



Normalisation of citation impact in economics

Lutz Bornmann¹ · Klaus Wohlrabe²

Received: 29 March 2019 / Published online: 14 June 2019
© The Author(s) 2019

Abstract

This study is intended to facilitate fair research evaluations in economics. Field- and time-normalisation of citation impact is the standard method in bibliometrics. Since citation rates for journal papers differ substantially across publication years and Journal of Economic Literature classification codes, citation rates should be normalised for the comparison of papers across different time periods and economic subfields. Without normalisation, both factors that are independent of research quality might lead to misleading results of citation analyses. We apply two normalised indicators in economics, which are the most important indicators in bibliometrics: (1) the mean normalised citation score (MNCS) compares the citation impact of a focal paper with the mean impact of similar papers published in the same economic subfield and publication year. (2) $PP_{top\ 10\ \%}$ is the share of papers that belong to the 10% most cited papers in a certain subfield and time period. Since the MNCS is based on arithmetic averages despite skewed citation distributions, we recommend using $PP_{top\ 10\ \%}$ for fair comparisons of entities in economics. In this study, we apply the normalisation methods to 294 journals (including normalised scores for 192,524 papers). We used the $PP_{top\ 10\ \%}$ results for assigning the journals to four citation impact classes. Seventeen journals have been identified as outstandingly cited. Two journals, *Quarterly Journal of Economics* and *Journal of Economic Literature*, perform statistically significantly better than all other journals. Thus, only two journals can be clearly separated from the rest in economics.

Keywords Bibliometrics · Citations · JEL codes · Journal ranking · Mean normalised citation score (MNCS) · Citation percentile · $PP_{top\ 10\ \%}$

✉ Klaus Wohlrabe
wohlrabe@ifo.de

Lutz Bornmann
bornmann@gv.mpg.de

¹ Division for Science and Innovation Studies, Administrative Headquarters of the Max Planck Society, Hofgartenstr. 8, 80539 Munich, Germany

² Ifo Institute for Economic Research, Poschingerstr. 5, 81679 Munich, Germany

Introduction

Research evaluation is the backbone of economic research; common standards in research and high-quality work cannot be achieved without such evaluations (Bornmann 2011; Moed and Halevi 2015). It is a sign of the current science system—with its focus on accountability—that quantitative methods of research evaluation complement qualitative assessments of research (i.e. peer review). Today, the most important quantitative method is bibliometrics with its measurements of research output and citation impact (Bornmann in press). Whereas in the early 1960s, only a small group of specialists was interested in bibliometrics (e.g. Eugene Garfield, the inventor of Clarivate Analytics' Journal Impact Factor, JIF), research activities in this area have substantially increased over the past two decades (Wouters et al. 2015). Today various bibliometric studies are being conducted based on data from individual researchers, scientific journals, universities, research organizations, and countries (Gevers 2014).

Citation impact is seen as a proxy of research quality, which measures one part of quality, namely usefulness (other parts are accuracy and importance, see Martin and Irvine 1983). Since impact measurements are increasingly used as a basis for funding or tenure decisions in science, citation impact indicators are the focus of bibliometric studies. In these studies it is often necessary to analyze citation impact across papers published in different fields and years. However, comparing counts of citations across fields and publication years leads to misleading results (see Council of Canadian Academies 2012). Since the average citation rates for papers published in different fields (e.g. mathematics and biology) and years differ significantly (independently of the quality of the papers) (Kreiman and Maunsell 2011; Opthof 2011), it is standard in bibliometrics to normalise citations. According to Abramo et al. (2011) and Waltman and Eck (2013b), field-specific differences in citation patterns arise for the following reasons: (1) different numbers of journals indexed for the fields in bibliometric databases (Marx and Bornmann 2015); (2) different citation and authorship practices, as well as cultures among fields; (3) different production functions across fields (McAllister et al. 1983); and (4) different numbers of researchers among fields (Kostoff 2002). The law of the constant ratios (Podlubny 2005) claims that the ratio of the numbers of citations in any two fields remains close to constant.

It is the aim of normalised bibliometric indicators “to correct as much as possible for the effect of variables that one does not want to influence the outcomes of a citation analysis” (Waltman 2016a, p. 375). In principle, normalised indicators compare the citation impact of a focal paper with a citation impact baseline defined by papers published in the same field and publication year. The recommendation to use normalised bibliometric indicators instead of bare citation counts is one of the ten guiding principles for research metrics listed in the Leiden manifesto (Hicks et al. 2015; Wilsdon et al. 2015).

This study is intended to introduce the approach of citation normalising in economics, which corresponds to the current state of the art in bibliometrics. “[Standard approaches in bibliometrics to normalise citation impact](#)” section presents two normalised citation indicators (see also “[Appendix 2](#)”): the mean normalised citation score (MNCS), which was the standard approach in bibliometrics over many years, and the current preferred alternative $PP_{top\ 10\%}$. The MNCS normalises the citation count of a paper with respect to a certain economic subfield. $PP_{top\ 10\%}$ further corrects for skewness in subfields' citation rates; the metric is based on percentiles. It determines whether a paper belongs to the 10% most frequently cited papers in a subfield. The subfield

definition used in this study relies on the Journal of Economic Literature (JEL) classification system. It is well-established in economics and most of the papers published in economics journals have JEL codes attached.

In “**Methods**” section we describe our dataset and provide several descriptive statistics. We extracted all of the papers from the Web of Science (WoS, Clarivate Analytics) economics subject category published between 1991 and 2013. We matched these papers with the corresponding JEL codes listed in EconLit. Using citation data from WoS, we realized that the citation rates substantially differ across economic subfields. As in many other disciplines, citation impact analyses can significantly inspire or hamper the career paths of researchers in economics, their salaries and reputation (Ellison 2013; Gibson et al. 2014, 2017). In a literature overview Hamermesh (2018) demonstrates that citations are related to the salaries earned by economists. Fair research evaluations in economics should therefore consider subfield-specific differences in citation rates, because the differences are not related to research quality.

In “**Results**” section we introduce a new economics journal ranking based on normalised citation scores. We calculated these scores for 192,524 papers published in 294 journals (see also “**Appendix 1**”). Although several top journals are similarly positioned to other established journal rankings in economics, we found large differences for many journals. In “**Discussion**” section, we discuss our results and give some direction for future research. The subfield-normalisation approach can be applied to other entities than journals, such as researchers, research groups, institutions and countries.

Methods

A key issue in the calculation of normalised citation scores is the definition of fields and subfields, which are used to compile the reference sets (Wilsdon et al. 2015; Wouters et al. 2015). The most common approach in bibliometrics is to use subject categories that are defined by Clarivate Analytics for WoS or Elsevier for Scopus. These subject categories are sets of journals publishing papers in similar research areas, such as biochemistry, condensed matter physics and economics. They shape a multidisciplinary classification system covering a broad range of research areas (Wang and Waltman 2016). However, this approach has been criticized in recent years because it is stretched to its limits with multi-disciplinary journals, e.g. *Nature* and *Science*, and field-specific journals with a broad scope, e.g. *Physical Review Letters* and *The Lancet*. “These journals do not fit neatly into a field classification system” (Waltman and van Eck 2013a, p. 700), because they cannot be assigned to a single field or publish research from a broad set of subfields (Haddow and Noyons 2013).

It is not only specific for fields, but also for subfields that they have different patterns of productivity and thus different numbers of citations (Crespo et al. 2014; National Research Council 2010). Thus, it is an obvious alternative for field-specific bibliometrics to use a mono-disciplinary classification system (Waltman 2016a). It is an advantage of these systems that they are specially designed to represent the subfield patterns in a single field (Boyack 2004) and are assigned to papers on the paper-level (and not journal-level). The assignment of subfields at the paper level protects the systems from problems with multi-disciplinary journals. In recent years, various bibliometric studies have used mono-disciplinary systems. Chemical Abstracts (CA) sections are used in chemistry and related areas (Bornmann and Daniel 2008; Bornmann et al. 2011), MeSH (Medical Subject Headings) terms in biomedicine (Bornmann et al. 2008; Leydesdorff and Opthof 2013; Strotmann and

Table 1 Main Journal of Economic Literature (JEL) codes

Code letter	Category
A	General Economics and Teaching
B	History of Economic Thought, Methodology, and Heterodox Approaches
C	Mathematical and Quantitative Methods
D	Microeconomics
E	Macroeconomics and Monetary Economics
F	International Economics
G	Financial Economics
H	Public Economics
I	Health, Education, and Welfare
J	Labour and Demographic Economics
K	Law and Economics
L	Industrial Organization
M	Business Administration and Business Economics; Marketing; Accounting; Personnel Economics
N	Economic History
O	Economic Development, Innovation, Technological Change, and Growth
P	Economic Systems
Q	Agricultural and Natural Resource Economics; Environmental and Ecological Economics
R	Urban, Rural, Regional, Real Estate, and Transportation Economics
Y	Miscellaneous Categories
Z	Other Special Topics

Zhao 2010), PACS (Physics and Astronomy Classification Scheme) codes in physics and related areas (Radicchi and Castellano 2011), and the MathSciNet's MSC (Mathematics subject classification) system in mathematics (Smolinsky and Lercher 2012).

The Journal of Econometric Literature (JEL) codes

In economics, the assignment of publications to subfields has a long history. Early classification attempts by the American Economic Association go back to the beginning of the 20th century when ten major categories were defined in the *American Economic Review*. These categories which have been developed to arrange publications to their subject matter have been subsequently revised several times and transferred to the EconLit system, including JEL codes. The majority of economics journals ask authors to provide JEL codes for their papers. A detailed overview of the history and meaning of JEL codes is provided by Cherrier (2017). In its current form (since 1991) all JEL codes—the main categories—are designed as “Exx”, i.e. a letter plus two stages of subcategories indicated by numbers (see <https://www.aeaweb.org/jel/guide/jel.php>). There are 20 categories at the main level, which are listed in Table 1. The main levels form the basis for the computation of the normalised scores in this study. The 133 categories at the first sub-level (e.g. E1) are used for robustness checks (see “Robustness” section; further disaggregated levels are not considered here).

According to Cherrier (2017) JEL codes are an important field-classification system in economics: “They provide a map with which to navigate the discipline on the American Economics Association (AEA) website. They are used to publish and search job offers, to

skim job offers, to assign grant applications and submitted papers to referees, and to search for book reviewers. Bibliometric studies of the characteristics of economists' publications, including size, age structure, co-authorship, subject-matter, methodology and citations patterns overwhelmingly rely on JEL codes to categorize papers" (p. 546). Kosnik (2018) used a dataset of articles which were published in the *American Economic Review* over twenty years to investigate whether the articles have been validly assigned to JEL codes. The results show that "JEL category codes do appear to represent papers that study topics and themes one would expect to be assigned to those codes" (p. 261). Thus, JEL codes seem to reflect research areas in economics validly.

Publication and citation data

WoS is the most important bibliographic database in bibliometrics. Most of the studies in this area are based on its publication and citation data. We downloaded all meta-data of the papers and the corresponding citations from the subject category "Economics", which were published between 1991 and 2013. The data are from an in-house version of the WoS database. We used 1991 as the first year, since JEL codes were established in its current form in 1991. We obtained data for 224,867 papers with the document type "article" or "review", which were published in 386 journals. With the exclusion of other document types (e.g. editorial material, notes, and comments), we focus in this study on substantial items.

We have made four adjustments to this dataset:

1. We excluded publications of the *Papers and Proceedings* issues from the *American Economic Review* and the *European Economic Review*. These papers are usually very short due to space considerations from the journal (usually five to six pages). They often represent an extension only that has been left out in full-length papers published elsewhere.
2. We only kept those papers published in journals that were listed in 2013 for at least four years. Thus, we excluded papers from journals that have stopped being listed (or reclassified) in WoS or deceased.
3. The journals in which the papers have appeared had to be listed in EconLit, since the JEL codes were obtained from the Econlit database. If we were not able to match a paper via EconLit (because the publishing journal was not listed), we used JEL codes data from RePEc (see Zimmermann 2013). For these papers we applied a similar matching procedure as described by Angrist, Azoulay, Ellison, Hill, and Lu (2017a).
4. Papers without JEL codes, or with JEL codes "Y" and "Z" were excluded from the study. The codes "Y" and "Z" are not related to a specific content.

The four adjustments ended up with 192,524 papers, which appeared in 294 journals. The citations of these papers refer to the time period between publication and the end of 2016. Thus, the citation counts of the papers are based on different citation windows (ranging between 4 and 26 years). The longer the citation window, the more the "true" impact of a paper can be determined (Research Evaluation and Policy Project 2005; Wang 2013). Glänzel (2008) and Glänzel et al. (2009) recommend using a citation window of at least three years. Johnston et al. (2013) show for papers published in the *American Economic Review* that the mean citation rate peaks in the fourth year after publication. Since the

Table 2 Descriptive statistics

Year	Journals	Papers	Citations	Share of papers with zero citations (%)	JEL codes
1991	108	4181	120,856	12.1	7748
1995	134	5145	149,439	10.1	9076
2000	165	6548	174,807	8.2	15,140
2005	192	8013	181,045	7.3	22,497
2010	293	13,474	139,462	13.2	43,649
2013	294	15,901	69,641	22.4	58,228
1991–2013	294	192,425	3,506,995	11.8	534,911

citations in our in-house database are counted until the end of 2016 (at the time when we conducted the study), papers that appeared after 2013 were not included in the study.

Descriptive statistics and differences in citation rates

Table 2 reports descriptive statistics for all papers in the dataset and for the papers from selected years in a 5 year time interval. The development over time shows that the number of economics journals increased. Correspondingly, the number of papers and assigned JEL codes also increased. Due to the diminishing citation window from 26 to 4 years, citation counts decrease and shares of non-cited papers increase over time. In Table 9 (see “Appendix 1”), we further report the number of papers, the time period covered in WoS, and descriptive citation statistics for each journal in our dataset. For 108 of all 294 journals in the set (37%), papers appeared across the complete time period from 1991 to 2013. For the other journals, the WoS coverage started later than 1991 (such as for the four American Economic journals). The results in Table 9 demonstrate that almost all journals published papers with zero citations. With an average of 145 citations, the highest citation rate was reached by the *Quarterly Journal of Economics* by way of comparison. Arellano and Bond (1991) is the most frequently cited paper in our set (with 4627 citations).

Table 3 shows average citation rates for papers assigned to different JEL codes. The results are presented for selected years in a five year time interval. It is clearly visible over all publication years that the average values differ substantially between the economics subfields. For example, papers published in 1991 in “General Economics and Teaching” (A) received on average 15.2 citations; with 49.5 citations this figure is more than three times larger in “Mathematical and Quantitative Methods” (C). Similar results for differences in citation rates of economic subfields have been published by van Leeuwen and Calero Medina (2012), Ellison (2013), Hamermesh (2018), and Perry and Reny (2016). The results in Table 3 also reveal that the average citation rates decline over time in most cases, as the citation window gets smaller.

The dependency of the average citations in economics on time and subfield, which is independent of research quality, necessitates the consideration of subfield and publication year in bibliometric studies. Without consideration of these differences, research evaluations are expected to be misleading and disadvantage economists newly publishing in the field or working in subfields with systematically low average citations (e.g. in subfield B “History of Economic Thought, Methodology, and Heterodox Approaches”).

Table 3 Average citation rates per JEL code and publication year

JEL-Code	1991	1995	2000	2005	2010	2013
A	15.2	8.7	16.3	15.7	5.3	2.9
B	4.7	7.9	11.6	7.4	5.4	1.9
C	49.5	54.6	28.0	25.3	10.8	4.3
D	35.4	28.3	26.5	21.1	9.4	4.0
E	23.9	19.9	23.8	18.9	7.3	3.7
F	17.2	25.8	18.8	18.6	8.3	3.5
G	46.4	36.7	43.1	27.8	12.8	4.9
H	18.8	19.0	21.4	17.2	8.6	4.0
I	35.1	37.3	32.4	28.6	12.1	4.7
J	31.9	26.2	25.3	21.8	9.6	4.0
K	37.7	22.1	29.3	16.4	6.5	3.2
L	18.8	30.6	22.6	22.5	10.1	4.5
M	25.6	38.7	41.4	35.7	14.0	5.4
N	13.0	12.2	15.0	17.1	8.3	3.7
O	37.3	38.0	32.2	22.5	10.5	4.1
P	11.2	15.4	16.4	20.1	9.1	3.9
Q	20.4	26.0	26.0	26.4	14.7	6.6
R	35.5	24.9	22.4	24.8	13.3	5.6

Standard approaches in bibliometrics to normalise citation impact

Economics was already part of a few bibliometric studies, which considered field-specific differences (e.g. Ruiz-Castillo 2012). Palacios-Huerta and Volij (2004) and Angrist et al. (2017b) generalized an idea for citation normalisation that goes back to Liebowitz and Palmer (1984), where citations are weighted with respect to the citing journal. Angrist et al. (2017a) constructed their own classification scheme featuring ten subfields in the spirit of Ellison (2002). The classification builds upon JEL codes, keywords, and abstracts. Using about 135,000 papers published in 80 journals, the authors construct time varying importance weights for journals that account for the subfield where a paper was published.

