study research articles published in academic journals or, simply, *articles*.² We begin with a large sample, consisting of 15,047,087 distinct articles published in the period 2000-2015. Since the construction of the data set follows RCC exactly, in this Sub-section we briefly discuss the solutions we have adopted for coping with the four methodological problems mentioned in the Introduction. A more detailed justification can be found in our previous contribution.

1. There are two main approaches to tackling the problem created by the assignment of publications to two or more journal subject categories, or simply categories, in WoS datasets. The first is a fractional strategy, where each publication is fractioned into as many equal pieces as necessary with each piece assigned to its corresponding category. The second approach follows a multiplicative strategy in which each paper is counted as many times as necessary in the several categories to which it is assigned. In this way, the space of articles is expanded as much as necessary beyond the initial size in what we call the *extended count*. Fortunately, previous results indicate that for many purposes, journals assigned to a single or several subject categories share similar characteristics, so that the choice between the two strategies is not that crucial (see RCC for references). In this paper we follow a multiplicative approach. Consequently, the number of articles in the extended count, denoted by N , is 21,202,678, or

² Following Waltman & van Eck (2013a, b), we exclude publications in local journals, as well as magazine and trade journals.

34.1% larger than the number of distinct articles. We adopt the classification system used in RCC, consisting of 30 broad fields, which is based in a partition of scientific activity into 35 fields introduced by Tijssen *et al.* (2010) and used in other publications (see RCC for references). However, in contrast with RCC, here we remove the heterogeneous ‘Multidisciplinary journals’ category by proportionally classifying these publications in the fields of the cited references. Therefore, we distinguish between 29 fields.³

2. For the assignment of articles to individual authors, we use the author disambiguation algorithm generated by Caron & van Eck (2014) for large bibliometric databases, whose main features are discussed in RCC. Overall, there are 18,526,987 distinct researchers associated to the 15 million distinct articles of the dataset.

3. For the purpose of analyzing the characteristics of individual citation distributions in a given field, as we do in this paper, researchers who write articles in several fields should be treated as independent, different authors in their respective fields. Therefore, the number of authors, denoted by I , goes up to 35,057,987 individuals, an 89.2% increase relative to the original number of distinct authors.

4. A fundamental difficulty in the study of scientists’ productivity is the definition of the individual contribution to an article in a world dominated by co-authorship in all fields (see the references in RCC, as well as the recent contributions by Waltman & Van Eck, 2015, and Perianes-Rodriguez & Ruiz-Castillo, 2015a). In this paper, we use a multiplicative strategy in which any article co-authored by two or more scholars is wholly assigned as many times as necessary to each of them. Of course, this means that the set of articles actually studied increases quite dramatically: the total number of articles in what we call the *double extended count*, denoted by N^D , becomes 105,289,384, or seven times larger than the number of distinct articles. The total number of citations in the double extended count

³ It is not claimed that this scheme provides the best possible representation of the structure of science. It is rather a convenient simplification for the discussion of field comparability issues in this paper.

is 2,102 million, or nine times larger than the initial number of citations for the 15 million distinct articles.

II. 2. Descriptive statistics

We denote by N_f and N_f^D the number of articles in each field in the extended and the double extended count, so that $\sum_f N_f = N = 21.1$ and $\sum_f N_f^D = N^D = 105.3$ million articles. Similarly, we denote by I_f the number of authors in each field, so that $\sum_f I_f = N = 35.1$ million authors. Table 1 presents the distribution of articles by field in the extended and double extended counts, as well as the distribution of authors by field, whereas Table 2 includes some evidence on the variability of co-authorship patterns within and between fields.

In this paper, the within- and between-field variation for all magnitudes is measured by the coefficient of variation (CV hereafter) over the 29 fields. The CV is defined as the ratio of the standard deviation over the mean. There is no generally agreed upon criterion in statistics concerning when a CV is “large” or “small”, possibly because this distinction is context dependent. Although any reader is free to apply a different criterion, in this paper we will use the following convention. We say that the within- or between-field variability of any characteristic is

- “Small”, if $CV \leq 0.10$, meaning that the dispersion of this characteristic measured by the standard deviation is smaller than or equal to 10% of the mean.
- “Intermediate”, if $0.10 < CV \leq 0.30$.
- “Large”, if $0.30 < CV \leq 0.60$.
- “Very large”, if $CV > 0.60$.

Tables 1 and 2 around here

The following three points should be noted. Firstly, according to the number of authors, fields can be classified into three groups (see column 3 in Table 1). (i) There are five fields with more than three million authors with at least 9.9% of the total number of authors (Clinical Medicine; Biomedical Sciences; Basic Life Sciences; Physics & Materials Science, and Chemistry & Chemical Engineering).

The largest is Clinical Medicine that has 6.4 million authors and 18.2% of the total. (ii) There are eleven intermediate fields with 528,000 to 1,815,000 authors, or 1.5% to 5.2% of the total. (iii) The remaining fifteen fields have fewer than 364,000 authors or 1.3% of the total. The smallest is Information & Communication Sciences with 94,965 authors, or 0.3% of the total. In view of this partition, the dispersion of field sizes is very large: the CV over the 29 fields is 1.3.⁴

Secondly, the average number of authors per article is 4.2 (column 1 in Table 2). However, the between-field variation is quite large: the coefficient of variation over the 29 fields is 0.46, and the range of variation goes from 2.2 and 2.3 authors per article in Mathematics and Management & Planning, up to 6.0 and 12.1 in Instruments & Instrumentation and Astronomy & Astrophysics. On the other hand, the within-field variation is very large indeed (column 2 in Table 2), ranging from a coefficient of variation of 0.51 in General & Industrial Engineering up to 8.50 and 8.54 in Physics & Materials Science and Astronomy & Astrophysics. Finally, the maximum number of authors per article (column 3 in Table 2) exhibits a phenomenal range of variation from 57 and 73 in General & Industrial Engineering and Management & Planning, up to 3,195 and 5,109 in Astronomy & Astrophysics and Physics & Materials Science.

Thirdly, comparing the percentage distributions in columns 2 and 4 in Table 1, we observe that some small fields (such as General & Industrial Engineering, Instruments & Instrumentation, and Energy Science & Technology) and some large ones (Clinical Medicine, Biomedical Sciences, and Basic Life Sciences) have relatively more authors than articles. The opposite is the case for some small fields (Mathematics; Astronomy & Astrophysics, and Economics & Business) as well as Physics & Materials Science. In turn, the increase in the total number of articles in the double extended count varies a lot across fields. Comparing columns 2 and 6 in Table 1, we observe that the percentage of the number of articles in the double extended count is greater than in the original count in only seven fields whose mean number of authors per article (column 1 in Table 2) is well above the average for all fields

⁴ Between-field variation when size is measured as the number of articles is also very high indeed: in these cases the coefficients of variation over the 29 fields are 1.2 in the extended count and 1.4 in the double extended count.

(Astronomy & Astrophysics; Basic Life Sciences; Basic Medical Sciences; Biomedical Sciences; Clinical Medicine; Instruments & Instrumentation, and Physics & Materials Science).

The construction of large datasets for the study of the research performance of individual authors is a daunting empirical exercise. As we have seen in Section II.1, our dataset, which has been constructed with the same criteria used in RCC, ends up being twice as large as the dataset used in that contribution. Therefore, it seems convenient to assess the reliability of the data used in this paper by comparing some characteristics of the two datasets. To facilitate the reading of the paper, this exercise is included in Appendix I. The high degree of consistency observed for all characteristics demonstrates the reliability of the present construction: having followed the same criteria in both cases, the two datasets seem to reflect the same world.

II. 3. Methods: the CSS approach

It is useful to provide a brief description of the CSS approach that will be repeatedly used in the sequel. Let N be the number of elements in any citation distribution X , indexed by $k = 1, \dots, N$, so that $X = (x_1, \dots, x_k, \dots, x_N)$ where x_k is the number of citations received by publication k . For later reference, let $I(X)$ be the total number of citations in X , i.e. $I(X) = \sum_k x_k$. Two *characteristic scores* will be used: m_1 , the mean of X , and m_2 , the second mean of X , or the mean of all elements in X with x_k greater than m_1 . Using m_1 and m_2 , we define the following three categories: category I consists of the proportion of *poorly cited* publications in X with x_k smaller than or equal to m_1 ; category II consists of the proportion of *fairly cited* publications in X with x_k greater than m_1 and smaller or equal to m_2 , and category III consists of the proportion of *remarkably* or *outstandingly cited* publications in X with x_k greater than m_2 . CSS results consists of six numbers, (p_1, p_2, p_3) and (s_1, s_2, s_3) , where $p_j, j = 1, 2, 3$ is the proportion of publications in X in categories I, II, and III, and $s_j, j = 1, 2, 3$ is the share of $I(X)$ accounted by categories I, II, and III. In many cases, we will typically have CSS results at the field level, say (p_{f1}, p_{f2}, p_{f3}) and (s_{f1}, s_{f2}, s_{f3}) , for $f = 1, \dots, 29$. We denote the average of the CSS results over the 29

fields by capital letters, i.e. (P_1, P_2, P_3) and (S_1, S_2, S_3) . As before, the between-field variation of these magnitudes is measured by means of the CV over the 29 fields.

III. WITHIN- AND BETWEEN-FIELD RESULTS CONCERNING INDIVIDUAL CITATION DISTRIBUTIONS

III.1. Very productive authors

In each field $f = 1, \dots, 29$, let $c_f(i)$ be the citation distribution of author i with $i = 1, \dots, I_f$, where I_f is the number of authors in field f . For each i and f , let $n_f(i)$ be the size of $c_f(i)$, i.e. the number of articles of author i in field f . For each f , the first and second means of distribution $\{n_f(i), i = 1, \dots, I_f\}$ are presented in Table 3. Note that in all fields, the average number of articles per author is very low indeed (column 1 in Table 3). As can be observed in Table 4, this is explained by the large percentage of authors with very few publications. On average, authors with a single publication in our 16-year publication period represent 71.3% of the total, whereas more than 90% of all authors have less than five publications. Possibly, the decision to treat researchers with publications in two or more fields as different authors increases the percentage of individuals with few publications in their minority field(s).⁵ Nevertheless, the low CV in columns 1 to 3 in Table 4, representing authors with less than five articles, indicates the existence of a surprising similarity across fields as far as low publication rates are concerned.⁶

Tables 3 and 4 around here

This poses a problem for the analysis of individual citation distributions: we are bound to restricting our attention to a very small percentage of authors with citation distributions of a certain minimum size. At any rate, how should we determine such a minimum size in each field? Note that differences in production practices at high publication rates give rise to considerable between-field variation in mean individual productivity: the CV over the 29 fields in column 1 in Table 3 is 0.61.

⁵ Moreover, as indicated in RCC, the Caron & van Eck (2014) name disambiguation algorithm promotes precision over recall. Thus, it should be acknowledged that when there is limited information to cluster the publications of a certain author, the algorithm may occasionally split the *oeuvre* of an author into clusters with only one publication.

⁶ The large percentage of authors with a single publication, the low between-field variation of this amount, as well as the low average number of articles per author are also observed in Table 1 in RCC.

Thus, for example, mean individual productivity is equal to 1.6 and 1.7 articles per author in Information & Communication Sciences and Social & Behavioral Sciences, while this magnitude is 3.3, 4.5, and 10.6 in Clinical Medicine, Physics & Materials Science, and Astronomy & Astrophysics. Therefore, it is natural to search for a benchmark that varies across fields.

In this vein, we define *very productive* authors in each field as those with a number of articles greater than the second mean in the distribution $\{n_i(i), i = 1, \dots, I_f\}$ (column 3 in Table 3). We denote by I_f^* the number of very productive authors in field $f = 1, \dots, 29$. Although the percentage of very productive authors is generally very small, they typically account for a relatively large percentage of all articles in the double extended count. In Agriculture and Food Science, for example, only 4.3% of all authors with a number of publications equal to or greater than nine will be considered very productive. However, this small percentage is responsible for 36.0% of all articles in the field. On average, very productive authors are 5.3% of the total, publish eleven or more articles *per capita*, and are responsible for 38.0% of all articles (for details, see columns 3 and 6 in Table AI.2 in Appendix I).

III.2. Within- and between-field variation of individual citation distributions for very productive authors

Recall that $n_i(i)$ is the size of the individual citation distribution $c_i(i)$. For every very productive author i in field f , let $m_{f1}(i)$ and $m_{f2}(i)$ be the first and second means of $c_i(i)$. For every field, denote the average of these three quantities over the I_f^* authors by *Mean-size_f*, m_{f1} and m_{f2} ; that is, *Mean-size_f* = $\sum_i n_i(i)/I_f^*$, and $m_{fj} = \sum_i m_{fj}(i)/I_f^*$ for $j = 1, 2$, where the sum in these expressions goes over the I_f^* very productive authors. In turn, the average of *Mean-size_f*, m_{f1} and m_{f2} over the 29 fields are denoted by *Mean-size*, M_1 and M_2 ; that is, *Mean-size* = $\sum_f \text{Mean-size}_f/29$, and $M_j = \sum_f m_{fj}/29$ for $j = 1, 2$. The results for all these magnitudes are in Table 5. Wide differences in citation impact among very productive authors will manifest themselves in large CVs of their mean citation rates. This is exactly what we observe for all fields in columns 4 and 6 in Table 5. On the other hand, large CVs over the 29 fields reflect large differences in citation practices across fields.

Table 5 around here

Of course, given the wide differences between authors' citation impact in each field, and in production and publication practices across fields, large within- and between-field differences in mean citation rates come as no surprise. The key question for our purposes concerns the skewness of individual citation distributions. For every very productive author i in field f , we denote the CSS results by $(p_{f1}(i), p_{f2}(i), p_{f3}(i))$ and $(s_{f1}(i), s_{f2}(i), s_{f3}(i))$, where $p_{fj}(i)$ is the proportion of articles in distribution $c_j(i)$ in category $j = \text{I, II, III}$, and $s_{fj}(i)$ is the share of total citations in distribution $c_j(i)$ accounted for by category $j = \text{I, II, III}$. For every field, we denote the average of these individual results over the I_f^* authors by (p_{f1}, p_{f2}, p_{f3}) and (s_{f1}, s_{f2}, s_{f3}) , that is, for every j ,

$$p_{fj} = \sum_i p_{fj}(i) / I_f^*, \quad (1)$$

and

$$s_{fj} = \sum_i s_{fj}(i) / I_f^*, \quad (2)$$

where the sum in expressions (1) and (2) goes over the I_f^* very productive authors. The corresponding CVs over the I_f^* very productive authors are denoted by $(cv_{f1}, cv_{f2}, cv_{f3})$ and $(cv_{f4}, cv_{f5}, cv_{f6})$, respectively. The results for (p_{f1}, p_{f2}, p_{f3}) , $(cv_{f1}, cv_{f2}, cv_{f3})$, (s_{f1}, s_{f2}, s_{f3}) and $(cv_{f4}, cv_{f5}, cv_{f6})$ in each field and are in columns 1 to 12 in Table 6. In turn, the average of p_{fj} and s_{fj} over the 29 fields for $j = \text{I, II, III}$ are denoted by P_j and S_j , respectively, whereas the average of $(cv_{f1}, cv_{f2}, cv_{f3})$ and $(cv_{f4}, cv_{f5}, cv_{f6})$ over the 29 fields are denoted by $(CV_{f1}, CV_{f2}, CV_{f3})$ and $(CV_{f4}, CV_{f5}, CV_{f6})$. The results on (P_1, P_2, P_3) , $(CV_{f1}, CV_{f2}, CV_{f3})$, (S_1, S_2, S_3) and $(CV_{f4}, CV_{f5}, CV_{f6})$, as well as their corresponding CVs over the 29 fields, are in the last two rows in Table 6. Finally, the information concerning (p_{f1}, p_{f2}, p_{f3}) for $f = 1, \dots, 29$ is illustrated in Figure 1 where fields are ordered by p_{f1} .

Table 6 and Figure 1 around here

There are three main results. Firstly, as expected, the CVs in columns 4 to 6 and 10 to 12 in Table 6 indicate that the skewness of individual citation distributions exhibit a very large within-field variability. Secondly, recall that uniform or normal distributions would yield percentages of articles in categories I, II, and III equal to 50%, 25%, and 25% in the first case, and 50%, 28.8% and 21.2% in the second one. However, on average over all fields, mean citation rates are approximately 16 points above the median, and less than 13% of articles in category III account for almost 43% of all citations. In brief, on average individual citation distributions within each field are considerably skewed. Thirdly, judging from the size of CVs over the 29 fields in columns 1 to 3 and 7 to 9, the degree of skewness across fields is very similar indeed. Figure 1 clearly illustrates this important result.

IV. WITHIN- AND BETWEEN-FIELD RESULTS CONCERNING FIELD CITATION DISTRIBUTIONS

IV.1. The extended *versus* the double extended count for all authors

In this Section we investigate the connection between the skewness at the individual and field levels. But to do this, we must determine which type of field citation distribution we wish to select: field citation distributions in the extended count or in the double extended count. In order to facilitate the reading of the text, a detailed discussion of this issue is relegated to Appendix II. Fortunately, the difference between the CSS results for field citation distributions in both counts is so small that, for all practical purposes, we may continue the analysis focusing on either case. In what follows, we will restrict ourselves to the double extended count.

IV.2. The gap between the skewness of the field citation distribution and the average skewness of the individual citation distributions for very productive authors

In Section III we considered very productive authors with citation distributions of a certain minimal size. In principle, it is natural to focus on field citation distributions consisting of articles published only by authors of this type. However, it is also important to consider the unrestricted field citation distributions of articles published by all authors. Although, as we will see, the difference is relatively small, we first study the field citation distributions consisting of articles published by very productive authors in the double extended count.

The information for the first and the second means for these distributions, denoted by μ_{fj}^{D*} for $j = 1, 2$, is in Table 7. It is interesting to compare the means of field citation distributions in the double extended count in Table 7 with the average of the mean citations of very productive authors (columns 3 and 5 in Table 5). For any f , the individual citation distributions of very productive authors form a partition of the corresponding field citation distribution. Consequently, for the first mean we have:

$$\mu_{f1}^{D*} = \sum_i w_{f1}^*(i) m_{f1}(i),$$

where $w_{f1}^*(i) = n_f(i)/N_f^{D*}$ is the proportion of the publications of author i , $n_f(i)$, with respect to the total number of publications in the double extended count, i.e. $N_f^{D*} = \sum_i n_f(i)$, where all summations are over the I_f^* very productive authors in the field. Instead, the average of the mean citations of very productive authors is

$$m_{f1} = [\sum_i m_{f1}(i)]/I_f^*.$$

Therefore, as long as $m_{f1}(i)$ tends to increase with $n_f(i)$, we expect $\mu_{f1}^{D*} > m_{f1}$. This is what we find for every f when we compare column 1 in Table 7 with column 3 in Table 5. Hence, on average over the 29 fields, we have $(\sum_f \mu_{f1}^{D*})/29 = 17.3 > (\sum_f m_{f1})/29 = 16.4$. However, the differences are relatively small, indicating that $m_{f1}(i)$ does not increase much with $n_f(i)$, i.e. that the scaling relationship between mean citations and the number of scientific publications among very productive authors is rather weak. Similar results hold for the second means.