Combes and Linnemer (2010) calculated normalised journal rankings for all EconLit journals. Although they considered JEL codes for the normalisation procedure, they calculated the normalisation at the journal, and not at the paper level. Linnemer and Visser (2016) document the most cited papers from the so called top-5 economics journals (Card and DellaVigna 2013), where they also account for time and JEL codes. With the focus on the top 5 journals, however, they considered only a small sample of journals and did not calculate bibliometric indicators.

In this study, we build upon the different normalization approaches published hitherto in economics by using, e.g. JEL codes as field-classification scheme for impact normalization and combine these approaches with recommendations from relevant metrics guidelines (e.g. Hicks et al. 2015).

Mean normalised citation score (MNCS)

The definition and use of normalised indicators in bibliometrics (based on mean citations) started in the mid-1980s with the papers by Schubert and Braun (1986) and Vinkler (1986). Here, normalised citation scores (NCSs) result from the division of the citation count of focal papers by the average citations of comparable papers in the same field or subfield. The denominator is the expected number of citations and constitutes the reference set of the focal papers (Mingers and Leydesdorff 2015; Waltman 2016a). Resulting impact scores larger than 1 indicate papers cited above-average in the field or subfield and scores below 1 denote papers with below-average impact.

Several variants of this basic approach have been introduced since the mid-1980s (Vinkler 2010) and different names have been used for the metrics, e.g. relative citation rate, relative subfield citedness, and field-weighted citation score. In the most recent past, the metric has been mostly used in bibliometrics under the label “MNCS”. Here the NCS for each paper in a publication set (of a researcher, institution, or country) are added up and divided by the number of papers in the set, which results in the mean NCS (MNCS). Since citation counts depend on the length of time between the publication year of the cited papers and the time point of the impact analysis (see Table 3), the normalisation is performed separately for each publication year.

Sandström (2014) published the following rules of thumb for interpreting normalised impact scores (of research groups):

“A. NCS_f [field-normalised citation score] ≤ 0.6 significantly far below international average (insufficient)

B. $0.60 < NCS_f \leq 1.20$ at international average (good)

C. $1.20 < NCS_f \leq 1.60$ significantly above international average (very good)

D. $1.60 < NCS_f \leq 2.20$ from an international perspective very strong (excellent)

E. $NCS_f > 2.20$ global leading excellence (outstanding)” (p. 66).

Thus, excellent research has been published by an entity (e.g. journal or researcher), if the MNCS exceeds 1.6.

The MNCS has an important property, which is required by established normalised indicators (Moed 2015; Waltman et al. 2011): The MNCS value of 1 has a specific statistical meaning: it represents average performance and below-average and above-average performance can be easily identified.

A detailed explanation of how the MNCS is calculated in this study can be found in “Appendix 2”.

PP_{top 10%}—a percentile based indicator as the better alternative to the MNCS

Although the MNSC has been frequently used as indicator in bibliometrics, it has an important disadvantage: it uses the arithmetic average as a measure of central tendency, although distributions of citation counts are skewed (Seglen 1992). As a rule, field-specific paper sets contain many lowly or non-cited papers and only a few highly-cited papers (Bornmann and Leydesdorff 2017). Therefore, percentile-based indicators have become popular in bibliometrics, which are robust against outliers. According to Hicks et al. (2015) in the Leiden Manifesto, “the most robust normalisation method is based on percentiles: each paper is weighted on the basis of the percentile to which it belongs in the citation distribution of its field (the top 1, 10 or 20%, for example)” (p. 430). The recommendation to use percentile-based indicators can also be found in the Metric Tide (Wilsdon et al. 2015).

Against the backdrop of these developments in bibliometrics, and resulting recommendations in the Leiden Manifesto and the Metric Tide, we use the $PP_{\text{top } 10\%}$ indicator in this study as the better alternative to the MNCS. Since we are especially interested in the excellent papers (or journals) and the 1% is too restrictive (resulting in too few papers in the group of highly cited papers), we focus in this study on the 10% most highly cited papers. Basically, the $PP_{\text{top } 10\%}$ indicator is calculated on the basis of the citation distribution in a specific subfield whereby the papers are sorted in decreasing order of citations. Papers belonging to the 10% most frequently cited papers are assigned the score 1 and the others the score 0 in a binary variable. The binary variables for all subfields can then be used to calculate the $P_{\text{top } 10\%}$ or $PP_{\text{top } 10\%}$ indicators. $P_{\text{top } 10\%}$ is the absolute number of papers published by an entity (e.g. journal or institution) belonging to the 10% most frequently cited papers and $PP_{\text{top } 10\%}$ the relative number. Here, $P_{\text{top } 10\%}$ is divided by the total number of papers in the set. Thus, it is the percentage of papers by an entity that are cited above-average in the corresponding subfields.

The detailed explanation of how the $PP_{\text{top } 10\%}$ indicator is calculated in this study can be found in “Appendix 2”.

Results

Comparison of citation counts, normalised citation scores (NCSs) and $P_{\text{top } 10\%}$

The normalisation of citations only makes sense in economics if the normalisation leads to meaningful differences between normalised scores and citations. However, one cannot expect complete independence, because both metrics measure impact based on the same data source.

Table 4 shows the papers with the largest NCSs in each subfield of economics. The listed papers include survey papers and methodological papers that are frequently used within and across subfields. We also find landmark papers in the table that have been continuously cited in the respective subfields. Linnemer and Visser (2016) published a similar list of most frequently cited papers in each subfield. For the JEL codes C, F, H, and R the same papers have been identified in agreement; differences are visible for the codes E, G, I, J, L, and O. Since Linnemer and Visser (2016) based their analyses on a different set of journals which is significantly smaller than our set, the differences are expectable.

The impact scores in Table 4 reveal that the papers are most frequently cited in the subfields with very different citation counts—between 344 citations in “General Economics and Teaching” (A) and 4627 citations in “Mathematical and Quantitative Methods” (C). Correspondingly, similar NCSs in the subfields reflect different citation counts. The list of papers also demonstrate that papers are assigned to more than one economic subfield. The paper by Acemoglu et al. (2001) is the most cited paper in four subfields. Since many other papers in the dataset are also assigned to more than one subfield, we considered a fractional counting approach of citation impact. The detailed explanation of how the fractional counting has been implemented in the normalisation can be found in “Appendix 2”.

Table 4 provides initial indications that normalisation is necessary in economics. However, the analysis could not include $P_{\text{top } 10\%}$, because this indicator is primarily a binary variable. To reveal the extent of agreement and disagreement between all metrics (citation counts, NCS, and $P_{\text{top } 10\%}$), we group the papers according to the Characteristics Scores and Scales (CSS) method, which is proposed by Glänzel and Schubert (1988). For each

Table 4 The most frequently cited paper in every subfield of economics based on normalised citation score (NCS)

JEL code	NCS	Citation count	Paper
A	37.6	344	Stefano DellaVigna (2009): "Psychology and Economics: Evidence from the Field", <i>Journal of Economic Literature</i> , 47(2), 315–72
B	39.4	526	John Sutton (1997): "Gibrat's Legacy", <i>Journal of Economic Literature</i> , 35(1), 40–59
C	119.2	4627	Manuel Arellano & Stephen Bond (1991): "Some Tests of Specification for Panel Data: Monte Carlo Evidence and an Application to Employment Equations", <i>Review of Economic Studies</i> , 58(2), 277–297
D	82.1	2985	Amos Tversky & Daniel Kahneman (1992): "Advances in Prospect Theory: Cumulative Representation of Uncertainty", <i>Journal of Risk and Uncertainty</i> , 5(4), 297–323
E	61.0	1584	Robert E. Hall and Charles I. Jones (1999): "Why do Some Countries Produce So Much More Output Per Worker than Others?", <i>The Quarterly Journal of Economics</i> , 114(1), 83–116
F	75.1	1917	Marc J. Melitz (2003): "The Impact of Trade on Intra-Industry Reallocations and Aggregate Industry Productivity", <i>Econometrica</i> , 71(6), 1695–1725
G	90.8	1644	Mitchell A. Petersen (2009): "Estimating Standard Errors in Finance Panel Data Sets: Comparing Approaches", <i>Review of Financial Studies</i> , 22(1), 435–480
H	48.7	1041	Simon Gächter & Ernst Fehr (2000): "Cooperation and Punishment in Public Goods Experiments", <i>American Economic Review</i> , 90(4), 980–994
I	77.6	1838	Daron Acemoglu, Simon Johnson, & James A. Robinson (2001): "The Colonial Origins of Comparative Development: An Empirical Investigation 2", <i>American Economic Review</i> , 91(5), 1369–1401
J	119.2	4627	Manuel Arellano & Stephen Bond (1991): "Some Tests of Specification for Panel Data: Monte Carlo Evidence and an Application to Employment Equations", <i>Review of Economic Studies</i> , 58(2), 277–297
K	66.7	3300	Andrei Shleifer, Florencio Lopez-de-Silanes, & Rafael La Porta (2008): "The Economic Consequences of Legal Origins", <i>Journal of Economic Literature</i> , 46(2), 285–332
L	75.1	1917	Marc J. Melitz (2003): "The Impact of Trade on Intra-Industry Reallocations and Aggregate Industry Productivity", <i>Econometrica</i> , 71(6), 1695–1725
M	29.3	568	Israel M. Kirzner (1997): "Entrepreneurial Discovery and the Competitive Market Process: An Austrian Approach", <i>Journal of Economic Literature</i> , 35(1), 60–85
N	77.6	1838	Daron Acemoglu, Simon Johnson, & James A. Robinson (2001): "The Colonial Origins of Comparative Development: An Empirical Investigation", <i>American Economic Review</i> , 91(5), 1369–1401
O	77.6	1838	Daron Acemoglu, Simon Johnson, & James A. Robinson (2001): "The Colonial Origins of Comparative Development: An Empirical Investigation", <i>American Economic Review</i> , 91(5), 1369–1401

Table 4 (continued)

JEL code	NCS	Citation count	Paper
P	77.6	1838	Daron Acemoglu, Simon Johnson, & James A. Robinson (2001): "The Colonial Origins of Comparative Development: An Empirical Investigation", <i>American Economic Review</i> , 91(5), 1369–1401
Q	53.8	1418	David Pimentel, Rodolfo Zuniga, & Doug Morrison (2005): "Update on the environmental and economic costs associated with alien-invasive species in the United States", <i>Ecological Economics</i> , 52(3), 273–288
R	58.4	2071	Paul Krugman (1991): "Increasing Returns and Economic Geography", <i>Journal of Political Economy</i> , 99(3), 483–499

The citation counts are also given for comparison in the table

Table 5 Agreement and disagreement in measuring citation impact by using citations, normalized citation score (NCS), and P_{top 10%}

	NCS				P _{top 10%}		Sum
	Poorly cited (1)	Fairly cited (2)	Remarkably cited (3)	Outstandingly cited (4)	≤ 0.9	> 0.9	
Citations							
(1)	134,564	13,843	705	2	148,216	898	149,114
(2)	7226	20,616	4182	557	26,694	5887	32,581
(3)	0	2139	4586	1108	1228	6605	7833
(4)	0	0	546	2352	7	2891	2898
Sum	141,790	36,598	10,019	4019	176,145	16,281	192,426
Percent agreement = 84.25%, Kappa = 0.601 [0.597, 0.604]							

metric (citation counts and NCS), CSS scores are obtained by truncating the publication set at their metric mean and recalculating the mean of the truncated part of the set until the procedure is stopped or no new scores are generated. We defined four classes which we labeled with “poorly cited”, “fairly cited”, “remarkably cited”, and “outstandingly cited” (Bornmann and Glänzel 2017). Whereas poorly cited papers fall below the average impact of all papers in the set, the other classes are above this average and further differentiate the high impact area.

Table 5 (left panel) shows how the papers in our set are classified according to CSS with respect to citations and NCS. 84% of the papers are positioned on the diagonal (printed in bold), i.e. the papers are equally classified. The Kappa coefficient is a more robust measure of agreement than the share of agreement, since the possibility of agreement occurring by chance is taken into account (Gwet 2014). The coefficient in Table 5 highlights that the agreement is not perfect (which is the case with Kappa = 1). According to the guidelines by Landis and Koch (1977), the agreement between citations and NCS is only moderate.¹

The results in Table 5 show that 16% of the papers in the set have different classifications based on citations and NCS. For example, 13,843 papers are cited below average according to citations (classified as poorly cited), but above average cited according to NCS (classified as fairly cited). Two papers clearly stand out by being classified as poorly cited with respect to citations, but outstandingly cited with respect to the NCS. These are Lawson (2013) with 15 citations and an NCS of 7.8, and Wilson and Gowdy (2013) with 13 citations and an NCS of 6.8. There are also numerous papers in the set that are downgraded in impact measurement by normalised citations: 7226 papers are cited above average (fairly cited) according to citations, but score below average according to NCR (poorly cited). 546 papers are outstandingly cited if citations are used; but they are remarkably cited on the base of the NCR, i.e. if the subfield is considered in impact measurement.

Table 5 (right panel) also includes the comparison of citations and P_{top 10%}. Several papers in this study are fractionally assigned to the 10% most-frequently cited papers in the corresponding subfields and publication years (see the explanation in “Appendix 2”). Since

¹ Their guidelines for categorizing Kappa values are as follows: <0 = no agreement, 0–0.20 = slight, 0.21–0.40 = fair, 0.41–0.60 = moderate, 0.61–0.80 = substantial, and 0.81–1 = almost perfect agreement.

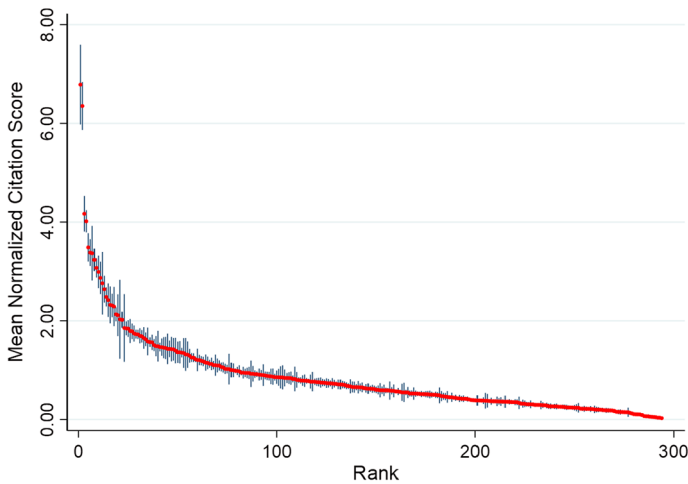


Fig. 1 Rank-distribution of 294 economics journals by mean normalized citation score (MNCS) with confidence intervals (CIs)

$P_{top\ 10\%}$ is not completely a binary variable (with the values 0 or 1), we categorized the papers in our set into two groups: $P_{top\ 10\%} \leq 0.9$ (being lowly cited) and $P_{top\ 10\%} > 0.9$ (being highly cited) for the statistical analysis. Nearly all of the papers classified as poorly cited on the basis of citations are also lowly cited on the basis of $P_{top\ 10\%}$. Thus, both indicators are more or less in agreement in this area. The results also show that some papers ($n=9,496$) that are highly cited by $P_{top\ 10\%}$ are classified differently by citations (remarkably or outstandingly cited). On the other hand, 898 papers are classified as poorly cited on the basis of citations, but are highly cited on the basis of $P_{top\ 10\%}$.

Taken together, the results in Table 5 demonstrate that normalisation leads to similar results as citations for many papers; however, there is also a moderate level of disagreement, which may lead to misleading results of impact analyses in economics based on citations.

New field- and time-normalised journal ranking

The first economics journal ranking was published by Coats (1971) who used readings from members of the American Economic Association as ranking criterion. With the emerging dominance of bibliometrics in research evaluation in recent decades, citations have become the most important source for ranking journals—in economics and beyond. The most popular current rankings in economics—besides conducting surveys among economists—are the relative rankings that are based on the approach of Liebowitz and Palmer (1984). Bornmann et al. (2018) provide a comprehensive overview of existing journal rankings in economics.

Since funding decisions and the offer of professorships in economics are mainly based on publications in reputable journals, journal rankings should not be influenced by different citation rates in economics subfields. Based on the NCS and the $P_{top\ 10\%}$ for each paper in our set, we therefore calculated journal rankings by aggregating the normalised paper impact across years. Figure 1 visualizes the MNCSs and confidence intervals (CIs) of the

294 journals in our publication set, which are rank-ordered by the MNCS. The CIs are generated by adding and subtracting $1.96 * \frac{\sigma}{\sqrt{N}}$ from the MNCS, where σ denotes the corresponding population standard deviation (Cumming and Calin-Jageman 2016). Thus, we are sampling from the population distribution of MNCSs. If the CIs of two journals do not overlap, they differ “statistically significantly” ($\alpha=1\%$) in their mean citation impact (Bornmann et al. 2014; Cumming 2012). The results should be interpreted against the backdrop of $\alpha=1\%$ (and not $\alpha=5\%$), because the publication numbers are generally high in this study. The chance of receiving statistically significant results grows with increasing sample sizes.