Table 7

We denote by $(P_{f1}^*, P_{f2}^*, P_{f3}^*), (S_{f1}^*, S_{f2}^*, S_{f3}^*), f = 1, \dots, 29$, the CSS results for field citation distributions of articles published by very productive authors in the double extended count. In turn, we denote by (P_1^*, P_2^*, P_3^*) and (S_1^*, S_2^*, S_3^*) the average of these quantities over the 29 fields. The CSS results are presented in Table 8.

Table 8 and Figure 2

We first note that, except for the proportion of articles in category III, the small CVs for the other five parameters over the 29 fields indicate that the skewness of field citation distributions is very similar indeed. This is clearly illustrated in Figure 2 representing the proportion of articles in the three categories ordered by P_{f1}^* .

Finally, we arrive to the most important comparison in this Section between the average skewness of individual citation distributions for very productive authors in a given field, i.e. (p_{f1}, p_{f2}, p_{f3}) and (s_{f1}, s_{f2}, s_{f3}) in Table 6 and Figure 1, and the skewness of the field citation distribution consisting of the articles these authors produce, i.e. $(P_{f1}^*, P_{f2}^*, P_{f3}^*)$ and $(S_{f1}^*, S_{f2}^*, S_{f3}^*)$ in Table 8 and Figure 2. The key observation is that the average skewness of the individual citation distributions in each field is considerably smaller than the skewness of field citation distributions. In Agriculture and Food Science, for example, in the first case the mean is 15.8 points greater than the median and 13.3% of highly cited articles account for 43.7% of all citations, whereas in the second case the mean is 23.8 points greater than the median and only 6.9% of highly cited articles account for 43.2% of all citations.

As we will presently see, a possible explanation is the following. Together with the skewness of individual citation distributions, the skewness of a field citation distribution may essentially arise from two additional factors: differences between individual productivity, measured by the number of articles per author, and differences between individual mean citation rates.

The following examples in Appendix III illustrate the situation. In the first place, differences in individual productivity in a given field may have no impact on the skewness of the field citation distribution. For example, if all individual citation distributions in a field have the same first and second mean and the same skewness, the average skewness of individual citation distributions will coincide with the skewness of the corresponding field citation distribution regardless of any difference in the size of individual citation distributions. However, when individual citation distributions have different skewness, within-field differences in individual productivity may affect the skewness of the field citation distribution. Example 1 in Appendix III illustrates this case for two individuals in a single field with the

same first mean. In the second place, when individuals are equally productive we may still have a skewness gap. Example 2 in Appendix III considers two individuals in a single field with citation distributions not only of equal size but also equal skewness. Naturally, in this case the average skewness coincides with the skewness of the individual citation distributions. However, the difference in individual mean citation rates causes the average skewness to be smaller than the skewness of the field citation distribution. More generally, in practice it is likely that individuals will have different number of publications, different means, and different skewness –a case illustrated in Example 3 in Appendix III.

Given the between-field results at the individual and field levels, the gap between the skewness of any field citation distribution and the average skewness of individual citation distributions in that field is of the same order of magnitude. Therefore, for simplifying purposes we will restrict ourselves to the skewness results for the average over the 29 fields, namely, (P_1, P_2, P_3) and (S_1, S_2, S_3) versus (P^*_1, P^*_2, P^*_3) and (S^*_1, S^*_2, S^*_3) , which are reproduced in rows I and II in Table 9.

Table 9 around here

Our aim is the following. We want to establish that the skewness gap between rows I and II in Table 9 can be mostly explained by differences in individual productivity and individual mean citations. We begin with the first factor. If the only source of the skewness gap were differences in individual productivity, then a solution would be to consider the weighted average skewness of individual citation distributions, with weights equal to the proportion that the number of articles of each author represents with respect to the total number of articles in the field. In this case, as illustrated in Example 1, the skewness gap would disappear. Therefore, given the CSS individual results $(p_{j1}(i), p_{j2}(i), p_{j3}(i)), (s_{j1}(i), s_{j2}(i), s_{j3}(i)), i = 1, \dots, I_j^*$ in every field, instead of the simple averages in equations 1 and 2 in Section III.2, we would estimate, for every $j = 1, 2, 3$,

$$p'_{jf} = \sum_i w_j^*(i) p_{jf}(i), \quad (3)$$

and

$$s'_{jf} = \sum_i w_j^*(i) s_{jf}(i), \quad (4)$$

where, as before, $w_f^*(i) = n_f(i)/N_f^{D*}$, $N_f^{D*} = \sum_i n_f(i)$, and the sum in expressions (3) and (4) goes over the I_f^* very productive authors. The average of $(p'_{f1}, p'_{f2}, p'_{f3})$ and $(s'_{f1}, s'_{f2}, s'_{f3})$ over all fields are denoted by (P', P'_2, P'_3) and (S', S'_2, S'_3) . The results for $(p'_{f1}, p'_{f2}, p'_{f3})$ and $(s'_{f1}, s'_{f2}, s'_{f3})$ in all fields are in Table A in Appendix III, while the results for (P', P'_2, P'_3) and (S', S'_2, S'_3) are reproduced in row III in Table 9.

Recall that the within-field variability of individual productivity is rather high (column 2 in Table 5). Therefore, as long as the skewness of individual citation distribution for author i increases with $n_f(i)$, we expect that (p'_{fj}, s'_{fj}) , $j = 1, 2, 3$, reflects a greater skewness than (p_{fj}, s_{fj}) , $j = 1, 2, 3$. However, by comparing the field results in Table 6 and Table A in Appendix III, we observe that the skewness of the weighted average is only slightly greater than the skewness of the simple average. This indicates that the skewness of citation distributions among very productive authors does not vary much with individual productivity. Therefore, we conclude that differences in individual productivity play a minor role in explaining the skewness gap in each field. Given the similarity across fields (see the low CVs in row III in Table 9), this is also what we observe by comparing rows I and III in Table 9.

There is another way of studying the role of differences in individual productivity. We can estimate the skewness of field citation distributions controlling for the within-field differences in individual productivity by equalizing the number of articles per author. Since the CSS technique is size-independent, the skewness of individual citation distributions is preserved. As illustrated in Example 1, if the only source of the skewness gap were differences in individual productivity, then after this normalization the skewness gap would disappear. In our case, we proceed by weighting every article of an individual i in field f by the quantity $[n^*/n_f(i)]$, where n^* is an arbitrary amount. In this way, individual productivity in each field becomes equal to n^* . The CSS results for the average over the 29 fields appear in row IV in Table 9 (detailed field results are available on request). Given the small role that differences in individual productivity have in explaining the skewness gap in each field, we expect

minor differences in the skewness in each field. This is exactly what we find when we compare rows II and IV in Table 9.

Next, we must study the role of within-field differences in mean citations in generating the skewness gap. For that purpose, we can estimate the skewness of field citation distributions controlling for these differences by equalizing the first mean of all authors in a given field. Since the CSS technique is scale-independent, the skewness of individual citation distributions is preserved. Instead, as a consequence of the equalization of individual mean citations, the skewness of the new field distribution should be reduced. The size of the reduction in skewness will inform us of the role of differences in mean citations in explaining the skewness gap. As Example 2 illustrates, when the main difference between authors is the difference in the first mean citation, this procedure completely eliminates the skewness gap. However, when authors also differ in their second mean citation by a sufficient amount - as it is the case in Example 3- the gap does not completely vanish.

For our dataset, the task is to explain the skewness gap at the field level between Table 8 and Table A in Appendix III, and between rows I and IV in the aggregate, *once* we have controlled the skewness of field citation distributions by differences in individual productivity. Thus, after weighting every article of an individual i in field f by the quantity $[n_*/\eta_f(i)]$, we now multiply the citation count $c_{jk}(i)$ for all k by the quantity $[\mu_*/\mu_{f1}(i)]$, where μ_* is an arbitrary amount. In this way, individual mean citations in each field become equal to μ_* .⁷ Note that the total number of articles in each field will be the product of I_f^* and n_* , the field mean citation will be equal to μ_* , and the percentage of articles in category I, as well as the percentage of total citations accounted by articles in this category, will coincide with the average of the corresponding individual percentages. The results for all fields are in Table B in Appendix III, while the results for the average over the 29 fields are reproduced in row V in Table 9. The resulting skewness after the double normalization is called the *basic skewness* of field citation distributions.

⁷ This normalization can only be applied for authors with a positive mean citation. However, very productive authors receiving no citations only represent 0.022% of the total (details by field are available on request).

Note that the within-field variability of the first mean among very productive authors is very large (column 3 in Table 5). Consequently, by comparing Tables A and B in Appendix III we observe that, as a consequence of the second normalization the skewness of field citation distributions is greatly reduced. Given the similarity across fields (see the low CVs in row V in Table 9), this is also the case when comparing rows II and IV in Table 9. The minimal resulting gap between the basic skewness in any field (row V) and the weighted or unweighted average skewness at the individual level (rows II and I) is due to differences in the second mean of the individual normalized citation distributions that cannot be eliminated without changing the original individual skewness.

The conclusion is that, in every field, the skewness of the field citation distribution can be essentially accounted for by the skewness of the average of individual citation distributions and the within-field differences in the size and the mean of the latter. Differences in individual mean citations are much more important than differences in individual productivity in explaining the initial skewness gap. Furthermore, the relative order of magnitude of these two sources of skewness is the same for all fields.

IV.3. The gap between the skewness of the field citation distribution and the average skewness of the individual citation distributions for merely productive authors

Very productive authors have been defined using a relative benchmark that takes into account field differences in production practices. An alternative is to define *merely productive* authors as those who publish at least five articles in the 2000-2015 period. The percentages of merely productive authors in each field are in column 4 in Table 4. Except for three fields where very productive authors publish only four or more articles –Information & Communication Sciences, Social & Behavioral Sciences, and Sociology & Anthropology–, the percentage of merely productive authors is greater than the percentage of very productive authors. Consequently, in 26 fields merely productive authors are responsible for greater percentages of articles than very productive researchers. In Agriculture and Food Science, for example, merely productive authors represent 9.1% of all authors and are responsible for 48.5% of all

articles in the field. On average, merely productive authors represent 9.4% of all authors and are responsible for 47.9% of all articles (for details, see column 6 in Table 4).

In order to facilitate the reading of the text, the CSS analysis of individual citation distributions for merely productive authors is relegated to Appendix IV. Interestingly, the results are very similar to those obtained for very productive authors. Three points should be noted. Firstly, individual citation distributions are slightly less skewed for merely productive authors than for very productive ones. Secondly, field citation distributions consisting of the articles published by merely productive authors turned out to be essentially as skewed as field citation distributions for very productive authors in Table 8. Thirdly, as a consequence of these two facts, the gap between the skewness of field citation distributions and the average skewness of the individual citation distributions for merely productive authors in each field is slightly greater than for very productive ones. However, controlling for differences in individual productivity and individual mean citations within each field, the skewness of field citation distributions for merely productive authors is essentially the same as for very productive authors.

In brief, the only difference between the two cases is that, since there are more authors involved, the within-field variation of individual productivity and individual mean citations is greater for merely productive authors than for very productive ones. Consequently, as we have seen, the gap between the skewness of field citation distributions and the average skewness of the individual citation distributions for merely productive authors in each field is slightly greater than for very productive ones. However, controlling for such differences, we arrive to a very similar basic skewness in each field. The consequence of this result is very helpful: for all practical purposes, our analysis can be equally conducted in terms of the two notions of productive authors. Generally, we will restrict our attention to very productive authors.

IV.4. The gap between the skewness of field citation distributions for all authors and the average skewness of the individual citation distributions for very productive authors in each field

As indicated before, it is also important to consider the unrestricted field citation distributions of articles published by all authors. Recall that the CSS results on the skewness of field citation distributions in this case are in Table AII.3 in Appendix II. CSS results for the average over the 29 fields are reproduced in row VI in Table 9.

Our task is to explain the skewness gap between rows I and VI in Table 9, which is slightly greater than the gap between rows I and II for very productive authors. In line with our previous argument, the explanation is that the within-field variation of individual productivity and individual mean citations is greater for all authors than for very productive authors. For individual productivity, this is exactly what we find when we compare column 2 in Table 3 and column 2 in Table 5. For individual mean citations, this is also what we find when we compare column 1 in Table AI.1 and column 4 in Table 5.

V. DISCUSSION

It will be useful to organize the discussion of the results around the following three issues: patterns of individual citation distributions, patterns of field citation distributions, and the relationship between the two.

V.1. Patterns of individual citation distributions

Within each field, individual scientists are extremely heterogeneous. In our dataset, we observe a well-known large within-field variation in the following three dimensions: individual productivity, measured by the number of articles; individual citation impact, measured by mean citation rates, and the pattern of co-authorship, measured by the number of authors per article.⁸ This large variability in individual productivity and citation impact is also present for very productive and merely productive authors).

In addition, we have investigated the skewness characteristics of individual citation distributions for very productive and merely productive authors that represent, on average, 5.2% and 9.4% of the

⁸ The within-field variability in individual productivity, individual citation impact, and the pattern of co-authorship are also characteristics of RCC's dataset.

total number of authors. Focusing on the former, two results stand out. Firstly, not surprisingly, we also find a large within-field variability for the CSS results of individual citation distributions. Secondly, the average of the individual CSS results in each field exhibit a clear skewness pattern. Furthermore, judging from the small size of CVs over the 29 fields, in spite of wide differences in production and citation practices across fields this skewness pattern is very similar for all of them. Note that this is at variance with the results in Costas *et al.* (2009), where field characteristics influence the research performance of individual authors in the sense that the size-dependent cumulative advantage for receiving citations tends to be larger in low citation-density fields. For later reference, the average results over the 29 fields for very productive authors are illustrated in Figure 3.

Figure 3 around here

It is important to emphasize that the CSS results concerning the within- and between-field variation just summarized, are only slightly less pronounced for merely productive authors (Table AIII.2). Essentially, this indicates that, within each field, the average CSS results for individual citation distributions conditional on the number of articles per person do not change much as we increase the authors' individual productivity.

The similarity of the average characteristics of individual citation distributions across fields has important conceptual and practical consequences: to explain the skewness of individual citation distributions we do not need a different model for each field. On the contrary, since a certain degree of average skewness seems to be generic, all we need is a single model for individual researchers in any scientific field. A good example can be found in Sinatra *et al.* (2016), where an author's high-impact work, resulting from a combination of her ability to take advantage of the available knowledge and a random element, is randomly distributed within her career.

V.2. Patterns of field citation distributions

Previous results for large WoS datasets for classification systems at different aggregation or granularity levels with a fixed five-year citation window indicate that field citation distributions are highly skewed, and that between-field variability is very reduced. For example, as documented in Li *et*

al. (2014), CSS results evolve smoothly during the 1980-2004 period. As the citation window increases from seven years for documents published in 2004 up to 31 years for documents published in 1980, sub-field citation distributions become somewhat more skewed (the increase in the degree of skewness with the length of the citation window is amply documented in Katz, 2016). The evidence for more than two decades is summarized in Li *et al.* (2014) by the following percentages of publications and total citations in categories I, II, and III: (70.9, 20.4, 8.7) and (22.7, 32.7, 44.6).⁹

In the extended count in our dataset, these percentages are (74.9, 18.6, 6.5) and (21.5, 32.0, 46.5), whereas in the double extended count the averages over the 29 fields are (76.8, 17.7, 5.5) and (21.6, 31.3, 47.1). As we have seen in Section Appendix II, the similarity between the extended and the double extended counts indicate that, essentially, the skewness characteristics of citation distributions of articles conditional on the number of authors do not change much as we vary the number of authors per publication. In any case, the skewness of field citation distributions in our dataset with a variable 16-year citation window is somewhat more pronounced, but still of a comparable order of magnitude, than the skewness documented in the previous contributions referred to.¹⁰

Publication and citation practices are very different across scientific disciplines at all aggregation levels. As a result, certain key statistics –such as the number of authors per paper, the first and second means of the number of publications per author and the mean citation rates, as well as the mean number of references or a variety of indicators of citation impact amply documented in the literature– exhibit a large range of variation across scientific fields. However, the reduced between-field variability of the CSS results presented in this paper and previous contributions indicate that the degree of skewness of field citation distributions is very similar indeed. Three comments are in order. Firstly, as emphasized in Albarrán *et al.* (2011) and Waltman *et al.* (2012), this similarity should not be confused

⁹ The situation closely resembles the one described in Albarrán *et al.* (2011) for 3.7 million articles with a common, five-year citation window published in 1998-2003 in a wide array of 219 WoS sub-fields. Similar results are also obtained for selected publication-level, algorithmically constructed classification systems consisting of 3.7 million articles classified into 2,272 and 4,161 significant clusters with at least 100 publications in Ruiz-Castillo & Waltman (2015).

¹⁰ As a matter of fact, our results are closer to those reported in Glänzel (2007) for 450,000 papers published in 1980 with a 20-year citation window, and classified into 12 major fields and 60 subfields according to the publication-level Leuven/Budapest classification system (Glänzel & Schubert, 2003). The proportion of the 450,000 publications in categories I, II, and III are (74.7, 18.5, 6.7).

with the universality claimed by Radicchi *et al.* (2008). Secondly, nevertheless, the similarity between field citation distributions opens the possibility of meaningful comparisons of citation counts across fields (Radicchi *et al.*, 2008, Glänzel, 2011, Radicci & Castellano, 2012, Crespo *et al.*, 2013, 2014, Li *et al.*, 2013, and Ruiz-Castillo, 2014). Thirdly, the similarity of the degree of skewness at the field level is at variance with the results in Van Raan (2006a, 2006b, 2008) concerning the scaling relationship between the number of citations and the number of scientific publications: in these contributions the size-dependent cumulative advantage for receiving citations tends to be larger in low citation-density fields (although this difference in the advantage between low and high field-citation-density for research groups is larger than the difference for individuals found in Costas *et al.*, 2009).

V.3. Relationships between individual and field citation distributions

It is useful to begin investigating the macro-micro relationship for very productive authors. The CSS results for the field citation distributions in the double extended count are in Table 8. The average results over the 29 fields, (75.1, 18.5, 6.4) and (22.3, 31.6, 46.1), are illustrated in Figure 4.

Figure 4 around here

The comparison between Figures 1 and 2 illustrates the extent of the skewness gap between the average of individual citation distributions and the citation distribution consisting of the articles published by very productive authors in each field, whereas the comparison of Figures 3 and 4 illustrates the skewness gap for the average over the 29 fields. However, a key result of this paper is that the skewness of field citation distributions can be explained in terms of the average skewness of individual citation distributions combined with the skewness in individual productivity and citation impact. Although the skewness gap is somewhat greater for merely productive authors, the explanation of the skewness of field citation distributions in terms of the average skewness of individual citation distributions combined with the differences in individual productivity and citation impact is maintained for merely productive authors.

We have established that differences in individual productivity are a very minor source of skewness at the field level, so that most of the skewness gap is accounted for by differences in

individual mean citations. What we cannot do in this framework is to measure the relative contribution to the total of the two main sources of skewness. That is, we cannot answer which part of the skewness of a field citation distribution can be attributed to the average skewness of individual citation distributions, and which part can be attributed to differences in individual mean citations. The reason, of course, is that the CSS technique is not decomposable by population subgroup. As a matter of fact, all real valued measures of skewness involve highly non-linear transformations of the data. Consequently, we do not know of any skewness index which is decomposable by population subgroup.

An alternative, of course, is to study the relationship between the citation inequality at the field and the individual level. There are size- and scale-independent citation inequality indices which are decomposable by population subgroup in the sense that for any partition of the population, for example the partition of a field into productive authors, the citation inequality at the field level can be expressed as the sum of a within-group and a between-group term. The within-group term is the weighted average of the citation inequality of individual authors, with weights equal to the proportion that the number of articles of any individual represent with respect to the total number of articles in the field. The between-group term is equal to the citation inequality of a field distribution in which the number of citations of any article is replaced by the mean citation of the author to which the article belongs.¹¹ In that case, it is possible to measure the relative contribution to the total of the within- and the between-group terms.

Coming back to the CSS approach, Thijs *et al.* (2017) indicate that the average CSS results for individual citation distributions in each field, (p_{f1}, p_{f2}, p_{f3}) and (s_{f1}, s_{f2}, s_{f3}) , constitute a natural benchmark for the assessment of the CSS results $(p_{fi}(i), p_{f2}(i), p_{f3}(i))$ and $(s_{fi}(i), s_{f2}(i), s_{f3}(i))$ for any individual author nevertheless i in that field. But the proximity which we have established between this average and the basic skewness of each field citation distribution, controlling for differences in individual productivity and individual mean citations, reinforces this choice of a benchmark.

¹¹ The Generalized Entropy (GE hereafter) family of inequality indices are the only measures of relative inequality that satisfy the usual properties required from any inequality index and, in addition, are decomposable by population subgroup (Bourguignon, 1978, and Shorrocks, 1980, 1984).

VI. CONCLUSIONS AND FURTHER RESEARCH

Given a classification system, field citation distributions consist of the citation counts for all publications in each field. Information available for large datasets for journal-based or publication-level classification systems indicates that field citation distributions are typically highly skewed. Furthermore, in spite of wide production and citation practices, the degree of skewness is very similar across fields at very different aggregation or granularity levels.

In so far as field citation distributions consist of articles published by individual authors, the three following questions naturally arise. Firstly, which are the basic characteristics of individual citation distributions? In particular, are they typically as skewed as field citation distributions, or are they normally, uniformly, or otherwise symmetrically distributed? Alternatively, are authors so different that it is impossible to assign them any systematic pattern at all? Secondly, is there any relationship between the characteristics of individual citation distributions in a given field and the characteristics of the field citation distribution of the articles they publish? Thirdly, is the situation common to all sciences, or is the authors' research experience quite different in more basic or more applied fields, in fields with a high or a low citation-density, or in the natural, the engineering and the social sciences?

These are key questions for understanding of the communication process in any science. However, the systematic study of the characteristics of individual citation distributions has traditionally been hampered by the lack of appropriate information. In this paper, we have largely overcome this difficulty by constructing a large dataset along the lines initiated in RCC. Our dataset consists of 35.1 million authors and 105.3 million articles published in the period 2000-2016, which are classified into 29 broad scientific fields. Citations of articles published in a given year are recorded up to the year 2016 in a variable citation window. Our main findings can be summarized in the following four points.

1. The vast majority of authors in all fields publish only less than five articles in the period 2000-2016. To study individual citation distributions of a certain size, recognizing the fact that individual productivity differs greatly between fields, it suffices to focus on very productive authors that, on average, represent only 5.2% of the population but are responsible for 38.0% of all articles. The

following results for this set of very productive authors are robust to an alternative notion of merely productive authors with a minimum of five articles *per capita*.

2. Very productive authors in a given field are very different from each other in many respects. Nevertheless, on average within each field, individual citation distributions exhibit a characteristic skewness pattern. Furthermore, the degree of average skewness is very similar across all fields.

3. As in previous contributions, we find that field citation distributions are highly skewed, and the degree of skewness is very similar across all fields. However, the typical pattern at the field level can be explained in terms of three ingredients: the average skewness of the field very productive authors' citation distributions, and the heterogeneity of individuals with respect to the number of publications per author and their mean citation rates.

4. The role of these ingredients is similar in all fields. Thus, the basic skewness of field citation distributions controlling for individual differences in productivity and mean citation rates is of the same order of magnitude in all fields. We emphasize that such basic skewness is already present at the individual level. Therefore, to understand the skewness of science at the field level we must simply explain the skewness of the individual citation distributions of very productive authors.

This view of science does not crucially depend on the way we have built our dataset. It is true that the disambiguation algorithm we have used admits further improvements. It is also true that it will be very instructive to experiment with other datasets and classification systems. However, previous results indicate that our conclusions are robust to the way we have solved the assignment of articles to two or more WoS subject categories, as well as to the assignment of responsibility in the case of multiple authorship.

As far as further research, we will mention two directions. Firstly, it might be useful to treat the skewness of citation distributions by means of a real valued skewness index that is robust to extreme observations. Although it is not decomposable by population subgroup, the Groeneveld & Meeden (1984) index, which we have used in other contributions (RCC, Albarrán *et al.*, 2015, and Perianes & Ruiz-Castillo, 2015b), might constitute an appropriate choice. Secondly, as long as one is interested in

investigating whether research questions receive similar or different answers across fields –as is the case in the present paper– it is advisable to treat distinct authors publishing in two or more fields as different authors. However, there are other contexts where this methodology is not adequate. For example, in many instances one may be interested in the ranking of universities or other research units which include distinct authors working in different fields. It would be interesting to extend our approach to this case.

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Table 1. Number of articles, and number of authors by scientific field, and average and coefficient of variation of these magnitudes over the 29 fields

	N_f (1)	% (2)	I_f (3)	% (4)	N_f^D (5)	% (6)
AGRICULTURE & FOOD SCIENCE	570,298	2.7	1,144,697	3.3	2,712,680	2.6
ASTRONOMY & ASTROPHYSICS	259,174	1.2	296,611	0.8	3,136,494	3.0
BASIC LIFE SCIENCES	1,918,845	9.1	3,827,339	10.9	11,123,668	10.6
BASIC MEDICAL SCIENCES	342,397	1.6	950,157	2.7	1,961,761	1.9
BIOLOGICAL SCIENCES	1,025,326	4.8	1,815,327	5.2	4,610,162	4.4
BIOMEDICAL SCIENCES	1,914,173	9.0	3,898,839	11.1	11,232,151	10.7
CHEMISTRY & CHEMICAL ENG.	2,337,898	11.0	3,466,432	9.9	10,521,685	10.0
CIVIL ENG. & CONSTRUCTION	155,484	0.7	259,045	0.7	501,431	0.5
CLINICAL MEDICINE	3,592,283	16.9	6,378,554	18.2	21,348,154	20.3
COMPUTER SCIENCES	575,536	2.7	828,925	2.4	1,812,869	1.7
EARTH SCIENCES & TECHNOLOGY	622,877	2.9	818,997	2.3	2,541,572	2.4
ECONOMICS & BUSINESS	262,412	1.2	240,653	0.7	588,016	0.6
EDUCATIONAL SCIENCES	146,675	0.7	245,618	0.7	422,657	0.4
ELECTRICAL ENG. & TELECOMM.	680,973	3.2	1,027,701	2.9	2,471,759	2.3
ENERGY SC. & TECHNOLOGY	330,577	1.6	676,827	1.9	1,582,464	1.5
ENVIRONMENTAL SCS & TECH.	867,052	4.1	1,368,240	3.9	3,461,632	3.3
GENERAL & INDUSTRIAL ENG.	190,700	0.9	337,448	1.0	592,217	0.6
HEALTH SCIENCES	455,500	2.1	891,728	2.5	1,983,543	1.9
INFORMATION & COMM. SCS.	63,516	0.3	94,965	0.3	156,287	0.1
INSTS. & INSTRUMENTATION	180,699	0.9	463,082	1.3	1,085,429	1.0
LAW & CRIMINOLOGY	75,480	0.4	107,371	0.3	193,701	0.2
MANAGEMENT & PLANNING	117,093	0.6	146,102	0.4	270,554	0.3
MATHEMATICS	536,382	2.5	410,133	1.2	1,204,809	1.1
MECHANICAL ENG. & AEROSPACE	433,086	2.0	625,175	1.8	1,383,929	1.3
PHYSICS & MATERIALS SCIENCE	2,696,102	12.7	3,514,815	10.0	15,699,147	14.9
PSYCHOLOGY	391,568	1.8	528,432	1.5	1,388,570	1.3
SOCIAL & BEHAVIORAL SCIENCES	93,762	0.4	167,312	0.5	278,710	0.3
SOCIOLOGY & ANTHROPOLOGY	136,174	0.6	208,419	0.6	367,223	0.3
STATISTICAL SCIENCES	230,636	1.1	319,043	0.9	656,110	0.6
TOTAL	21,202,678	100.0	35,057,987	100.0	105,289,384	100.0
Average over the 29 fields	731,127		1,208,896		3,630,668	
Coefficient of variation	1.2		1.3		1.4	

N_f = Number of articles in field f in the extended count according to the multiplicative approach, where each article is counted as many times as the number of fields to which it is assigned in the Web of Science

I_f = Number of authors in field f when researchers with articles in tow or more fields are treated as different authors

N_f^D = Number of articles in field f in the double extended count according to the multiplicative approach, where each article in the extended count is counted as many times as the number of its authors

Table 2. Average, coefficient of variation, and maximum number of authors per article, and average and coefficient of variation of these magnitudes over the 29 fields

	Number of authors per article:		
	Average (1)	CV (2)	Maximum (3)
AGRICULTURE & FOOD SCIENCE	4.8	0.67	479
ASTRONOMY & ASTROPHYSICS	12.1	8.50	3195
BASIC LIFE SCIENCES	5.8	0.82	1014
BASIC MEDICAL SCIENCES	5.7	0.64	404
BIOLOGICAL SCIENCES	4.5	0.91	769
BIOMEDICAL SCIENCES	5.9	0.69	1010
CHEMISTRY & CHEMICAL ENG.	4.5	0.54	341
CIVIL ENG. & CONSTRUCTION	3.2	0.54	133
CLINICAL MEDICINE	5.9	0.86	2458
COMPUTER SCIENCES	3.1	1.89	3035
EARTH SCIENCES & TECHNOLOGY	4.1	0.84	496
ECONOMICS & BUSINESS	2.2	0.77	479
EDUCATIONAL SCIENCES	2.9	0.75	192
ELECTRICAL ENG. & TELECOMM.	3.6	0.66	385
ENERGY SC. & TECHNOLOGY	4.8	2.41	3035
ENVIRONMENTAL SCS & TECH.	4.0	0.74	415
GENERAL & INDUSTRIAL ENG.	3.1	0.51	57
HEALTH SCIENCES	4.4	0.69	283
INFORMATION & COMM. SCS.	2.5	0.77	86
INSTS. & INSTRUMENTATION	6.0	7.01	3035
LAW & CRIMINOLOGY	2.6	0.97	163
MANAGEMENT & PLANNING	2.3	0.63	73
MATHEMATICS	2.2	0.62	287
MECHANICAL ENG. & AEROSPACE	3.2	0.55	174
PHYSICS & MATERIALS SCIENCE	5.8	8.54	5109
PSYCHOLOGY	3.5	0.80	479
SOCIAL & BEHAVIORAL SCIENCES	3.0	0.89	198
SOCIOLOGY & ANTHROPOLOGY	2.7	0.94	201
STATISTICAL SCIENCES	2.8	1.33	598
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Average over the 29 fields	4.2	1.6	985.6
Coefficient of variation	0.46	1.39	1.29

Table 3. First and second means of the field individual productivity (number of articles per author) distributions for all authors in the double extended count, and average and coefficient of variation of these magnitudes over the 29 fields

	First mean (1)	CV (2)	Second mean (3)	CV (4)
AGRICULTURE & FOOD SCIENCE	2.4	2.22	8.3	1.32
ASTRONOMY & ASTROPHYSICS	10.6	3.37	65.4	1.15
BASIC LIFE SCIENCES	2.9	2.34	9.2	1.37
BASIC MEDICAL SCIENCES	2.1	1.97	6.9	1.25
BIOLOGICAL SCIENCES	2.5	2.08	8.3	1.22
BIOMEDICAL SCIENCES	2.9	2.38	9.4	1.38
CHEMISTRY & CHEMICAL ENG.	3.0	3.14	14.0	1.60
CIVIL ENG. & CONSTRUCTION	1.9	1.85	4.7	1.36
CLINICAL MEDICINE	3.3	3.08	15.2	1.52
COMPUTER SCIENCES	2.2	1.97	7.5	1.18
EARTH SCIENCES & TECHNOLOGY	3.1	2.39	13.0	1.17
ECONOMICS & BUSINESS	2.4	1.71	7.6	0.96
EDUCATIONAL SCIENCES	1.7	1.63	4.0	1.26
ELECTRICAL ENG. & TELECOMM.	2.4	2.33	8.4	1.39
ENERGY SC. & TECHNOLOGY	2.3	1.99	7.7	1.18
ENVIRONMENTAL SCS & TECH.	2.5	2.18	8.5	1.27
GENERAL & INDUSTRIAL ENG.	1.8	1.62	4.2	1.22
HEALTH SCIENCES	2.2	2.12	7.9	1.27
INFORMATION & COMM. SCS.	1.6	1.50	3.9	1.16
INSTS. & INSTRUMENTATION	2.3	1.96	8.1	1.11
LAW & CRIMINOLOGY	1.8	1.79	4.4	1.34
MANAGEMENT & PLANNING	1.9	1.39	4.2	0.99
MATHEMATICS	2.9	2.27	9.5	1.30
MECHANICAL ENG. & AEROSPACE	2.2	2.08	7.8	1.24
PHYSICS & MATERIALS SCIENCE	4.5	5.41	27.3	2.37
PSYCHOLOGY	2.6	2.27	9.0	1.31
SOCIAL & BEHAVIORAL SCIENCES	1.7	1.30	3.8	0.98
SOCIOLOGY & ANTHROPOLOGY	1.8	1.37	4.0	1.02
STATISTICAL SCIENCES	2.1	1.98	7.5	1.19
<hr/>				
Average over the 29 fields	2.7	2.19	10.3	1.28
Coefficient of variation	0.60	0.35	1.10	0.20
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Table 4. Percentage distribution of authors classified by the number of articles they have published, and average and coefficient of variation of these magnitudes over the 29 fields

	Number of publications per author:					Percentage of total articles published by authors in column (4)
	One	Two	Three or four	Five or more	Total	
	(1)	(2)	(3)	(4)	(5)	(6)
AGRICULTURE & FOOD SCIENCE	70.7	12.0	8.1	9.1	100.0	48.5
ASTRONOMY & ASTROPHYSICS	65.1	7.7	6.4	20.7	100.0	90.3
BASIC LIFE SCIENCES	66.4	11.8	9.3	12.5	100.0	58.2
BASIC MEDICAL SCIENCES	71.8	12.3	8.3	7.6	100.0	39.9
BIOLOGICAL SCIENCES	68.0	12.5	8.9	10.6	100.0	51.5
BIOMEDICAL SCIENCES	67.5	11.5	8.9	12.1	100.0	58.2
CHEMISTRY & CHEMICAL ENG.	70.6	10.3	8.0	11.1	100.0	61.1
CIVIL ENG. & CONSTRUCTION	74.4	11.8	7.3	6.5	100.0	36.8
CLINICAL MEDICINE	69.5	10.5	7.7	12.3	100.0	65.2
COMPUTER SCIENCES	71.3	12.4	8.1	8.3	100.0	43.7
EARTH SCIENCES & TECHNOLOGY	67.9	11.1	8.3	12.7	100.0	62.0
ECONOMICS & BUSINESS	68.0	11.9	8.8	11.4	100.0	50.3
EDUCATIONAL SCIENCES	76.3	11.7	6.7	5.2	100.0	28.9
ELECTRICAL ENG. & TELECOMM.	70.7	11.9	8.2	9.2	100.0	49.2
ENERGY SC. & TECHNOLOGY	69.8	12.0	8.6	9.6	100.0	47.5
ENVIRONMENTAL SCS & TECH.	69.3	11.9	8.5	10.4	100.0	52.0
GENERAL & INDUSTRIAL ENG.	76.3	11.4	6.8	5.4	100.0	30.6
HEALTH SCIENCES	71.9	12.0	7.7	8.3	100.0	45.2
INFORMATION & COMM. SCS.	78.0	11.0	6.3	4.7	100.0	26.4
INSTS. & INSTRUMENTATION	70.6	12.3	7.9	9.2	100.0	48.0
LAW & CRIMINOLOGY	76.5	11.1	6.6	5.8	100.0	33.0
MANAGEMENT & PLANNING	73.4	12.0	7.7	6.8	100.0	33.4
MATHEMATICS	66.7	11.8	8.6	12.9	100.0	59.4
MECHANICAL ENG. & AEROSPACE	72.0	11.8	7.9	8.3	100.0	44.8
PHYSICS & MATERIALS SCIENCE	71.8	9.0	7.0	12.2	100.0	74.6
PSYCHOLOGY	68.9	12.2	8.4	10.5	100.0	53.8
SOCIAL & BEHAVIORAL SCIENCES	75.9	12.1	7.0	4.9	100.0	25.8
SOCIOLOGY & ANTHROPOLOGY	74.3	12.3	7.6	5.8	100.0	29.5
STATISTICAL SCIENCES	73.9	11.7	7.1	7.3	100.0	41.1
Average over the 29 fields	71.3	11.5	7.8	9.4	100.0	47.9
Coefficient of variation	0.05	0.09	0.10	0.36	0.00	0.31

Table 5. Average and coefficient of variation of the mean productivity and the first and second mean citations for very productive authors in each field, and average and coefficient of variation of these magnitudes over the 29 fields

	<i>Mean-size_f</i> (1)	<i>CV</i> (2)	<i>m₁</i> (3)	<i>CV</i> (4)	<i>m₂</i> (5)	<i>CV</i> (6)
AGRICULTURE & FOOD SCIENCE	19.9	0.85	15.6	1.23	41.6	1.93
ASTRONOMY & ASTROPHYSICS	153.0	0.55	29.4	0.63	126.1	0.94
BASIC LIFE SCIENCES	22.6	0.87	32.4	1.41	96.4	2.43
BASIC MEDICAL SCIENCES	14.4	0.93	16.7	1.14	43.5	2.21
BIOLOGICAL SCIENCES	19.3	0.78	25.3	1.97	79.3	2.95
BIOMEDICAL SCIENCES	22.8	0.87	24.7	1.24	69.1	2.00
CHEMISTRY & CHEMICAL ENG.	36.0	1.01	18.0	1.30	49.5	3.82
CIVIL ENG. & CONSTRUCTION	10.9	0.93	10.9	1.21	24.9	1.42
CLINICAL MEDICINE	39.4	0.92	25.0	1.17	77.3	2.23
COMPUTER SCIENCES	16.7	0.79	9.3	2.05	30.7	3.28
EARTH SCIENCES & TECHNOLOGY	28.4	0.73	19.8	0.94	55.7	1.69
ECONOMICS & BUSINESS	14.9	0.63	17.5	1.20	47.2	1.60
EDUCATIONAL SCIENCES	9.6	0.92	12.6	1.35	30.9	1.63
ELECTRICAL ENG. & TELECOMM.	20.5	0.91	9.4	2.23	30.3	6.61
ENERGY SC. & TECHNOLOGY	16.5	0.81	11.7	1.31	33.4	2.34
ENVIRONMENTAL SCS & TECH.	19.5	0.83	20.2	1.11	55.6	1.96
GENERAL & INDUSTRIAL ENG.	9.9	0.85	9.2	1.02	21.6	1.33
HEALTH SCIENCES	17.5	0.86	15.7	1.06	42.4	2.18
INFORMATION & COMM. SCS.	7.6	0.90	12.9	1.34	33.0	2.27
INSTS. & INSTRUMENTATION	18.9	0.66	9.2	1.47	32.7	2.39
LAW & CRIMINOLOGY	10.3	0.95	10.2	1.09	23.9	2.56
MANAGEMENT & PLANNING	9.0	0.65	20.5	1.16	52.1	1.59
MATHEMATICS	21.4	0.85	6.5	1.43	18.7	2.43
MECHANICAL ENG. & AEROSPACE	17.4	0.83	10.0	1.10	25.3	1.85
PHYSICS & MATERIALS SCIENCE	93.9	1.31	17.1	1.11	55.5	2.57
PSYCHOLOGY	20.6	0.83	20.7	1.07	58.5	1.91
SOCIAL & BEHAVIORAL SCIENCES	7.2	0.74	15.1	1.53	36.6	2.47
SOCIOLOGY & ANTHROPOLOGY	7.4	0.76	15.4	1.40	37.1	1.81
STATISTICAL SCIENCES	16.6	0.80	13.8	1.98	45.2	3.09
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	<i>Mean-size</i>	<i>CV</i>	<i>M₁</i>	<i>CV</i>	<i>M₂</i>	<i>CV</i>
Average over the 29 fields	24.9	0.84	16.4	1.32	47.4	2.33
Coefficient of variation	1.18	0.16	0.40	0.27	0.51	0.44