We use CIs to receive indications of the “true” level of citation impact (differences), although there is a considerable disagreement among bibliometricians about the correctness of the use of confidence intervals and statistical significance when working with bibliometric indicators (Waltman 2016b; Williams and Bornmann 2016).² We follow the general argument by Claveau (2016) “that these observations [citations] are realizations of an underlying data generating process constitutive of the research unit [here: journals]. The goal is to learn properties of the data generating process. The set of observations to which we have access, although they are all the actual realizations of the process, do not constitute the set of all possible realizations. In consequence, we face the standard situation of having to infer from an accessible set of observations—what is normally called the sample—to a larger, inaccessible one—the population. Inferential statistics are thus pertinent” (p. 1233).

There are two groups including two journals each in Fig. 1, which are clearly separated from the other journals: *Journal of Economic Literature* and *Quarterly Journal of Economics* in the first group—confirming the result by Stern (2013)—and *Journal of Political Economy* and *American Economic Review* in the second group. The very high impact of the journals in the first group is especially triggered by a few very frequently cited papers appearing in these journals: 26 papers in these journals are among the 100 papers with the highest NCSs. Excluding this small group of papers, the CIs of the journals would overlap with many other journals. All other economic journals in the figure are characterized by overlaps of CIs (more or less clearly pronounced). Most of the journals in Fig. 1 do not differ statistically significantly from similarly ranked journals.

The alternative $PP_{\text{top } 10\%}$ journal ranking is based on the premise that the impact results for scientific entities (here: journals) should not be influenced by a few outliers, i.e. the few very highly-cited papers. Figure 2 shows the rank distribution of the journals on the basis of $PP_{\text{top } 10\%}$ and the corresponding CIs. The shape of the distribution exhibits a similar convexity as the distribution in Fig. 1. For the calculation of the CIs in Fig. 2 we defined three quantities: $A = 2r + z^2$, $B = z\sqrt{z^2 + 4rq}$ and $C = 2(n + z^2)$, where r is the number of $P_{\text{top } 10\%}$, $q = 1 - 1/N$ and z the corresponding value from the standard normal distribution. The CI for the population proportion is given by $(A - B)/C$ to $(A + B)/C$ (Altman et al. 2013).

In agreement with the MNCS results, we find the same two journals (*Quarterly Journal of Economics* and *Journal of Economic Literature*) at the top that are clearly separated from the others. The results confirm thus the previous results which are based on the MNCS. It seems that only two journals in economics (and not five journals as always supposed) can be clearly separated from the rest (in terms of field-normalised citations).

² The disagreement is reflected in the various comments following the paper by Williams and Bornmann (2016) on sampling issues in bibliometrics. These comments either argue for the use of inference statistics (e.g. Mutz 2016) or against (e.g. Schneider 2016).

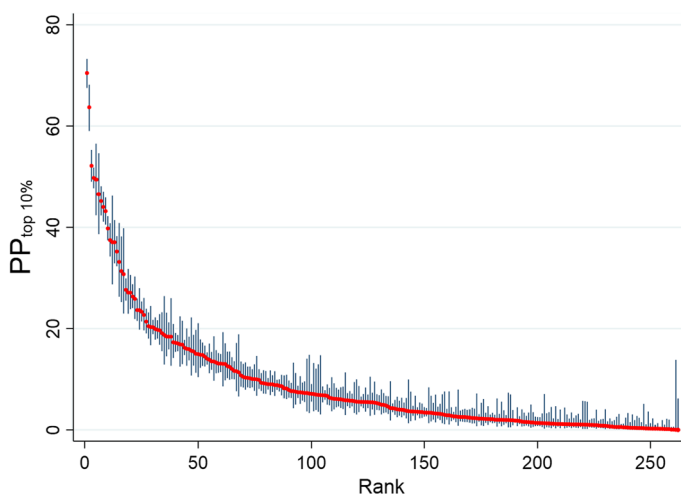


Fig. 2 Rank-distribution of 294 economics journals by $PP_{top\ 10\%}$ with confidence intervals (CIs)

The overlaps of the CIs for the rest of the journals in Fig. 2 make it impossible to unambiguously identify specific performance groups of economics journals in terms of citation impact. We therefore used another (robust) method to classify the journals into certain impact groups and separate an outstandingly cited group (which include *Quarterly Journal of Economics* and *Journal of Economic Literature*). In “Comparison of citation counts, normalised citation scores (NCSs) and $P_{top\ 10\%}$ ” section we applied the CSS method to assign the papers in our set to four impact classes. Since the method can also be used with aggregated scores (Bornmann and Glänzel 2017), we assigned the journals in our set to four impact classes based on $PP_{top\ 10\%}$. Table 9 in “Appendix 1” shows all journals ($n = 294$) with their assignments to the four groups: 205 journals are poorly cited, 62 journals are fairly cited, 14 journals are remarkably cited, and 13 journals are outstandingly cited.

Table 6 shows the 13 economics journals in the outstandingly cited group. Four additional journals are considered in the table. Their CIs include the threshold that separates the outstandingly cited journal group from the remarkably cited journal group. Thus, one cannot exclude the possibility that these journals also belong to the outstandingly cited group.

The three top journals in Table 6 are *Quarterly Journal of Economics*, *Journal of Economic Literature*, and *Journal of Political Economy*. With $PP_{top\ 10\%}$ of 70.48, 63.71, and 52.16, respectively, (significantly) more than half of the papers published in these journals are $P_{top\ 10\%}$. All journals in the table are able to publish significantly more papers in the corresponding subject categories and publication years than can be expected—the expected value is 10%. $PP_{top\ 10\%}$ of each journal is greater than 30%; thus, the journals published at least three times more $P_{top\ 10\%}$ than can be expected.

In order to investigate the stability of journals in the outstandingly cited group, we annually assigned each economics journal in our set to the four citation impact classes (following the CSS approach). No journal falls in every year into the outstandingly cited group. *Quarterly Journal of Economics*, *Journal of Political Economy*, and *Journal of*

Table 6 Outstandingly cited economics journals (according to $PP_{\text{top } 10\%}$) with confidence intervals (CIs)

Rank	Journal	$PP_{\text{top } 10\%}$	CI	
1	Quarterly Journal of Economics	70.48	67.52	73.27
2	Journal of Economic Literature	63.71	59.02	68.16
3	Journal of Political Economy	52.16	49.00	55.29
4	American Economic Review	49.75	47.70	51.81
5	American Economic Journal-Applied Economics	49.44	42.37	56.52
6	American Economic Journal-Macroeconomics	46.55	38.65	54.62
7	Journal of Finance	45.20	42.35	48.09
8	Journal of Economic Perspectives	44.04	41.10	47.02
9	Econometrica	43.20	40.53	45.91
10	Journal of Financial Economics	39.79	37.41	42.21
11	Review of Financial Studies	37.46	34.23	40.81
12	Annual Review of Economics	37.07	28.73	46.26
13	Transportation Research Part B-Methodological	37.05	32.88	41.43
14	Review of Economic Studies	35.22	32.30	38.26
15	American Economic Journal-Economic Policy	33.17	26.28	40.86
16	Journal of Economic Growth	31.35	25.24	38.19
17	Review of Environmental Economics and Policy	30.75	22.96	39.82

The so called top five economics journals are printed in bold

Economic Literature missed the outstandingly cited category in only one year. *American Economic Review* is classified as outstandingly cited from 1992 to 2010 but not in the other four years. *Review of Economics Studies* and *Econometrica* are listed at 8 and 15, respectively, out of 23 years in the outstandingly cited category. The other journals in Table 6 are either classified as outstandingly or remarkably cited over the years; some journals are only fairly cited in certain years.

Comparisons with other journal rankings

How is the $PP_{\text{top } 10\%}$ journal ranking related to the results of other rankings in economics? The most simple form of ranking the journals is by their mean citation rate. The JIF is one of the most popular journal metric, which is based on the mean citation rate of papers within one year received by papers in the two previous years (Garfield 2006). In the comparison with $PP_{\text{top } 10\%}$ we use the mean citation rate for each journal. Since the citation window is not restricted to certain years in the calculation of $PP_{\text{top } 10\%}$, we consider all citations from publication year until the end of 2016 in the calculation of the mean citation rate.

The RePEc website (see www.repec.org) has become an essential source for various rankings in economics. Based on a large and still expanding bibliometric database, RePEc publishes numerous rankings for journals, authors, economics departments and institutions. RePEc covers more journals and additional working papers, chapters and books compared to WoS (further details can be found in Zimmermann 2013). For the comparison with the $PP_{\text{top } 10\%}$ journal ranking, we consider two popular journal metrics from RePEc: the simple and the recursive Impact Factor (IF). The simple IF is the ratio of all citations to a specific

Table 7 Comparison of the $PP_{top\ 10\%}$ journal ranking with rankings based on the mean citation rate, simple IF, and recursive IF

Other rankings	$PP_{top\ 10\%}$			
	Journal classification			
	Outstandingly cited (1)	Remarkably cited (2)	Fairly cited (3)	Poorly cited (4)
<i>Mean citation rate (WoS)</i>				
(1)	8	1	2	0
(2)	2	8	4	0
(3)	0	7	41	11
(4)	0	0	12	175
Percent agreement = 85.61%, Kappa = 0.742 [0.668, 0.821]				
<i>RePEC simple IF</i>				
(1)	7	4	0	0
(2)	2	5	7	0
(3)	0	7	25	27
(4)	0	0	19	168
Percent agreement = 75.65%, Kappa = 0.576 [0.478, 0.670]				
<i>RePEC recursive IF</i>				
(1)	4	7	0	0
(2)	1	5	5	3
(3)	0	4	22	33
(4)	0	0	14	173
Percent agreement = 75.28%, Kappa = 0.494 [0.379, 0.589]				

journal and the number of listed papers in RePEc. The recursive IF also takes the prestige of the citing journal into account (Liebowitz and Palmer 1984). Whereas the simple and recursive IFs are based on citations from the RePEc database, the citations for calculating the mean citation rates (see above) are from WoS.

The results of the comparisons are reported in Table 7. Twenty three journals in our sample are not listed in RePEc, thus, we excluded these journals from all comparisons. We used the CSS method to classify all journals on the basis of the mean citation rate, $PP_{top\ 10\%}$, as well as simple and recursive IFs, as outstandingly, remarkably, fairly, and poorly cited. In “[Comparison of citation counts, normalised citation scores \(NCSs\) and \$P_{top\ 10\%}\$](#) ” section we applied the CSS method to assign the papers in our set to four impact classes. Since the method can also be used with aggregated scores (Bornmann and Glänzel 2017), we assigned the journals in our set to four impact classes based on the different indicators.

The Kappa coefficients in the table highlight a moderate agreement between the RePEc simple/recursive IF and $PP_{top\ 10\%}$ and a substantial agreement between $PP_{top\ 10\%}$ and mean citation rate (Landis and Koch 1977). Thus, the results reveal that there is considerable agreement, but also disagreement between the rankings. This finding can be expected if subfield-normalized citation metrics and citation metrics are compared. Both metrics groups are based on citation impact, why a considerable agreement is expectable. Since subfield-normalization correct citation impact in many cases

Table 8 Robustness checks with respect to JEL codes, as well as top-cited and lowly-cited papers in the set

PP _{top 10%}	PP _{top 10%} —all papers				
	Journal classification				
	Outstand- ingly cited (1)	Remarkably cited (2)	Fairly cited (3)	Poorly cited (4)	
<i>First robustness check</i>					
JEL codes first subfield level	(1)	12	2	0	0
	(2)	1	10	5	0
	(3)	0	2	53	7
	(4)	0	0	4	198
	Percent agreement = 92.86%, Kappa = 0.872 [0.813, 0.923]				
<i>Second robustness check</i>					
Excluding top-cited papers	(1)	12	1	0	0
	(2)	1	13	4	0
	(3)	0	0	54	4
	(4)	0	0	4	201
	Percent agreement = 95.24%, Kappa = 0.914 [0.852, 0.956]				
<i>Third robustness check</i>					
Excluding lowly-cited papers	(1)	11	0	0	0
	(2)	2	14	3	0
	(3)	0	0	56	6
	(4)	0	0	3	199
	Percent agreement = 95.24%, Kappa = 0.913 [0.858, 0.953]				
<i>Fourth robustness check</i>					
PP _{top 50%}	(1)	13	13	4	0
	(2)	0	1	37	2
	(3)	0	0	21	58
	(4)	0	0	0	145
	Percent agreement = 61.22%, Kappa = 0.480 [0.405, 0.568]				

moderately, but in a few cases substantially, the Kappa coefficients tend to be closer to almost perfect agreement than to no agreement.

Robustness

JEL codes are available on different levels. We used the main level with 18 categories in this study to normalise the data (see “[The journal of econometric literature \(JEL\) codes](#)” section). The first sub-level includes 122 categories. In a first robustness check of our new journal ranking in “[New field- and time-normalised journal ranking](#)” section we calculated PP_{top 10%} for all journals by using the 122 sub-levels, instead of the 18 main levels for normalisation. Again, we used the CSS method to classify the journals as outstandingly, remarkably, fairly, and poorly cited on the basis of PP_{top 10%} (see “[Comparison of citation counts, normalised citation scores \(NCSs\) and P_{top 10%}](#)” section). Table 8 (see the part with

the first robustness check) shows the comparison of two different $PP_{top\ 10\%}$ journal rankings, whereby one ranking was calculated on the basis of the JEL codes main level and the other on the basis of the JEL codes first subfield level. The Kappa coefficient and the percent agreement highlight a very high level of agreement between the rankings based on the two different subfield definitions. Thus, the journal results are robust to the use of the JEL code level for normalisation.

In two further robustness checks, we tested the results against the influence of extreme values: are the journals similarly classified as outstandingly, remarkably, fairly, and poorly cited, if the most-cited and lowly-cited papers in the journals are removed? The most-cited papers refer in the check to the most-cited papers of each journal in each year, which reduce the publication numbers by 4863 papers. The lowly-cited papers are defined as papers with zero citations or one citation (this reduced the publication numbers by almost one-fourth). The results of the further robustness checks are presented in Table 8 (see the parts with the second and third robustness checks). If the top-cited papers are excluded, the agreement is 95% and Kappa equals 0.91. Almost the same figures are obtained when we exclude lowly-cited papers, whereas the change in the classification scheme is slightly different. According to the guidelines of Landis and Koch (1977) the agreement in both cases is almost perfect, i.e. our results are robust.

In a final robustness check, we compared the $PP_{top\ 10\%}$ to the corresponding $PP_{top\ 50\%}$ journal ranking (see the results in Table 8). The $PP_{top\ 50\%}$ indicator is the percentage of papers (published by a journal) which are among the 50% most frequently cited papers in the corresponding economic subfields and publication years. As the $PP_{top\ 10\%}$ ranking is more selective than the $PP_{top\ 50\%}$ ranking, more journals in the $PP_{top\ 50\%}$ ranking are grouped in better categories than in the $PP_{top\ 10\%}$ ranking. As a consequence, the percent agreement between both rankings and the corresponding Kappa coefficient are only on a moderate level. All journals listed in Table 6 are also outstandingly cited based on the $PP_{top\ 50\%}$ ranking.

Discussion

Field- and time-normalisation of citation impact is the standard method in bibliometrics (Hicks et al. 2015), which should be applied in citation impact analyses across different time periods and subfields in economics. The most important reason is that there are different publication and citation cultures, which lead to subfield- and time-specific citation rates: for example, the mean citation rate in “General Economics and Teaching” decreases from 12 citations in 2000 to 5 citations in 2009. There is a low rate of only 7 citations in “History of Economic Thought, Methodology, and Heterodox Approaches”, but a high rate of 31 citations in “Financial Economics” (for papers published in 2001). Anauati et al. (2016) and other studies have confirmed the evidence that citation rates in subfields of economics differs. Without consideration of time- and subfield-specific differences in citation impact analysis, fair comparisons between scientific entities (e.g. journals, single researchers, research groups, and institutions) are impossible and entities with publication sets from recent time periods and in specific subfields are at a disadvantage.

In this study, we applied two normalised indicators in economics, which are the most important indicators in bibliometrics. The MNCS compares the citation impact of a focal paper with the mean impact of similar papers published in the same subfield and publication year. Thomson Reuters (2015) published a list of recommendations, which should be

considered in the use of this indicator: for example, “use larger sets of publications when possible, for example, by extending the time period or expanding the number of subjects to be covered” (p. 15). We strongly encourage the consideration of the listed points in bibliometric studies in economic using the MNCS. However, Thomson Reuters (2015) and many bibliometricians view the influence of very highly cited papers on the mean as a measure of central tendency as a serious problem of the MNCS: “In our view, the sensitivity of the MNCS indicator to a single very highly cited publication is an undesirable property” (Waltman et al. 2012, p. 2425).

In recent years, percentiles have become popular as a better alternative to mean-based normalised indicators. The share of papers belonging to the x % most cited papers is regarded as the most important citation impact indicator in the Leiden Ranking (Waltman et al. 2012). According to Li and Ruiz-Castillo (2014), the percentile rank indicator is robust to extreme observations. In this study, we used the $PP_{top\ 10\%}$ indicator to identify highly cited papers in a certain subfield and time period. Besides focusing on the 10% most frequently cited papers, it is also possible to focus on the 1% or 20% most frequently cited papers ($PP_{top\ 1\%}$ or $PP_{top\ 20\%}$). As the results of Waltman et al. (2012) show, however, the focus on another percentile rank is expected to lead to similar results. Besides percentiles, the use of log-transformed citations instead of citations in the MNCS formula has also been proposed as an alternative (Thelwall 2017). However, this alternative has not reached the status of a standard in bibliometrics yet.

In this study, we calculated normalised scores for each paper. The normalisation leads to similar impact assignments for many papers; however, there is also a high level of disagreement. There are several cases in the data that demonstrate unreasonable advantages or disadvantages for the papers if the impact is measured by citation counts without consideration of subfield- and time-specific baselines. For example, we can expect that papers published in “History of Economic Thought, Methodology, and Heterodox Approaches” and papers published recently are systematically disadvantaged in research evaluations across different subfields and time (because of their low mean citation rates). By contrast, papers from “Financial Economics” and papers published several years ago are systematically advantaged, since more citations can be expected. Thus, we attach importance to the consideration of normalisation in economic impact studies, which is strongly recommended by experts in bibliometrics (Hicks et al. 2015).

In this study, we introduce a new journal ranking, which is based on subfield-normalized citation scores. The results of the study reveal that only two journals can be meaningfully separated from the rest of economics journals (in terms of both indicators MNCS and $PP_{top\ 10\%}$): *Quarterly Journal of Economics* and *Journal of Economic Literature*. This selection is based on field- and time-normalised impact indicators which are the best available indicators in bibliometrics for the quality assessment of journals. According to Bornmann and Marx (2014b), the benefit of citation analysis is based on what Galton (1907) called the “wisdom of crowds”. In the next few years, future studies should investigate with field-normalised indicators whether both journals can hold this position or will be replaced by other journals.

The ideal way of assessing entities in science, such as journals, is to combine quantitative (metrics) and qualitative (peer review) assessments to overcome the disadvantages of both approaches each. For example, the most-reputable journals that are used for calculating the Nature Index (NI, see <https://www.natureindex.com>) are identified by two expert panels (Bornmann and Haunschild 2017; Haunschild and Bornmann 2015). The NI counts the publications in these most-reputable journals; the index is used by the Nature Publishing Group (NPG) to rank institutions and countries. To apply the ideal method of research

evaluation in economics, peer review and metrics should be combined to produce a list of top-journals in economics: a panel of economists uses the recommendations from our study with the two separated (based on CIs) and further 15 outstandingly cited journals and compare them with the rest of the journals according to their importance in economics. Ferrara and Bonaccorsi (2016) offer advice on how a journal ranking can be produced by using expert panels.

In this study we used a dataset with normalised scores on the paper level to identify the most frequently cited papers and journals. The dataset can be further used for various other entities in economics. The most frequently cited researchers, research groups, institutions, and countries can be determined subfield- and time-normalised. On the level of single researchers, we recommend that the normalised scores should be used instead of the popular h index proposed by Hirsch (2005). Like citation counts, the h index is not time- and subfield normalised. It is also dependent on the academic age of the researcher. Thus, Bornmann and Marx (2014a) recommended calculating the sum of $P_{\text{top } 10\%}$ for a researcher and dividing it by the number of his or her academic years. This results in a subfield-, time-, and age-normalised impact score. In future studies, we will apply citation impact normalisation on different entities in economics. It would be helpful for these studies if normalised impact scores were to be regularly included in RePec, although it is a sophisticated task to produce these scores.

Acknowledgements Open access funding provided by Max Planck Society. The bibliometric data used in this paper are from an in-house database developed and maintained by the Max Planck Digital Library (MPDL, Munich) and derived from the Science Citation Index Expanded (SCI-E), Social Sciences Citation Index (SSCI), Arts and Humanities Citation Index (AHCI) prepared by Clarivate Analytics (Philadelphia, Pennsylvania, USA). We would like to thank Ludo Waltman at the Centre for Science and Technology Studies (CWTS) of the Leiden University for his support in calculating the field-normalised indicators. We would also like to thank David Card, Daniel Hamermesh, Stefano DellaVigna and participants at the workshop “Economics of Science and Innovation” at the Summer Forum in Barcelona 2017 for their valuable feedback on previous versions of the manuscript.

Open Access This article is distributed under the terms of the Creative Commons Attribution 4.0 International License (<http://creativecommons.org/licenses/by/4.0/>), which permits unrestricted use, distribution, and reproduction in any medium, provided you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons license, and indicate if changes were made.

Appendix 1

Table 9 Descriptive statistics for the journals included in this study and journal rankings based on the mean normalised citation scores (MNCS) and PP_{top 10%}

Journal	Descriptive statistics						PP _{top 10%}					MNCS					
	Start	Papers	Total citations	Citations per paper	Paper citations			Confidence interval			Confidence interval						
					Min	Max	Rank	Score	Lower	Upper	Rank	Mean	Lower	Upper			
CSS: outstandingly cited journals																	
<i>Quarterly Journal of Economics</i>	1991	966	140,461	145.4	0	2,369	1	70.48	67.52	73.27	2	6.35	5.87	6.84			
<i>Journal of Economic Literature</i>	1991	422	54,936	130.2	0	1,242	2	63.71	59.02	68.16	1	6.79	5.98	7.59			
<i>Journal of Political Economy</i>	1991	966	96,697	100.1	0	3,300	3	52.16	49.00	55.29	3	4.17	3.81	4.53			
<i>American Economic Review</i>	1991	2,274	177,182	77.9	0	1,838	4	49.75	47.70	51.81	4	4.02	3.79	4.24			
<i>American Economic Journal-Applied Economics</i>	2009	188	4618	24.6	1	95	5	49.44	42.37	56.52	10	2.99	2.66	3.32			
<i>American Economic Journal-Macroeconomics</i>	2009	146	3681	25.2	0	137	6	46.55	38.65	54.62	7	3.37	2.82	3.92			
<i>Journal of Finance</i>	1991	1,151	92,389	80.3	0	2,089	7	45.20	42.35	48.09	8	3.24	3.01	3.46			
<i>Journal of Economic Perspectives</i>	1991	1,078	71,984	66.8	0	1,273	8	44.04	41.10	47.02	6	3.38	3.11	3.65			
<i>Econometrica</i>	1991	1,296	111,278	85.9	0	2,306	9	43.20	40.53	45.91	5	3.49	3.20	3.77			
<i>Journal of Financial Economics</i>	1991	1,596	108,964	68.3	0	3,707	10	39.79	37.41	42.21	9	3.07	2.88	3.27			
<i>Review of Financial Studies</i>	1991	827	32,980	39.9	0	1,644	11	37.46	34.23	40.81	13	2.64	2.37	2.91			
<i>Annual Review of Economics</i>	2009	113	1973	17.5	0	107	12	37.07	28.73	46.26	18	2.29	1.89	2.68			
<i>Transportation Research Part B-Methodological</i>	2001	487	12,381	25.4	0	288	13	37.05	32.88	41.43	14	2.48	2.29	2.68			
CSS: remarkably cited journals																	
<i>Review of Economic Studies</i>	1991	984	60,027	61.0	0	4,627	14	35.22	32.30	38.26	11	2.87	2.54	3.20			
<i>American Economic Journal-Economic Policy</i>	2009	157	2658	16.9	0	134	15	33.17	26.28	40.86	16	2.32	1.95	2.69			
<i>Journal of Economic Growth</i>	1999	194	10,491	54.1	0	937	16	31.35	25.24	38.19	12	2.76	2.13	3.39			
<i>Review of Environmental Economics and Policy</i>	2007	112	2672	23.9	0	166	17	30.75	22.96	39.82	20	2.11	1.69	2.54			
<i>Review of Economics and Statistics</i>	1991	1,593	65,089	40.9	0	978	18	27.66	25.52	29.90	19	2.13	1.99	2.27			
<i>Journal of Economic Geography</i>	2002	386	13,648	35.4	0	627	19	27.12	22.93	31.77	15	2.42	2.08	2.76			

Table 9 (continued)

Journal	Descriptive statistics						PP _{top10%}				MNCS			
	Start	Papers	Total citations	Citations per paper	Paper citations		Rank	Score	Confidence interval		Rank	Mean	Confidence interval	
					Min	Max			Lower	Upper			Lower	Upper
<i>Journal of Accounting & Economics</i>	1991	664	37,842	57.0	0	718	20	27.08	23.84	30.59	17	2.31	2.07	2.55
<i>Journal of International Economics</i>	1991	1337	43,370	32.4	0	749	21	26.34	24.05	28.77	22	2.02	1.86	2.18
<i>Transportation Research Part E-Logistics and Transportation Review</i>	1997	416	7070	17.0	0	93	22	25.87	21.90	30.29	25	1.84	1.67	2.01
<i>Journal of Monetary Economics</i>	1991	1410	48,025	34.1	0	700	23	23.65	21.51	25.94	24	1.85	1.71	1.99
<i>Transportation Research Part A-Policy and Practice</i>	2005	410	6207	15.1	0	142	24	23.64	19.79	27.99	27	1.79	1.63	1.95
<i>Energy Economics</i>	1991	1766	40,225	22.8	0	289	25	23.31	21.40	25.34	30	1.72	1.63	1.82
<i>Journal of Labor Economics</i>	1991	655	25,143	38.4	0	511	26	22.69	19.65	26.05	31	1.70	1.54	1.85
<i>Journal of Environmental Economics and Management</i>	1991	1088	37,856	34.8	0	650	27	21.38	19.05	23.92	32	1.69	1.56	1.81
CSS: fairly cited journals														
<i>Rand Journal of Economics</i>	1991	928	34,587	37.3	0	634	28	20.49	18.01	23.20	34	1.63	1.49	1.76
<i>Economic Geography</i>	1991	412	13,070	31.7	0	570	29	20.31	16.71	24.46	33	1.65	1.43	1.87
<i>Ecological Economics</i>	1993	3125	93,398	29.9	0	1418	30	20.27	18.90	21.71	29	1.73	1.64	1.82
<i>Journal of Health Economics</i>	1991	1258	44,976	35.8	0	1515	31	19.93	17.81	22.23	28	1.74	1.58	1.91
<i>Economic Journal</i>	1991	1850	56,975	30.8	0	615	32	19.77	18.02	21.64	36	1.57	1.47	1.67
<i>Journal of The European Economic Association</i>	2003	661	15,007	22.7	0	754	33	19.66	16.81	22.86	26	1.79	1.55	2.04
<i>Journal of Economic Surveys</i>	2000	400	9609	24.0	0	687	34	19.09	15.54	23.23	35	1.58	1.30	1.86
<i>Economic Systems Research</i>	2008	127	1940	15.3	0	124	35	18.74	12.91	26.40	21	2.03	1.23	2.83
<i>Brookings Papers On Economic Activity</i>	1991	309	7708	24.9	0	413	36	18.44	14.52	23.15	48	1.42	1.19	1.65
<i>Journal of Human Resources</i>	1991	809	27,617	34.1	0	420	37	18.41	15.89	21.23	38	1.51	1.39	1.63
<i>Socio-Economic Review</i>	2009	128	1446	11.3	0	124	38	18.41	12.65	26.01	40	1.49	1.17	1.80
<i>World Bank Economic Review</i>	1991	487	15,524	31.9	0	783	39	17.30	14.20	20.91	44	1.45	1.22	1.67
<i>Journal of Development Economics</i>	1991	1583	37,002	23.4	0	329	40	17.19	15.41	19.13	52	1.35	1.26	1.45
<i>Journal of Public Economics</i>	1991	2115	55,282	26.1	0	688	41	17.07	15.53	18.74	41	1.47	1.38	1.56

Table 9 (continued)

Journal	Descriptive statistics						PP _{top10%}						MNCS		
	Start	Papers	Total citations	Citations per paper	Paper citations		Rank	Score	Confidence interval		Rank	Mean	Confidence interval		
					Min	Max			Lower	Upper			Lower	Upper	
<i>Economic Policy</i>	1997	217	3860	17.8	0	265	42	16.88	12.48	22.43	51	1.36	1.14	1.59	
<i>Economics & Human Biology</i>	2006	292	4536	15.5	0	166	43	16.83	12.97	21.54	46	1.43	1.26	1.60	
<i>Journal of Econometrics</i>	1991	2332	89,458	38.4	0	3168	44	16.22	14.78	17.77	37	1.56	1.41	1.72	
<i>Journal of Economic Surveys</i>	1991	380	8243	21.7	0	307	45	16.00	12.66	20.03	50	1.36	1.15	1.58	
<i>Journal of Business & Economic Statistics</i>	1991	943	33,586	35.6	0	1613	46	15.92	13.73	18.40	39	1.49	1.28	1.69	
<i>American Economic Journal-Microeconomics</i>	2009	153	1642	10.7	0	145	47	15.62	10.72	22.20	45	1.43	1.12	1.74	
<i>Industrial and Corporate Change</i>	2002	543	12,620	23.2	0	527	48	15.48	12.68	18.76	42	1.47	1.29	1.64	
<i>Experimental Economics</i>	2000	273	6925	25.4	0	1848	49	15.03	11.27	19.75	23	1.85	1.17	2.54	
<i>Review of Finance</i>	2008	172	2151	12.5	0	124	50	14.97	10.41	21.06	47	1.43	1.18	1.67	
<i>Journal of Law & Economics</i>	1991	614	16,119	26.3	0	588	51	14.83	12.24	17.86	56	1.29	1.15	1.43	
<i>Journal of Applied Econometrics</i>	1991	914	27,414	30.0	0	1235	52	14.79	12.63	17.24	43	1.46	1.24	1.67	
<i>Journal of Urban Economics</i>	1991	1125	27,772	24.7	0	447	53	14.45	12.52	16.63	58	1.25	1.16	1.34	
<i>World Development</i>	1991	3180	74,997	23.6	0	694	54	14.06	12.89	15.31	59	1.22	1.16	1.28	
<i>Journal of Law Economics & Organization</i>	1991	537	16,575	30.9	0	1318	55	13.85	11.19	17.03	49	1.40	1.14	1.66	
<i>Health Economics</i>	1996	1446	35,343	24.4	0	502	56	13.53	11.86	15.39	57	1.26	1.17	1.35	
<i>World Bank Research Observer</i>	1993	233	6774	29.1	0	614	57	13.50	9.70	18.48	55	1.31	1.05	1.58	
<i>Journal of Financial and Quantitative Analysis</i>	1991	846	22,764	26.9	0	392	58	13.29	11.17	15.75	62	1.19	1.10	1.28	
<i>Small Business Economics</i>	1992	1142	22,570	19.8	0	273	59	13.09	11.26	15.17	66	1.13	1.05	1.21	
<i>European Economic Review</i>	1991	1369	30,462	22.3	0	1003	60	13.09	11.40	14.98	61	1.21	1.10	1.31	
<i>Cambridge Journal of Regions Economy and Society</i>	2008	153	1869	12.2	0	146	61	13.03	8.59	19.29	54	1.32	0.99	1.65	
<i>European Journal of Political Economy</i>	2008	325	2927	9.0	0	73	62	13.01	9.78	17.11	67	1.13	1.00	1.25	
<i>Journal of Industrial Economics</i>	1991	622	15,427	24.8	0	447	63	12.60	10.22	15.44	65	1.15	1.02	1.28	
<i>Journal of Policy Analysis and Management</i>	1993	539	10,362	19.2	0	175	64	12.42	9.90	15.47	71	1.09	0.99	1.20	

Table 9 (continued)

	Descriptive statistics						PP _{top10%}				MNCS			
	Start	Papers	Total citations	Citations per paper	Paper citations		Rank	Score	Confidence interval		Rank	Mean	Confidence interval	
					Min	Max			Lower	Upper			Lower	Upper
<i>Journal</i>														
<i>Food Policy</i>	1996	880	15,280	17.4	0	301	65	12.02	10.04	14.34	63	1.18	1.08	1.27
<i>Journal of Money Credit and Banking</i>	1991	1246	22,364	17.9	0	281	66	11.65	9.98	13.55	68	1.10	1.02	1.19
<i>Journal of Financial Stability</i>	2008	175	1627	9.3	0	93	67	11.57	7.64	17.16	60	1.21	0.98	1.43
<i>Review of International Organizations</i>	2008	104	827	8.0	0	50	68	11.35	6.59	18.87	69	1.10	0.89	1.31
<i>Journal of Risk and Uncertainty</i>	1991	599	19,528	32.6	0	2985	69	10.76	8.52	13.49	53	1.35	1.05	1.65
<i>Regional Studies</i>	1994	667	9867	14.8	0	892	70	10.47	8.37	13.03	64	1.15	0.99	1.32
<i>Journal of Evolutionary Economics</i>	1996	491	6471	13.2	0	180	71	10.32	7.93	13.32	86	0.94	0.82	1.07
<i>Journal of Comparative Economics</i>	1991	847	13,302	15.7	0	208	72	10.27	8.40	12.50	80	0.98	0.90	1.07
<i>International Journal of Forecasting</i>	1992	780	14,921	19.1	0	1038	73	10.19	8.26	12.51	70	1.09	0.95	1.24
<i>Journal of Economic Psychology</i>	1991	1100	18,075	16.4	0	474	74	10.05	8.41	11.97	75	1.02	0.94	1.11
<i>Resource and Energy Economics</i>	1993	509	8903	17.5	0	187	75	10.03	7.71	12.94	81	0.98	0.86	1.10
<i>Econometric Reviews</i>	2005	225	2555	11.4	0	248	76	9.99	6.72	14.60	93	0.90	0.68	1.12
<i>Land Economics</i>	1991	910	20,863	22.9	0	459	77	9.86	8.09	11.97	72	1.06	0.97	1.16
<i>Cambridge Journal of Economics</i>	1991	1040	13,129	12.6	0	577	78	9.30	7.68	11.22	95	0.88	0.80	0.97
<i>Energy Journal</i>	1991	605	9487	15.7	0	226	79	9.22	7.16	11.79	85	0.94	0.84	1.05
<i>Journal of Economics & Management Strategy</i>	1992	630	11,219	17.8	0	234	80	9.13	7.12	11.63	92	0.90	0.81	1.00
<i>Journal of Economic Behavior & Organization</i>	1991	2431	36,138	14.9	0	685	81	9.03	7.95	10.23	90	0.92	0.86	0.97
<i>Jems-Journal of Common Market Studies</i>	2005	534	4933	9.2	0	85	82	9.02	6.87	11.75	73	1.06	0.95	1.16
<i>International Economic Review</i>	1991	1164	25,146	21.6	0	658	83	9.00	7.49	10.79	74	1.03	0.92	1.13
<i>Journal of Regional Science</i>	1991	524	8170	15.6	0	169	84	8.84	6.70	11.58	83	0.95	0.83	1.07
<i>Journal of Banking & Finance</i>	1991	3161	54,633	17.3	0	787	85	8.84	7.90	9.88	79	0.99	0.94	1.03
<i>Journal of Economic Theory</i>	1991	2144	42,260	19.7	0	609	86	8.71	7.59	9.98	82	0.95	0.90	1.00
<i>Oxford Review of Economic Policy</i>	1991	740	11,768	15.9	0	246	87	8.61	6.80	10.86	96	0.88	0.79	0.97