Mean-size_f = average of the sizes of distributions $c_{ji}(i)$, or individual productivities $n_j(i)$, over the I_f^* very productive authors in field f

m₁ = average of the individual first means $m_{j1}(i)$ of distributions $c_{ji}(i)$ over the I_f^* very productive authors in field f

m₂ = average of the individual second means $m_{j2}(i)$ of distributions $c_{ji}(i)$ over the I_f^* very productive authors in field f

Table 6. The skewness of individual citation distributions according to the CSS approach for very productive authors in the double extended count, and average and coefficient of variation of the CSS results over the 29 fields

Within-field results: average and coefficient of variation of the CSS results for all individuals in each field

Between-field results: average and coefficient of variation of the CSS results over the 29 fields

	\hat{p}_1	\hat{p}_2	\hat{p}_3	cv_1	cv_2	cv_3	s_1	s_2	s_3	cv_4	cv_5	cv_6
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
AGRICULTURE & FOOD SCIENCE	65.8	20.8	13.3	0.14	0.36	0.44	24.2	32.2	43.7	0.34	0.36	0.27
ASTRONOMY & ASTROPHYSICS	76.9	17.3	5.8	0.11	0.26	0.73	25.8	28.6	45.7	0.13	0.20	0.13
BASIC LIFE SCIENCES	67.4	20.1	12.5	0.14	0.36	0.45	25.2	31.1	43.7	0.30	0.34	0.26
BASIC MEDICAL SCIENCES	64.5	21.2	14.3	0.17	0.39	0.47	25.9	32.7	41.4	0.38	0.40	0.34
BIOLOGICAL SCIENCES	66.8	20.3	12.9	0.15	0.37	0.46	24.0	32.0	44.0	0.35	0.38	0.29
BIOMEDICAL SCIENCES	66.7	20.5	12.7	0.14	0.35	0.44	25.4	31.3	43.3	0.30	0.33	0.25
CHEMISTRY & CHEMICAL ENG.	67.1	20.8	12.1	0.12	0.30	0.41	25.0	31.7	43.4	0.27	0.27	0.21
CIVIL ENG. & CONSTRUCTION	63.5	22.0	14.5	0.19	0.39	0.58	22.8	37.1	40.0	0.51	0.48	0.47
CLINICAL MEDICINE	68.8	20.1	11.1	0.12	0.29	0.42	24.9	31.1	44.0	0.24	0.27	0.20
COMPUTER SCIENCES	68.5	19.4	12.2	0.15	0.41	0.52	18.8	34.4	46.5	0.52	0.47	0.35
EARTH SCIENCES & TECHNOLOGY	67.4	20.5	12.1	0.13	0.31	0.42	24.6	31.4	44.0	0.27	0.29	0.22
ECONOMICS & BUSINESS	66.7	20.1	13.2	0.15	0.38	0.46	22.4	32.6	45.0	0.39	0.39	0.30
EDUCATIONAL SCIENCES	63.8	21.7	14.5	0.19	0.39	0.58	23.2	36.9	39.8	0.50	0.48	0.49
ELECTRICAL ENG. & TELECOMM.	67.4	20.1	12.5	0.15	0.38	0.47	20.7	33.3	46.0	0.44	0.40	0.29
ENERGY SC. & TECHNOLOGY	66.6	20.2	13.1	0.16	0.41	0.48	22.2	33.3	44.4	0.46	0.43	0.33
ENVIRONMENTAL SCS & TECH.	66.3	20.6	13.1	0.15	0.36	0.45	24.6	31.8	43.6	0.33	0.36	0.27
GENERAL & INDUSTRIAL ENG.	63.4	22.2	14.4	0.19	0.41	0.59	22.9	37.6	39.1	0.53	0.49	0.50
HEALTH SCIENCES	65.6	20.8	13.6	0.15	0.37	0.45	24.0	32.3	43.7	0.36	0.37	0.29
INFORMATION & COMM. SCS.	63.1	23.5	13.4	0.21	0.39	0.75	22.6	42.9	34.2	0.58	0.50	0.68
INSTS. & INSTRUMENTATION	71.0	17.8	11.2	0.15	0.45	0.52	19.2	33.1	47.7	0.49	0.44	0.32
LAW & CRIMINOLOGY	62.9	22.4	14.6	0.19	0.39	0.56	23.4	37.2	39.2	0.51	0.46	0.47
MANAGEMENT & PLANNING	64.0	21.5	14.5	0.19	0.39	0.57	23.2	36.6	40.1	0.47	0.46	0.48
MATHEMATICS	67.5	20.2	12.3	0.14	0.37	0.46	18.5	34.0	47.5	0.50	0.39	0.28
MECHANICAL ENG. & AEROSPACE	65.2	21.1	13.7	0.15	0.37	0.45	22.9	33.0	44.1	0.41	0.38	0.29
PHYSICS & MATERIALS SCIENCE	71.1	19.4	9.5	0.10	0.24	0.42	23.0	31.3	45.7	0.24	0.21	0.16
PSYCHOLOGY	66.9	20.2	12.9	0.14	0.36	0.45	23.7	31.7	44.6	0.33	0.35	0.26
SOCIAL & BEHAVIORAL SCIENCES	62.4	23.5	14.1	0.21	0.37	0.73	23.7	41.6	34.6	0.54	0.48	0.66
SOCIOLOGY & ANTHROPOLOGY	62.6	23.4	14.0	0.21	0.37	0.73	23.4	41.6	34.9	0.54	0.48	0.65
STATISTICAL SCIENCES	67.3	19.9	12.8	0.15	0.40	0.48	21.1	33.2	45.7	0.44	0.42	0.32
Average over the 29 fields	66.4	20.8	12.8	0.16	0.36	0.52	23.2	34.0	42.8	0.40	0.39	0.34
Coefficient of variation	0.05	0.07	0.14	0.19	0.12	0.20	0.08	0.10	0.09	0.28	0.21	0.42

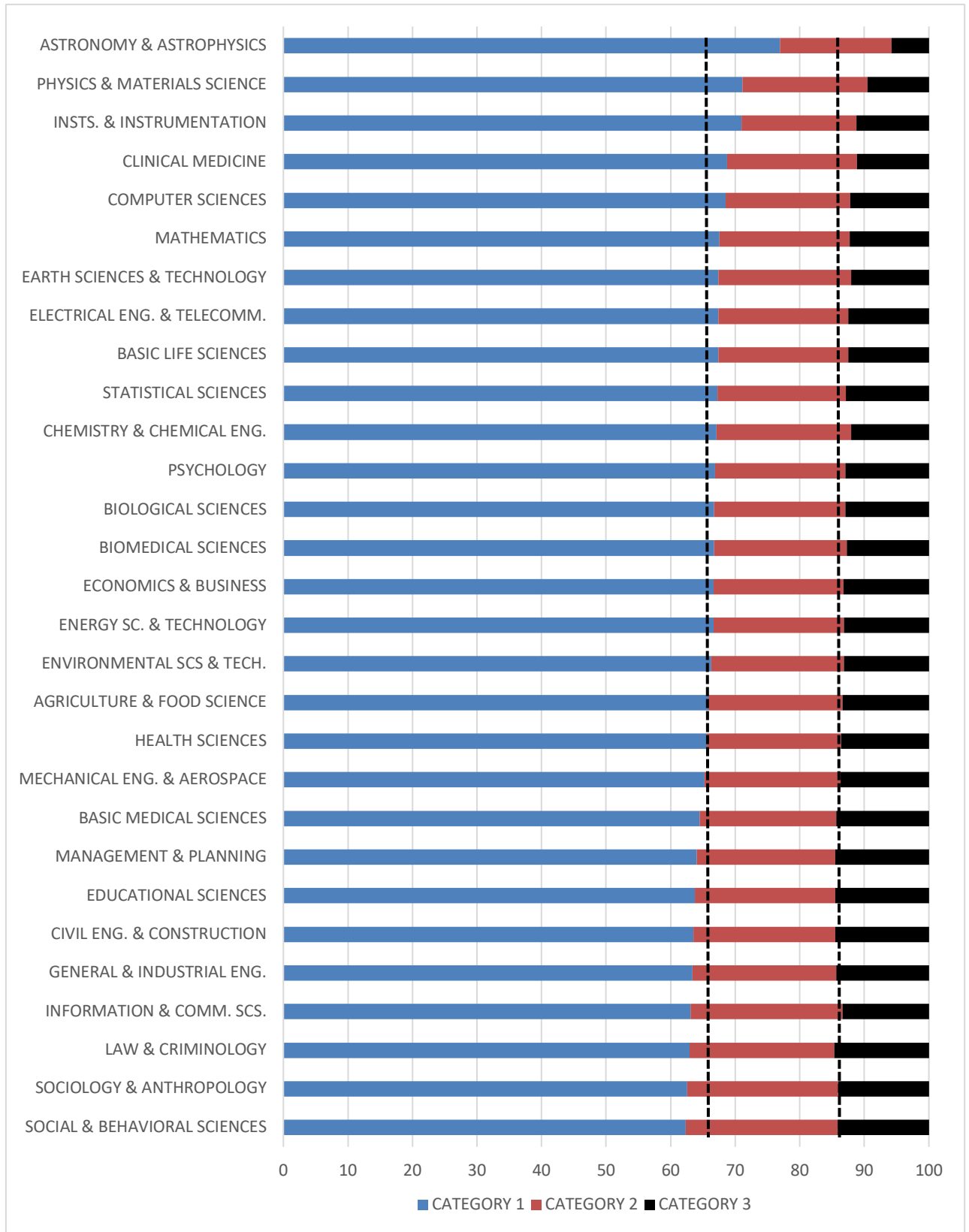


Figure 1. The skewness of average individual citation distributions in each field according to the CSS approach. Very productive authors in the double extended count.

Table 7. First and second means of field citation distributions for very productive authors in the double extended count, and average and coefficient of variation of these magnitudes over the 29 fields

FIELDS	μ_{f1}^{D*} (1)	CV (2)	μ_{f2}^{D*} (3)	CV (4)
AGRICULTURE & FOOD SCIENCE	16.0	2.46	46.3	1.47
ASTRONOMY & ASTROPHYSICS	29.2	5.37	116.1	3.01
BASIC LIFE SCIENCES	35.3	3.43	122.5	1.94
BASIC MEDICAL SCIENCES	17.0	2.58	47.8	1.61
BIOLOGICAL SCIENCES	25.3	4.48	96.6	2.47
BIOMEDICAL SCIENCES	26.1	2.79	81.8	1.63
CHEMISTRY & CHEMICAL ENG.	19.6	3.36	58.5	2.08
CIVIL ENG. & CONSTRUCTION	10.9	1.98	30.4	1.09
CLINICAL MEDICINE	27.5	3.40	89.5	2.00
COMPUTER SCIENCES	9.5	4.50	36.1	2.42
EARTH SCIENCES & TECHNOLOGY	21.6	2.51	63.6	1.48
ECONOMICS & BUSINESS	18.5	2.55	61.9	1.34
EDUCATIONAL SCIENCES	13.5	2.28	40.6	1.27
ELECTRICAL ENG. & TELECOMM.	9.8	5.55	32.2	3.32
ENERGY SC. & TECHNOLOGY	11.4	3.21	37.2	1.83
ENVIRONMENTAL SCS & TECH.	21.8	2.60	64.1	1.55
GENERAL & INDUSTRIAL ENG.	10.0	1.94	26.5	1.11
HEALTH SCIENCES	16.8	2.70	47.1	1.67
INFORMATION & COMM. SCS.	14.4	2.79	44.8	1.60
INSTS. & INSTRUMENTATION	8.3	3.59	31.9	1.85
LAW & CRIMINOLOGY	10.8	2.35	28.0	1.47
MANAGEMENT & PLANNING	23.4	2.33	73.4	1.26
MATHEMATICS	7.0	3.11	22.0	1.76
MECHANICAL ENG. & AEROSPACE	10.5	2.43	30.0	1.44
PHYSICS & MATERIALS SCIENCE	19.6	4.71	66.1	2.81
PSYCHOLOGY	22.4	2.74	67.1	1.62
SOCIAL & BEHAVIORAL SCIENCES	15.9	2.65	44.5	1.63
SOCIOLOGY & ANTHROPOLOGY	16.6	2.42	49.6	1.39
STATISTICAL SCIENCES	13.8	5.02	50.4	2.82
Average over the 29 fields	17.3	3.2	55.4	1.8
Coefficient of variation	0.41	0.32	0.47	0.32

Table 8. The skewness of field citation distributions according to the CSS approach for very productive authors in the double extended count, and average and coefficient of variation of the CSS results over the 29 fields

	% of articles in category:			% of total citations in category:		
	I (1)	II (2)	III (3)	I (4)	II (5)	III (6)
AGRICULTURAL & FOOD SCIENCES	73.8	19.3	6.9	24.2	32.6	43.2
ASTRONOMY & ASTROPHYSICS	81.4	15.7	2.9	25.8	27.7	46.4
BASIC LIFE SCIENCES	78.1	16.8	5.1	23.8	29.8	46.4
BASIC MEDICAL SCIENCES	73.5	19.5	7.0	25.7	32.0	42.3
BIOLOGICAL SCIENCES	79.8	16.1	4.1	23.0	29.2	47.8
BIOMEDICAL SCIENCES	75.9	18.1	6.0	24.7	30.9	44.4
CHEMISTRY & CHEMICAL ENG.	74.5	19.0	6.5	23.9	31.7	44.5
CIVIL ENG. & CONSTRUCTION	71.8	19.9	8.3	21.0	32.4	46.6
CLINICAL MEDICINE	76.6	18.0	5.4	23.8	31.0	45.1
COMPUTER SCIENCES	78.6	16.6	4.7	18.9	31.3	49.8
EARTH SCIENCES & TECHNOLOGY	74.2	19.1	6.6	23.9	31.8	44.3
ECONOMICS & BUSINESS	76.2	17.4	6.4	20.7	31.1	48.3
EDUCATIONAL SCIENCES	74.3	18.8	6.9	22.5	32.0	45.5
ELECTRICAL ENG. & TELECOMM.	75.8	18.5	5.7	20.2	32.3	47.5
ENERGY SC. & TECHNOLOGY	75.6	18.0	6.3	20.9	32.2	46.9
ENVIRONMENTAL SCS & TECH.	74.0	19.4	6.6	23.6	32.0	44.4
GENERAL & INDUSTRIAL ENG.	70.4	21.0	8.6	21.4	33.1	45.5
HEALTH SCIENCES	72.4	20.3	7.2	23.0	33.0	44.1
INFORMATION & COMM. SCS.	74.9	18.5	6.7	21.9	31.8	46.2
INSTS. & INSTRUMENTATION	78.8	15.5	5.7	18.7	30.5	50.9
LAW & CRIMINOLOGY	69.9	21.1	9.0	22.0	32.8	45.1
MANAGEMENT & PLANNING	75.2	18.0	6.9	22.1	31.3	46.7
MATHEMATICS	73.9	19.0	7.1	17.8	31.7	50.5
MECHANICAL ENG. & AEROSPACE	72.7	19.5	7.7	22.5	32.0	45.4
PHYSICS & MATERIALS SCIENCE	77.2	17.9	4.8	23.2	31.4	45.4
PSYCHOLOGY	74.4	18.9	6.7	23.2	32.2	44.6
SOCIAL & BEHAVIORAL SCIENCES	72.5	20.0	7.5	22.8	32.8	44.4
SOCIOLOGY & ANTHROPOLOGY	74.1	19.1	6.8	22.7	31.8	45.6
STATISTICAL SCIENCES	78.2	17.3	4.4	20.3	31.2	48.5
Average over the 29 fields	75.1	18.5	6.4	22.3	31.6	46.1
Coefficient of variation	0.04	0.08	0.21	0.09	0.04	0.05

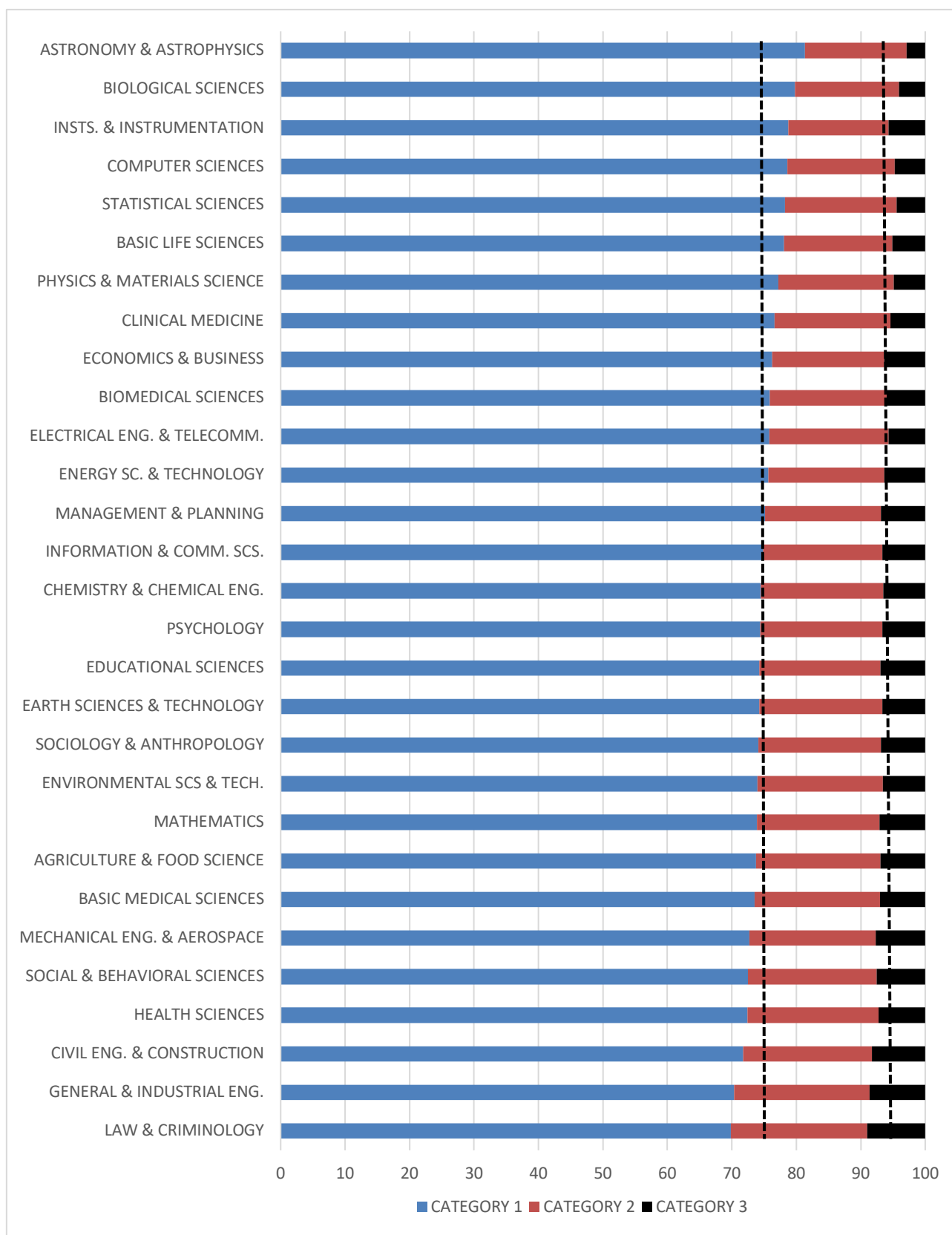


Figure 2. The skewness of field citation distributions according to the CSS approach. Articles published by very productive authors in the double extended count.