Table 9 (continued)

	Descriptive statistics						PP _{top10%}				MNCS			
	Start	Papers	Total citations	Citations per paper	Paper citations		Rank	Score	Confidence interval		Rank	Mean	Confidence interval	
					Min	Max			Lower	Upper			Lower	Upper
<i>Journal of Productivity Analysis</i>	1995	575	9142	15.9	0	325	88	8.22	6.24	10.75	105	0.84	0.74	0.94
<i>Review of Economic Dynamics</i>	2001	502	6643	13.2	0	296	89	8.16	6.07	10.88	78	1.00	0.86	1.14
CSS: poorly cited journals														
<i>Regional Science and Urban Economics</i>	1991	971	15,480	15.9	0	422	90	7.87	6.34	9.74	99	0.86	0.79	0.93
<i>Environmental & Resource Economics</i>	1995	1219	20,523	16.8	0	363	91	7.65	6.29	9.28	89	0.92	0.84	0.99
<i>Journal of Economic Inequality</i>	2008	139	1001	7.2	0	59	92	7.63	4.27	13.27	87	0.93	0.66	1.19
<i>Papers In Regional Science</i>	2000	282	3050	10.8	0	158	93	7.54	5.00	11.22	100	0.86	0.72	0.99
<i>Games and Economic Behavior</i>	1991	1829	34,829	19.0	0	1200	94	7.41	6.30	8.70	94	0.89	0.81	0.97
<i>European Journal of Health Economics</i>	2007	359	3072	8.6	0	155	95	7.41	5.13	10.59	91	0.91	0.79	1.03
<i>Oxford Bulletin of Economics and Statistics</i>	1991	842	16,283	19.3	0	1254	96	7.37	5.79	9.33	97	0.87	0.72	1.03
<i>China Economic Review</i>	1995	631	7997	12.7	0	152	97	7.28	5.50	9.57	84	0.94	0.86	1.03
<i>Annual Review of Resource Economics</i>	2009	101	908	9.0	0	73	98	7.28	3.63	14.06	101	0.86	0.69	1.03
<i>Annual Review of Financial Economics</i>	2009	83	646	7.8	0	52	99	7.18	3.32	14.82	102	0.85	0.63	1.08
<i>Oxford Economic Papers-New Series</i>	1991	857	13,843	16.2	0	840	100	7.13	5.59	9.05	114	0.79	0.66	0.91
<i>Theoretical Economics</i>	2007	117	957	8.2	0	44	101	7.07	3.66	13.20	77	1.00	0.85	1.15
<i>Emerging Markets Review</i>	2009	155	1020	6.6	0	38	102	6.97	3.91	12.10	88	0.93	0.77	1.08
<i>One-Quantitative Marketing and Economics</i>	2006	119	1244	10.5	0	62	103	6.91	3.57	12.94	104	0.84	0.67	1.01
<i>Journal of Economic Interaction and Coordination</i>	2008	78	349	4.5	0	37	104	6.87	3.05	14.73	164	0.55	0.36	0.74
<i>Econometric Theory</i>	1991	999	19,172	19.2	0	842	105	6.84	5.43	8.57	120	0.76	0.68	0.84
<i>American Journal of Agricultural Economics</i>	1991	2724	44,923	16.5	0	402	106	6.81	5.93	7.82	116	0.78	0.74	0.82
<i>Kyklos</i>	1991	605	7277	12.0	0	317	107	6.71	4.97	8.98	132	0.71	0.61	0.81
<i>Economica</i>	1991	778	10,438	13.4	0	412	108	6.35	4.84	8.29	129	0.73	0.65	0.80
<i>Journal of Business Economics and Management</i>	2009	206	1402	6.8	0	69	109	6.23	3.66	10.40	117	0.77	0.63	0.91

Table 9 (continued)

Journal	Descriptive statistics						PP _{top10%}				MNCS			
	Start	Papers	Total citations	Citations per paper	Paper citations		Rank	Score	Confidence interval		Rank	Mean	Confidence interval	
					Min	Max			Lower	Upper			Lower	Upper
<i>Journal of Empirical Finance</i>	2008	342	2750	8.0	0	99	110	6.14	4.05	9.20	125	0.74	0.64	0.83
<i>Review of International Political Economy</i>	1994	608	8003	13.2	0	915	111	6.11	4.47	8.31	98	0.87	0.70	1.03
<i>Labour Economics</i>	2000	793	9288	11.7	0	190	112	6.07	4.61	7.95	108	0.82	0.75	0.89
<i>European Review of Agricultural Economics</i>	1993	530	7555	14.3	0	286	113	6.05	4.32	8.41	121	0.76	0.67	0.84
<i>Mathematical Finance</i>	1997	450	11,525	25.6	0	1828	114	5.95	4.11	8.53	76	1.02	0.71	1.33
<i>Journal of Institutional Economics</i>	2009	128	613	4.8	0	53	115	5.85	2.93	11.34	118	0.77	0.57	0.96
<i>International Journal of Industrial Organization</i>	1991	1291	19,921	15.4	0	477	116	5.84	4.69	7.26	115	0.78	0.73	0.84
<i>Journal of Economic History</i>	1991	750	8922	11.9	0	155	117	5.76	4.31	7.67	112	0.79	0.73	0.86
<i>Economic Inquiry</i>	1991	1324	16,900	12.8	0	288	118	5.76	4.63	7.15	127	0.73	0.67	0.79
<i>Econometrics Journal</i>	2005	230	2038	8.9	0	202	119	5.65	3.33	9.43	137	0.67	0.53	0.81
<i>North American Journal of Economics and Finance</i>	2008	178	914	5.1	0	46	120	5.57	3.04	9.97	107	0.84	0.70	0.97
<i>New Political Economy</i>	2003	311	2499	8.0	0	67	121	5.54	3.49	8.66	128	0.73	0.62	0.84
<i>Journal of Economic Dynamics & Control</i>	1991	2148	28,401	13.2	0	468	122	5.52	4.63	6.57	130	0.72	0.68	0.77
<i>Journal of Development Studies</i>	1991	1043	12,565	12.0	0	396	123	5.52	4.29	7.08	124	0.74	0.68	0.81
<i>Feminist Economics</i>	1998	363	4066	11.2	0	281	124	5.51	3.59	8.35	122	0.75	0.65	0.85
<i>Economics of Education Review</i>	1995	1111	14,708	13.2	0	221	125	5.48	4.29	6.98	111	0.80	0.74	0.86
<i>Journal of Financial Economics</i>	2007	142	1415	10.0	0	225	126	5.45	2.76	10.49	103	0.85	0.61	1.10
<i>Economic Development and Cultural Change</i>	1991	740	11,588	15.7	0	407	127	5.45	4.03	7.33	123	0.75	0.67	0.82
<i>Industry and Innovation</i>	2008	192	1535	8.0	0	68	128	5.38	2.97	9.53	106	0.84	0.72	0.96
<i>Journal of Population Economics</i>	1992	809	10,419	12.9	0	126	129	5.28	3.94	7.04	119	0.76	0.70	0.83
<i>Scandinavian Journal of Economics</i>	1991	846	10,853	12.8	0	410	130	5.15	3.85	6.85	134	0.70	0.63	0.77
<i>Economics and Philosophy</i>	1991	310	2862	9.2	0	179	131	4.97	3.06	7.99	141	0.65	0.53	0.78
<i>Journal of Transport Economics and Policy</i>	1991	480	6589	13.7	0	159	132	4.93	3.33	7.25	133	0.71	0.63	0.78

Table 9 (continued)

Journal	Descriptive statistics						PP _{top10%}				MNCS			
	Start	Papers	Total citations	Citations per paper	Paper citations		Rank	Score	Confidence interval		Rank	Mean	Confidence interval	
					Min	Max			Lower	Upper			Lower	Upper
<i>Economic History Review</i>	1991	678	6013	8.9	0	142	133	4.81	3.44	6.69	135	0.68	0.62	0.75
<i>International Tax and Public Finance</i>	1998	564	5111	9.1	0	136	134	4.58	3.14	6.63	148	0.62	0.54	0.70
<i>Spatial Economic Analysis</i>	2008	118	982	8.3	0	124	135	4.30	1.86	9.61	109	0.82	0.57	1.07
<i>Public Choice</i>	1991	2158	20,660	9.6	0	355	136	4.27	3.50	5.21	156	0.59	0.54	0.64
<i>Review of World Economics</i>	1991	732	6699	9.2	0	100	137	4.19	2.97	5.90	150	0.60	0.53	0.67
<i>Journal of Housing Economics</i>	1995	379	4252	11.2	0	120	138	4.06	2.49	6.55	161	0.57	0.50	0.64
<i>Journal of Agricultural Economics</i>	1991	692	8391	12.1	0	131	139	4.01	2.79	5.74	140	0.65	0.59	0.72
<i>International Review of Economics & Finance</i>	2008	410	2208	5.4	0	64	140	3.99	2.48	6.35	126	0.73	0.65	0.82
<i>Agricultural Economics</i>	2000	914	10,827	11.8	0	330	141	3.92	2.84	5.38	136	0.68	0.62	0.73
<i>Information Economics and Policy</i>	2000	361	3532	9.8	0	125	142	3.79	2.25	6.29	143	0.65	0.55	0.75
<i>Clometrica</i>	2007	88	436	5.0	0	51	143	3.69	1.31	9.95	157	0.59	0.40	0.77
<i>German Economic Review</i>	2007	176	854	4.9	0	43	144	3.65	1.73	7.56	169	0.53	0.40	0.65
<i>Canadian Journal of Economics-Revue Canadienne D Economique</i>	1991	1428	13,028	9.1	0	189	145	3.63	2.78	4.73	179	0.50	0.46	0.54
<i>Post-Soviet Affairs</i>	1992	245	1998	8.2	0	63	146	3.60	1.89	6.74	145	0.63	0.55	0.72
<i>Economic Modelling</i>	1991	2099	11,007	5.2	0	280	147	3.58	2.87	4.47	160	0.57	0.53	0.61
<i>Explorations In Economic History</i>	1991	556	4952	8.9	0	194	148	3.53	2.28	5.41	142	0.65	0.57	0.73
<i>Journal of Regulatory Economics</i>	1991	663	7457	11.2	0	337	149	3.50	2.35	5.19	158	0.58	0.51	0.66
<i>Insurance Mathematics & Economics</i>	1991	1461	17,754	12.2	0	310	150	3.43	2.61	4.49	144	0.64	0.59	0.68
<i>World Economy</i>	1992	1860	13,491	7.3	0	293	151	3.39	2.66	4.32	162	0.56	0.52	0.60
<i>International Finance</i>	2007	118	489	4.1	0	31	152	3.39	1.33	8.39	186	0.46	0.34	0.58
<i>Journal of Forecasting</i>	2002	214	1392	6.5	0	205	153	3.37	1.66	6.73	159	0.58	0.45	0.71
<i>Economics of Transition</i>	1997	465	4702	10.1	0	151	154	3.29	2.01	5.33	149	0.62	0.55	0.69
<i>Journal of Forest Economics</i>	2005	180	1602	8.9	0	61	155	3.24	1.48	6.96	131	0.71	0.62	0.81
<i>Quantitative Finance</i>	2001	890	7274	8.2	0	556	156	3.20	2.23	4.57	175	0.51	0.45	0.58

Table 9 (continued)

	Descriptive statistics						PP _{top10%}				MNCS			
	Paper citations						Confidence interval				Confidence interval			
	Start	Papers	Total citations	Citations per paper	Min	Max	Rank	Score	Lower	Upper	Rank	Mean	Lower	Upper
<i>Journal</i>														
<i>Review of International Economics</i>	2007	480	2228	4.6	0	58	157	3.16	1.93	5.14	167	0.53	0.47	0.59
<i>Journal of Real Estate Research</i>	2006	162	1138	7.0	0	51	158	3.08	1.32	7.02	151	0.60	0.51	0.69
<i>Journal of Pension Economics & Finance</i>	2008	126	520	4.1	0	67	159	2.96	1.12	7.58	182	0.49	0.33	0.65
<i>National Tax Journal</i>	1991	959	9806	10.2	0	425	160	2.89	2.00	4.15	165	0.55	0.49	0.60
<i>Journal of Consumer Affairs</i>	1998	121	869	7.2	0	66	161	2.87	1.05	7.59	138	0.66	0.55	0.77
<i>Macroeconomic Dynamics</i>	1998	576	4215	7.3	0	354	162	2.78	1.72	4.46	180	0.50	0.44	0.56
<i>Economic Theory</i>	1995	1682	14,915	8.9	0	420	163	2.67	2.00	3.56	172	0.52	0.48	0.56
<i>Australian Journal of Agricultural and Resource Economics</i>	1997	457	4806	10.5	0	184	164	2.65	1.52	4.56	153	0.59	0.52	0.67
<i>American Law and Economics Review</i>	2008	97	626	6.5	0	65	165	2.58	0.80	7.97	110	0.80	0.64	0.96
<i>Journal of Real Estate Finance and Economics</i>	1993	866	9429	10.9	0	462	166	2.54	1.68	3.82	174	0.51	0.46	0.56
<i>Empirical Economics</i>	2002	691	3909	5.7	0	171	167	2.54	1.60	4.00	168	0.53	0.46	0.59
<i>Journal of Risk and Insurance</i>	1995	594	7085	11.9	0	141	168	2.54	1.54	4.14	155	0.59	0.53	0.65
<i>Review of Income and Wealth</i>	1991	675	5608	8.3	0	128	169	2.48	1.55	3.95	181	0.49	0.43	0.55
<i>Journal of The Japanese and International Economics</i>	1991	551	4758	8.6	0	225	170	2.40	1.42	4.05	184	0.47	0.40	0.54
<i>Real Estate Economics</i>	1996	396	5624	14.2	0	85	171	2.35	1.26	4.36	147	0.62	0.56	0.69
<i>Review of Development Economics</i>	2005	444	2758	6.2	0	138	172	2.32	1.27	4.18	176	0.51	0.45	0.58
<i>International Journal of Health Care Finance & Economics</i>	2008	107	528	4.9	0	26	173	2.29	0.71	7.19	166	0.55	0.44	0.66
<i>Journal of Economic Issues</i>	1991	1507	6614	4.4	0	252	174	2.22	1.59	3.10	216	0.36	0.33	0.39
<i>Metroeconomica</i>	2008	198	480	2.4	0	33	175	2.22	0.90	5.35	213	0.36	0.28	0.45
<i>Review of Economics of The Household</i>	2008	144	809	5.6	0	47	176	2.19	0.77	6.10	146	0.63	0.52	0.73
<i>Journal of Institutional and Theoretical Economics-Zeitschrift Fur Die Gesamte Staatswissenschaft</i>	1991	828	5000	6.0	0	207	177	2.16	1.37	3.39	208	0.37	0.30	0.44
<i>International Labour Review</i>	1991	270	1732	6.4	0	75	178	2.15	0.98	4.67	188	0.45	0.36	0.54

Table 9 (continued)