Table 9. Average and coefficient of variation of the CSS results over the 29 fields. Selected individual and field citation distributions

A. Very productive authors in the double extended count

	% of articles in category:			% of total citations in category:		
	I	II	III	I	II	III
I. Average individual citation distributions of very productive authors in each field						
Average over 29 fields	66.4	20.8	12.8	23.2	34.0	42.8
Coefficient of variation	0.05	0.07	0.14	0.08	0.10	0.09
II. Field citation distributions						
Average over 29 fields	75.1	18.5	6.4	22.3	31.6	46.1
Coefficient of variation	0.04	0.08	0.21	0.09	0.04	0.05
III. Weighted average individual citation distributions of very productive authors in each field						
Average over 29 fields	67.5	20.5	12.0	23.3	33.0	43.7
Coefficient of variation	0.05	0.06	0.16	0.08	0.07	0.06
IV. Field citation distributions controlling for individual productivity within each field						
Average over 29 fields	75.4	18.3	6.3	22.3	31.5	46.2
Coefficient of variation	0.03	0.08	0.20	0.08	0.04	0.04
V. Field citation distributions controlling for individual productivity and individual mean citations within each field						
Average over 29 fields	66.4	21.9	11.7	23.2	33.2	43.6
Coefficient of variation	0.05	0.05	0.17	0.08	0.02	0.04

B. All authors in the double extended count

VI. Field citation distributions

Average over 29 fields	76.8	17.7	5.5	21.6	31.3	47.1
Coefficient of variation	0.03	0.08	0.22	0.08	0.05	0.05

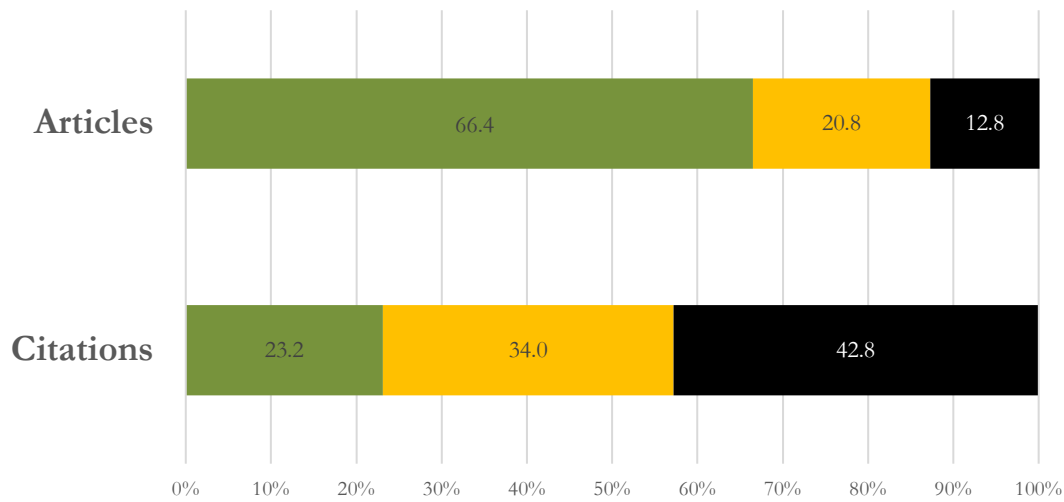


Figure 3. Percentage of articles in categories I, II, III, and percentage of citations accounted for by each category. Individual citation distributions of very productive authors in each field. Average of CSS results over the 29 fields

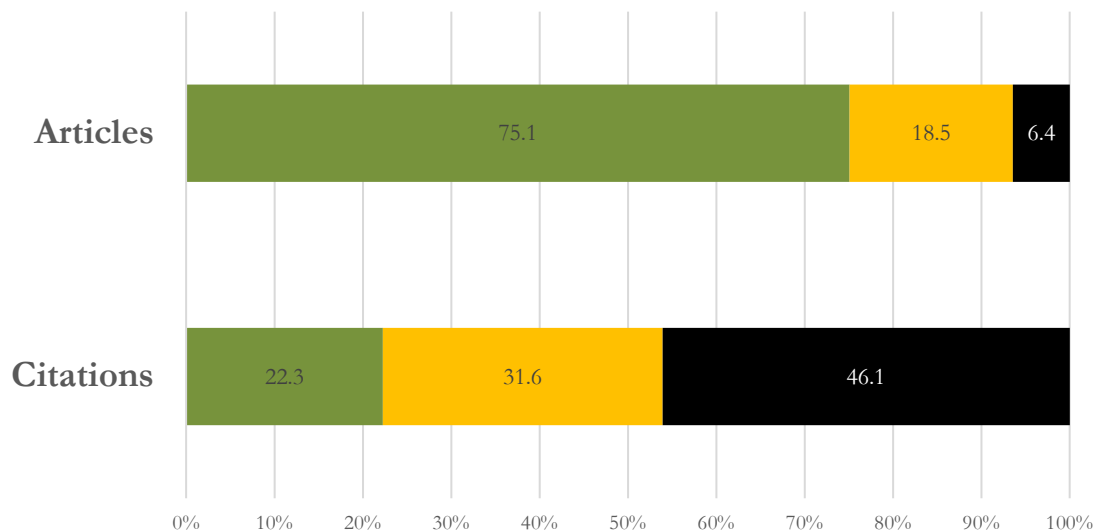


Figure 4. Percentage of articles in categories I, II, III, and percentage of citations accounted for by each category. Field citation distributions in the double extended count for very productive authors. Average of CSS results over the 29 fields

APPENDIX I. THE RELIABILITY OF OUR DATASET

As we know, our dataset, which has been constructed with the same criteria used in RCC, ends up being twice as large as the dataset used in that contribution. Therefore, it seems advisable to assess the reliability of the data used in this paper by comparing some characteristics of the two datasets. We do this in two instances: the descriptive statistics included in our Tables 1 and 2, as well as the CSS results for the two key magnitudes investigated at length in RCC, namely, the number of articles and the mean citation per author in the double extended count.

Descriptive statistics

We begin by considering the number of authors per field and the number of articles per field in the extended and the double extended counts (Table 1), as well as the number of articles per author in each field (Table 2). The corresponding descriptive statistics for the RCC's dataset are in Table C in the Appendix and Table 1 in the Working Paper version of RCC, Ruiz-Castillo & Costas (2014b).

We verify that the following characteristics are extremely similar in both datasets: the classification of fields into three groups according to the number of authors per field; the average number of authors per article and the between-field variability of this quantity, and the relationship between the distribution of articles by field in the extended and the double extended counts.

The skewness of the individual productivity and mean citation distributions in the two datasets

In each field $f = 1, \dots, 29$ in our dataset, let $c_f(i)$ be the citation distribution of author i with $i = 1, \dots, I_f$, where I_f is the number of authors in field f . For each i and f , let $n_f(i)$ and $\mu_f(i)$ be the number of articles and the mean citation of $c_f(i)$. The first and second means for individual productivity i.e. for distributions $\{n_f(i), i = 1, \dots, I_f\}$ are presented in Table 3 in the text, whereas the first and second means for citation distributions, i.e. for distributions $\{\mu_f(i), i = 1, \dots, I_f\}$ are in Table AI. We expect a large within-field variability in both distributions. This is confirmed by the very large CVs for these characteristics in all fields, which are always greater than 0.96 and often greater than 2. In turn, large

differences in production and citation practices across fields are expected to give rise to large between-field variability. This is confirmed by the large size of the CVs over the 29 fields in Table 3 and AI.1.

Table AI.1 around here

Tables AI.2 and AI.3 present the complete CSS results for the individual productivity and mean citation distributions in our dataset.¹² These results can be compared to those presented in Table B and E in the Appendix in Ruiz-Castillo & Costas (2014b). A summary of the comparison between the two datasets is illustrated in Table AI.4 for the average results.

Tables AI.2, AI.3 and AI.4 around here

There are three main findings. Firstly, the CVs over the 29 fields indicate that between-field variation is typically small, and of the same order of magnitude as in RCC. Secondly, the skewness of individual productivity is essentially the same as in RCC. Thirdly, generally the skewness in our dataset is somewhat greater than in RCC in all fields, and hence on the average. For example, on average the percentage of articles in category I and III in RCC is 71.0% and 8.3%, whereas these magnitudes are 76.7% and 5.3% in our dataset.

We conclude that the high degree of consistency between the descriptive statistics and the CSS results in the two datasets demonstrate the reliability of the data used in this paper: having followed the same criteria in both cases, the two datasets seem to reflect the same world.

It should be mentioned that RCC establish that the results concerning the skewness of the individual productivity and mean citation distributions when we use the multiplicative or the fractional counting method for assigning responsibility in co-authored publications are essentially the same. Therefore, we are confident that the results in this paper are robust to the method adopted for treating the co-authorship problem without replicating the RCC exercise in this respect for our dataset.

¹² The CSS approach is summarized in Section II.3 in the text.

Table AI.1. First and second means of the mean citation distribution for all authors in each field in the double extended count, and average and coefficient of variation of these magnitudes over the 29 fields

	First mean (1)	<i>CV</i> (2)	Second mean (3)	<i>CV</i> (4)
AGRICULTURE & FOOD SCIENCE	14.6	3.34	45.9	2.01
ASTRONOMY & ASTROPHYSICS	21.6	4.57	75.4	2.67
BASIC LIFE SCIENCES	24.1	3.37	78.1	2.00
BASIC MEDICAL SCIENCES	15.5	3.65	47.3	2.27
BIOLOGICAL SCIENCES	22.8	4.47	92.3	2.38
BIOMEDICAL SCIENCES	19.3	2.92	58.9	1.76
CHEMISTRY & CHEMICAL ENG.	16.2	3.15	49.4	1.92
CIVIL ENG. & CONSTRUCTION	10.3	2.13	29.7	1.19
CLINICAL MEDICINE	20.1	3.56	65.9	2.11
COMPUTER SCIENCES	11.4	6.90	52.5	3.47
EARTH SCIENCES & TECHNOLOGY	13.8	2.79	41.2	1.66
ECONOMICS & BUSINESS	12.1	2.65	39.3	1.46
EDUCATIONAL SCIENCES	9.6	5.87	30.6	3.67
ELECTRICAL ENG. & TELECOMM.	8.4	5.56	29.8	3.23
ENERGY SC. & TECHNOLOGY	12.7	4.87	39.4	3.03
ENVIRONMENTAL SCS & TECH.	16.1	3.49	50.9	2.11
GENERAL & INDUSTRIAL ENG.	7.1	2.44	20.6	1.43
HEALTH SCIENCES	12.8	2.65	36.2	1.63
INFORMATION & COMM. SCS.	10.2	5.13	33.5	3.11
INSTS. & INSTRUMENTATION	12.1	3.69	44.5	2.02
LAW & CRIMINOLOGY	12.2	8.82	53.3	4.72
MANAGEMENT & PLANNING	11.7	2.58	35.9	1.46
MATHEMATICS	7.2	4.70	28.7	2.50
MECHANICAL ENG. & AEROSPACE	8.9	3.66	27.0	2.25
PHYSICS & MATERIALS SCIENCE	12.8	3.83	42.1	2.24
PSYCHOLOGY	16.4	3.03	50.7	1.80
SOCIAL & BEHAVIORAL SCIENCES	11.4	3.05	32.9	1.89
SOCIOLOGY & ANTHROPOLOGY	12.1	2.96	39.0	1.71
STATISTICAL SCIENCES	22.3	6.09	137.6	2.57
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Average over the 29 fields	14.0	4.00	48.6	2.28
Coefficient of variation	0.33	0.37	0.48	0.34

Table AI.2. The skewness of the field individual productivity distributions for all authors in the double extended count according to the CSS approach, and average and coefficient of variation of the CSS results over the 29 fields

	% of individuals in category:			% of total articles in category:		
	I	II	III	I	II	III
	(1)	(2)	(3)	(4)	(5)	(6)
AGRICULTURE & FOOD SCIENCE	82.8	12.9	4.3	40.0	23.9	36.1
ASTRONOMY & ASTROPHYSICS	86.1	9.7	4.2	14.2	25.2	60.7
BASIC LIFE SCIENCES	78.2	16.3	5.5	31.0	26.3	42.7
BASIC MEDICAL SCIENCES	84.2	11.3	4.6	46.7	21.3	32.0
BIOLOGICAL SCIENCES	80.5	14.4	5.0	36.6	25.1	38.2
BIOMEDICAL SCIENCES	79.0	15.6	5.4	31.4	25.4	43.2
CHEMISTRY & CHEMICAL ENG.	85.9	10.6	3.6	35.0	22.6	42.4
CIVIL ENG. & CONSTRUCTION	74.4	19.1	6.5	38.4	24.8	36.8
CLINICAL MEDICINE	84.8	11.4	3.8	31.4	24.1	44.5
COMPUTER SCIENCES	83.7	12.0	4.4	43.9	22.6	33.4
EARTH SCIENCES & TECHNOLOGY	84.3	11.1	4.7	34.1	22.9	43.0
ECONOMICS & BUSINESS	79.8	13.9	6.3	37.5	24.3	38.2
EDUCATIONAL SCIENCES	76.3	18.5	5.2	44.4	26.7	28.9
ELECTRICAL ENG. & TELECOMM.	82.6	13.0	4.3	39.3	23.7	37.0
ENERGY SC. & TECHNOLOGY	81.7	13.1	5.2	40.1	23.5	36.5
ENVIRONMENTAL SCS & TECH.	81.1	13.8	5.0	36.8	24.3	39.0
GENERAL & INDUSTRIAL ENG.	76.3	18.2	5.4	43.5	25.9	30.6
HEALTH SCIENCES	83.9	11.6	4.5	43.1	21.7	35.2
INFORMATION & COMM. SCS.	78.0	15.2	6.8	47.4	21.0	31.6
INSTS. & INSTRUMENTATION	82.9	12.7	4.4	40.6	23.9	35.6
LAW & CRIMINOLOGY	76.5	17.7	5.8	42.4	24.5	33.0
MANAGEMENT & PLANNING	73.4	19.7	6.8	39.7	27.0	33.4
MATHEMATICS	78.5	15.4	6.1	30.7	24.9	44.4
MECHANICAL ENG. & AEROSPACE	83.8	11.8	4.4	43.2	22.0	34.8
PHYSICS & MATERIALS SCIENCE	87.8	9.7	2.5	25.4	22.8	51.7
PSYCHOLOGY	81.1	13.6	5.3	35.5	23.0	41.5
SOCIAL & BEHAVIORAL SCIENCES	75.9	16.8	7.3	45.6	22.9	31.5
SOCIOLOGY & ANTHROPOLOGY	74.3	17.3	8.4	42.2	22.5	35.4
STATISTICAL SCIENCES	85.6	10.5	3.9	47.3	21.0	31.7
Average over the 29 fields	80.8	14.0	5.2	38.2	23.8	38.0
Coefficient of variation	0.05	0.21	0.24	0.19	0.07	0.18

Category I = individuals with low productivity, smaller than or equal to the first mean in Table 3

Category II = individuals with a fair productivity, between the first and the second mean in Table 3

Category III = individuals with a remarkable or outstanding productivity above the second mean in Table 3

Individual productivity is measured by the number of articles per author

Table AI.3. The skewness of the field mean citation distributions for all authors in the double extended count according to the CSS approach, and average and coefficient of variation of the CSS results over the 29 fields

	% of individuals in category:			% of total citations in category:		
	I (1)	II (2)	III (3)	I (4)	II (5)	III (6)
AGRICULTURE & FOOD SCIENCE	75.9	18.9	5.3	24.0	31.7	44.3
ASTRONOMY & ASTROPHYSICS	78.1	18.1	3.8	23.4	31.3	45.3
BASIC LIFE SCIENCES	76.7	18.0	5.3	24.6	30.9	44.4
BASIC MEDICAL SCIENCES	75.4	19.2	5.4	24.8	32.0	43.2
BIOLOGICAL SCIENCES	80.9	15.3	3.7	22.8	28.1	49.1
BIOMEDICAL SCIENCES	75.4	18.7	5.9	24.8	31.3	43.9
CHEMISTRY & CHEMICAL ENG.	75.3	18.7	6.0	24.6	31.5	43.8
CIVIL ENG. & CONSTRUCTION	73.0	19.1	7.9	22.0	32.2	45.8
CLINICAL MEDICINE	77.0	18.0	4.9	24.5	31.1	44.4
COMPUTER SCIENCES	82.5	14.1	3.4	19.3	28.1	52.6
EARTH SCIENCES & TECHNOLOGY	74.1	19.6	6.3	22.6	32.5	44.8
ECONOMICS & BUSINESS	76.2	17.7	6.2	22.4	31.2	46.4
EDUCATIONAL SCIENCES	75.6	18.9	5.5	22.7	32.4	44.9
ELECTRICAL ENG. & TELECOMM.	77.8	17.6	4.7	21.5	31.3	47.3
ENERGY SC. & TECHNOLOGY	75.0	19.2	5.7	22.3	32.8	44.8
ENVIRONMENTAL SCS & TECH.	76.0	18.5	5.4	24.1	31.6	44.2
GENERAL & INDUSTRIAL ENG.	73.2	19.3	7.5	22.5	32.9	44.6
HEALTH SCIENCES	73.0	20.3	6.7	23.6	32.9	43.5
INFORMATION & COMM. SCS.	76.3	18.3	5.4	22.5	32.4	45.2
INSTS. & INSTRUMENTATION	78.9	16.5	4.6	22.5	30.1	47.4
LAW & CRIMINOLOGY	82.4	15.7	2.0	22.8	29.0	48.3
MANAGEMENT & PLANNING	74.6	18.5	6.9	21.9	31.4	46.7
MATHEMATICS	79.9	16.0	4.1	20.2	30.3	49.5
MECHANICAL ENG. & AEROSPACE	74.2	20.0	5.8	21.9	33.2	44.9
PHYSICS & MATERIALS SCIENCE	76.1	18.4	5.5	21.5	31.9	46.6
PSYCHOLOGY	75.3	18.8	5.9	23.4	31.8	44.8
SOCIAL & BEHAVIORAL SCIENCES	73.3	19.6	7.1	22.9	32.5	44.6
SOCIOLOGY & ANTHROPOLOGY	76.2	18.2	5.6	23.5	31.4	45.1
STATISTICAL SCIENCES	86.9	10.8	2.3	19.3	23.3	57.5
Average over the 29 fields	76.7	17.9	5.3	22.7	31.1	46.1
Coefficient of variation	0.04	0.11	0.26	0.06	0.06	0.07

Category I = individuals with low citation impact, smaller than or equal to the first mean in Table AI.1

Category II = individuals with a fair citation impact, between the first and the second mean in Table AI.1

Category III = individuals with a remarkable or outstanding citation impact above the second mean in Table AI.1

Table AI.4. The skewness of field individual productivity and individual mean citation distributions according to the CSS approach. Average and coefficient of variation of CSS results over all fields in RCC for the period 2003-2011 (Table 3) and in the present dataset for the period 2000-2015 (Tables AI.2 and AI.3)

<u>Individual productivity = number of articles per person</u>						
	% of authors in category:			% of articles in category:		
	I	II	III	I	II	III
I. RCC results						
Average	79.3	14.8	5.9	40.4	24.5	35.1
Coefficient of variation	0.04	0.17	0.19	0.17	0.07	0.18
II. Present dataset						
Average	80.8	14.0	5.2	38.2	23.8	38.0
Coefficient of variation	0.05	0.21	0.24	0.19	0.07	0.18

<u>Individual mean citation per article per person</u>						
	% of authors in category:			% of total mean citations in category:		
	I	II	III	I	II	III
III. RCC						
Average	71.0	20.7	8.3	22.8	33.2	44.0
Coefficient of variation	0.03	0.06	0.13	0.09	0.02	0.05
IV. Present dataset						
Average	76.7	17.9	5.3	22.7	31.1	46.1
Coefficient of variation	0.04	0.11	0.26	0.06	0.06	0.06

APPENDIX II. THE EXTENDED VERSUS THE DOUBLE EXTENDED COUNT

To understand the relationship between the extended and the double extended count, it is useful to introduce some notation. For each f , the field citation distribution in the extended count is denoted by $C_f = \{c_f(i), i = 1, \dots, I_f\}$. Let A_f be the maximum number of authors for an article in field f , and let $N_f(a)$ be the number of articles in the extended count with a authors, where $a = 1, \dots, A_f$. Of course, $N_f = \sum_a N_f(a)$. We denote by $C_f(a)$ the citation distribution of articles with a authors, so that $C_f = \{C_f(a), a = 1, \dots, A_f\}$. For every f and a , the mean of $C_f(a)$ is denoted by $\mu_{f1}(a)$, whereas for every f , the mean of C_f is denoted by μ_{f1} . Since the distributions $C_f(a)$, $a = 1, \dots, A_f$, form a partition of C_f , we have:

$$\mu_{f1} = \sum_a w_f(a) \mu_{f1}(a), \quad (1)$$

where $w_f(a) = N_f(a)/N_f$ is the proportion of articles with a authors in field f in the extended count.

After following a multiplicative strategy for assigning responsibility to individual authors in case of co-authorship, the number of articles in each field is increased up to $N_f^D = \sum_a a N_f(a)$. Let $C_f^D(a)$ be the a -replica of $C_f(a)$. Thus, for example, if $a = 3$ and $C_f(3) = (3, 7)$, then $C_f^D(3) = (3, 3, 3, 7, 7, 7)$. We can write the citation distribution of field f in the double extended count as $C_f^D = \{C_f^D(a), a = 1, \dots, A_f\}$. For every f and a , let $\mu_{f1}^D(a)$ be the mean of $C_f^D(a)$. Since $C_f^D(a)$ is the a -replica of $C_f(a)$, we have $\mu_{f1}^D(a) = \mu_{f1}(a)$ for every f and a . For every f , if μ_{f1}^D is the mean of C_f^D we have:

$$\mu_{f1}^D = \sum_a w_f^D(a) \mu_{f1}(a), \quad (2)$$

where $w_f^D(a) = a N_f(a)/N_f^D$ is the proportion of articles with a authors in field f in the double extended count. Thus, the only difference between expressions (1) and (2) is the weighting scheme, i.e. the proportion of articles with a authors in field f in the single and the double extended count. For any a , let $\mu_{f2}(a)$, μ_{f2} , and μ_{f2}^D be the second mean of distributions $C_f^D(a)$, C_f , and C_f^D . The relationship between μ_{f2} and μ_{f2}^D is similar to the one between μ_{f1} and μ_{f1}^D in expressions (1) and (2). The information for the two means of C_f and C_f^D is in Table AIII.1.

Table AII.1 around here

For each f , given expressions (1) and (2), under the assumption that $\mu_{f1}(a)$ increases with a we expect $\mu_{f1} < \mu_{f1}^D$. A similar argument leads us to expect $\mu_{f2} < \mu_{f2}^D$. For all f , we observe in Table AIII.1 that $\mu_{fj} < \mu_{fj}^D$ for $j = 1, 2$, but the difference is small. This means that for any two numbers of authors a and a' , $\mu_{fj}(a)$ is quite similar to $\mu_{fj}(a')$ for $j = 1, 2$. The key question, however, is how different the skewness of field citation distributions is in the extended and the double extended counts. The CSS results in both cases are in Tables AIII.2 and AIII.3.

Table AII.2 and AII.3 around here

We want to emphasize the following two findings. Firstly, the small size of CVs over 29 fields in Tables AIII.2 and AIII.3 indicate that in both counts the between-field variation is very small. Secondly, recall that, in every field f , for every a the distribution $C_f^D(a)$ is the a -replica of $C_f(a)$. Consequently, since the CSS technique is replication invariant, the skewness of $C_f^D(a)$ and $C_f(a)$ coincide. However, the difference between the weights $w_f^D(a)$ and $w_f(a)$ would tend to increase with a . Thus, under the assumption that the skewness of $C_f(a)$ increases with a (for reasons of space, the evidence concerning this conjecture is available on request), we expect that the degree of skewness of field citation distributions in the double extended count is somewhat greater than in the extended count. Except for three fields –Civil Engineering & Construction, Economics & Business, and Mathematics– this is what we find in the remaining 26 fields in Tables AII.2 and AII.3, and hence in the average. However, the difference is so small that for all practical purposes we may continue the analysis focusing on either the extended or the double extended count.

Table II.1. First and second means of field citation distributions for all authors in the extended and the double extended counts, and average and coefficient of variation of these magnitudes over the 29 fields

FIELDS	Extended count		Double extended count	
	μ_{f1} (1)	μ_{f2} (2)	μ_{f1}^D (3)	μ_{f2}^D (4)
AGRICULTURE & FOOD SCIENCE	13.8	39.1	15.1	47.9
ASTRONOMY & ASTROPHYSICS	17.4	51.8	26.1	98.9
BASIC LIFE SCIENCES	23.8	73.1	29.2	102.6
BASIC MEDICAL SCIENCES	14.7	43.1	16.0	49.5
BIOLOGICAL SCIENCES	17.2	56.7	24.1	100.5
BIOMEDICAL SCIENCES	19.8	58.6	22.6	72.7
CHEMISTRY & CHEMICAL ENG.	16.7	50.1	17.7	54.4
CIVIL ENG. & CONSTRUCTION	10.2	30.4	10.6	31.1
CLINICAL MEDICINE	18.6	56.1	23.6	79.4
COMPUTER SCIENCES	8.3	32.4	10.5	46.1
EARTH SCIENCES & TECHNOLOGY	15.2	45.2	17.4	54.7
ECONOMICS & BUSINESS	13.4	46.3	14.5	50.5
EDUCATIONAL SCIENCES	9.7	30.2	10.7	34.4
ELECTRICAL ENG. & TELECOMM.	8.6	29.4	9.0	31.1
ENERGY SC. & TECHNOLOGY	11.7	35.7	12.3	41.0
ENVIRONMENTAL SCS & TECH.	16.2	48.0	18.4	58.8
GENERAL & INDUSTRIAL ENG.	8.1	23.9	8.0	24.1
HEALTH SCIENCES	13.0	38.5	14.3	43.0
INFORMATION & COMM. SCS.	10.4	34.2	11.4	38.1
INSTS. & INSTRUMENTATION	10.0	33.0	10.6	40.2
LAW & CRIMINOLOGY	7.6	22.2	11.4	40.7
MANAGEMENT & PLANNING	14.8	49.1	15.5	52.1
MATHEMATICS	6.0	22.1	6.9	24.5
MECHANICAL ENG. & AEROSPACE	9.0	25.8	9.5	29.3
PHYSICS & MATERIALS SCIENCE	12.8	42.8	16.5	57.2
PSYCHOLOGY	17.2	53.3	19.0	59.6
SOCIAL & BEHAVIORAL SCIENCES	11.7	35.0	12.8	38.0
SOCIOLOGY & ANTHROPOLOGY	11.4	34.6	13.6	43.8
STATISTICAL SCIENCES	11.0	38.2	19.2	104.4

Average over the 29 fields	13.0	40.7	15.4	53.4
Coefficient of variation	0.32	0.30	0.37	0.44

Table II.2. The skewness of field citation distributions according to the CSS approach for all authors in the single extended count, and average and coefficient of variation of the CSS results over the 29 fields

	% of articles in category:			% of total citations in category:		
	I (1)	II (2)	III (3)	I (4)	II (5)	III (6)
AGRICULTURE & FOOD SCIENCE	72.4	20.1	7.4	22.1	32.9	45.0
ASTRONOMY & ASTROPHYSICS	73.9	19.0	7.1	22.5	31.9	45.6
BASIC LIFE SCIENCES	75.0	18.8	6.2	23.2	31.6	45.1
BASIC MEDICAL SCIENCES	73.9	19.5	6.6	23.1	32.5	44.4
BIOLOGICAL SCIENCES	76.6	17.8	5.5	23.0	31.3	45.7
BIOMEDICAL SCIENCES	74.1	19.2	6.8	23.3	31.7	45.1
CHEMISTRY & CHEMICAL ENG.	74.2	19.2	6.6	22.8	32.2	45.0
CIVIL ENG. & CONSTRUCTION	73.8	18.5	7.7	21.6	32.2	46.3
CLINICAL MEDICINE	74.4	19.0	6.5	23.0	32.1	44.9
COMPUTER SCIENCES	79.2	16.3	4.6	18.7	31.3	50.0
EARTH SCIENCES & TECHNOLOGY	74.0	19.1	6.9	22.8	32.4	44.7
ECONOMICS & BUSINESS	77.0	17.0	6.1	20.3	31.0	48.7
EDUCATIONAL SCIENCES	74.7	18.8	6.5	21.1	32.3	46.6
ELECTRICAL ENG. & TELECOMM.	76.5	17.9	5.5	19.7	32.1	48.2
ENERGY SC. & TECHNOLOGY	73.9	19.2	6.9	20.8	32.4	46.8
ENVIRONMENTAL SCS & TECH.	74.1	18.8	7.1	23.2	31.8	45.0
GENERAL & INDUSTRIAL ENG.	73.6	18.6	7.8	22.2	32.1	45.7
HEALTH SCIENCES	74.0	18.9	7.1	23.3	32.5	44.3
INFORMATION & COMM. SCS.	76.1	17.9	6.0	21.1	32.0	46.9
INSTS. & INSTRUMENTATION	75.9	18.6	5.5	20.0	32.2	47.8
LAW & CRIMINOLOGY	72.6	19.6	7.8	20.4	32.8	46.8
MANAGEMENT & PLANNING	76.0	17.7	6.3	20.5	31.2	48.3
MATHEMATICS	78.1	16.5	5.4	19.6	32.2	48.2
MECHANICAL ENG. & AEROSPACE	72.2	20.1	7.7	20.3	32.5	47.3
PHYSICS & MATERIALS SCIENCE	76.1	18.0	5.9	20.0	31.4	48.6
PSYCHOLOGY	75.0	18.4	6.6	22.3	32.0	45.7
SOCIAL & BEHAVIORAL SCIENCES	73.6	19.0	7.4	21.0	31.8	47.1
SOCIOLOGY & ANTHROPOLOGY	74.4	18.7	6.9	21.9	32.2	45.9
STATISTICAL SCIENCES	76.9	17.9	5.2	19.4	31.6	49.0
Average over the 29 fields	74.9	18.6	6.5	21.5	32.0	46.5
Coefficient of variation	0.02	0.05	0.13	0.06	0.01	0.03

Table II.3. The skewness of field citation distributions according to the CSS approach for all authors in the double extended count, and average and coefficient of variation of the CSS results over the 29 fields

	% of articles in category:			% of total citations in category:		
	I	II	III	I	II	III
	(1)	(2)	(3)	(4)	(5)	(6)
AGRICULTURE & FOOD SCIENCE	75.8	18.4	5.8	23.4	31.9	44.8
ASTRONOMY & ASTROPHYSICS	80.1	16.6	3.3	24.8	29.6	45.6
BASIC LIFE SCIENCES	78.2	16.8	5.0	23.3	30.0	46.7
BASIC MEDICAL SCIENCES	75.6	18.6	5.8	24.6	31.9	43.5
BIOLOGICAL SCIENCES	81.3	15.0	3.7	22.2	28.4	49.5
BIOMEDICAL SCIENCES	76.2	18.1	5.7	23.4	31.0	45.6
CHEMISTRY & CHEMICAL ENG.	75.0	18.8	6.2	23.1	31.7	45.2
CIVIL ENG. & CONSTRUCTION	73.0	19.3	7.7	20.7	32.8	46.5
CLINICAL MEDICINE	77.1	17.8	5.1	22.9	31.0	46.1
COMPUTER SCIENCES	81.3	15.0	3.6	18.1	29.8	52.1
EARTH SCIENCES & TECHNOLOGY	75.3	18.5	6.2	22.5	31.6	45.9
ECONOMICS & BUSINESS	77.0	17.0	6.0	20.2	30.8	49.0
EDUCATIONAL SCIENCES	75.5	18.6	5.9	21.3	32.1	46.6
ELECTRICAL ENG. & TELECOMM.	76.6	18.3	5.1	18.9	32.1	49.0
ENERGY SC. & TECHNOLOGY	76.5	17.8	5.7	21.3	32.2	46.5
ENVIRONMENTAL SCS & TECH.	75.8	18.3	5.8	22.8	31.6	45.6
GENERAL & INDUSTRIAL ENG.	74.3	18.6	7.1	22.5	33.0	44.5
HEALTH SCIENCES	74.3	19.2	6.4	23.0	32.9	44.1
INFORMATION & COMM. SCS.	76.4	17.9	5.7	21.2	31.9	46.9
INSTS. & INSTRUMENTATION	78.6	16.4	4.9	19.1	30.8	50.1
LAW & CRIMINOLOGY	78.2	18.1	3.7	22.1	32.5	45.4
MANAGEMENT & PLANNING	76.4	17.4	6.2	20.8	31.1	48.1
MATHEMATICS	76.7	17.8	5.4	17.2	31.8	50.9
MECHANICAL ENG. & AEROSPACE	74.6	19.0	6.4	21.7	32.7	45.7
PHYSICS & MATERIALS SCIENCE	77.4	17.6	5.0	21.8	31.3	46.9
PSYCHOLOGY	75.1	18.7	6.3	21.6	31.7	46.7
SOCIAL & BEHAVIORAL SCIENCES	73.6	19.2	7.2	21.6	31.9	46.5
SOCIOLOGY & ANTHROPOLOGY	75.6	18.3	6.0	21.5	31.5	47.0
STATISTICAL SCIENCES	85.2	12.4	2.4	19.4	25.8	54.8
Average over the 29 fields	76.8	17.7	5.5	21.6	31.3	47.1
Coefficient of variation	0.03	0.08	0.22	0.08	0.05	0.05

APPENDIX III

In the following three examples, it suffices to study the first part of the CSS results for any citation distribution, namely, the percentages of articles (p_1, p_2, p_3) in categories I, II, and III.

EXAMPLE 1

1. Individual citation distributions

Consider two authors with the following citation distributions:

$$\text{Author 1: } c(1) = (0, 0, 5, 6, 6, 8, 10)$$

$$\text{Author 2: } c(2) = (0, 2, 2, 2, 3, 3, 4, 6, 6, 6, 8, 8, 10, 10)$$

The number of articles, and the two means for both authors and the field citation distribution are under *Raw data* in Table I.

Table I. Number of articles, first and second mean for authors and the field citation distribution. Raw and productivity normalized data

	Author 1	Author 2	Field	
<i>Raw data</i>				
$n(i)$	7	14	N	21
$m_1(i)$	5	5	M_1	5
$m_2(i)$	7.5	7.714	M_2	7.636
<i>Productivity normalized data</i>				
$n'(i)$	14	14	N	28
$m'_1(i)$	5	5	M'_1	5
$m'_2(i)$	7.5	7.714	M'_2	7.6

The CSS results on the skewness of the authors' distributions are:

$$p(1) = (p_1(1), p_2(1), p_3(1)) = (3/7, 2/7, 2/7) = (42.8, 28.6, 28.6),$$

$$p(2) = (p_1(2), p_2(2), p_3(2)) = (7/14, 3/14, 4/14) = (50.0, 21.4, 28.6).$$

Therefore, the average skewness is:

$$p = (p_1, p_2, p_3) = (46.4, 25.0, 28.6),$$

where $p_j = (\sum_i p_j(i))/2, j = 1, 2, 3$.

2. Field citation distribution

As indicated under *Raw data* in Table II, the number of articles, as well as the first and the second means of the field citation distribution are:

$$N = 7 + 14 = 21, M_1 = 5, M_2 = 7.636.$$

Therefore, the skewness of the field citation distribution is:

$$P = (P_1, P_2, P_3) = (10/21, 5/21, 6/21) = (47.6, 23.8, 28.6).$$

The difference in skewness between p and P is due to differences in their individual productivity.

3. Correcting for differences in individual productivity

There are two ways of correcting for differences in individual productivity. In the first place, we can consider the weighted average skewness where, for every j ,

$$p'_j = \sum_i w(i) p_j(i),$$

where $w(i) = n(i)/N$. In this example, $w(1) = 7/21 = 1/3$, and $w(2) = 14/21 = 2/3$. Therefore, the weighted average skewness will be:

$$p' = (p'_1, p'_2, p'_3) = (1/3) (42.8, 28.6, 28.6) + (2/3) (50.0, 21.4, 28.6) = (47.6, 23.8, 28.6),$$

which coincides with the skewness of the field citation distribution. Therefore, in this case the skewness gap is entirely explained by the difference in individual productivity.

In the second place, we will correct for productivity differences replicating the first individual citation distribution so that it has 14 articles, that is:

$$c''(1) = (0, 0, 0, 0, 5, 5, 6, 6, 6, 6, 8, 8, 10, 10).$$

The mean citations for both authors and the field citation distribution are presented under *Productivity normalized data* in Table II.