	Descriptive statistics						PP _{top10%}				MNCS			
	Paper citations						Confidence interval				Confidence interval			
	Start	Papers	Total citations	Citations per paper	Min	Max	Rank	Score	Lower	Upper	Rank	Mean	Lower	Upper
<i>Journal</i>														
<i>Southern Economic Journal</i>	1991	1411	10,860	7.7	0	127	179	2.13	1.50	3.03	197	0.41	0.38	0.44
<i>Journal of Economic Education</i>	1991	773	4110	5.3	0	202	180	2.11	1.31	3.38	193	0.42	0.37	0.48
<i>Social Choice and Welfare</i>	1991	1191	9076	7.6	0	106	181	2.02	1.36	2.99	191	0.43	0.40	0.46
<i>European Review of Economic History</i>	2007	126	723	5.7	0	47	182	2.00	0.62	6.21	139	0.66	0.54	0.77
<i>Contemporary Economic Policy</i>	1994	858	6608	7.7	0	110	183	1.99	1.25	3.16	198	0.39	0.36	0.43
<i>Econ Journal Watch</i>	2004	157	663	4.2	0	49	184	1.99	0.69	5.58	183	0.47	0.37	0.58
<i>Economic Development Quarterly</i>	1996	394	3892	9.9	0	713	185	1.97	0.99	3.87	163	0.55	0.38	0.72
<i>Journal of Policy Modeling</i>	1991	1170	8056	6.9	0	196	186	1.95	1.30	2.92	187	0.45	0.41	0.49
<i>Economics & Politics</i>	2009	90	392	4.4	0	40	187	1.94	0.49	7.33	152	0.59	0.45	0.74
<i>Asian Economic Papers</i>	2008	97	266	2.7	0	26	188	1.92	0.51	7.00	215	0.36	0.24	0.48
<i>World Trade Review</i>	2008	355	1286	3.6	0	34	189	1.90	0.91	3.92	170	0.52	0.46	0.59
<i>Economics Letters</i>	1991	6379	44,076	6.9	0	1401	190	1.83	1.53	2.19	202	0.39	0.36	0.41
<i>Cesifo Economic Studies</i>	2005	235	1177	5.0	0	44	191	1.82	0.73	4.46	185	0.46	0.39	0.54
<i>Journal of Macroeconomics</i>	1991	1117	5966	5.3	0	121	192	1.75	1.13	2.70	199	0.39	0.36	0.43
<i>Fiscal Studies</i>	2001	256	1791	7.0	0	77	193	1.64	0.65	4.05	177	0.51	0.43	0.58
<i>Annals of Regional Science</i>	1993	380	1797	4.7	0	117	194	1.58	0.73	3.40	192	0.43	0.36	0.50
<i>Journal of Agricultural and Resource Economics</i>	1992	699	6557	9.4	0	93	195	1.55	0.86	2.77	189	0.44	0.40	0.48
<i>Studies in Nonlinear Dynamics and Econometrics</i>	1997	358	2011	5.6	0	188	196	1.54	0.68	3.42	238	0.27	0.21	0.33
<i>Review of Radical Political Economics</i>	2009	147	377	2.6	0	50	197	1.47	0.42	5.00	201	0.39	0.28	0.49
<i>China & World Economy</i>	2006	310	1429	4.6	0	104	198	1.45	0.59	3.49	194	0.42	0.36	0.49
<i>Politická Ekonomie</i>	1994	649	1434	2.2	0	33	199	1.38	0.73	2.61	249	0.24	0.20	0.28
<i>International Journal of Game Theory</i>	1991	713	5103	7.2	0	143	200	1.38	0.75	2.53	214	0.36	0.32	0.41
<i>Manchester School</i>	1998	705	3531	5.0	0	87	201	1.35	0.72	2.50	229	0.31	0.27	0.34
<i>Pacific Economic Review</i>	2005	332	1248	3.8	0	51	202	1.33	0.54	3.23	221	0.35	0.29	0.41

Table 9 (continued)

Journal	Descriptive statistics						PP _{top10%}				MNCS			
	Start	Papers	Total citations	Citations per paper	Paper citations		Score	Confidence interval		Rank	Mean	Confidence interval		
					Min	Max		Lower	Upper			Lower	Upper	
<i>Geneva Risk and Insurance Review</i>	2005	76	330	4.3	0	90	203	1.32	0.23	7.08	0.34	0.21	0.46	
<i>Journal of Public Economic Theory</i>	2007	308	1119	3.6	0	93	204	1.29	0.50	3.28	0.37	0.30	0.43	
<i>Ekonomika Istrazivanja-Economic Research</i>	2007	266	383	1.4	0	23	205	1.28	0.46	3.49	0.20	0.15	0.26	
<i>Journal of Applied Economics</i>	2005	153	579	3.8	0	47	206	1.23	0.33	4.52	0.31	0.23	0.39	
<i>Review of Industrial Organization</i>	1994	795	6718	8.5	0	169	207	1.23	0.66	2.26	0.42	0.38	0.46	
<i>Annals of Economics and Finance</i>	2007	141	259	1.8	0	40	208	1.18	0.29	4.66	0.21	0.13	0.28	
<i>History of Political Economy</i>	1991	859	2442	2.8	0	22	209	1.17	0.64	2.14	0.38	0.34	0.41	
<i>Panoeconomicus</i>	2008	183	394	2.2	0	20	210	1.17	0.33	4.01	0.32	0.25	0.38	
<i>Applied Economics</i>	1991	5121	33,649	6.6	0	423	211	1.16	0.90	1.49	0.38	0.36	0.40	
<i>China Agricultural Economic Review</i>	2009	151	435	2.9	0	43	212	1.13	0.28	4.41	0.38	0.24	0.52	
<i>Tourism Economics</i>	2008	305	1248	4.1	0	32	213	1.12	0.40	3.05	0.50	0.45	0.56	
<i>Theory and Decision</i>	1995	673	3645	5.4	0	111	214	1.11	0.55	2.23	0.37	0.33	0.41	
<i>European Journal of Law and Economics</i>	2008	231	455	2.0	0	20	215	1.08	0.34	3.43	0.28	0.23	0.34	
<i>Open Economies Review</i>	1995	557	2195	3.9	0	93	216	1.08	0.50	2.34	0.32	0.28	0.37	
<i>Amfiteatru Economic</i>	2008	339	1039	3.1	0	49	217	1.08	0.41	2.85	0.36	0.31	0.41	
<i>Journal of Competition Law & Economics</i>	2005	269	1179	4.4	0	35	218	1.07	0.36	3.17	0.44	0.38	0.50	
<i>Economic Record</i>	1991	799	5092	6.4	0	163	219	1.06	0.55	2.05	0.38	0.31	0.45	
<i>Portuguese Economic Journal</i>	2005	95	280	2.9	0	48	220	1.05	0.19	5.72	0.24	0.14	0.33	
<i>Journal of Cultural Economics</i>	2007	96	518	5.4	0	39	221	1.04	0.18	5.67	0.59	0.48	0.71	
<i>History of Economic Ideas</i>	2009	98	72	0.7	0	30	222	1.02	0.18	5.56	0.15	0.05	0.24	
<i>Journal of Economics</i>	1991	756	3375	4.5	0	101	223	0.99	0.49	1.98	0.26	0.23	0.29	
<i>Defence and Peace Economics</i>	1994	545	2839	5.2	0	76	224	0.98	0.43	2.21	0.35	0.31	0.40	
<i>Economist-Netherlands</i>	1991	490	2145	4.4	0	183	225	0.95	0.40	2.27	0.26	0.21	0.30	
<i>Journal of Economic Policy Reform</i>	2007	159	419	2.6	0	31	226	0.94	0.22	3.98	0.32	0.24	0.40	

Table 9 (continued)

Journal	Descriptive statistics						PP _{top10%}				MNCS			
	Start	Papers	Total citations	Citations per paper	Paper citations		Rank	Score	Confidence interval		Rank	Mean	Confidence interval	
					Min	Max			Lower	Upper			Lower	Upper
<i>Journal of Post Keynesian Economics</i>	1991	794	3490	4.4	0	70	227	0.87	0.42	1.79	235	0.29	0.26	0.32
<i>Journal of African Economics</i>	1997	515	3653	7.1	0	257	228	0.85	0.34	2.08	196	0.42	0.36	0.47
<i>Emerging Markets Finance and Trade</i>	2003	459	2436	5.3	0	32	229	0.84	0.32	2.17	171	0.52	0.48	0.57
<i>Prague Economic Papers</i>	2008	140	352	2.5	0	18	230	0.79	0.15	4.07	211	0.37	0.27	0.46
<i>Scottish Journal of Political Economy</i>	1991	721	5129	7.1	0	122	231	0.75	0.33	1.69	209	0.37	0.33	0.41
<i>Mathematical Social Sciences</i>	1998	335	1377	4.1	0	181	232	0.66	0.19	2.25	220	0.35	0.29	0.41
<i>International Review of Law and Economics</i>	1995	578	4003	6.9	0	89	233	0.65	0.25	1.71	200	0.39	0.35	0.43
<i>Transformations In Business & Economics</i>	2005	403	1374	3.4	0	43	234	0.58	0.17	1.92	230	0.30	0.26	0.34
<i>Journal of Sports Economics</i>	2007	237	1292	5.5	0	34	235	0.57	0.12	2.59	173	0.52	0.45	0.58
<i>Finanzarchiv</i>	2005	177	504	2.8	0	59	236	0.56	0.10	3.13	241	0.27	0.21	0.32
<i>Journal of Mathematical Economics</i>	1991	1233	7485	6.1	0	183	237	0.56	0.27	1.16	231	0.30	0.27	0.32
<i>American Journal of Economics and Sociology</i>	1991	915	3542	3.9	0	60	238	0.51	0.21	1.22	242	0.26	0.23	0.29
<i>European Journal of The History of Economic Thought</i>	2005	245	422	1.7	0	23	239	0.49	0.10	2.41	226	0.31	0.26	0.36
<i>International Journal of Transport Economics</i>	2005	149	326	2.2	0	16	240	0.45	0.06	3.33	262	0.20	0.15	0.25
<i>B E Journal of Economic Analysis & Policy</i>	2007	471	1189	2.5	0	64	241	0.42	0.12	1.53	257	0.22	0.18	0.25
<i>Investigacion Economica</i>	2006	163	84	0.5	0	16	242	0.41	0.05	3.05	287	0.06	0.03	0.10
<i>Australian Economic Review</i>	2006	315	695	2.2	0	28	243	0.41	0.08	1.92	254	0.22	0.17	0.26
<i>Economics-The Open Access Open-Assessment E-Journal</i>	2009	182	74	0.4	0	15	244	0.37	0.05	2.74	290	0.05	0.02	0.09
<i>Independent Review</i>	2005	220	346	1.6	0	19	245	0.36	0.05	2.36	276	0.15	0.12	0.18
<i>Post-Communist Economics</i>	1999	427	1480	3.5	0	31	246	0.35	0.08	1.51	239	0.27	0.23	0.30
<i>Asian Economic Policy Review</i>	2007	81	560	6.9	0	29	247	0.33	0.02	5.13	113	0.79	0.68	0.90
<i>Japan and The World Economy</i>	1993	644	2626	4.1	0	142	248	0.29	0.08	1.09	248	0.24	0.21	0.28
<i>Review of Network Economics</i>	2008	120	363	3.0	0	29	249	0.28	0.02	3.62	233	0.29	0.21	0.38

Table 9 (continued)

	Descriptive statistics						PP _{top10%}				MNCS			
	Start	Papers	Total citations	Citations per paper	Paper citations		Rank	Score	Confidence interval		Rank	Mean	Confidence interval	
					Min	Max			Lower	Upper			Lower	Upper
<i>Journal</i>														
<i>Japanese Economic Review</i>	1999	440	1465	3.3	0	60	250	0.25	0.05	1.32	259	0.21	0.18	0.24
<i>Australian Economic History Review</i>	1991	208	534	2.6	0	15	251	0.24	0.03	2.26	256	0.22	0.17	0.26
<i>Journal of World Trade</i>	1992	679	1962	2.9	0	70	252	0.24	0.06	0.98	263	0.20	0.18	0.22
<i>Developing Economies</i>	1991	424	1440	3.4	0	40	253	0.24	0.04	1.32	269	0.18	0.15	0.21
<i>Applied Economics Letters</i>	1995	4189	12,663	3.0	0	110	254	0.21	0.11	0.41	265	0.19	0.18	0.20
<i>Computational Economics</i>	2008	247	672	2.7	0	16	255	0.21	0.02	1.92	227	0.31	0.26	0.35
<i>Canadian Journal of Agricultural Economics-Revue Canadienne D Agroeconomie</i>	1991	533	2637	4.9	0	43	256	0.17	0.03	1.03	232	0.30	0.26	0.33
<i>Acta Oeconomica</i>	1991	202	253	1.3	0	25	257	0.17	0.01	2.19	279	0.12	0.09	0.16
<i>Journal of International Trade & Economic Development</i>	2007	218	457	2.1	0	18	258	0.15	0.01	2.02	237	0.27	0.22	0.32
<i>Jahrbucher Fur Nationalokonomie Und Statistik</i>	1991	887	1372	1.5	0	39	259	0.08	0.01	0.57	283	0.10	0.09	0.12
<i>South African Journal of Economics</i>	1991	778	1977	2.5	0	38	260	0.07	0.01	0.62	275	0.15	0.13	0.17
<i>Socio-economic Planning Sciences</i>	2013	24	42	1.8	0	9	261	0.01	0.00	13.82	205	0.38	0.22	0.54
<i>Asian Economic Journal</i>	2007	133	370	2.8	0	18	262	0.00	0.00	2.81	234	0.29	0.23	0.35
<i>Asian-Pacific Economic Literature</i>	2007	97	170	1.8	0	12	262	0.00	0.00	3.81	258	0.21	0.16	0.27
<i>Asia-Pacific Journal of Accounting & Economics</i>	2008	114	131	1.1	0	21	262	0.00	0.00	3.26	278	0.13	0.09	0.17
<i>Australian Economic Papers</i>	2007	137	267	1.9	0	16	262	0.00	0.00	2.73	267	0.19	0.14	0.23
<i>B E Journal of Macroeconomics</i>	2007	266	510	1.9	0	27	262	0.00	0.00	1.42	268	0.18	0.15	0.22
<i>B E Journal of Theoretical Economics</i>	2007	237	301	1.3	0	15	262	0.00	0.00	1.60	282	0.11	0.09	0.13
<i>Baltic Journal of Economics</i>	2007	60	69	1.2	0	7	262	0.00	0.00	6.02	273	0.15	0.10	0.20
<i>Bulletin of Economic Research</i>	2008	154	286	1.9	0	12	262	0.00	0.00	2.43	246	0.25	0.19	0.30
<i>Bulletin of Indonesian Economic Studies</i>	1991	354	2290	6.5	0	36	262	0.00	0.00	1.07	219	0.35	0.32	0.39
<i>Cepal Review</i>	2007	144	64	0.4	0	4	262	0.00	0.00	2.60	291	0.05	0.03	0.06
<i>Eastern European Economics</i>	1991	533	1116	2.1	0	24	262	0.00	0.00	0.72	271	0.16	0.14	0.18

Table 9 (continued)

Journal	Descriptive statistics						PP _{top10%}				MNCS			
	Start	Papers	Total citations	Citations per paper	Paper citations		Score	Confidence interval		Rank	Mean	Confidence interval		
					Min	Max		Lower	Upper			Lower	Upper	
<i>Economia Mexicana-Nueva Epoca</i>	2009	58	14	0.2	0	3	0.00	0.00	6.21	293	0.04	0.01	0.06	
<i>Economia Politica</i>	2007	132	235	1.8	0	23	0.00	0.00	2.83	250	0.24	0.18	0.29	
<i>Economic and Social Review</i>	1991	307	1002	3.3	0	34	0.00	0.00	1.24	253	0.22	0.19	0.25	
<i>Estudios De Economia</i>	2007	76	116	1.5	0	9	0.00	0.00	4.81	272	0.15	0.11	0.20	
<i>Global Economic Review</i>	2008	145	233	1.6	0	14	0.00	0.00	2.58	264	0.19	0.15	0.24	
<i>Hitoisubashi Journal of Economics</i>	1991	262	305	1.2	0	18	0.00	0.00	1.45	286	0.07	0.05	0.08	
<i>Ikrisat Isteme Ve Finans</i>	2009	249	219	0.9	0	8	0.00	0.00	1.52	280	0.11	0.09	0.13	
<i>International Journal of Economic Theory</i>	2008	118	245	2.1	0	13	0.00	0.00	3.15	240	0.27	0.22	0.32	
<i>Journal of Australian Political Economy</i>	2007	112	160	1.4	0	13	0.00	0.00	3.32	274	0.15	0.12	0.18	
<i>Journal of Behavioral Finance</i>	2008	130	311	2.4	0	15	0.00	0.00	2.87	245	0.25	0.20	0.30	
<i>Journal of The Asia Pacific Economy</i>	2007	215	487	2.3	0	21	0.00	0.00	1.76	247	0.25	0.20	0.29	
<i>Recherches Economiques De Louvain-Louvain Economic Review</i>	2008	95	56	0.6	0	4	0.00	0.00	3.89	285	0.07	0.05	0.09	
<i>Review of Derivatives Research</i>	2008	65	151	2.3	0	14	0.00	0.00	5.58	255	0.22	0.16	0.28	
<i>Review of Economic Design</i>	2008	102	189	1.9	0	13	0.00	0.00	3.63	251	0.24	0.17	0.31	
<i>Revista de Economia Mundial</i>	1999	309	74	0.2	0	12	0.00	0.00	1.23	294	0.03	0.02	0.04	
<i>Revista de Historia Economica</i>	2008	89	137	1.5	0	13	0.00	0.00	4.14	266	0.19	0.13	0.24	
<i>Revista de Historia Industrial</i>	2009	67	28	0.4	0	6	0.00	0.00	5.42	288	0.06	0.03	0.09	
<i>Revue d'Economie Politique</i>	2005	304	203	0.7	0	19	0.00	0.00	1.25	289	0.05	0.04	0.07	
<i>Romanian Journal of Economic Forecasting</i>	2008	277	398	1.4	0	15	0.00	0.00	1.37	270	0.17	0.14	0.20	
<i>Singapore Economic Review</i>	2007	212	224	1.1	0	14	0.00	0.00	1.78	281	0.11	0.08	0.13	
<i>South African Journal of Economic and Management Sciences</i>	2006	267	257	1.0	0	11	0.00	0.00	1.42	284	0.09	0.07	0.11	
<i>Trimestre Economico</i>	1993	486	318	0.7	0	36	0.00	0.00	0.78	292	0.04	0.03	0.05	

The table is sorted in decreasing order by PP_{top 10%}. The journals are classified into four CSS classes: outstandingly, remarkably, fairly, and poorly cited

Appendix 2

Calculation of the Mean Normalised Citation Score (MNCS)

For the calculation of the MNCS, each paper’s citations in a paper set (of a journal, researcher, institution, or country) are divided by the mean citation impact in a corresponding reference set; the received NCSs are averaged to the MNCS. Table 10 shows how the MNCs are calculated for two fictitious journals. For example, the NCS for paper number 2 is $3/10.67=0.28$; the MNCS for journal B is $(0.28 + 1.00)/2=0.64$. The MNCS is formally defined as (Waltman et al. 2011)

$$MNCS = \frac{1}{n} \sum_{i=1}^n \frac{c_i}{e_i}$$

where c_i is the citation count of a focal paper and e_i is the corresponding expected number of citations in the economic subfield (JEL code). The MNCS is defined similar to the item-oriented field-normalised citation score average indicator (Lundberg 2007; Rehn et al. 2007). Since citation counts depend on the length of time between the publication year of the cited papers and the time point of the impact analysis (see Table 3), the normalisation is performed separately for each publication year.