Since the CSS technique is replication invariant, the values

$$p''(1) = (p''_1(1), p''_2(1), p''_3(1)) = (42.8, 28.6, 28.6),$$

and

$$p'' = (p''_1, p''_2, p''_3) = (46.4, 25.0, 28.6),$$

remain invariant. However, at the field level we have:

$$P'' = (P''_1, P''_2, P''_3) = (13/28, 7/28, 8/28) = (46.4, 25.0, 28.6).$$

Therefore, the skewness of the field citation distribution controlling for differences in individual productivity, i. e. the *basic field skewness* coincides with the simple average skewness. As before, we conclude that the skewness gap is entirely explained by the difference in individual productivity.

EXAMPLE 2

1. Individual citation distributions

Consider two authors with the following citation distributions:

$$\underline{\text{Author 1:}} \quad c(1) = (0, 0, 5, 6, 6, 8, 10)$$

$$\underline{\text{Author 2:}} \quad c(2) = (0, 3, 7, 12, 13, 18, 24)$$

The number of articles, and the two means for both authors and the field citation distribution are under *Raw data* in Table II.

Table II. Number of articles, first and second mean for authors and the field citation distribution. Raw and citation impact normalized data

	Author 1	Author 2	Field	
<i>Raw data</i>				
$n(i)$	7	7	N	14
$m_1(i)$	5	11	M_1	8
$m_2(i)$	7.5	16.75	M_2	12.125
<i>Citation impact normalized data</i>				
$m'_1(i)$	8	8	M'_1	8
$m'_2(i)$	12	12.18	M'_2	12.09

For both authors, we have

$$p(i) = (p_1(i), p_2(i), p_3(i)) = (3/7, 2/7, 2/7) = (42.8, 28.6, 28.6), i = 1, 2.$$

Therefore, the average skewness is also

$$p = (p_1, p_2, p_3) = (42.8, 28.6, 28.6),$$

where $p_j = (\sum_i p_j(i))/2, j = 1, 2, 3$.

2. Field citation distribution

As indicated under *Raw data* in Table II, the number of articles, as well as the first and the second means of the field citation distribution are:

$$N = 7 + 7 = 14, M_1 = (5 + 11)/2 = 8, M_2 = (7.5 + 16.75)/2 = 12.125.$$

Therefore, the skewness of the field citation distribution is:

$$P = (P_1, P_2, P_3) = (9/14, 2/14, 3/14) = (64.3, 14.3, 21.4).$$

3. Correcting for differences in individual mean citations

Why is there such a difference in skewness between p and P ? Because individuals differ in their citation impact. Therefore, to establish the basic skewness of the field citation distribution we must control for individual differences in the first mean citation.

Consider multiplying every citation count $c_k(i)$ for article $k = 1, \dots, 7$ published by author $i = 1, 2$ by the quantity $\mu^*/m_1(i)$, where $\mu^* = 8$. The normalized citation distributions become:

$$\text{Author 1: } c'(1) = (0, 0, 8, 9.6, 9.6, 12.8, 16)$$

$$\text{Author 2: } c'(2) = (0, 2.18, 5.09, 8.73, 9.45, 13.09, 17.45)$$

The mean citations for both authors and the field citation distribution are presented under *Citation impact normalized data* in Table II. Since the CSS technique is scale independent, the skewness of the individual citation distributions does not vary. However, when all individual citation distributions have the same size, the same first mean citation, and the same skewness in spite of a very small difference in the second mean citation, it can be verified that the basic skewness of the field citation distribution, $P' = (P'_1, P'_2, P'_3)$, becomes equal to the average skewness of the individual distributions:

$$P' = (P'_1, P'_2, P'_3) = (6/14, 4/14, 4/14) = (42.8, 28.6, 28.6).$$

Therefore, in this case we conclude that the skewness gap is entirely explained by the difference between the individuals' first mean citations.

EXAMPLE 3

1. Individual citation distributions

Consider three authors with the following citation distributions:

$$\text{Author 1: } c(1) = (0, 0, 5, 6, 6, 8, 10),$$

$$\text{Author 2: } c(2) = (0, 0, 0, 0, 3, 3, 4, 4, 6, 6, 6, 10, 14, 21),$$

Author 3: $c(3) = (0, 5, 5, 5, 5, 5, 5, 5, 5, 7, 7, 7, 7, 15, 15, 15, 15, 15, 15, 15, 36, 36, 38, 38)$.

The number of articles, and the two means for the three authors and the field citation distribution are under *Raw data* in Table III.

Table III. Number of articles, first and second mean for authors and the field citation distribution. Raw data, citation impact normalized data, and both citation impact and productivity normalized data

	Author 1	Author 2	Author 3	Field	
<i>Raw data</i>					
$n(i)$	7	14	42	N	63
$m_1(i)$	5	5.5	8	M_1	7.11
$m_2(i)$	7.5	10.5	22.33	M_2	19.47
<i>First mean citation normalized data</i>					
$n'(i)$	7	14	42	N'	63
$m'_1(i)$	8	8	8	M'_1	8
$m'_2(i)$	12	15.27	22.33	M'_2	18.52
<i>First mean citation and productivity normalized data</i>					
$n''(i)$	14	14	14	N''	42
$m''_1(i)$	8	8	8	M''_1	8
$m''_2(i)$	12	15.27	22.33	M''_2	15.39

The CSS results on the skewness of the authors' distributions are:

$$p(1) = (p_1(1), p_2(1), p_3(1)) = (3/7, 2/7, 2/7) = (42.8, 28.6, 28.6),$$

$$p(2) = (p_1(2), p_2(2), p_3(2)) = (8/14, 4/14, 2/14) = (57.1, 28.6, 14.3),$$

$$p(3) = (p_1(3), p_2(3), p_3(3)) = (30/42, 8/42, 4/42) = (71.4, 19.1, 9.5).$$

Thus, productivity, mean citation, and skewness increase as we go from author 1 to author 3. On the other hand, the average skewness is:

$$p = (p_1, p_2, p_3) = (57.1, 25.4, 17.5),$$

where $p_j = (\sum_i p_j(i))/3, j = 1, 2, 3$.

2. Field citation distribution

As indicated under *Raw data* in Table III, the number of articles, as well as the first and the second means of the field citation distribution are:

$$N = 7 + 14 + 42 = 63, M_1 = 7.11, M_2 = 18.05.$$

Therefore, the skewness of the field citation distribution is:

$$P = (P_1, P_2, P_3) = (46/63, 12/63, 5/63) = (73.02, 19.05, 7.93).$$

Why is there such a difference in skewness between p and P ? Because individuals differ in their productivity (number of articles per author) and their citation impact (first and second mean citations per author). Therefore, to establish the basic skewness of the field citation distribution preserving the skewness of the individual citation distributions, we must control for individual differences in productivity and in the first mean citation.

3. Correcting for differences in individual mean citations

We will correct for mean citation differences giving all authors the same first mean citation, say 8 citations per article. Thus, citations received by authors 1 and 2 must be multiplied by the following factors: $8/5 = 1.6$ and $8/5.5 = 1.45$, respectively. Thus,

$$c'(1) = (0, 0, 8, 9.6, 9.6, 12.8, 16)$$

$$c'(2) = (0, 0, 0, 0, 4.36, 4.36, 5.8, 5.8, 8.7, 8.7, 14.5, 20.4, 30.54).$$

The number of articles, and the two means for the three authors and the field citation distribution are under *First mean citation normalized data* in Table III.

Since the CSS technique is scale-invariant, the values

$$p'(i) = (p'_1(i), p'_2(i), p'_3(i)) = (p_1(i), p_2(i), p_3(i)), i = 1, 2,$$

do not change. Therefore, the average skewness remains constant too. However, as indicated under *First mean citation normalized data* in Table III, in the new field citation distribution we have:

$$N' = \sum_i n'(i) = 63, \mu_1 = 8, \mu_2 = 19.602,$$

so that

$$P' = (P'_1, P'_2, P'_3) = (41/63, 16/63, 6/63) = (65.08, 25.4, 9.52).$$

As expected, the skewness of this new distribution is smaller than before, and the difference between P and P' is due to individual differences in the first mean citation.

4. Correcting for differences in individual productivity

We will correct for productivity differences modifying individual citation distributions so that all have $63/3 = 21$ articles. Thus, citations received by the three authors are now weighted by a factor $21/n(i)$, $i = 1, 2, 3$, so that they all have 21 articles:

$$c''(1) = (0, 0, 0, 0, 0, 0, 8, 8, 8, 9.6, 9.6, 9.6, 9.6, 9.6, 12.8, 12.8, 12.8, 16, 16, 16),$$

$c''(2) = (0 \text{ a total of 6 times, } 4.36 \text{ a total of 3 times, } 5.8 \text{ a total of 3 times, } 8.7 \text{ a total of 4.5 times, and } 14.5, 20.3, 30.4 \text{ a total of 1.5 times each}),$

$$c''(3) = (0, 0, 0, 0, 0, 0, 0, 0, 0, 5, 5, 5, 5, 7, 7, 15, 15, 15, 15, 36, 38).$$

The number of articles, and the two means for the three authors and the field citation distribution are under *First mean citation and productivity normalized data* in Table III.

Since the CSS technique is replication invariant, the values

$$p''(i) = (p''_1(i), p''_2(i), p''_3(i)) = (p_1(i), p_2(i), p_3(i)), i = 1, 2, 3,$$

remain invariant for all authors. Therefore, the average skewness is the same as before. However, at the field level we now have:

$$P'' = (36/63, 22/63, 5/63) = (57.1, 30.2, 12.7).$$

The difference between P' and P'' is due to individual differences in productivity. On the other hand, the skewness of the field citation distribution controlling for differences in the first mean citation and productivity, i. e. the *basic field skewness* in $P'' = (57.1, 30.2, 12.7)$ is better approximated by the average skewness $p = (57.1, 25.4, 17.5)$. Nevertheless, the difference between P'' and p is due to the remaining differences in the authors' second means that cannot be eliminated without changing the original individual skewness.

Table A. The weighted average skewness of individual citation distributions for very productive authors in the double extended count, and average and coefficient of variation of the CSS results over the 29 fields

	% of articles in category:			% of total citations in category:		
	I (1)	II (2)	III (3)	I (4)	II (5)	III (6)
AGRICULTURE & FOOD SCIENCE	66.6	20.9	12.5	24.2	32.0	43.7
ASTRONOMY & ASTROPHYSICS	78.1	16.9	5.0	25.8	28.4	45.8
BASIC LIFE SCIENCES	68.5	20.0	11.5	25.2	31.0	43.8
BASIC MEDICAL SCIENCES	65.4	21.2	13.4	25.9	32.2	41.9
BIOLOGICAL SCIENCES	67.6	20.3	12.1	24.1	31.8	44.1
BIOMEDICAL SCIENCES	67.7	20.5	11.8	25.5	31.3	43.2
CHEMISTRY & CHEMICAL ENG.	68.0	20.8	11.2	25.1	31.7	43.1
CIVIL ENG. & CONSTRUCTION	64.8	21.5	13.7	22.8	34.8	42.4
CLINICAL MEDICINE	70.0	19.9	10.2	24.9	31.1	44.0
COMPUTER SCIENCES	69.4	19.3	11.4	19.1	33.5	47.5
EARTH SCIENCES & TECHNOLOGY	68.2	20.4	11.3	24.8	31.4	43.8
ECONOMICS & BUSINESS	67.5	20.0	12.5	22.5	32.2	45.4
EDUCATIONAL SCIENCES	65.0	21.2	13.8	23.2	34.9	41.9
ELECTRICAL ENG. & TELECOMM.	68.5	20.0	11.5	21.0	32.7	46.3
ENERGY SC. & TECHNOLOGY	67.5	20.2	12.3	22.3	32.7	45.0
ENVIRONMENTAL SCS & TECH.	67.2	20.5	12.2	24.7	31.7	43.6
GENERAL & INDUSTRIAL ENG.	64.4	21.8	13.8	23.3	35.4	41.3
HEALTH SCIENCES	66.7	20.7	12.6	24.1	32.0	43.8
INFORMATION & COMM. SCS.	64.5	22.2	13.2	22.8	38.5	38.7
INSTS. & INSTRUMENTATION	72.2	17.6	10.2	18.6	32.0	49.4
LAW & CRIMINOLOGY	63.9	22.1	14.0	23.6	35.3	41.2
MANAGEMENT & PLANNING	65.1	21.1	13.9	23.4	34.8	41.9
MATHEMATICS	68.2	20.1	11.7	19.0	33.4	47.6
MECHANICAL ENG. & AEROSPACE	66.2	21.0	12.8	23.1	32.6	44.3
PHYSICS & MATERIALS SCIENCE	73.6	18.9	7.5	24.0	30.9	45.0
PSYCHOLOGY	67.9	20.2	11.9	23.8	31.6	44.6
SOCIAL & BEHAVIORAL SCIENCES	63.6	22.4	14.0	23.8	38.0	38.2
SOCIOLOGY & ANTHROPOLOGY	63.9	22.3	13.8	23.6	37.9	38.5
STATISTICAL SCIENCES	68.1	19.8	12.1	21.3	32.6	46.1
----- Average over the 29 fields	67.5	20.5	12.0	23.3	33.0	43.7
----- Coefficient of variation	0.05	0.06	0.16	0.08	0.07	0.06

Table B. The skewness of field citation distributions for very productive authors in the double extended count after normalizing for differences in individual productivity and individual mean citations, and average and coefficient of variation of the CSS results over the 29 fields

	% of articles in category:			% of total citations in category:		
	I (1)	II (2)	III (3)	I (4)	II (5)	III (6)
AGRICULTURE & FOOD SCIENCE	65.8	22.3	11.9	24.1	33.3	42.5
ASTRONOMY & ASTROPHYSICS	76.9	18.6	4.5	25.8	31.4	42.8
BASIC LIFE SCIENCES	67.4	21.8	10.8	25.2	32.9	41.9
BASIC MEDICAL SCIENCES	64.5	23.0	12.5	25.9	33.3	40.8
BIOLOGICAL SCIENCES	66.7	22.1	11.2	24.0	33.4	42.6
BIOMEDICAL SCIENCES	66.7	22.1	11.2	25.4	33.0	41.6
CHEMISTRY & CHEMICAL ENG.	67.1	22.0	10.9	25.0	33.1	41.9
CIVIL ENG. & CONSTRUCTION	63.4	23.1	13.5	22.8	34.1	43.1
CLINICAL MEDICINE	68.8	21.4	9.9	24.9	32.8	42.3
COMPUTER SCIENCES	68.4	20.6	11.0	18.9	33.6	47.6
EARTH SCIENCES & TECHNOLOGY	67.4	21.8	10.8	24.6	33.0	42.4
ECONOMICS & BUSINESS	66.7	21.4	11.9	22.4	33.1	44.6
EDUCATIONAL SCIENCES	63.7	22.5	13.7	23.2	33.3	43.5
ELECTRICAL ENG. & TELECOMM.	67.4	21.4	11.2	20.7	33.7	45.6
ENERGY SC. & TECHNOLOGY	66.6	22.0	11.4	22.2	33.7	44.0
ENVIRONMENTAL SCS & TECH.	66.3	22.1	11.6	24.6	33.1	42.3
GENERAL & INDUSTRIAL ENG.	63.2	23.0	13.8	23.0	33.7	43.3
HEALTH SCIENCES	65.6	22.3	12.1	24.0	33.3	42.6
INFORMATION & COMM. SCS.	63.0	22.7	14.3	22.6	33.4	43.9
INSTS. & INSTRUMENTATION	71.0	19.2	9.8	19.2	32.0	48.8
LAW & CRIMINOLOGY	62.8	23.3	13.9	23.4	33.9	42.7
MANAGEMENT & PLANNING	64.0	22.5	13.5	23.2	33.3	43.5
MATHEMATICS	67.5	21.3	11.2	18.5	34.3	47.2
MECHANICAL ENG. & AEROSPACE	65.2	22.4	12.4	22.9	33.6	43.6
PHYSICS & MATERIALS SCIENCE	71.1	20.3	8.6	23.0	32.7	44.3
PSYCHOLOGY	66.9	21.7	11.4	23.7	33.1	43.3
SOCIAL & BEHAVIORAL SCIENCES	62.3	23.1	14.6	23.7	33.4	42.8
SOCIOLOGY & ANTHROPOLOGY	62.5	23.0	14.4	23.4	33.6	42.9
STATISTICAL SCIENCES	67.3	21.4	11.4	21.1	33.5	45.4
-----	66.4	21.9	11.7	23.2	33.3	43.6
Average over the 29 fields	0.05	0.05	0.17	0.08	0.02	0.04

Coefficient of variation						

APPENDIX IV. MERELY PRODUCTIVE AUTHORS

As indicated in the text, *merely productive* authors are defined as those who publish at least five articles in the 2000-2016 period. We denote by I_f^{**} the number of merely productive authors in field $f = 1, \dots, 29$. The results for n_f , m_{f1} and m_{f2} in each field, as well as for *Mean-size*, M_1 and M_2 are in Table AIV.1. Wide differences in individual productivity and citation impact among merely productive authors give rise again to large CVs in columns 4 to 6 and 10 to 12 in all fields. On the other hand, large CVs over the 29 fields in columns 1 to 3 and 7 to 9 again reflect the existence of large differences in production and citation practices across fields.

Table AIV.1 around here

The CSS results for (p_{f1}, p_{f2}, p_{f3}) and (s_{f1}, s_{f2}, s_{f3}) in each field and the corresponding CVs, $(cv_{f1}, cv_{f2}, cv_{f3})$ and $(cv_{f4}, cv_{f5}, cv_{f6})$ are in columns 1 to 12 in Table AIV.2. In turn, the results on (P_1, P_2, P_3) , $(CV_{f1}, CV_{f2}, CV_{f3})$, (S_1, S_2, S_3) , and $(CV_{f4}, CV_{f5}, CV_{f6})$ over the 29 fields are in the last two rows in Table AIV.2. The results are remarkable. Not surprisingly, since there are more authors involved, within-field variability is somewhat larger than before. However, on average, the degree of skewness is only slightly smaller than what is obtained for very productive authors. Less productive authors seem to have average individual citation distributions less skewed than more productive authors¹³. Finally, between-field variability is somewhat smaller than before.

Table AIV.2 around here

Next, the CSS results for $(P_{f1}^*, P_{f2}^*, P_{f3}^*)$ and $(S_{f1}^*, S_{f2}^*, S_{f3}^*)$ in each field and the corresponding CVs, as well as the results on (P_1^*, P_2^*, P_3^*) and (S_1^*, S_2^*, S_3^*) are in Table AIV.3. Field citation distributions turned out to be essentially as skewed as field citation distributions for very productive authors in Table 8. Again, between-field variability is somewhat smaller than before.

Table AIV.3 around here

¹³ In agreement with this remark, in the three fields where merely productive authors are less numerous than very productive authors because researchers with less than five articles are eliminated –Information & Communication Sciences, Social & Behavioral Sciences, and Sociology & Anthropology– the average skewness of individual citation distributions is slightly greater in Table AIV.2 than in Table 6.

Comparing Tables AIV.2 and AIV.3, we observe that the gap between the skewness of field citation distributions and the unweighted average skewness of the individual citation distributions for merely productive authors in each field is slightly greater than for very productive ones. Restricting ourselves to the skewness results for the average over the 29 fields, the amounts (P_1, P_2, P_3) and (S_1, S_2, S_3) versus (P^*_1, P^*_2, P^*_3) and (S^*_1, S^*_2, S^*_3) are reproduced in rows I and II in Table AIV.4. Thus, in the first case the mean is 14.8 points greater than the median and 13.8% of highly cited articles account for 40.2% of all citations, whereas in the second case the mean is 25.3 points greater than the median and only 6.3% of highly cited articles account for 46.2% of all citations.

Table AIV.4 around here

The conjecture is that, as we saw for very productive authors, this skewness gap can be essentially explained by individual differences in mean citation rates and the number of articles per authors. In the first place, consider the skewness of field citation distributions controlling for the within-field differences in individual productivity. The CSS results for the average over the 29 fields appear in row III in Table AIV.4 (detailed field results are available on request). We verify that, as for very productive authors, the skewness of field citation distributions remains basically unchanged.

In the second place, consider the skewness of field citation distributions controlling for the within-field differences in individual mean citations.¹⁴ The CSS aggregate results appear in row IV in Table AIV.4, whereas detailed field results are in table IV.5. As before, the skewness of field citation distributions is greatly reduced. The minimal resulting gap between the basic skewness in any field (row IV) and the average skewness at the individual level (row I) is due to differences in the second mean of the individual normalized citation distributions that cannot be eliminated without changing the original individual skewness.

Table AIV.5 around here

¹⁴ As with very productive authors, this normalization can only be applied for authors with a positive mean citation. However, merely productive authors receiving no citations only represent 0.114% of the total (details by field are available on request).

Note that what we have called the basic skewness of field citation distributions is of the same order of magnitude than what we found for very productive authors: comparing the row IV in Table AIV.4 and row V in Table 9, we observe that, on average over the 29 fields, the basic skewness in this case is (64.7, 22.5, 12.8) and (23.1, 33.5, 43.4), whereas for very productive authors it is equal to (66.4, 21.9, 11.7) and (23.2, 33.2, 43.6). The only difference between the two cases is that, since there are more merely productive than very productive authors, the within-field variation of individual productivity and individual mean citations is greater for the former than for the latter (detailed evidence is available on request). Consequently, as we have seen, the gap between the skewness of field citation distributions and the average skewness of the individual citation distributions for merely productive authors in each field is slightly greater than for very productive ones. However, once we control for such differences, we arrive to a very similar basic skewness in each field.

The consequence of this result is very helpful: for all practical purposes, in our analysis we may focus on either of the two notions of productive authors.

Table AIV.1. Average and coefficient of variation of the mean productivity and the first and second mean citations for merely productive authors in each field, and average and coefficient of variation of these magnitudes over the 29 fields

	<i>Mean-size_f</i> (1)	<i>CV</i> (2)	<i>m_{f1}</i> (3)	<i>CV</i> (4)	<i>m_{f2}</i> (5)	<i>CV</i> (6)
AGRICULTURE & FOOD SCIENCE	12.6	1.07	15.2	1.49	38.7	2.14
ASTRONOMY & ASTROPHYSICS	46.1	1.46	22.1	1.52	77.1	2.49
BASIC LIFE SCIENCES	13.6	1.13	29.5	1.63	80.7	2.48
BASIC MEDICAL SCIENCES	10.8	1.04	16.4	1.37	41.0	2.31
BIOLOGICAL SCIENCES	12.4	0.99	25.3	2.32	73.9	3.07
BIOMEDICAL SCIENCES	13.8	1.14	23.2	1.54	60.8	2.14
CHEMISTRY & CHEMICAL ENG.	16.6	1.48	17.1	1.58	43.0	3.18
CIVIL ENG. & CONSTRUCTION	10.9	0.93	10.9	1.21	24.9	1.42
CLINICAL MEDICINE	17.8	1.40	22.4	1.72	62.1	2.55
COMPUTER SCIENCES	11.6	0.96	9.5	3.08	29.3	4.52
EARTH SCIENCES & TECHNOLOGY	15.2	1.07	17.3	1.31	46.1	2.08
ECONOMICS & BUSINESS	10.8	0.77	15.8	1.36	41.5	1.74
EDUCATIONAL SCIENCES	9.6	0.92	12.6	1.35	30.9	1.63
ELECTRICAL ENG. & TELECOMM.	12.9	1.14	9.2	3.07	27.6	6.64
ENERGY SC. & TECHNOLOGY	11.5	0.97	12.1	1.55	32.3	2.42
ENVIRONMENTAL SCS & TECH.	12.6	1.04	18.7	1.45	49.0	2.25
GENERAL & INDUSTRIAL ENG.	9.9	0.85	9.2	1.02	21.6	1.33
HEALTH SCIENCES	12.1	1.03	14.9	1.29	38.4	2.29
INFORMATION & COMM. SCS.	9.3	0.83	13.8	1.30	36.7	2.30
INSTS. & INSTRUMENTATION	12.3	0.88	10.0	1.79	31.6	2.67
LAW & CRIMINOLOGY	10.3	0.95	10.2	1.09	23.9	2.56
MANAGEMENT & PLANNING	9.0	0.65	20.5	1.16	52.1	1.59
MATHEMATICS	13.5	1.07	6.2	1.85	17.4	2.77
MECHANICAL ENG. & AEROSPACE	12.0	1.01	9.6	1.34	23.5	2.29
PHYSICS & MATERIALS SCIENCE	27.3	2.37	14.3	1.81	41.5	3.08
PSYCHOLOGY	13.4	1.06	19.1	1.32	51.6	2.11
SOCIAL & BEHAVIORAL SCIENCES	8.7	0.67	15.7	1.24	39.2	2.15
SOCIOLOGY & ANTHROPOLOGY	8.9	0.69	16.4	1.35	40.8	1.79
STATISTICAL SCIENCES	11.6	0.96	15.2	2.53	47.0	3.37
<hr/>						
	<i>Mean-size</i>	<i>CV</i>	<i>M₁</i>	<i>CV</i>	<i>M₂</i>	<i>CV</i>
Average over the 29 fields	13.7	1.05	15.6	1.61	42.2	2.53
Coefficient of variation	0.52	0.31	0.35	0.33	0.39	0.40

Mean-size_f = average of the sizes of distributions $c_f(i)$, or individual productivities $n_f(i)$, over the I_f^{**} very productive authors in field f

m_{f1} = average of the individual first means $m_{f1}(i)$ of distributions $c_f(i)$ over the I_f^{**} very productive authors in field f

m_{f2} = average of the individual second means $m_{f2}(i)$ of distributions $c_f(i)$ over the I_f^{**} very productive authors in field f

Table AIV.2. The skewness of individual citation distributions according to the CSS approach for merely productive authors in the double extended count, and average and coefficient of variation of the CSS results over the 29 fields

Within-field results: average and coefficient of variation of the CSS results for all individuals in each field

Between-field results: average and coefficient of variation of the CSS results over the 29 fields

	\hat{p}_1	\hat{p}_2	\hat{p}_3	cv_1	cv_2	cv_3	s_1	s_2	s_3	cv_4	cv_5	cv_6
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
AGRICULTURE & FOOD SCIENCE	63.9	21.7	14.4	0.18	0.38	0.56	24.1	35.8	40.1	0.46	0.46	0.44
ASTRONOMY & ASTROPHYSICS	68.7	20.0	11.4	0.16	0.36	0.63	24.7	33.1	42.1	0.37	0.43	0.35
BASIC LIFE SCIENCES	64.8	21.1	14.1	0.18	0.38	0.56	25.4	34.5	40.1	0.41	0.45	0.43
BASIC MEDICAL SCIENCES	63.2	22.0	14.8	0.19	0.38	0.56	25.8	35.6	38.6	0.45	0.45	0.46
BIOLOGICAL SCIENCES	64.8	21.2	13.9	0.18	0.39	0.58	23.9	36.1	40.0	0.46	0.48	0.46
BIOMEDICAL SCIENCES	64.4	21.4	14.2	0.18	0.38	0.55	25.4	34.7	39.9	0.41	0.44	0.42
CHEMISTRY & CHEMICAL ENG.	64.0	21.8	14.2	0.18	0.38	0.55	25.2	34.9	39.9	0.43	0.44	0.42
CIVIL ENG. & CONSTRUCTION	63.5	22.0	14.5	0.19	0.39	0.58	22.9	37.1	40.0	0.52	0.48	0.47
CLINICAL MEDICINE	65.2	21.1	13.7	0.17	0.37	0.56	25.0	34.4	40.6	0.41	0.44	0.40
COMPUTER SCIENCES	66.8	20.8	12.4	0.19	0.43	0.66	19.3	40.2	40.5	0.71	0.55	0.55
EARTH SCIENCES & TECHNOLOGY	64.8	21.3	13.9	0.18	0.38	0.56	24.3	35.0	40.6	0.43	0.45	0.42
ECONOMICS & BUSINESS	65.2	21.1	13.6	0.18	0.40	0.59	22.1	37.0	40.9	0.51	0.49	0.47
EDUCATIONAL SCIENCES	63.8	21.7	14.5	0.19	0.39	0.58	23.3	36.9	39.8	0.51	0.48	0.49
ELECTRICAL ENG. & TELECOMM.	65.5	21.4	13.1	0.18	0.42	0.62	20.9	38.6	40.5	0.61	0.51	0.50
ENERGY SC. & TECHNOLOGY	64.9	21.4	13.7	0.19	0.42	0.60	22.6	37.6	39.8	0.57	0.51	0.50
ENVIRONMENTAL SCS & TECH.	64.3	21.5	14.3	0.18	0.38	0.56	24.5	35.3	40.2	0.44	0.45	0.43
GENERAL & INDUSTRIAL ENG.	63.4	22.2	14.4	0.19	0.41	0.59	23.3	37.6	39.1	0.55	0.49	0.50
HEALTH SCIENCES	64.1	21.6	14.3	0.18	0.38	0.56	24.1	35.7	40.2	0.46	0.45	0.44
INFORMATION & COMM. SCS.	64.2	21.7	14.2	0.19	0.39	0.59	22.6	37.8	39.6	0.52	0.48	0.51
INSTS. & INSTRUMENTATION	67.8	20.0	12.2	0.19	0.45	0.67	20.2	39.1	40.7	0.63	0.53	0.53
LAW & CRIMINOLOGY	62.9	22.4	14.6	0.19	0.39	0.56	23.6	37.2	39.2	0.52	0.46	0.47
MANAGEMENT & PLANNING	64.0	21.5	14.5	0.19	0.39	0.57	23.3	36.6	40.1	0.47	0.46	0.48
MATHEMATICS	66.3	21.2	12.4	0.18	0.43	0.64	18.7	40.0	41.2	0.78	0.54	0.52
MECHANICAL ENG. & AEROSPACE	63.7	22.1	14.2	0.19	0.40	0.57	22.8	37.1	40.1	0.53	0.48	0.47
PHYSICS & MATERIALS SCIENCE	66.2	20.9	12.8	0.17	0.39	0.59	22.6	35.9	41.6	0.51	0.47	0.43
PSYCHOLOGY	65.0	21.1	13.9	0.18	0.38	0.57	23.8	35.4	40.8	0.44	0.45	0.43
SOCIAL & BEHAVIORAL SCIENCES	63.6	21.5	14.9	0.19	0.38	0.58	23.8	36.6	39.6	0.48	0.47	0.49
SOCIOLOGY & ANTHROPOLOGY	63.7	21.6	14.7	0.19	0.38	0.58	23.6	36.8	39.6	0.48	0.47	0.49
STATISTICAL SCIENCES	65.7	21.0	13.2	0.18	0.41	0.61	21.1	38.2	40.7	0.56	0.51	0.50
<hr/>												
	P_1	P_2	P_3	CV_1	CV_2	CV_3	S_1	S_2	S_3	CV_4	CV_5	CV_6
Average over the 29 fields	64.8	21.4	13.8	0.18	0.39	0.59	23.2	36.6	40.2	0.51	0.47	0.46
Coefficient of variation	0.02	0.03	0.06	0.04	0.05	0.05	0.08	0.05	0.02	0.18	0.07	0.09

Table AIV.3. The skewness of field citation distributions according to the CSS approach for merely productive authors in the double extended count, and average and coefficient of variation of the CSS results over the 29 fields

	% of articles in category:			% of total citations in category:		
	I (1)	II (2)	III (3)	I (4)	II (5)	III (6)
AGRICULTURAL & FOOD SCIENCES	73.1	20.0	6.9	23.0	32.8	44.2
ASTRONOMY & ASTROPHYSICS	79.6	17.1	3.4	24.9	29.9	45.3
BASIC LIFE SCIENCES	78.1	16.8	5.1	23.8	29.9	46.3
BASIC MEDICAL SCIENCES	72.4	20.5	7.1	24.3	32.7	43.0
BIOLOGICAL SCIENCES	80.3	15.7	3.9	22.7	28.8	48.5
BIOMEDICAL SCIENCES	76.2	18.0	5.8	24.6	30.9	44.5
CHEMISTRY & CHEMICAL ENG.	74.4	19.2	6.4	23.5	32.0	44.5
CIVIL ENG. & CONSTRUCTION	71.8	19.9	8.3	21.0	32.4	46.6
CLINICAL MEDICINE	76.7	18.0	5.4	23.5	31.0	45.5
COMPUTER SCIENCES	78.9	16.4	4.6	18.7	30.7	50.6
EARTH SCIENCES & TECHNOLOGY	74.1	19.2	6.7	23.1	31.6	45.3
ECONOMICS & BUSINESS	76.5	17.2	6.3	20.7	30.9	48.3
EDUCATIONAL SCIENCES	74.3	18.8	6.9	22.5	32.0	45.5
ELECTRICAL ENG. & TELECOMM.	76.5	18.0	5.4	20.4	31.9	47.6
ENERGY SC. & TECHNOLOGY	75.2	18.3	6.5	20.5	32.0	47.5
ENVIRONMENTAL SCS & TECH.	74.4	19.1	6.5	23.5	31.7	44.8
GENERAL & INDUSTRIAL ENG.	70.4	21.0	8.6	21.4	33.1	45.5
HEALTH SCIENCES	73.6	19.4	7.0	23.8	32.3	43.9
INFORMATION & COMM. SCS.	75.4	18.1	6.5	22.6	31.9	45.5
INSTS. & INSTRUMENTATION	77.6	16.5	5.9	17.9	30.3	51.8
LAW & CRIMINOLOGY	69.9	21.1	9.0	22.0	32.8	45.1
MANAGEMENT & PLANNING	75.2	18.0	6.9	22.1	31.3	46.7
MATHEMATICS	75.2	18.5	6.3	18.3	32.6	49.1
MECHANICAL ENG. & AEROSPACE	73.6	19.0	7.4	23.1	32.0	44.9
PHYSICS & MATERIALS SCIENCE	77.0	17.9	5.1	22.1	31.2	46.6
PSYCHOLOGY	74.8	18.6	6.6	23.1	31.7	45.2
SOCIAL & BEHAVIORAL SCIENCES	73.1	19.5	7.3	23.7	32.9	43.4
SOCIOLOGY & ANTHROPOLOGY	74.7	18.6	6.7	23.3	31.6	45.0
STATISTICAL SCIENCES	79.8	16.3	3.9	20.5	30.3	49.2
----- Average over the 29 fields	75.3	18.4	6.3	22.2	31.6	46.2
----- Coefficient of variation	0.03	0.08	0.21	0.08	0.03	0.05

Table AIV.4. Average and coefficient of variation of the CSS results over the 29 fields. Selected individual and field citation distributions for merely productive authors

	% of articles in category:			% of total citations in category:		
	I	II	III	I	II	III
I. Average individual citation distributions of merely productive authors in each field						
Average over 29 fields	64.8	21.4	13.8	23.2	36.6	40.2
Coefficient of variation	0.02	0.03	0.06	0.08	0.04	0.02
II. Field citation distributions						
Average over 29 fields	75.3	18.4	6.3	22.2	31.6	46.2
Coefficient of variation	0.03	0.07	0.21	0.08	0.03	0.05
III. Field citation distributions controlling for individual productivity within each field						
Average over 29 fields	75.7	18.2	6.1	22.2	31.4	46.3
Coefficient of variation	0.03	0.08	0.20	0.07	0.04	0.05
IV. Field citation distributions controlling for individual productivity and individual mean citations within each field						
Average over 29 fields	64.7	22.5	12.8	23.1	33.5	43.4
Coefficient of variation	0.02.	0.03	0.07	0.08	0.01	0.04

Table IV.5. The skewness of field citation distributions for merely productive authors in the double extended count after normalizing for differences in individual productivity and individual mean citations, and average and coefficient of variation of the CSS results over the 29 fields

	% of articles in category:			% of total citations in category:		
	I	II	III	I	II	III
AGRICULTURE & FOOD SCIENCE	63.9	23.0	13.1	24.1	33.7	42.2
ASTRONOMY & ASTROPHYSICS	68.6	21.8	9.5	24.7	33.8	41.6
BASIC LIFE SCIENCES	64.8	22.7	12.5	25.4	33.2	41.4
BASIC MEDICAL SCIENCES	63.2	23.5	13.3	25.8	33.6	40.6
BIOLOGICAL SCIENCES	64.8	22.6	12.6	23.9	33.6	42.6
BIOMEDICAL SCIENCES	64.4	22.9	12.7	25.4	33.3	41.3
CHEMISTRY & CHEMICAL ENG.	64.0	23.3	12.8	25.1	33.7	41.1
CIVIL ENG. & CONSTRUCTION	63.5	23.1	13.5	22.8	34.1	43.1
CLINICAL MEDICINE	65.2	22.6	12.2	24.9	33.4	41.6
COMPUTER SCIENCES	66.5	21.2	12.3	18.7	33.8	47.5
EARTH SCIENCES & TECHNOLOGY	64.8	22.7	12.5	24.3	33.5	42.2
ECONOMICS & BUSINESS	65.2	21.8	13.0	22.0	33.3	44.7
EDUCATIONAL SCIENCES	63.7	22.5	13.7	23.2	33.3	43.5
ELECTRICAL ENG. & TELECOMM.	65.4	22.0	12.6	20.5	33.9	45.5
ENERGY SC. & TECHNOLOGY	64.8	22.6	12.6	22.3	34.0	43.7
ENVIRONMENTAL SCS & TECH.	64.3	22.8	13.0	24.5	33.4	42.1
GENERAL & INDUSTRIAL ENG.	63.2	23.0	13.8	23.0	33.7	43.3
HEALTH SCIENCES	64.1	22.7	13.2	24.1	33.4	42.5
INFORMATION & COMM. SCS.	64.1	22.3	13.6	22.6	33.2	44.3
INSTS. & INSTRUMENTATION	67.7	21.0	11.3	20.0	33.4	46.7
LAW & CRIMINOLOGY	62.8	23.3	13.9	23.4	33.9	42.7
MANAGEMENT & PLANNING	64.0	22.5	13.5	23.2	33.3	43.5
MATHEMATICS	66.0	21.5	12.5	17.8	34.4	47.8
MECHANICAL ENG. & AEROSPACE	63.6	23.0	13.4	22.7	34.0	43.3
PHYSICS & MATERIALS SCIENCE	66.1	22.2	11.7	22.4	33.9	43.8
PSYCHOLOGY	65.0	22.3	12.8	23.7	33.2	43.1
SOCIAL & BEHAVIORAL SCIENCES	63.6	22.5	13.9	23.7	33.0	43.3
SOCIOLOGY & ANTHROPOLOGY	63.7	22.6	13.8	23.6	33.2	43.3
STATISTICAL SCIENCES	65.7	21.7	12.7	20.9	33.4	45.7
----- Average over the 29 fields	64.7	22.5	12.8	23.1	33.5	43.4
----- Coefficient of variation	0.02	0.03	0.07	0.08	0.01	0.04