It is a nice property of the MNCS that it leads to an average value of 1. However, this is only valid in a paper set (with papers from one year) if each paper is assigned to one field. However, many of the field classification systems (e.g. JEL codes) assign papers to more than one field. Table 10 shows a simple example that illustrates the problem with the multi-assignment of papers. Paper number 5 is assigned to two fields. The obvious solution for the calculation of the NCS would be to calculate an average of two ratios for this paper: $((9/10.67) + (9/8.5))/2=0.95$. However, this solution leads to an average value of greater than 1 (1.01) across the five papers in Table 10.

In order to solve this problem, Waltman et al. (2011) propose the following two calculations, which ensure a mean value of 1 (see Table 11): (1) The expected number of

Table 10 Case study demonstrating the calculation of the mean normalized citation score (MNCS) and the problems with the normalisation of citation counts if papers are assigned to more than one field

Paper number	Number of papers	Field	Citations	Number of fields	Normalised citation score	Journal A	Journal B
1	1	X	20	1	1.88	1	
2	1	X	3	1	0.28		1
3	1	Y	8	1	0.94	1	
4	1	Z	6	1	1.00		1
5	1	X and Y	9	2	0.95	1	
Total					1.01	3	2
Expected number of citations		X	10.67				
		Y	8.50				
		Z	6.00				
MNCS Journal A						1.26	
MNCS Journal B							0.64

Table 11 Example to demonstrate the solution of the problems illustrated in Table 10

Paper	Number of papers	Field	Citations	Field fraction	Normalised impact (NCS)	Journal A	Journal B
1	1	X	20	1	1.82	1	
2	1	X	3	1	0.27		1
3	1	Y	8	1	0.96	1	
4	1	Z	6	1	1.00		1
5	1	X and Y	9	0.5	0.95	1	
					1.00	3	2
Expected number of citations		X	11.00				
		Y	8.33				
		Z	6.00				
MNCS Journal A						1.24	
MNCS Journal B							0.64

citations for field X is calculated as follows: $(20 + 3 + (9 \cdot 0.5)) / (1 + 1 + 0.5) = 11$. Thus, the citations of paper 5 are fractionally counted; the calculation for field Y is correspondingly: $(8 + (9 \cdot 0.5)) / (1 + 0.5)$. (2) The NCS for paper 5 also considers its fractional assignment to two fields and is calculated as follows: $(9/11 \cdot 0.5) + (9/8.33 \cdot 0.5)$. Both calculations lead to the desired property of the indicator that it results in a mean value of 1 across all papers in a field—although the papers might be assigned to more than one field.

Calculation of the percentile based indicator: $PP_{top\ 10\%}$ (and $PP_{top\ 50\%}$)

Table 12 uses an example dataset to demonstrate how the $PP_{top\ 10\%}$ indicator is calculated. Basically, the indicator is generated on the basis of the citation distribution in a field (here: field A) whereby the papers are sorted in decreasing order of citations. Papers belonging to the 10% most frequently cited papers are assigned the score 1 and the others the score 0 in a binary variable. The binary variable can then be used to calculate the $P_{top\ 10\%}$ or $PP_{top\ 10\%}$ indicators. $P_{top\ 10\%}$ is the absolute number of papers published in field A belonging to the 10% most frequently cited papers (here: 1) and $PP_{top\ 10\%}$ the relative number whereas $P_{top\ 10\%}$ is divided by the total number of papers ($1/10 \cdot 100 = 10$). If a journal (here: journal X) had published 6 papers from field A (and no further papers in other fields), $P_{top\ 10\%} = 1$ and $PP_{top\ 10\%} = 16.7\%$ ($1/6 \cdot 100$).

The $PP_{top\ 10\%}$ indicator is concerned by two problems, whereby the solution for the first problem is outlined in Table 13. Citation distributions are characterized by ties, i.e. papers having the same number of citations. The ties lead to problems in identifying the 10% most frequently cited papers, if the ties concern papers around the threshold of 10% in a citation distribution. We explain the problem and the solution based on the $PP_{top\ 50\%}$ indicator, because the use of this indicator needs to include fewer papers in an example than the $PP_{top\ 10\%}$. However, the procedure is the same with $PP_{top\ 10\%}$.

In Table 13, the 7 papers with 20 citations can be clearly assigned to the 50% most frequently cited papers and the 5 papers with 0 citations to the rest. However, this is not

Table 12 Fictitious example with 10 papers in field A to demonstrate the calculation of $PP_{top\ 10\%}$

Citations	Paper	$P_{top\ 10\%}$	Journal X
25	1	1	1
21	1	0	1
19	1	0	1
17	1	0	1
16	1	0	
14	1	0	1
12	1	0	1
9	1	0	
1	1	0	
0	1	0	
Number	10	10	6
$P_{top\ 10\%}$		1	1
$PP_{top\ 10\%}$		10	16.7

possible for the 6 papers with 10 citations; they cannot be clearly assigned to one of both groups. Waltman and Schreiber (2013) propose a solution for this problem, which leads to exactly 50% most frequently cited papers in a field despite the existence of papers with the same number of citations (around the threshold). We explain their solution using the example data in Table 13.

Each of the 18 papers in field B represents $1/18 = 5.56\%$ of the field-specific citation distribution. Hence, together the 7 papers with 20 citations represent $7 * 5.56\% = 38.92\%$ of the citation distribution, the 6 papers with 10 citations represent $6 * 5.56\% = 33.36\%$ of

Table 13 Fictitious example with 18 papers in field B to demonstrate the calculation of $PP_{top\ 50\%}$ following the approach of Waltman and Schreiber (2013)

Citations	Paper	$P_{top\ 50\%}$
20	1	1
20	1	1
20	1	1
20	1	1
20	1	1
20	1	1
20	1	1
10	1	0.33
10	1	0.33
10	1	0.33
10	1	0.33
10	1	0.33
10	1	0.33
0	1	0
0	1	0
0	1	0
0	1	0
0	1	0
Sum	18	9

the citation distribution, and the 5 papers with 0 citations represent $5 * 5.56\% = 27.8\%$. We would like to identify the 50% most frequently cited papers, whereby the 6 papers with 10 citations are still unclear. Waltman and Schreiber (2013) fractionally assign these papers to the 50% most frequently cited papers, so that we end up with 50% 50% most frequently cited papers.

The 7 papers with 20 citations cover 38.92% of the 50% most frequently cited papers. The rest ($50\% - 38.92\% = 11.08\%$) needs to be covered by the 6 papers with 10 citations. In order to reach this goal, the segment of the citation distribution covered by the papers with 10 citations must be split into two parts, one part covering 11.08% of the distribution, the other part covering the remaining $33.36 - 11.08\% = 22.28\%$. This other part (22.28%) belongs to the bottom 50% of the citation distribution. Splitting the segment of the distribution covered by papers with 10 citations is done by assigning each of the 6 papers to the 50% most frequently cited papers with a fraction of $11.08\%/33.36\% = 0.33$. The value 11.08% represents the share of the papers with 10 citations, which belong to the 50% most frequently cited papers; 33.36% is the percentage of papers in the field with 10 citations.

In this way, we obtain 50% 50% most frequently cited papers, since $((0.33 * 6) + 7) / 18$ equals 50%. There are 6 papers in the field with 10 citations, which are fractionally assigned to the 50% most frequently cited papers, and 7 papers with 20 citations that clearly belong to the 50% most frequently cited papers.

Table 14 shows an example that reveals the second problem with the $PP_{top\ 50\%}$ indicator: papers are assigned not only to one, but to two or more fields. The example in Table 14 consists of 16 papers whereby 1 paper belongs to two fields. In these cases, the papers in multiple fields are fractionally counted for the calculation of $PP_{top\ 50\%}$ following the approach of Waltman et al. (2011).

We explain the approach using the example in Table 14. Since 1 paper in the table belongs to two fields (B and C), it is weighted by 0.5 instead of 1 (the other papers in the sets which belong to one field each are weighted with 1). This leads to 15.5 papers in field B and 10.5 papers in field C.

In field B, the papers with 20 citations represent 29.03% of the citation distribution ($4.5/15.5$), the papers with 10 citations 38.71% ($6/15.5$), and the papers with 0 citations 32.26% ($5/15.5$). Thus, the papers with 20 citations cover 29.03% of the 50% most frequently cited papers. The rest with 20.97% ($50\% - 29.03\%$) should be covered by the 6 papers with 10 citations. Splitting the segment of the distribution covered by papers with 10 citations is done by assigning each of the 6 papers to the 50% most frequently cited papers with a fraction of $20.97\%/38.71\% = 0.54$. Thus, we obtain 50% 50% most frequently cited papers since $((0.54 * 6) + 4.5) / 15.5$ equals 50%.

In field C with a total of 10.5 papers, we have 3 papers with 50 citations (28.57% of the citation distribution), 4.5 papers with 20 citations (42.86% of the distribution), and 3 papers with 10 citations (28.57%). 21.43% of the citation distribution ($50 - 28.57\%$) should be covered by the papers with 20 citations: $21.43\%/42.86\% = 0.5$. We receive the value of 50% with $((0.5 * 4.5) + 3) / 10.5$.

In Table 15, the data from Table 14 are used to transfer the calculations for two different fields (B and C) towards a small world example in which only two journals exist (Y and Z) publishing all the papers in fields B and C. Journal Y has published 16 papers and journal Z 10 (the small world consists of 26 papers). Whereas 25 papers belong to one field (B or C), 1 paper belongs to two fields (B and C). If papers belong to multiple fields, $P_{top\ 50\%}$ from both fields is added up by considering the paper fractions. For the paper belonging to both fields in Table 15, $P_{top\ 50\%}$ is calculated as follows: $(0.5 * 1) + (0.5 * 0.5) = 0.75$.

Table 14 Fictitious example for field B with 16 papers and field C with 11 papers to demonstrate the calculation of $PP_{top\ 50\%}$ if papers are fractionally counted following the approach of Waltman and Schreiber (2013) and Waltman et al. (2011)

Field B	Citations	Paper	Number of fields	Paper fraction	$P_{top\ 50\%}$
	20	1	2	0.5	1
	20	1	1	1	1
	20	1	1	1	1
	20	1	1	1	1
	20	1	1	1	1
	10	1	1	1	0.54
	10	1	1	1	0.54
	10	1	1	1	0.54
	10	1	1	1	0.54
	10	1	1	1	0.54
	10	1	1	1	0.54
	0	1	1	1	0
	0	1	1	1	0
	0	1	1	1	0
	0	1	1	1	0
	0	1	1	1	0
Sum		16		15.5	7.75
Field C	Citations	Paper	Number of fields	Paper fraction	$P_{top\ 50\%}$
	50	1	1	1	1
	50	1	1	1	1
	50	1	1	1	1
	20	1	2	0.5	0.5
	20	1	1	1	0.5
	20	1	1	1	0.5
	20	1	1	1	0.5
	20	1	1	1	0.5
	10	1	1	1	0
	10	1	1	1	0
	10	1	1	1	0
Sum		11		10.5	5.25

If the $P_{top\ 50\%}$ scores for the papers belonging to journals Y and Z are added up each, this gives the $P_{top\ 50\%}$ scores for the journals. It equals 8.42 for journal Y. Thus, 8.42 papers published by the journal belong to the 50% most frequently cited papers. This results in $PP_{top\ 50\%} = 52.6\%$ ($8.42/16$). The results in Table 15 show that journal Y has published more above-average papers than journal Z with 45.83%. Together, both journals have published 50% 50% most frequently cited papers: $((52.6\% * 16) + (45.83\% * 10)) / (16 + 10)$.

Table 15 Fictitious example including two journals (Y and Z) using data from Table 14

Field	$P_{\text{top } 50\%}$	Journal Y	Journal Z
B/C	0.75	1	
B	1		1
B	1	1	
B	1	1	
B	1		1
B	0.54	1	
B	0.54		1
B	0.54	1	
B	0.54		1
B	0.54	1	
B	0.54	1	
B	0		1
B	0		1
B	0	1	
B	0		1
B	0	1	
C	1		1
C	1	1	
C	1	1	
C	0.5	1	
C	0.5		1
C	0.5	1	
C	0.5	1	
C	0	1	
C	0	1	
C	0		1
Sum	13	16	10
$P_{\text{top } 50\%}$		8.42	4.58
$PP_{\text{top } 50\%}$		52.6	45.83

The table is sorted in decreasing order by $PP_{\text{top } 10\%}$. The journals are classified into four CSS classes: outstandingly, remarkably, fairly, and poorly cited

References

- Abramo, G., Cicero, T., & D'Angelo, C. A. (2011). Assessing the varying level of impact measurement accuracy as a function of the citation window length. *Journal of Informetrics*, 5(4), 659–667. <https://doi.org/10.1016/j.joi.2011.06.004>.
- Acemoglu, D., Johnson, S., & Robinson, J. A. (2001). The colonial origins of comparative development: An empirical investigation. *American Economic Review*, 91(5), 1369–1401. <https://doi.org/10.1257/aer.91.5.1369>.
- Altman, D., Machin, D., Bryant, T., & Gardner, M. (2013). *Statistics with confidence: Confidence intervals and statistical guidelines*. London: BMJ Books.
- Anauati, V., Galliani, S., & Galvez, R. H. (2016). Quantifying the life cycle of scholarly articles across field of economic research. *Economic Inquiry*, 54(2), 1339–1355.
- Angrist, J., Azoulay, P., Ellison, G., Hill, R., & Lu, S. F. (2017a). Economic research evolves: Fields and styles. *American Economic Review*, 107(5), 293–297. <https://doi.org/10.1257/aer.p20171117>.

- Angrist, J., Azoulay, P., Ellison, G., Hill, R., & Lu, S. F. (2017b). *Inside job or deep impact? Using extramural citations to assess economic scholarship*. NBER working paper no. 23698. Cambridge, MA: National Bureau of Economic Research.
- Arellano, M., & Bond, S. (1991). Some tests of specification for panel data—monte-carlo evidence and an application to employment equations. *Review of Economic Studies*, 58(2), 277–297. <https://doi.org/10.2307/2297968>.
- Bornmann, L. (2011). Scientific peer review. *Annual Review of Information Science and Technology*, 45, 199–245.
- Bornmann, L. (in press). Bibliometric indicators—Methods for measuring science. In R. Williams (Ed.), *Encyclopedia of research methods*. Thousand Oaks, CA: Sage.
- Bornmann, L., Butz, A., & Wohlrabe, K. (2018). What are the top five journals in economics? A new meta-ranking. *Applied Economics*, 50(6), 659–675.
- Bornmann, L., & Daniel, H.-D. (2008). Selecting manuscripts for a high impact journal through peer review: A citation analysis of communications that were accepted by *Angewandte Chemie-International Edition*, or rejected but published elsewhere. *Journal of the American Society for Information Science and Technology*, 59(11), 1841–1852. <https://doi.org/10.1002/asi.20901>.
- Bornmann, L., & Glänzel, W. (2017). Applying the css method to bibliometric indicators used in (university) rankings. *Scientometrics*, 110(2), 1077–1079. <https://doi.org/10.1007/s11192-016-2198-5>.
- Bornmann, L., & Haunschild, R. (2017). An empirical look at the nature index. *Journal of the Association of Information Science and Technology*, 68(3), 653–659. <https://doi.org/10.1002/asi.23682>.
- Bornmann, L., & Leydesdorff, L. (2017). Skewness of citation impact data and covariates of citation distributions: A large-scale empirical analysis based on web of science data. *Journal of Informetrics*, 11(1), 164–175.
- Bornmann, L., & Marx, W. (2014a). How to evaluate individual researchers working in the natural and life sciences meaningfully? A proposal of methods based on percentiles of citations. *Scientometrics*, 98(1), 487–509. <https://doi.org/10.1007/s11192-013-1161-y>.
- Bornmann, L., & Marx, W. (2014b). The wisdom of citing scientists. *Journal of the American Society of Information Science and Technology*, 65(6), 1288–1292.
- Bornmann, L., Mutz, R., Neuhaus, C., & Daniel, H.-D. (2008). Use of citation counts for research evaluation: Standards of good practice for analyzing bibliometric data and presenting and interpreting results. *Ethics in Science and Environmental Politics*, 8, 93–102. <https://doi.org/10.3354/esepe00084>.
- Bornmann, L., Schier, H., Marx, W., & Daniel, H.-D. (2011). Is interactive open access publishing able to identify high-impact submissions? A study on the predictive validity of *Atmospheric Chemistry and Physics* by using percentile rank classes. *Journal of the American Society for Information Science and Technology*, 62(1), 61–71.
- Bornmann, L., Stefaner, M., de Moya Anegón, F., & Mutz, R. (2014). What is the effect of country-specific characteristics on the research performance of scientific institutions? Using multi-level statistical models to rank and map universities and research-focused institutions worldwide. *Journal of Informetrics*, 8(3), 581–593. <https://doi.org/10.1016/j.joi.2014.04.008>.
- Boyack, K. W. (2004). Mapping knowledge domains: Characterizing PNAS. *Proceedings of the National Academy of Sciences of the United States of America*, 101, 5192–5199.
- Card, D., & DellaVigna, S. (2013). Nine facts about top journals in economics. *Journal of Economic Literature*, 51(1), 144–161. <https://doi.org/10.1257/jel.51.1.144>.
- Cherrier, B. (2017). Classifying economics: A history of JEL codes. *Journal of Economic Literature*, 55(2), 545–579.
- Claveau, F. (2016). There should not be any mystery: A comment on sampling issues in bibliometrics. *Journal of Informetrics*, 10(4), 1233–1240. <https://doi.org/10.1016/j.joi.2016.09.009>.
- Coats, A. (1971). The role of scholarly journals in the history of economics: An essay. *Journal of Economic Literature*, 9(1), 29–44.
- Combes, P.-P., & Linnemer, L. (2010). *Inferring missing citations: A quantitative multi-criteria ranking of all journals in economics*. Working papers halshs-00520325, hal.
- Council of Canadian Academies. (2012). *Informing research choices: Indicators and judgment. The expert panel on science performance and research funding*. Ottawa: Council of Canadian Academies.
- Crespo, J. A., Herranz, N., Li, Y., & Ruiz-Castillo, J. (2014). The effect on citation inequality of differences in citation practices at the web of science subject category level. *Journal of the Association for Information Science and Technology*, 65(6), 1244–1256. <https://doi.org/10.1002/asi.23006>.
- Cumming, G. (2012). *Understanding the new statistics: Effect sizes, confidence intervals, and meta-analysis*. London: Routledge.
- Cumming, G., & Calin-Jageman, R. (2016). *Introduction to the new statistics: Estimation, open science, and beyond*. London: Taylor & Francis.

- Ellison, G. (2002). The slowdown of the economics publishing process. *Journal of Political Economy*, 110(5), 947–993. <https://doi.org/10.1086/341868>.
- Ellison, G. (2013). How does the market use citation data? The hirsch index in economics. *American Economic Journal-Applied Economics*, 5(3), 63–90. <https://doi.org/10.1257/app.5.3.63>.
- Ferrara, A., & Bonaccorsi, A. (2016). How robust is journal rating in humanities and social sciences? Evidence from a large-scale, multi-method exercise. *Research Evaluation*, 25(3), 279–291. <https://doi.org/10.1093/reseval/rvv048>.
- Galton, F. (1907). Vox populi. *Nature*, 75, 450–451. <https://doi.org/10.1038/075450a0>.
- Garfield, E. (2006). The history and meaning of the journal impact factor. *Journal of the American Medical Association*, 295(1), 90–93.
- Gevers, M. (2014). Scientific performance indicators: A critical appraisal and a country-by-country analysis. In W. Blockmans, L. Engwall, & D. Weaire (Eds.), *Bibliometrics: Use and abuse in the review of research performance* (pp. 43–53). London: Portland Press.
- Gibson, J., Anderson, D. L., & Tressler, J. (2014). Which journal rankings best explain academic salaries? Evidence from the University of California. *Economic Inquiry*, 52(4), 1322–1340. <https://doi.org/10.1111/ecin.12107>.
- Gibson, J., Anderson, D. L., & Tressler, J. (2017). Citations or journal quality: Which is rewarded more in the academic labor market? *Economic Inquiry*, 55(4), 1945–1965. <https://doi.org/10.1111/ecin.12455>.
- Glänzel, W. (2008). Seven myths in bibliometrics. About facts and fiction in quantitative science studies. *Paper presented at the fourth international conference on webometrics, informetrics and scientometrics (WIS 2008) & ninth COLLNET meeting, Berlin, Germany*.
- Glänzel, W., & Schubert, A. (1988). Characteristic scores and scales in assessing citation impact. *Journal of Information Science*, 14(2), 123–127.
- Glänzel, W., Thijs, B., Schubert, A., & Debackere, K. (2009). Subfield-specific normalized relative indicators and a new generation of relational charts: Methodological foundations illustrated on the assessment of institutional research performance. *Scientometrics*, 78, 165–188.
- Gwet, K. L. (2014). *Handbook of inter-rater reliability, 4th edition: The definitive guide to measuring the extent of agreement among raters*. Washington DC: Advanced Analytics LLC.
- Haddow, G., & Noyons, E. (2013). Misfits? Research classification in research evaluation: Visualizing journal content within fields of research codes. *Paper presented at the 14th International Conference of the International Society for Scientometrics and Informetrics Vienna (ISSI 2013), Austria, July 15-19, 2013*.
- Hamermesh, D. S. (2018). Citations in economics: measurement, uses, and impacts. *Journal of Economic Literature*, 56(1), 115–156.
- Haunschild, R., & Bornmann, L. (2015). Criteria for nature index questioned. *Nature*, 517(7532), 21.
- Hicks, D., Wouters, P., Waltman, L., de Rijcke, S., & Rafols, I. (2015). Bibliometrics: The Leiden manifesto for research metrics. *Nature*, 520(7548), 429–431.
- Hirsch, J. E. (2005). An index to quantify an individual's scientific research output. *Proceedings of the National Academy of Sciences of the United States of America*, 102(46), 16569–16572. <https://doi.org/10.1073/pnas.0507655102>.
- Johnston, D. W., Piatti, M., & Torgler, B. (2013). Citation success over time: Theory or empirics? *Scientometrics*, 95(3), 1023–1029. <https://doi.org/10.1007/s11192-012-0910-7>.
- Kosnik, L.-R. (2018). A survey of JEL codes: What do they mean and are they used consistently? *Journal of Economic Surveys*, 32(1), 249–272. <https://doi.org/10.1111/joes.12189>.
- Kostoff, R. N. (2002). Citation analysis of research performer quality. *Scientometrics*, 53(1), 49–71.
- Kreiman, G., & Maunsell, J. H. R. (2011). Nine criteria for a measure of scientific output. *Frontiers in Computational Neuroscience*, 5, 48. <https://doi.org/10.3389/fncom.2011.00048>.
- Landis, J. R., & Koch, G. G. (1977). Measurement of observer agreement for categorical data. *Biometrics*, 33(1), 159–174.
- Lawson, T. (2013). What is this 'school' called neoclassical economics? *Cambridge Journal of Economics*, 37(5), 947–983. <https://doi.org/10.1093/cje/bet027>.
- Leydesdorff, L., & Opthof, T. (2013). Citation analysis with medical subject headings (MeSH) using the web of knowledge: A new routine. *Journal of the American Society for Information Science and Technology*, 64(5), 1076–1080. <https://doi.org/10.1002/asi.22770>.
- Li, Y., & Ruiz-Castillo, J. (2014). The impact of extreme observations in citation distributions. *Research Evaluation*, 23(2), 174–182. <https://doi.org/10.1093/reseval/rvu006>.
- Liebowitz, S. J., & Palmer, J. P. (1984). Assessing the relative impact of economics journals. *Journal of Economic Literature*, 22(1), 77–88.
- Linnemer, L., & Visser, M. (2016). The most cited articles from the top-5 journals (1991–2015). In *CESifo working paper series nr. 5999*. Munich: CESifo Group.

- Lundberg, J. (2007). Lifting the crown—Citation z -score. *Journal of Informetrics*, 1(2), 145–154.
- Martin, B. R., & Irvine, J. (1983). Assessing basic research—Some partial indicators of scientific progress in radio astronomy. *Research Policy*, 12(2), 61–90.
- Marx, W., & Bornmann, L. (2015). On the causes of subject-specific citation rates in web of science. *Scientometrics*, 102(2), 1823–1827. <https://doi.org/10.1007/s11192-014-1499-9>.
- McAllister, P. R., Narin, F., & Corrigan, J. G. (1983). Programmatic evaluation and comparison based on standardized citation scores. *IEEE Transactions on Engineering Management*, 30(4), 205–211.
- Mingers, J., & Leydesdorff, L. (2015). A review of theory and practice in scientometrics. *European Journal of Operational Research*, 246(1), 1–19. <https://doi.org/10.1016/j.ejor.2015.04.002>.
- Moed, H. (2015). Comprehensive indicator comparisons intelligible to non-experts: The case of two SNIP versions. *Scientometrics*, 106, 1–15. <https://doi.org/10.1007/s11192-015-1781-5>.
- Moed, H. F., & Halevi, G. (2015). Multidimensional assessment of scholarly research impact. *Journal of the Association for Information Science and Technology*, 66(10), 1988–2002. <https://doi.org/10.1002/asi.23314>.
- Mutz, R. (2016). Some further aspects of sampling: Comment on Williams and Bornmann. *Journal of Informetrics*, 10(4), 1241–1242. <https://doi.org/10.1016/j.joi.2016.09.007>.
- National Research Council. (2010). *A revised guide to the methodology of the data-based assessment of research-doctorate programs in the united states*. Washington, DC: The National Academies Press.
- Ophthof, T. (2011). Differences in citation frequency of clinical and basic science papers in cardiovascular research. *Medical & Biological Engineering & Computing*, 49(6), 613–621. <https://doi.org/10.1007/s11517-011-0783-6>.
- Palacios-Huerta, I., & Volij, O. (2004). The measurement of intellectual influence. *Econometrica*, 72(3), 963–977. <https://doi.org/10.1111/j.1468-0262.2004.00519.x>.
- Perry, M., & Reny, P. J. (2016). How to count citations if you must. *American Economic Review*, 106(9), 2722–2741.
- Podlubny, I. (2005). Comparison of scientific impact expressed by the number of citations in different fields of science. *Scientometrics*, 64(1), 95–99.
- Radicchi, F., & Castellano, C. (2011). Rescaling citations of publications in physics. *Physical Review E*, 83(4), 046116. <https://doi.org/10.1103/physreve.83.046116>.
- Rehn, C., Kronman, U., & Wadskog, D. (2007). *Bibliometric indicators—Definitions and usage at karolinska institutet*. Stockholm: Karolinska Institutet University Library.
- Research Evaluation and Policy Project. (2005). *Quantitative indicators for research assessment—A literature review (REPP discussion paper 05/1)*. Canberra, Australia: Research Evaluation and Policy Project, Research School of Social Sciences, The Australian National University.
- Ruiz-Castillo, J. (2012). The evaluation of citation distributions. *SERIEs: Journal of the Spanish Economic Association*, 3(1), 291–310. <https://doi.org/10.1007/s13209-011-0074-3>.
- Sandström, U. (2014). *Bibliometric evaluation of sepa-funded large research programs 2003–2013*. Stockholm: Swedish Environmental Protection Agency.
- Schneider, J. W. (2016). The imaginarium of statistical inference when data are the population: Comments to Williams and Bornmann. *Journal of Informetrics*, 10(4), 1243–1248. <https://doi.org/10.1016/j.joi.2016.09.011>.
- Schubert, A., & Braun, T. (1986). Relative indicators and relational charts for comparative assessment of publication output and citation impact. *Scientometrics*, 9(5–6), 281–291.
- Seglen, P. O. (1992). The skewness of science. *Journal of the American Society for Information Science*, 43(9), 628–638.
- Smolinsky, L., & Lercher, A. (2012). Citation rates in mathematics: A study of variation by subdiscipline. *Scientometrics*, 91(3), 911–924. <https://doi.org/10.1007/s11192-012-0647-3>.
- Stern, D. I. (2013). Uncertainty measures for economics journal impact factors. *Journal of Economic Literature*, 51(1), 173–189. <https://doi.org/10.1257/jel.51.1.173>.
- Strotmann, A., & Zhao, D. (2010). Combining commercial citation indexes and open-access bibliographic databases to delimit highly interdisciplinary research fields for citation analysis. *Journal of Informetrics*, 4(2), 194–200. <https://doi.org/10.1016/j.joi.2009.12.001>.
- Thelwall, M. (2017). Three practical field normalised alternative indicator formulae for research evaluation. *Journal of Informetrics*, 11(1), 128–151. <https://doi.org/10.1016/j.joi.2016.12.002>.
- Thomson Reuters. (2015). *InCites indicators handbook*. Philadelphia, PA: Thomson Reuters.
- van Leeuwen, T. N., & Calero Medina, C. (2012). Redefining the field of economics: Improving field normalization for the application of bibliometric techniques in the field of economics. *Research Evaluation*, 21(1), 61–70. <https://doi.org/10.1093/reseval/rvr006>.

- Vinkler, P. (1986). Evaluation of some methods for the relative assessment of scientific publications. *Scientometrics*, 10(3–4), 157–177.
- Vinkler, P. (2010). *The evaluation of research by scientometric indicators*. Oxford: Chandos Publishing.
- Waltman, L. (2016a). A review of the literature on citation impact indicators. *Journal of Informetrics*, 10(2), 365–391.
- Waltman, L. (2016b). Special section on statistical inference in citation analysis. *Journal of Informetrics*, 10(4), 1224. <https://doi.org/10.1016/j.joi.2016.09.008>.
- Waltman, L., Calero-Medina, C., Kosten, J., Noyons, E. C. M., Tijssen, R. J. W., van Eck, N. J., et al. (2012). The Leiden ranking 2011/2012: Data collection, indicators, and interpretation. *Journal of the American Society for Information Science and Technology*, 63(12), 2419–2432.
- Waltman, L., & Schreiber, M. (2013). On the calculation of percentile-based bibliometric indicators. *Journal of the American Society for Information Science and Technology*, 64(2), 372–379.
- Waltman, L., & van Eck, N. J. (2013a). Source normalized indicators of citation impact: An overview of different approaches and an empirical comparison. *Scientometrics*, 96(3), 699–716. <https://doi.org/10.1007/s11192-012-0913-4>.
- Waltman, L., & van Eck, N. J. (2013b). A systematic empirical comparison of different approaches for normalizing citation impact indicators. *Journal of Informetrics*, 7(4), 833–849.
- Waltman, L., van Eck, N. J., van Leeuwen, T. N., Visser, M. S., & van Raan, A. F. J. (2011). Towards a new crown indicator: Some theoretical considerations. *Journal of Informetrics*, 5(1), 37–47. <https://doi.org/10.1016/j.joi.2010.08.001>.
- Wang, J. (2013). Citation time window choice for research impact evaluation. *Scientometrics*, 94(3), 851–872. <https://doi.org/10.1007/s11192-012-0775-9>.
- Wang, Q., & Waltman, L. (2016). Large-scale analysis of the accuracy of the journal classification systems of web of science and scopus. *Journal of Informetrics*, 10(2), 347–364.
- Williams, R., & Bornmann, L. (2016). Sampling issues in bibliometric analysis. *Journal of Informetrics*, 10(4), 1253–1257.
- Wilsdon, J., Allen, L., Belfiore, E., Campbell, P., Curry, S., Hill, S., et al. (2015). *The metric tide: Report of the independent review of the role of metrics in research assessment and management*. Bristol: Higher Education Funding Council for England (HEFCE).
- Wilson, D. S., & Gowdy, J. M. (2013). Evolution as a general theoretical framework for economics and public policy. *Journal of Economic Behavior & Organization*, 90, S3–S10. <https://doi.org/10.1016/j.jebo.2012.12.008>.
- Wouters, P., Thelwall, M., Kousha, K., Waltman, L., de Rijcke, S., Rushforth, A., et al. (2015). *The metric tide: Literature review (supplementary report i to the independent review of the role of metrics in research assessment and management)*. London: Higher Education Funding Council for England (HEFCE).
- Zimmermann, C. (2013). Academic rankings with repec. *Econometrics*, 1(3), 249–280.